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Working Paper

Gender Identity and Economic Decision Making

CESifo Working Paper, No. 11612

Provided in Cooperation with:

Ifo Institute – Leibniz Institute for Economic Research at the University of Munich

Suggested Citation: Brenøe, Anne Ardila; Eyibak, Zeynep; Heursen, Lea; Ranehill, Eva; Weber, Roberto A. (2025) : Gender Identity and Economic Decision Making, CESifo Working Paper, No. 11612, CESifo GmbH, Munich

This Version is available at:

<https://hdl.handle.net/10419/314651>

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Impressum:

CESifo Working Papers

ISSN 2364-1428 (electronic version)

Publisher and distributor: Munich Society for the Promotion of Economic Research - CESifo GmbH

The international platform of Ludwigs-Maximilians University's Center for Economic Studies and the ifo Institute

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Editor: Clemens Fuest

<https://www.cesifo.org/en/wp>

An electronic version of the paper may be downloaded

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Gender Identity and Economic Decision Making

Abstract

Economic research on gender gaps has focused on variation based on the binary classification of “men” and “women”. We explore whether a self-reported continuous measure of gender identity (CGI) explains variation in economic decisions and outcomes beyond the relationship with binary gender. We analyze data from four diverse populations (N=8,018), including measures of economic preferences and educational and labor market outcomes. We find that CGI is significantly associated with economic outcomes, with stronger relationships for men than women. Our results indicate that incorporating measures of self-reported gender identity could enhance our understanding of gender gaps in economic behavior and outcomes.

JEL-Codes: C910, J160, J200.

Keywords: gender identity, non-binary gender, economic preferences, economic outcomes.

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18th December 2024

We are grateful to participants at several conferences and seminars for helpful comments and suggestions. We thank Johanna Croton, Lukas Diethelm, Asma Hussein, Andia Kozmai, Andreas Lercher, Maximilian Mähr, Viviane Meyer, Chi Nguyen, and Liaoliang Zhang for outstanding research assistance. We thank yousty.ch for generously supporting our data collection among Swiss teenagers and sharing their administrative data with us. This work was supported by the University Research Priority Program “URPP Equality of Opportunity” of the University of Zurich. Lea Heursen thanks the Deutsche Forschungsgemeinschaft through CRC TRR 190 (project number 280092119) for financial support. Eva Ranehill thanks the Tore Browaldh and the Torsten Söderberg foundation for generous support.

1. Introduction

A large and growing body of literature documents a variety of gender gaps in economic behaviors and outcomes. Women tend to select different college majors, career tracks and industries, earn less than men, and are less likely to attain leadership positions in industry or government (Bertrand and Hallock, 2001; Olivetti and Petrongolo, 2008; Blau and Kahn, 2017; Bertrand, 2018; Patnaik, Wiswall and Zafar, 2021). There are also gaps in measures of economic preferences: men often exhibit less risk-aversion, more willingness to compete, and have a stronger tendency to prioritize efficiency over equality (Andreoni and Vesterlund, 2001; Barber and Odean, 2001; Dohmen et al., 2011; Eckel and Grossman, 2008; Filippin and Crosetto, 2016; Fisman, Jakiela and Kariv, 2017; Niederle and Vesterlund, 2011).¹ An important feature of this research is that it studies gender gaps from the perspective of a *binary* classification of individuals as men or women. The focus on this binary categorization of gender is natural given the availability of datasets containing it and the salience of this distinction in social, educational, and economic contexts.

However, a binary classification of individuals as male or female is not the only way in which we may conceptualize the broader notion of *gender*, nor the only way it has been studied in academic research. In the area of gender studies—primarily rooted in sociology and social psychology—gender is often conceptualized as a more complex pattern of behaviors, perceived characteristics, and aspects of individuals’ identities, including perceptions about what traits define the concepts of *masculinity* and *femininity* (Hawkesworth, 1997; Pryzgoda and Chrisler, 2000; Westbrook and Saperstein, 2015). As described in a highly influential paper by West and Zimmerman (1987), gender is the collection of behaviors—what an individual “does”—and the degree to which an individual’s behavior reflects identification with expectations about what it means to be male or female.

In this study, we draw upon research from these fields to explore whether such broader notions of gender are valuable for understanding gender gaps of interest to economists. Specifically, we provide evidence on whether non-binary, self-reported measures of *gender identity* are helpful for understanding gender gaps beyond any relationship with an indicator for binary gender. We initially present a conceptual framework that provides a basis for why an individual’s identity as “masculine” or “feminine” may correlate with gender-stereotypical behaviors, preferences and traits, even after accounting for binary gender classifications. We then empirically investigate these relationships across several measures of preferences, behaviors, and outcomes for which earlier research in economics has found gaps between men and women.

¹These preference gaps have been associated with gaps in a variety of important economic outcomes, including career choices and salary attainment (Barber and Odean, 2001; Buser, Niederle and Oosterbeek, 2014; Dohmen and Falk, 2011; Dohmen et al., 2011; Fisman, Jakiela and Kariv, 2017; Gärtner, Mollerstrom and Seim, 2017; Zhang, 2013).

There are at least two reasons why measures of self-perceived gender identity may help understand questions of interest to economists. First, despite the observation of differences between men and women in behaviors, such as risk-taking and competitiveness, and in the attainment of a variety of educational and labor-market outcomes, there is also substantial within-gender variation. Better understanding whether men and women who identify as more feminine are less likely to engage in risk taking, enter STEM fields, or obtain high income provides potential value for understanding the sources of economic gender disparities and for improved targeting of policy. Second, non-traditional forms of gender identification are growing in many Western societies. For example, according to the Pew Center, 1.6 percent of U.S. adults identify as transgender or non-binary and this percentage is 5.1 percent for adults under 30 (Brown, 2022; see also, Flores et al., 2016). More nuanced and inclusive measures of gender identity recognize this diversity, including individuals who identify as women or men. Indeed, many datasets widely used by economists are starting to incorporate self-reported gender identity, often with non-binary measures.² Because gender identity may motivate individuals' choices and behaviors (Akerlof and Kranton, 2000), it is important to be able to reliably measure such diverse gender identities, in addition to economic outcomes.

As a first step in our analysis, we test the relationship between several measures of non-binary gender identity used in gender studies research and a novel unidimensional question that simply asks individuals where they place themselves on a scale from “very masculine” to “very feminine.” This single-item measure of continuous gender identity (CGI) captures a large part of the variation in several richer scales that measure gender identity along separate dimensions for masculinity and femininity, sometimes using scales constructed from dozens of individual items (Bem, 1974; Kachel, Steffens and Niedlich, 2016; Magliozzi, Saperstein and Westbrook, 2016). This part of our analysis replicates and extends our preliminary investigation of the same question in a sample of Swiss students reported in Brenøe et al. (2022), to larger and broader samples. We also document that our CGI measure is fairly stable over two weeks, by eliciting CGI a second time in one of our samples. Given the concordant results across samples and over time, we employ this validated one-question measure of CGI as the primary measure of gender identity in our subsequent research. However, we also elicit and explore other measures of continuous gender identity employed in the broader literature of gender studies.

We include our measure of CGI in four separate data collections using diverse pop-

²This includes the 2021 UK Census (<https://www.bbc.com/news/uk-wales-55721123>, accessed May 9, 2024) and the U.S Census Bureau's Household Pulse Survey (<https://www.census.gov/library/stories/2021/08/household-pulse-survey-updates-sex-question-now-asks-sexual-orientation-and-gender-identity.html>, accessed May 9, 2024).

ulations.³ The four samples comprise (i) a sample of Swiss students from the university subject pool in Zurich (*Swiss Uni*), (ii) a broader sample of adult U.S. residents recruited through Prolific Academic (*U.S. Adults*), (iii) a sample of Swiss adolescents recruited from an online platform for apprenticeships (*Swiss Teens*), and (iv) a sample of young Swedish adults recruited through contacts with Swedish secondary schools (*Swedish Youths*). Recruiting from such diverse populations allows us to observe a wide range of economic choices and behaviors for which previous research has found gender gaps, ranging from economic preferences to educational choices and labor market outcomes. The considerable variation in age, from 13 years to 60 years, allows us to observe potential generational differences in reported gender identity or in its association with reported outcomes.

In each sample, we elicit our measure of CGI along with various preference and outcome variables for which previous research has documented gender gaps. In the Swiss Uni sample, we elicit several measures of incentivized and unincentivized preferences. We collected the U.S. Adults sample in two waves. In the first wave, we elicit an incentivized measure of risk-taking, several unincentivized preference measures, and measures of labor market and educational outcomes. In the second wave, we elicit a subset of the economic outcome measures collected in the first wave, together with various alternative measures of gender identity that we use for the re-validation of our novel CGI measure. In the Swiss Teens sample, we use a combination of administrative and survey data to observe occupational search and categorize the occupations by their gender composition and math and language skill requirements as proxies for the associated gender stereotypes. For a subset of these adolescents, we also elicit unincentivized preference measures. In the sample of Swedish Youths, we measure incentivized preferences for equality over efficiency, as well as unincentivized preference measures. We also elicit a measure of the respondents' intended fields of study at university and categorize fields by gender composition to measure the associated gender stereotypes.

In all four samples, we observe heterogeneous responses to the CGI question among both men and women. Not surprisingly, the distributions of responses differ substantially between men and women, with women generally reporting a more feminine and men a

³We also had the opportunity to add our CGI measure to a pharmacological experiment implemented by researchers at the Institute of Experimental Psychology, Heinrich Heine University Düsseldorf, which also measured sex hormones. This provides suggestive evidence on the relationship between gender identity and hormones in a sample of students (N=130, 65 women, mean age 24 years). Regressing reported CGI on testosterone, estradiol and progesterone, separately for those identifying as male and female, two relationships among the six tested approached significance at conventional levels. Our point estimates indicate a positive relationship between testosterone levels and masculinity among those identifying as female ($p=0.053$) and a positive relationship between progesterone and masculinity among those who reported being male ($p=0.050$). Due to the underpowered nature of the tests, the relationship should be interpreted cautiously. Hormones were measured in saliva pooled across three samples taken five minutes apart at the onset of the experiment. CGI was elicited after the saliva samples were taken, before the onset of the pharmacological experiment. Due to two participants identifying as non-binary, as well as one missing value for testosterone, two for progesterone, and one for gender, our tests comprise between 62 and 64 individuals. Details on the analysis or data are available on request.

more masculine identity. However, the distributions of responses for men and women exhibit substantial variation and overlap.

We then analyze these datasets to identify whether our novel CGI measure correlates with economic preferences as well as educational and labor market outcomes, *beyond* any relationships accounted for by a standard binary gender measure. For each outcome variable and sample, our analysis follows the same basic structure.⁴ We first test for a difference in preference and outcome measures between men and women, and only continue the analysis for those outcomes for which we find a statistically significant gap. For these outcomes, we proceed to test the degree to which replacing the binary gender measure with CGI yields a statistically significant relationship of the same sign as the gender gap. A key part of our analysis is to then investigate the degree to which CGI relates to the outcome measure when also controlling for binary gender. Finally, we examine the extent to which reported gender identity correlates with economic choices and outcomes, separately, for men and women.

Expanding the scope of our approach, we also investigate correlations between other non-binary measures of gender identity and a subset of the outcomes we elicit. This allows us to test whether the broad conclusions that arise when using CGI are similar across various continuous measures of gender identity, or if they are specific to our novel and simple measure.

We provide four broad sets of results. First, we consider the statistical significance of CGI in explaining gender gaps in our outcome variables across all 33 tests we conducted. We strongly reject the null hypothesis that CGI has no relationship with the outcomes beyond what can be explained by binary gender alone. These findings are demonstrated in Figure 2, which shows the distributions of p -values that result from the key regressions for each outcome measure, clearly indicating a tendency toward lower p -values than one would expect by chance. Further, while we strongly reject the null hypothesis that CGI has no relationship with the preference and outcome measures we study, the added explanatory power provided is generally modest in absolute terms; that is, introducing CGI into a model that already contains binary gender generally yields modest increases in R^2 .

Second, we identify those gender gaps for which we observe the strongest statistical relationships between the relevant outcome and CGI when accounting for binary gender. Given the exploratory nature of our work, we note only those relationships for which CGI is statistically significant at $p < 0.005$ while controlling for binary gender, which is true for 10 of the 33 tests. These include measures of self-reported willingness to take risk, engage in competition, redistributive preferences, the choice of a stereotypically female educational track, income, being a full-time homemaker and weekly average work hours.

⁴We pre-registered the approach for our analysis of the Swiss Uni dataset (<https://osf.io/phyt6/>). The analysis of the remaining datasets follows the same approach.

Interestingly, we find no significant relationships for incentivized preference measures, including risk and competitiveness, once controlling for binary gender. Thus, our findings suggest that relationships with CGI tend to be strongest for choices and outcomes related to the labor market and for unincentivized preference measures, and weakest for the type of incentivized choices that are widely used to measure economic preferences. The associations between CGI and human capital investment decisions are particularly noteworthy, because those investments happen relatively early in life and have far-reaching implications for subsequent life outcomes.

Third, we study the relationships between CGI and our outcome measures separately for men and women, finding that CGI shows substantially stronger correlations with economic behaviors and outcomes for men than for women. This provides suggestive evidence that the relationship between norms governing gender identity and behavior may be stronger for men. One potential, though speculative, interpretation for the stronger relationship for men is that norms of appropriate economic behavior for women have evolved and relaxed substantially in the past several decades, relative to those for men, at least in the samples we study.

Fourth, we also investigate the relationships between other measures of gender identity used in gender studies research and a subset of our outcome measures related to the labor market, incentivized risk-taking and unincentivized preference measures. In line with the results presented above, we find that scales measuring masculinity show stronger correlations with the type of economic choices and outcomes that we study, after controlling for binary gender, than measures of femininity. Measures of feminine identity add little explanatory value.

Taken together, our findings suggest that measures of continuous gender identity are statistically significantly related to preference and outcome measures often studied in (gender) economics. While the strength of the relationships varies across outcome measures and they rarely provide much additional explanatory power, there are some domains in which the relationships are substantial in magnitude and of economic significance. So, while our results do not indicate that including measures like CGI is critical for having a rich accounting of all the relationships between “gender” and economic preferences and outcomes, there may be populations and decision contexts in which such relationships are particularly important and valuable for improved understanding, prediction, and for policy targeting. For example, we observe particularly strong relationships for men, suggesting that gender identity may be more important for economic behaviors and outcomes in this population.

It is important to note that our work is exploratory and correlational. It is not based on a theory of which behaviors or outcomes should be correlated with gender identity, or on induced variation in such identity to show a causal relationship. Instead, we investigate whether there are relationships between self-reported gender identity and

a wide variety of behaviors and outcomes as an initial step in determining the potential value in studying such relationships.⁵ In this regard, our work largely follows much work in gender economics, which documents correlations between binary gender and economic preferences, behavior and outcomes, as a starting point for learning more about these relationships. Because of the exploratory nature of our work and the large number of tests implemented, we employ a pre-specified approach for identifying variables of interest and limit our conclusions about specific relationships involving gender identity to those that are statistically significant at more conservative levels ($p < 0.005$) than those typically employed in economic research (Benjamin et al., 2018).

By focusing on a continuous measure of gender identity, our approach contrasts with most prior economic research on gender, which primarily focuses on differences between men and women.⁶ A few notable exceptions acknowledge the potential roles of variation in gender identity for economic behavior. For example, in their influential work on identity, Akerlof and Kranton (2000) recognize that individuals may experience disutility from behaviors that depart from what is expected from members of a social category, such as “woman” and “man.” Other empirical research studies whether variation in norms governing gender conduct—e.g., because of variation across cultures or households—influences women’s tendencies to demonstrate stereotypically female behavior (Gneezy, Leonard and List, 2009; Bertrand, Kamenica and Pan, 2015; Brenøe, 2022; Bursztyn, González and Yanagizawa-Drott, 2020).

A handful of papers in economics investigate similar questions as the one underlying our research. Burn and Martell (2022) use respondents’ answers to various survey items that correlate with binary gender to construct a proxy measure of “gender typicality” (see Fleming, Harris and Halpern, 2017) and then explore its relationship to labor market outcomes, with a focus on explaining differential outcomes for gays and lesbians. They find that gender typicality is related to labor market outcomes for men but not for women, but do not find that this measure explains gaps in outcomes between homosexual and heterosexual individuals. Banan, Santavirta and Sarzosa (2023) take a similar approach, using survey responses on preferences and interests to construct a measure of gender typicality in youth and correlating it with occupational choices, family planning, and health outcomes later in life.

Two recent papers investigate relationships between economic preferences and self-identified gender categories. Fornwagner et al. (2022) focus on differences among cismen, ciswomen, transmen, and transwomen—as well as priming of gender identity—to study

⁵For examples of recent research following a similar empirical approach by systematically investigating correlations between various measures of economic preferences and behaviors, see Chapman et al. (2023) and Stango and Zinman (2023).

⁶An exception to the literature emphasizing gender gaps in means is also provided by Nelson (2015), who emphasizes within-gender variation and the often substantial overlap in the distributions of measures of risk preferences of women and men.

gender and biological sex differences in economic preferences.⁷ Overall, they find little evidence of either correlational or causal relationships between both sex and gender and behavior. Coffman, Coffman and Ericson (2024) focus on the economic preferences and beliefs of individuals who identify as non-binary and neither as a woman nor a man on a discrete measure of gender identity. While this research indicates some differences between non-binary individuals and men and women, the results reject simple explanations such as the preferences and beliefs of non-binary individuals lying between those of men and women.

To our knowledge, no other work in economics or related fields measures individuals' self-reported continuous gender identity in samples of men and women and across a variety of populations and investigates the relationships between such identification and a broad set of economic behaviors and outcomes.⁸ Moreover, we also provide evidence on the relative value of different gender identity measures, including various measures used in gender studies research.

The rest of this paper is structured as follows. Section 2 provides a brief discussion of the measurement of gender identity outside of economics and a simple framework for how such identity potentially relates to economic behavior and outcomes. This section also reports the results of our validation exercise for our CGI measure. In Section 3, we briefly outline our empirical approach and hypotheses. Thereafter, in Section 4, we describe the different data sets and associated measures, before we present the results in Section 5. Section 6 concludes.

2. Gender Identity: Measurement and Implications for Behavior

Psychologists and other social scientists have long recognized that awareness of gender roles plays an important role in human cognitive and social development (Kohlberg and Kramer, 1969; Gilligan, 1977). While earlier research conceptualized gender as behaviors and identity reflecting concordance with *either* male or female norms, Sandra Bem (1974) advocated for viewing masculinity and femininity as distinct constructs, with the possibility that an individual could exhibit high (or low) concordance with both masculinity and femininity.

Bem (1974) also introduced a scale for measuring the distinct dimensions of masculinity and femininity, which subsequently became widely adopted and referred to as the Bem Sex Role Inventory (BSRI). The inventory contains 60 items eliciting the re-

⁷A small literature in economics studies whether priming experimental participants with gender impacts gender gaps in economic preferences, finding mixed results (see, e.g., Benjamin, Choi and Strickland, 2010, Boschini, Muren and Persson, 2012, Boschini et al., 2018).

⁸There is, however, a growing body of research exploring relationships between non-binary gender and outcomes in other fields in the social sciences, see, for example, Alexander, Bolzendahl and Wängnerud (2021) for a discussion of the use of non-binary measures of gender in political science.

spondents’ compliance with traditional gender roles along both masculine (e.g., having a strong personality) and feminine (e.g., loving children) dimensions. Based on the self-reported traits and behaviors, respondents are classified as either masculine, feminine, androgynous, or undifferentiated. Much subsequent research measuring gender identity or conformity with gender norms identifies masculine and feminine dimensions separately, by eliciting respondents’ self-reported tendencies to exhibit gender stereotypical behaviors or traits (Mahalik et al., 2003; Mahalik et al., 2005).

More recently, researchers have started questioning the need to rely on lengthy inventories that evaluate masculinity or femininity by measuring adherence to, or self-identification with, characteristics or behaviors whose relationship to gender norms may evolve over time. Thus, some recent approaches measure gender identity directly, by asking individuals the degree to which they perceive themselves as masculine and feminine (Magliozzi, Saperstein and Westbrook, 2016; Solevid et al., 2021). Furthermore, some researchers propose constructions of gender identity with masculinity and femininity as opposite ends of a unidimensional spectrum (Kachel, Steffens and Niedlich, 2016). As an alternative to self-identification, researchers have also proposed data-driven approaches to measure variation in conformity with gendered behavior and norms. For example, Fleming, Harris and Halpern (2017) use the degree to which an individual exhibits traits or behaviors most frequently associated with male or female respondents in a dataset as a measure of gender typicality.

Following the introduction of Bem’s measure, as well as other measures of gender identity, several studies have investigated the relationship between gender identity and the tendency to exhibit other behaviors or characteristics typically associated with men or women. For example, self-reported gender identity has been found to correlate with social behaviors like aggressiveness and conformity (Eagly, 1978; Bernard, Bernard and Bernard, 1985; Weisbuch, Beal and O’Neal, 1999), approaches to decision-making (Nezu and Nezu, 1987; Brems and Johnson, 1989), eating disorders (Meyer, Blissett and Oldfield, 2001; Griffiths, Murray and Touyz, 2015) and psychological well-being (Taylor and Hall, 1982; Feather, 1985; Whitley, 1985). We take the existence of these relationships as a starting point for our investigation of whether a measure of continuous gender identity can be helpful for understanding behaviors, traits, and outcomes typically of interest to economists.

2.1. (Re-)Validating a Single-Item Measure of Continuous Gender Identity

Brenøe et al. (2022), introduced a novel measure of continuous gender identity, based on an individual’s response to a single question, *“In general, how do you see yourself? Where would you put yourself on this scale from ‘Very masculine’ to ‘Very feminine’?”* where higher scores correspond to a more feminine identity. Brenøe et al. found that responses

to this question correlated positively (negatively) with the feminine (masculine) scores of several two-dimensional measures of gender identity: the Bem Sex Role Inventory (BSRI), the Open Sex Role Inventory (an updated version of the BSRI) and a two-dimensional scale introduced by Magliozzi, Saperstein and Westbrook (2016). They also correlated positively with femininity scores from a unidimensional index proposed by Kachel, Steffens and Niedlich (2016). While this provides evidence that the single-item CGI measure has value for capturing an important part of more complex measures of gender identity, the data in Brenøe et al. were based on a relatively small and homogeneous student sample ($N = 99$).

To provide additional evidence on the relationship between our CGI measure and other scales employed in prior research, we conducted a study of 2,659 adult respondents (aged 20–60) in the U.S. recruited from the platform Prolific Academic. In addition to the measures of gender identity studied in our earlier paper, we collected additional measures used in prior work. Specifically, we included the following eight measures of gender identity:

1. Our novel *Single-item CGI* (1 item, 11-point scale, **CGI**), based on the single question described above, with higher scores indicating a more *feminine* identity.
2. The *Traditional Masculinity-Femininity* scale (Kachel, Steffens and Niedlich, 2016; 6 items, 6-point scales, **TMF**), on which individuals provide several responses ranging from very masculine to very feminine on a unidimensional scale, measuring how respondents personally identify and how their behaviors, attitudes, interests, and appearance would be socially interpreted. The six items are averaged to obtain a score reflecting *femininity*.
3. A *Two-Dimensional Masculinity-Femininity* scale (Magliozzi, Saperstein and Westbrook, 2016; 2 items, 6-point scales, **Magliozzi**), in which participants respond to “how do you see yourself?” separately for feminine and masculine dimensions, with responses ranging from “not at all” to “very.” This yields separate scores for *masculinity* and *femininity*.
4. A *Trait-Based Self-Categorization* scale (Solevid et al., 2021; 2 items, 10-point scales, **SOM**), which asks participants the extent to which they believe that they possess masculine and feminine traits, yielding separate scores for *masculinity* and *femininity*.
5. An adapted version of the *Bem Sex-Role Inventory* (Bem, 1979; 10 feminine, 10 masculine, 10 neutral items, 5-point scales, **BSRI**), in which participants self-report the degree to which several characteristics apply personally. We follow the standard scoring approach to obtain separate measures of *masculinity* and *femininity*.
6. The *Open Sex-Role Inventory* (<https://openpsychometrics.org/tests/OSRI/>; 11 fem-

inine, 11 masculine items, 5-point scales, **OSRI**), which asks similar questions to the BSRI, though with an updated interpretation of masculine and feminine behaviors. Similarly to BSRI, this yields separate scores for *masculinity* and *femininity*.

7. The *Conformity to Masculine Norms Inventory* (Mahalik et al., 2003; 30 items, 5-point scales, **CMNI**), which asks respondents to rate the degree to which they exhibit stereotypically male traits or behaviors (e.g., “I put myself in risky situations”). The items are combined into a single scale reflecting *masculinity*.
8. The *Conformity to Feminine Norms Inventory* (Mahalik et al., 2005; 45 items, 5-point scales, **CFNI**), which asks respondents to rate the degree to which they exhibit stereotypically female traits or behaviors (e.g., “I regularly wear make-up”). The items are combined into a single scale reflecting *femininity*.

All respondents completed the first four measures as well as one from the remaining four inventory-based measures, administered in random order. The choice to let participants complete only one of the inventory-based measures was due to their lengthy nature and substantial overlap in the type of questions asked.

Table 1 shows the correlations between the above measures. As in Brenøe et al. (2022), all the correlations are highly statistically significant and in the anticipated directions. The single-item CGI scale correlates most strongly with the other measures that directly elicit self-reported masculinity and femininity (TMF, Magliozzi, SOM) and less strongly with the measures that elicit gender identity indirectly through evaluations of the applicability of gendered characteristics (BSRI, OSRI, CMNI, CFNI). The final row reports correlations with the first factor from a principal components analysis of all the measures except for CGI, revealing that this aggregated measure correlates highly with CGI. Appendix Figure A1 illustrates the relationships between CGI and each of the other measures in binned scatter plots. This analysis also shows that when elicited separately, the masculinity and femininity dimensions are highly (negatively) correlated, indicating that a large share of the variation between these measures can be captured by a unidimensional scale. Taken together, this analysis corroborates the preliminary evidence from Brenøe et al. (2022), that the single-item CGI measure captures a substantial part of individuals’ gender identity, though with a much larger and broader sample of respondents.

To further evaluate the reliability of our CGI measure, we use data from our sample of Swiss university students (N=584), who completed two separate elicitations of the CGI measure two weeks apart. In the first elicitation, completed as part of our main study using this sample, respondents reported CGI on an 11-point scale from “very masculine” to “very feminine.” In the second elicitation, they responded on an inverted 12-point scale, from “very feminine” to “very masculine.” Despite these changes in the response format, the responses exhibit a high degree of stability (see Appendix Figure A2), with

Table 1: Correlation Matrix of all Gender Scales

Variables	CGI	TMF	Magliozzi (f)	Magliozzi (m)	SOM (f)	SOM (m)
TMF	0.923*** (0.007)					
Magliozzi (f)	0.901*** (0.008)	0.889*** (0.009)				
Magliozzi (m)	-0.899*** (0.008)	-0.880*** (0.009)	-0.846*** (0.010)			
SOM (f)	0.879*** (0.009)	0.882*** (0.009)	0.890*** (0.009)	-0.822*** (0.011)		
SOM (m)	-0.853*** (0.010)	-0.854*** (0.010)	-0.806*** (0.011)	0.880*** (0.009)	-0.798*** (0.012)	
OSRI (f)	0.570*** (0.032)	0.600*** (0.031)	0.627*** (0.030)	-0.500*** (0.034)	0.632*** (0.030)	-0.478*** (0.034)
OSRI (m)	-0.462*** (0.035)	-0.450*** (0.035)	-0.428*** (0.035)	0.478*** (0.034)	-0.409*** (0.035)	0.508*** (0.033)
CFNI	0.335*** (0.037)	0.362*** (0.036)	0.348*** (0.037)	-0.338*** (0.037)	0.342*** (0.037)	-0.368*** (0.036)
CMNI	-0.431*** (0.035)	-0.433*** (0.035)	-0.362*** (0.036)	0.445*** (0.035)	-0.375*** (0.036)	0.446*** (0.035)
BSRI (f)	0.211*** (0.038)	0.220*** (0.038)	0.252*** (0.037)	-0.193*** (0.038)	0.278*** (0.037)	-0.201*** (0.038)
BSRI (m)	-0.264*** (0.037)	-0.263*** (0.037)	-0.212*** (0.038)	0.297*** (0.037)	-0.227*** (0.037)	0.297*** (0.037)
First comp	0.942*** (0.007)	0.953*** (0.006)	0.936*** (0.007)	-0.935*** (0.007)	0.929*** (0.007)	-0.918*** (0.008)

Notes: The table presents pairwise correlations between gender identity measures with standard errors reported in parentheses. U.S. Adults wave 2 (age 20–60) sample is used. CGI and TMF are unidimensional scales, with higher values indicating higher femininity. “f” refers to the femininity and “m” refers to the masculinity score of the two-dimensional scales BSRI, OSRI, Magliozzi and SOM. CFNI and CMNI are the conformity to feminine and masculine norm inventories. *First comp* combines the first factor from a principal component analysis of all alternate gender identity scales in a given sub-sample excluding CGI. N=2,659 for CGI, TMF, Magliozzi and SOM measures; N=662 for OSRI measures; N=659 for CFNI; N=662 for CMNI; N= 676 for BSRI measures. The significance levels are: *** p<0.005, ** p<0.01, * p<0.05.

a correlation of 0.874 (once the second set of responses is inverted). Thus, aside from strong correlations with alternative measures of gender identity used in prior research, our CGI measure exhibits stability over time.

2.2. Gender Identity and Behavior

Before proceeding to our empirical analysis of the relationships between CGI and economic behaviors and outcomes, we provide a framework for understanding why such relationships may exist, beyond those with an individual’s binary categorization as a man or woman. We employ a simple utility framework based on the richer model in Akerlof and Kranton (2000), building on their premise that identity is based on social categories, c , to which an individual belongs. For example, an individual responding to

a survey question eliciting their binary gender classification, may belong either to the category “male” or “female.”

Assume that an individual derives utility from the material consequences produced by actions taken, a , as well as from the individual’s sense of identity, I ,

$$U = U(a, I).$$

The actions correspond, for example, to selecting a risky option or choosing a profession. These actions can impact the individual’s utility through standard consequential considerations. Actions can also impact utility through their effect on the individual’s identity. Following Akerlof and Kranton (2000), we assume that an individual’s identity is influenced by several factors:

$$I = I(a, c, P, \gamma_c).$$

These factors include the individual’s actions (a), the social categories to which the individual belongs (c), and social prescriptions (P) regarding appropriate behavior for someone belonging to these social categories. While an individual may belong to several social categories, we are interested in the categorization of individuals as a “woman” or a “man”. With respect to how prescriptions influence the individual’s gender identity, assume that P represents actions that men or women “should” take.

The final parameter, $\gamma_c \in [0, 1]$, captures the strength of an individual’s identification with their binary gender category. That is, a woman can vary in the degree to which she identifies as “feminine” (γ_f) and a man in the degree to which he perceives himself as “masculine” (γ_m). We assume that an individual can strongly identify with a category ($\gamma_c = 1$), strongly reject that category ($\gamma_c = 0$), or exhibit varying degrees of intermediate identification ($\gamma_c \in (0, 1)$). This self-perception, γ_c , determines the degree to which the individual derives utility or disutility from following the behavioral prescriptions for someone with the individual’s gender category. For the case of binary gender categories, and in our primary empirical approach, we assume that identification with one’s own gender category is inversely related to identification with the other category, such that $\gamma_f = (1 - \gamma_m)$.⁹ Gender identity is then a continuous measure with two poles representing either strong identification with one’s own or with the other binary category.

To concretely illustrate the above concepts, consider someone selecting between two potential professional tracks—one stereotypically male ($a = 1$), like construction, and one stereotypically female ($a = 0$), like nursing. The prescription (P) is that someone belonging to the category “man” should choose $a = 1$, while someone belonging to the category “woman” should choose $a = 0$. In this case, the individual’s sense of gender

⁹The correlation patterns in Table 1 provide support for this assumption. However, it need not necessarily be the case that an individual who identifies as more masculine must also identify as less feminine, as reflected in notions of “androgynous” gender identity (Bem, 1974). In Subsection 5.7, we empirically investigate the value of separate measures of γ_f and γ_m .

identity might be described as $I = 1 - |\gamma_c - 1_{a=P}|$, with γ_c corresponding to the degree to which a man perceives himself as masculine or a woman perceives herself as feminine and the indicator function $1_{a=P}$ taking the value of 1 if the individual takes the action prescribed for his or her category and 0 otherwise.¹⁰ In this example, a man who identifies as very masculine ($\gamma_m = 1$) will obtain a more positive sense of identity when following the behavioral prescriptions for a man ($a = 1$). Conversely, another man who rejects masculine identity ($\gamma_m = 0$) will obtain a stronger sense of identity by rejecting gender stereotypes and adopting actions traditionally prescribed for a woman ($a = 0$). As γ_c varies, the emphasis that an individual places on acting in the manner prescribed for his or her gender category also changes.

Based on this simple framework, we propose two separate channels through which we may observe a relationship between an individual’s behavior (a) and the degree to which that individual identifies (γ_c) with their binary gender category. First, if we assume that the individual’s actions are (largely) exogenous but identity is malleable, then identity can shift to maximize the utility from following or violating behavioral prescriptions. Thus, a man forced (e.g., by societal expectations) into a stereotypically male career track may, *ceteris paribus*, more strongly identify as masculine to increase his identity-based utility. Second, if the individual’s identity is (largely) fixed but the individual has agency over actions, then the actions may be influenced by the strength of gender identity. That is, a man who views himself as very masculine will be more likely to select stereotypically male career paths holding constant the economic benefit from doing so. Regardless of whether the individual adjusts actions to concord with gender identity or vice versa—or, as is more likely the case, that both forces are at play in a setting where actions and identity are determined endogenously—this framework illustrates why we may observe a correlation between behavior and strength of identification with the social categories woman and man, even after accounting for the role played by assignment to one of these categories.

We compare this prediction with one based on a version of this framework in which an individual’s strength of identification with gender categories are irrelevant—e.g., when γ_c is absent or when its effect on I is independent of the individual’s actions. Under this interpretation, which is closer in spirit to that of Akerlof and Kranton (2000), the individual’s utility from identity is determined mainly by social prescriptions for how someone who belongs to their assigned gender category should behave and the individual’s actions. Our empirical tests, therefore, aim to identify the degree to which individuals’ self-reported strength of identification with their binary gender categories ($\gamma_f = 1 - \gamma_m$) are related to their behavior and outcomes (a), against the null hypothesis of no such

¹⁰In this simple illustrative example, the individual’s overall utility might be represented as $U = v(a) + I$, reflecting additively separable utility from the direct consequences of actions and an additional identity-based component of utility. We provide this example only for illustration, recognizing that there are varied and richer ways of capturing the relationships of interest.

relationship.

3. Empirical Strategy

To explore whether measures of CGI correlate with variation in economic decisions and outcomes, *beyond* what can be accounted for by a binary indicator of gender, we apply the same analytical approach across all outcome variables from the diverse samples. We pre-registered this approach for the analysis of our first sample, Swiss Uni.¹¹

We begin by identifying gender gaps of the kind typically studied by economists, by estimating the following regression:

$$Y = \alpha + \beta \text{ FemBin} + \epsilon. \quad (1)$$

We test the null hypothesis that $\beta = 0$, which captures whether a binary classification of gender (*FemBin*) correlates with the relevant behavior, preference, or outcome, Y . *FemBin* takes the value one for women and zero for men (corresponding to c in the framework in Section 2.2). Our one-sided alternative hypothesis is that β takes the same sign as indicated by prior research on gender preference gaps. For the next steps of our analysis, we retain only those variables for which we reject the null hypothesis at $p < 0.05$.

Conditional on a statistically significant relationship for binary gender, we next investigate the relationship between the outcome measure and CGI. First, we test whether our CGI measure correlates with the elicited outcome variable when it replaces the binary gender measure. To do so, we run regressions of the form,

$$Y = \alpha' + \gamma \text{ FemCGI} + \epsilon. \quad (2)$$

We test the null hypothesis that $\gamma = 0$ against the alternative hypothesis that γ takes the same sign as β , for each outcome Y . We only proceed to the next step when we can reject the null at $p < 0.05$.

Finally, for our principal test, we investigate whether CGI is statistically significantly correlated with the outcome measure when we additionally account for the relationship with binary gender. To test this, we run regressions of the form,

$$Y = \alpha'' + \beta' \text{ FemBin} + \gamma' \text{ FemCGI} + \epsilon, \quad (3)$$

using the same dependent variable as above. Our null hypothesis is that CGI only operates through binary gender, i.e., $\gamma' = 0$. The alternative hypothesis—and our principal test—is that γ' takes the same sign as β from regression (1) for a given outcome Y . This provides evidence on whether CGI correlates with measures of economic preferences and

¹¹Our pre-analysis plan is available at <https://osf.io/phyt6/>.

outcomes, even when accounting for binary gender.

4. Design and Sample Descriptions

This section briefly describes the four different datasets we collected, including the main preference and outcome measures. For a detailed description of each separate part of the data collection, see Appendix B.¹²

Table 2 provides an overview of the main outcome measures elicited in each data collection. A more detailed overview of these variables is available in Appendix Table A1.

Table 2: Overview of Main Outcome Variables

Outcome measures	Dataset			
	Swiss Uni	U.S. Adults	Swiss Teens	Swedish Youths
<i>Incentivized preferences</i>				
Risk	Y	Y		
Competitiveness	Y			
Efficiency	Y			Y
Overconfidence	Y			
<i>Non-Incentivized preferences</i>				
Staircase risk	Y			
Risk	Y	Y	Y	Y
Financial risk		Y		
Competitiveness	Y	Y	Y	Y
Redistribution	Y	Y		
Altruism	Y	Y	Y	Y
<i>Educational and labor market outcomes</i>				
Female educational track share		Y	Y	Y
Math/Language skill requirement			Y	
Income		Y		
Full-time homemaker		Y		
Weekly average work hours		Y		
Female industry share		Y		
Managerial responsibilities		Y		
Performance pay		Y		
Wage negotiation		Y		
Work flexibility		Y		

Notes: Appendix Table A1 provides more details on each variable.

We elicited our main variable of interest, our *single-item measure of CGI*, in all samples.¹³ In Swiss Uni and U.S. Adults, we additionally measured how a person de-

¹²Full instructions for three of the four data collections are available in Appendix C. The data collection using Swedish Youths was implemented as part of another ongoing study by one of the authors of this paper (Ranehill). The full instructions for this data collection will be made available when the main study is completed.

¹³Across the studies, the timing of when CGI and binary gender were elicited relative to the outcome variables varied (see Appendix B for details).

scribes being seen by others on our masculinity-femininity scale. In U.S Adults, we also elicited two secondary CGI measures asking participants to place themselves on our masculinity-femininity scale relative to men and relative to women, separately. In the Swiss Uni, U.S. Adults, and Swiss Teens samples, we also elicited participants’ *binary gender* by asking them to report one of two categories, “male” or “female” which we code as a binary variable taking the value 1 for women and 0 for men.¹⁴ For Swiss Uni, U.S. adults and Swedish Youth we additionally elicited non-binary but discrete gender identity, where participants could select the category “non-binary” or “other.”¹⁵ Due to the small number of individuals providing non-binary gender classifications, we omit participants selecting a category other than “male” or “female” from our primary sample.¹⁶ Further, it reflects our primary research interest in studying the extent to which differences in gender identity may explain heterogeneous economic behavior for individuals who identify as a woman or a man. Across all samples, we also exclude individuals for whom we have multiple binary measures if there is any inconsistency. For the U.S. adults sample, for example, we also observe participants’ gender previously reported to Prolific Academic. For either one of these two reasons (non-binary or inconsistency), we exclude a total of 95 participants (less than 2% of all observations). Appendix Table A2 provides a detailed overview of the exclusions made for each sample.

Our first data collection, Swiss Uni, focused primarily on the correlation between CGI and incentivized measures of economic preferences.¹⁷ This dataset was collected in September and October 2021 through an online experiment and comprises 584 student participants from the University of Zurich and the Swiss Federal Institute of Technology (ETH). This dataset primarily elicited preference measures through widely-used incentivized tasks for which earlier studies have reliably documented gender gaps. *Risk preferences* were elicited through a one-shot investment task used by Gneezy and Potters (1997). Our measure of *Competitiveness* follows the design introduced by Niederle and Vesterlund (2007), with slight modifications due to the online setting. Preferences

¹⁴For U.S. adults, we took the wording from the U.S. Census questionnaire and asked “What is your sex?”. In Swiss Uni and Swiss Teens, we asked “Are you male or female?”

¹⁵For Swiss Uni we asked “What is your current gender?” with the answer options Woman/Man/Transgender/Non-binary/Other/Prefer not to answer. In U.S. Adults (Wave 2) we asked “What gender are you currently?” with the answer options Man (including Trans Male/Trans Man)/Woman (including Trans Female/Trans Woman)/Non-binary/Would rather not say. For Swedish Youth we asked “Do you identify as a woman or a man?” to which participants could answer Woman, Man or Other

¹⁶In the Appendix, we show that our main results for the U.S. Adults and Swiss Uni samples are very similar when including the responses of individuals who identify as non-binary for our analysis (see Appendix Tables A28 - A31). The U.S. Adults and Swiss Uni samples are the ones in which we elicited both a binary measure of gender and a non-binary but discrete measure of gender identity.

¹⁷We implemented changes relative to the pre-registration (<https://osf.io/phyt6/>) mainly because we adapted the experiment, which was originally intended as a laboratory experiment, to an online format due to COVID-19 restrictions. A detailed description of the departure from the initial pre-registration is available in the document “Updates to Pre-registration Final.pdf” available at <https://osf.io/phyt6/> and summarized in Appendix B.

for *Equality versus efficiency* were elicited by implementing 15 graphical budget sets involving inter-personal allocations, similarly to Fisman, Kariv and Markovits (2007). Finally, our measure of *Overconfidence* is based on three measures of relative overplacement (Moore and Healy, 2008); for example, participants had to rank their performance on matrix reasoning items (similar to Raven’s matrices) relative to a reference group of other participants.

As part of the Swiss Uni data collection, we administered a follow-up survey two weeks after the main experiment that elicited non-incentivized preference measures, such as self-reported measures of risk seeking, competitiveness, attitudes towards redistribution, and altruism. The main purpose of the follow-up survey was to duplicate the elicitation of the CGI measure using a different (reversed) scale to provide the opportunity to account for possible measurement error in our statistical analysis, following the ORIV procedure of Gillen, Snowberg and Yariv (2019).

Our second data collection, U.S. Adults, focused primarily on the correlation between CGI and economic choices and outcomes in a broader sample of U.S. residents of working age. For this analysis, we recruited 3,902 respondents aged 30–60 years in two waves (in March 2022 and March 2023) through Prolific Academic. We restrict our sample to individuals at least 30 years old in order to target a population likely to have at least a few years of work experience.¹⁸ In this sample, we collected a broad set of variables related to demographics, family, education, labor market participation and job attributes, such as study field of college major, employment status, income, and past wage negotiations. We further elicited the incentivized risk measure and the same unincentivized preference measures as in our Swiss Uni sample. As a proxy for real-world (financial) risk taking, we asked whether a respondent actively trades in securities.

Our third data collection, Swiss Teens, focused on educational choices and unincentivized economic preferences among adolescents. We conducted two surveys comprising 1,740 Swiss teenagers. The respondents were recruited through two newsletters sent from the largest Swiss online platform for apprenticeship search. Vocational education and training is the most common type of education after compulsory schooling (9th grade) in Switzerland. Most respondents (91 percent) were 8th and 9th grade students (with an average age of 14.8 years) who were planning to do an apprenticeship after 9th grade. At the time of our surveys (December 2021 and March 2022), the 8th graders were considering potential future apprenticeships. This means exploring different professions at firms through trial apprenticeships, which typically last 1-5 days. The 9th graders were further along—52 percent of them had already signed a contract for their apprenticeships with a specific company. In the first survey, we elicited unincentivized preference measures

¹⁸In Wave 2, which also serves as the dataset for the validation of our CGI measure reported in Section 2.1, we additionally recruited 751 participants aged 20–29 only for use in this validation analysis. For these respondents, we did not collect the full set of outcome measures.

for risk-seeking, competitiveness, and altruism. Moreover, for respondents in the first survey, we were able to merge their survey responses to profile data from the platform, covering real life decisions in terms of apprenticeship applications. In the second survey, we elicited detailed information on all trial apprenticeships the respondents had done. Based on these data, we construct three main variables that characterize the respondents' occupational preferences in terms of gender composition (female share) and skill requirements (math and language) of their chosen apprenticeship profession.¹⁹

Our fourth data collection, Swedish Youths, includes 1,041 Swedish students (age 18-19 years) in the final year of the natural science track in Sweden and was part of a larger experiment. The experiment took place in the beginning of 2022, in the weeks before the students applied for their preferred educational field for subsequent university studies. This dataset comprises a measure of incentivized preferences for equality versus efficiency (based on 20 allocation decisions), as well as the unincentivized preference measures for risk, competitiveness, and altruism. Finally, the dataset comprises information on participants' intended fields of undergraduate studies, allowing us to construct a measure of the gender composition of the students' intended field of study using administrative data.

5. Results

In this section, we first describe the distributions of self-reported CGI across our four samples. Thereafter, we study the specific relationships between CGI and the elicited preference measures and educational and labor market outcomes. To provide a broad test of our main hypothesis regarding the general value of CGI, we then present an overview of the statistical significance across all the tests presented in the paper involving CGI and our outcome measures. Finally, we compare the relative strength of the relationship between CGI and the outcome measures to those for the other gender identity scales described in Section 2.

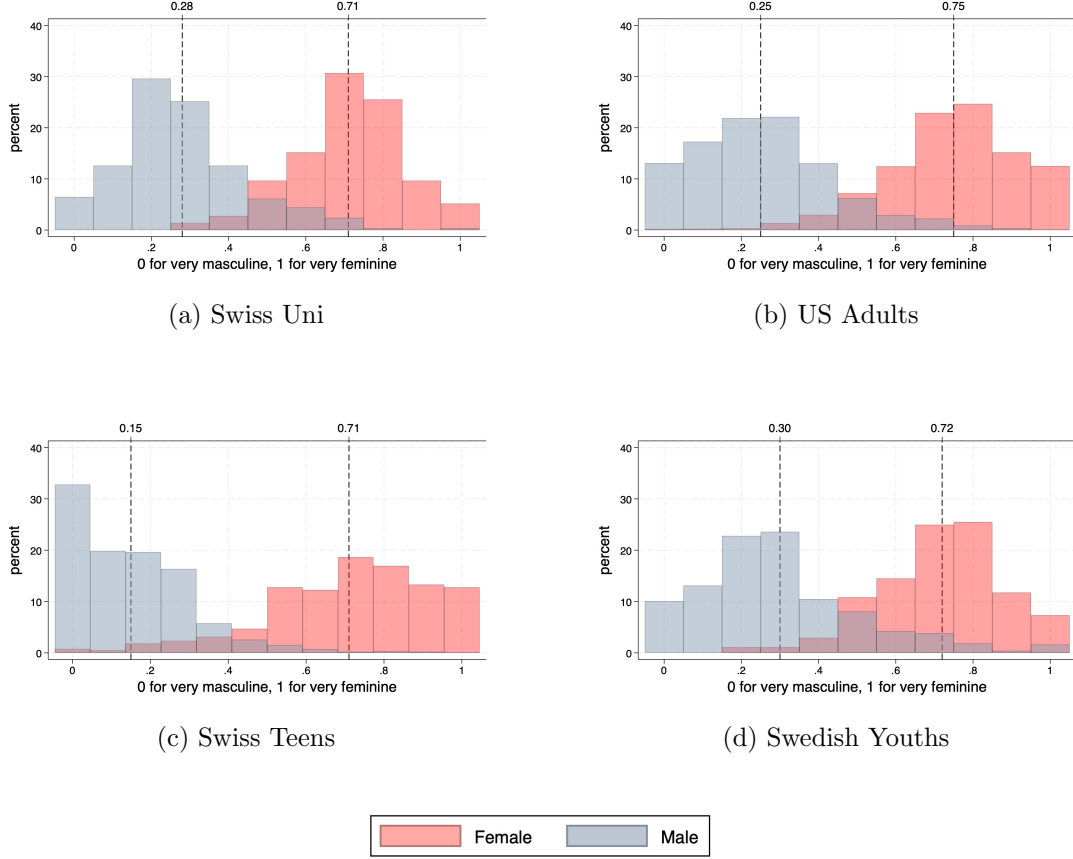
5.1. Continuous Gender Identity in Four Samples

Figure 1 presents the distributions of self-reported gender identity by binary gender across our four samples.²⁰ Two notable patterns are visible. First, in each sample, as expected, there is a substantial gap in the average gender identity of women and men, with women tending to report stronger feminine identity. Second, in all samples, the within-gender variation is substantial. In three samples, the distributions of CGI for each gender are

¹⁹We obtain the gender-composition measure for each occupational field using administrative data and the skill requirements from an independent set of expert ratings. For details, see Appendix B.

²⁰The distribution of CGI for all participants who report a non-binary gender category in any of our studies (N=78) is shown in Appendix Figure A4. The modal answer is at the mid-point of our scale, with nearly 45% of non-binary respondents choosing this response.

Figure 1: Continuous Gender Identity by Binary Gender



Notes: Continuous Gender Identity (CGI) represents our single question, ranging from “very masculine” to “very feminine”. CGI is rescaled from 0 to 1 where 1 is very feminine. The dashed lines mark the sample means for women and men. The sample of Swiss Uni (N=584) is collected through online incentivized experiments. The sample of U.S. adults (N=4,653) aged 20–60 is collected through the company Prolific Academic. The sample of Swiss Teens (N=1,740) is collected through the online “job board” company Yousty. The sample of Swedish Youths (N=1,041) is collected through contacts with Swedish high schools.

generally symmetrically distributed and there is substantial overlap in the distributions between women and men. The sample of Swiss Teens (Panel C) differs in this regard, with fat tails in the extremes for both genders but particularly for men.

5.2. Continuous Gender Identity and Economic Preferences

5.2.1 Incentivized Behavioral Measures

Table 3 presents the relationships between CGI and our incentivized behavioral measures. In this table and henceforth, we standardize CGI within each sample to have a mean of zero and a standard deviation of one, with a higher value indicating a more feminine identity. As much of our analysis will be shown using similar tables, we describe the structure of Table 3 carefully here. Each column of Panel A tests the relationship between binary gender (female) and the corresponding outcome variable, as specified in Equation (1). We document statistically significant gender differences with the expected sign for our measures of risk, competitiveness, and a tendency to prioritize efficiency over

Table 3: CGI and Incentivized Behavioral Measures

Measure	Risk		Competitiveness	Efficiency		Overconfidence
Sample	Swiss Uni (1)	US Adults (2)	Swiss Uni (3)	Swiss Uni (4)	Swedish Youths (5)	Swiss Uni (6)
Panel A. Binary gender						
Female	-0.378*** (0.081)	-0.215*** (0.032)	-0.128*** (0.041)	-0.145*** (0.023)	-0.161*** (0.017)	-0.002 (0.083)
R ²	0.036	0.012	0.017	0.062	0.078	0.000
Panel B. Gender identity						
CGI (feminine)	-0.169*** (0.040)	-0.092*** (0.016)	-0.056*** (0.020)	-0.066*** (0.012)	-0.068*** (0.009)	
R ²	0.028	0.008	0.013	0.052	0.055	
Panel C. Binary gender and gender identity						
Female	-0.305* (0.153)	-0.195*** (0.057)	-0.110 (0.072)	-0.110** (0.043)	-0.135*** (0.026)	
CGI (feminine)	-0.045 (0.074)	-0.012 (0.029)	-0.011 (0.036)	-0.022 (0.021)	-0.017 (0.013)	
R ²	0.036	0.012	0.017	0.064	0.079	
Panel D. Sample split by binary gender						
Female: CGI (feminine)	-0.039 (0.060)	-0.001 (0.023)	-0.029 (0.030)	-0.042** (0.017)	-0.014 (0.011)	
R ²	0.002	0.000	0.004	0.023	0.003	
Male: CGI (feminine)	-0.017 (0.063)	-0.012 (0.023)	0.013 (0.030)	0.013 (0.017)	-0.010 (0.014)	
R ²	0.000	0.000	0.001	0.002	0.001	
Test: male sample split CGI = female sample split CGI						
p-value	0.794	0.717	0.295	0.016	0.696	
Observations	584	3,902	584	584	1,041	584
Mean of Dependent Variable	0	0	0.413	0.538	0.550	0

Notes: The table presents results from regressing the incentivized measures of risk, competitiveness, equality versus efficiency and overconfidence on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30–60), and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of participants' investment decisions (mean=0, SD=1). Competitiveness is a dummy that takes the value 1 for those who chose to compete in the competitive task. Our measure of preferences for equality versus efficiency is measured in deciles, with increasing numbers indicating higher priority for efficiency. Overconfidence is measured as relative overplacement and is standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. Equality of sample split coefficients in Panel D are tested with a Wald test. *** p<0.005, ** p<0.01, * p<0.05.

equality. For overconfidence, we fail to find a significant gender difference and therefore omit this measure from the remaining analysis, as pre-registered.²¹

In Panel B, we regress the remaining outcome measures on CGI (see Equation (2)). Consistent with the gender differences in Panel A and the relationships between CGI and gender in Figure 1, we see that people who identify as more feminine are also less risk-

²¹While not pre-registered, it is still interesting to know whether reported gender identity correlates with being overconfident. Regressing overconfidence on CGI (standardized), the estimated correlation coefficient is 0.039 with a standard error of 0.044. Also within-gender, we find no evidence that being overconfident correlates with CGI. While some earlier studies find that women tend to be less overconfident, the absence of such a relationship is consistent with the results of a recent meta-study (Bandiera et al., 2022).

loving, less likely to enter competition, and have a lower preference for efficiency relative to equality. These associations are all statistically significant at the $p < 0.005$ level.

Next, in Panel C, we implement our principal test and explore whether CGI has any statistically significant association with our preference measures once we control for binary gender. As indicated in Panel C, the CGI coefficients all have the expected sign, but are small in magnitude and none of them are statistically significantly different from zero, even at the $p < 0.05$ level.

Finally, In Panel D, we investigate the relationships between CGI and the preference measures separately for men and women. Out of the ten estimates in Panel D, only one is statistically significant at $p < 0.01$ (efficiency preferences for women in the Swiss Uni sample). The Wald tests reported at the bottom of Panel D test whether the relationships between CGI and the relevant outcome measures differ significantly for men and women, generally finding no significant differences.²²

Based on the above analysis, we conclude that there is at best a weak relationship between our measure of CGI and variation in incentivized behavioral measures, once controlling for the portion of the relationship accounted for by binary gender. This is further reinforced by a comparison of the R^2 values in Panels A and C of Table 3, which show little change, if any, with the introduction of our CGI measure.

5.2.2 *Unincentivized Preference Measures*

Table 4 shows the relationships between CGI and unincentivized risk preference and competitiveness measures. Panel A replicates (at $p < 0.05$) gender gaps commonly found in the literature for all measures and samples. Panel B further shows that CGI strongly correlates with all unincentivized risk and competitiveness outcomes in the expected direction.

In contrast to the results for the incentivized behavioral measures, Panel C shows that CGI can account for some of the variation in several of the unincentivized behavioral outcomes beyond binary gender. Across the ten regressions in Table 4, the coefficient for CGI has the expected sign and is statistically significant at $p < 0.05$ in eight cases, four of which have a significance level of $p < 0.005$. In particular, we observe strong relationships between CGI and self-reported risk-seeking and competitiveness for both the U.S. Adults and Swiss Teens samples. We interpret these findings as providing evidence that gender identity can exhibit substantial correlation with unincentivized measures of risk and competitiveness, even when accounting for the relationships with binary gender. This is further supported by the increases in R^2 when comparing Panels A and C in Table 4, which are larger than those observed in Table 3, though still small in absolute

²²As a complement to the Wald tests in each of our main results tables (Table 3 through Table 7), Appendix Table A32 through Table A36 report complementary regressions of the interactions between CGI and binary gender. The results are generally similar across the two approaches.

Table 4: CGI and Unincentivized Risk and Competitiveness Measures

Measure	Staircase Risk	Risk				Financial Risk	Competitiveness			
Sample	Swiss Uni	Swiss Uni	U.S Adults	Swiss Teens	Swedish Youths	US Adults	Swiss Uni	U.S Adults	Swiss Teens	Swedish Youths
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Panel A. Binary gender										
Female	-0.211* (0.083)	-0.444*** (0.081)	-0.387*** (0.031)	-0.285*** (0.071)	-0.243*** (0.062)	-0.268*** (0.015)	-0.388*** (0.081)	-0.405*** (0.031)	-0.311*** (0.071)	-0.153* (0.062)
R ²	0.011	0.049	0.038	0.020	0.015	0.075	0.038	0.041	0.024	0.006
Panel B. Gender identity										
CGI (feminine)	-0.137*** (0.044)	-0.234*** (0.040)	-0.206*** (0.016)	-0.171*** (0.036)	-0.109*** (0.031)	-0.119*** (0.008)	-0.194*** (0.040)	-0.218*** (0.016)	-0.181*** (0.036)	-0.109*** (0.029)
R ²	0.019	0.055	0.042	0.029	0.012	0.059	0.038	0.048	0.033	0.012
Panel C. Binary gender and gender identity										
Female	0.031 (0.148)	-0.188 (0.148)	-0.152*** (0.056)	0.124 (0.166)	-0.182* (0.103)	-0.224*** (0.027)	-0.212 (0.138)	-0.145** (0.057)	0.087 (0.168)	0.022 (0.101)
CGI (feminine)	-0.149* (0.078)	-0.157* (0.073)	-0.144*** (0.029)	-0.227*** (0.085)	-0.041 (0.051)	-0.027* (0.013)	-0.108 (0.068)	-0.159*** (0.029)	-0.220*** (0.085)	-0.117** (0.048)
R ²	0.019	0.058	0.044	0.030	0.015	0.076	0.042	0.049	0.033	0.012
Panel D. Sample split by binary gender										
Female: CGI (feminine)	-0.050 (0.064)	-0.081 (0.057)	-0.018 (0.025)	-0.051 (0.060)	0.058 (0.046)	-0.001 (0.010)	-0.015 (0.054)	0.068 (0.024)	0.021 (0.055)	0.015 (0.044)
R ²	0.003	0.007	0.000	0.003	0.003	0.000	0.000	0.005	0.000	0.000
Male: CGI (feminine)	-0.123* (0.068)	-0.105 (0.064)	-0.143*** (0.024)	-0.154*** (0.045)	-0.097* (0.052)	-0.029*** (0.011)	-0.110* (0.062)	-0.253*** (0.024)	-0.238*** (0.047)	-0.163*** (0.050)
R ²	0.015	0.011	0.020	0.024	0.009	0.003	0.012	0.064	0.056	0.027
Test: male sample split CGI = female sample split CGI										
p-value	0.548	0.867	0.000	0.062	0.027	0.073	0.297	0.000	0.000	0.018
Observations	584	584	3,902	786	1,041	3,902	584	3,902	792	1,041
Mean of Dependent Variable	0	0	0	0	0	0.401	0	0	0	0

Notes: The table presents results from regressing unincentivized measures of risk attitudes and competitiveness on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens, and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Staircase risk is a categorical certainty equivalence measure of risk-taking based on a series of hypothetical allocation decisions. Risk is a self-reported measure of risk-taking. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk taking or competitiveness. All measures except financial risk are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. Equality of sample split coefficients in Panel D are tested with a Wald test. *** $p < 0.005$, ** $p < 0.01$, * $p < 0.05$.

terms.

Splitting the sample by binary gender (Panel D) reveals that the relationships between CGI and the unincentivized preference measures are driven by men. All the coefficients for men have the expected sign, nine of ten are statistically significant at $p < 0.05$ and only one is not statistically significant at conventional levels ($p = 0.052$). Six of ten coefficients are highly statistically significant at $p < 0.005$. The Wald tests reported at the bottom of Panel D in Table 4 show that the relationships between CGI and the preference measures are often highly statistically significantly different for men and for women (see, also, Appendix Table A33).

We next consider self-reported attitudes toward redistribution in society and self-reported altruism (Table 5). For all but one regression (altruism in the U.S. sample,

Table 5: CGI and Unincentivized Distributional Preferences

Measure	Redistribution		Altruism			
Sample	Swiss Uni (1)	U.S. Adults (2)	Swiss Uni (3)	U.S. Adults (4)	Swiss Teens (5)	Swedish Youths (6)
Panel A. Binary gender						
Female	0.201* (0.083)	0.063* (0.032)	0.249*** (0.082)	0.001 (0.032)	0.210*** (0.071)	0.551*** (0.060)
R ²	0.010	0.001	0.016	0.000	0.011	0.076
Panel B. Gender identity						
CGI (feminine)	0.106*** (0.041)	0.053*** (0.017)	0.095* (0.041)		0.074* (0.035)	0.224*** (0.032)
R ²	0.011	0.003	0.009		0.005	0.050
Panel C. Binary gender and gender identity						
Female	0.082 (0.155)	-0.072 (0.055)	0.279* (0.166)		0.412** (0.165)	0.489*** (0.099)
CGI (feminine)	0.073 (0.076)	0.083*** (0.029)	-0.018 (0.084)		-0.112 (0.082)	0.042 (0.052)
R ²	0.012	0.003	0.016		0.013	0.077
Panel D. Sample split by binary gender						
Female: CGI (feminine)	-0.136 (0.063)	-0.115 (0.023)	-0.047 (0.071)		-0.051 (0.054)	0.052 (0.046)
R ²	0.019	0.013	0.002		0.013	0.003
Male: CGI (feminine)	0.193*** (0.062)	0.196*** (0.024)	0.020 (0.068)		-0.046 (0.046)	0.013 (0.053)
R ²	0.037	0.038	0.000		0.002	0.000
Test: male sample split CGI = female sample split CGI						
p-value	0.000	0.000	0.477		0.916	0.534
Observations	584	3,902	584	3,902	798	1,041
Mean of Dependent Variable	0	0	0	0	0	0

Notes: The table presents results from regressing unincentivized distributional measures on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Redistribution is a measure of how much economic redistribution one wants in society. Altruism is a measure of how much one would donate out of a windfall gain. A higher value means a greater willingness to redistribute or donate. All outcome measures are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. Equality of sample split coefficients in Panel D are tested with a Wald test. *** $p < 0.005$, ** $p < 0.01$, * $p < 0.05$.

Column 4 in Panel A), we replicate statistically significant gender gaps commonly found in the literature. Further, Panel B indicates that CGI correlates with attitudes towards redistribution and altruism in the expected direction for the five outcome measures that move to this stage of the analysis. However, once controlling for binary gender in Panel C, the relationships between CGI and stated distributional preferences are statistically significant at $p < 0.005$ for only one measure, attitudes toward redistribution among U.S. Adults, and are not statistically significant for other measures. Moreover, the sign for the coefficient for CGI in Panel C has the wrong sign in two cases. As in Table 3, we again observe very small increases in R^2 when comparing Panels A and C.

Splitting the sample by gender in Panel D reveals stronger relationships for men. Two of five regressions for men and none for women yield statistically significant ($p < 0.005$)

coefficients for CGI in the expected direction (redistribution in the Swiss Uni and U.S. Adults samples). These gender differences are statistically significant (see the reported p -values of Wald tests in Table 5 and Appendix Table A34). This is consistent with our observation in Table 4 of stronger relationships between CGI and self-reported preference measures for men than for women.

5.3. Continuous Gender Identity and Educational Choices

Table 6 shows the relationships between CGI and educational outcome measures. Each outcome measure corresponds to the choice of a particular educational track, coded either for the share of women in a study field or apprenticeship (in each case, obtained separately from administrative data), or the math and language requirements of the apprenticeship (evaluated by an independent panel of experts).

Not surprisingly, given the construction of these outcome measures, we find substantial gender gaps in the gender composition (share of women) of the chosen or intended educational track in all three samples (see Panel A).²³ We also find that Swiss female adolescents tend to select apprenticeship tracks with lower math and higher language requirements compared to males. In Panel B, CGI is strongly correlated with all educational outcomes in all samples.

When we control for binary gender in Panel C, all the coefficients for CGI have the expected signs. The relationships with choosing a predominantly female educational track are highly statistically significant at $p < 0.005$ for two samples (U.S. Adults and Swiss Teens). The relationship with lower math and higher language requirements are statistically significant at $p < 0.05$ and $p < 0.01$, respectively. Looking at the changes in R^2 between Panels A and C, we again observe generally small increases, even when the coefficients are highly statistically significant. We thus conclude that CGI appears to be correlated with educational track choices and can account for additional variation in gender-typical educational choices beyond the one already accounted for by binary gender, though the strength and size of these relationships vary across our samples.

Turning to the within-gender relationships in Panel D, the association between CGI and educational choices is statistically significant at $p < 0.05$ in five of ten cases. Three out of these five coefficients of CGI are highly statistically significant at $p < 0.005$. We see no clear pattern that the relationships between CGI and educational choices in Panel C are primarily driven by one of the two genders. Consistently, all Wald tests fail to reject the null hypothesis that the estimated correlation coefficients are the same for boys and girls (see also Appendix Table A35).

²³Although it is worth noting that the gender share measures are constructed independently from administrative data, and not from the responses in our data. For the intended college major chosen by Swedish Youths, the gap is relatively small, which is likely due to the sample only including students in the science-math high school track.

Table 6: CGI and Educational Choice

Measure	Female Educational Track Share			Skill Requirements	
				Math	Language
Sample	U.S Adults (1)	Swiss Teens (2)	Swedish Youths (3)	Swiss Teens (4)	Swiss Teens (5)
Panel A. Binary gender					
Female	0.125*** (0.007)	0.383*** (0.012)	0.093*** (0.009)	-0.917*** (0.047)	0.913*** (0.047)
R ²	0.108	0.404	0.090	0.204	0.203
Panel B. Gender identity					
CGI (feminine)	0.061*** (0.004)	0.167*** (0.006)	0.039*** (0.005)	-0.403*** (0.024)	0.404*** (0.023)
R ²	0.104	0.315	0.063	0.163	0.163
Panel C. Binary gender and gender identity					
Female	0.075*** (0.014)	0.329*** (0.023)	0.079*** (0.014)	-0.762*** (0.084)	0.745*** (0.083)
CGI (feminine)	0.031*** (0.007)	0.033*** (0.011)	0.009 (0.007)	-0.092* (0.042)	0.100** (0.040)
R ²	0.117	0.408	0.092	0.207	0.206
Panel D. Sample split by binary gender					
Female: CGI (feminine)	0.009* (0.005)	0.028*** (0.009)	0.007 (0.006)	-0.028 (0.040)	0.135*** (0.038)
R ²	0.003	0.014	0.002	0.001	0.018
Male: CGI (feminine)	0.024*** (0.006)	0.009 (0.009)	0.005 (0.007)	-0.087** (0.035)	0.005 (0.035)
R ²	0.016	0.002	0.001	0.008	0.000
Test: male sample split CGI = female sample split CGI					
p-value	0.098	0.331	0.690	0.105	0.062
Observations	2,279	1,448	1,041	1,433	1,433
Mean of Dependent Variable	0.541	0.442	0.521	0	0

Notes: The table presents results from regressing the educational track measures on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. U.S. Adults (age 30–60), Swiss Teens and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor’s degree in the chosen field of study in 2020 for U.S. Adults; the share of women from past cohorts of graduates from the apprenticeship for Swiss Teens; the share of women accepted for undergraduate studies in that field the year before our sample made their educational choices for Swedish Youths. Skills requirements are a standardized (mean=0, SD=1) measure based on expert evaluation of the job content in occupations chosen by Swiss Teens to start apprenticeships. Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. Equality of sample split coefficients in Panel D are tested with a Wald test. *** p<0.005, ** p<0.01, * p<0.05.

5.4. Continuous Gender Identity and Labor Market Outcomes

To investigate the relationships between CGI and labor market outcomes, we asked participants in the U.S. Adults sample to provide information on several measures related to labor market outcomes and behaviors (Table A3). To discipline our selection of outcome measures for further study, we implement a selection procedure to minimize false positive relationships between outcome measures and gender for the first step of our analysis. Specifically, for every survey item related to employment and labor market outcomes, we

test whether there is a significant gender gap, correcting for multiple-hypothesis testing (MHT) using the Romano and Wolf (2005) approach. If there is a significant gender gap at the $p < 0.05$ level after this correction, we retain that outcome variable for the second and third stages of our analysis involving CGI and other gender identity measures (in Subsection 5.7). Of the 11 independent labor market outcome measures we collected for U.S. Adults, we obtain eight statistically significant gender gaps using this approach.²⁴ We then perform the same analysis as for our earlier outcome measures on these eight measures of labor market outcomes. Table 7 presents the results.

Given the procedure for selecting these eight variables, we unsurprisingly observe highly statistically significant gender gaps for all outcomes in Panel A. Compared to men, women earn less, work fewer hours, and are less likely to have managerial responsibilities, be rewarded with variable incentives and to have engaged in wage negotiation. Women are also more likely to be full-time homemakers, work in industries with high female shares, and have jobs that involve flexible work hours. In Panel B, we observe that all the relationships between CGI and these outcomes are also highly statistically significant and have the expected sign.

Panel C shows that, conditional on binary gender, CGI correlates with three outcomes at $p < 0.005$. Accounting for binary gender, individuals who report a more feminine identity earn less, are more likely to be full-time homemakers, and work fewer hours. Two additional outcomes are statistically significant at respectively $p < 0.01$ and $p < 0.05$. This indicates that at least some categories of labor market outcomes have highly statistically significant associations with gender identity, beyond the relationship with binary gender. However, once again we observe that even the relationships that are statistically significant yield modest increases in R^2 in absolute terms—typically less than half a percentage point—when adding CGI as an explanatory variable.

The split-sample analysis in Panel D indicates that the correlation of CGI with the likelihood of being a full-time homemaker is driven primarily by women, while the remaining significant correlations are driven by men.²⁵ In fact, for men, six out of eight relationships with CGI are significant at $p < 0.05$, of which five are significant at $p < 0.005$. Thus, consistent with earlier observations, we again observe relationships that are stronger for

²⁴Appendix Table A3 shows the results from this exploration of potential gender gaps in labor market outcomes for Wave 1 of the U.S. Adults sample, reporting p -values corrected for MHT. We adopted the above selection procedure in Wave 1 of data collection in our U.S. sample; for Wave 2, we only collected labor market-related outcomes that had survived the MHT correction for Wave 1. For the category of variables, *Employment Status*, we find two response categories with substantial gender differences after the MHT correction: whether a respondent is *employed* (by someone else) and whether the respondent is a *full-time homemaker*. Given the relationship between these two response categories (participants could only select one), we retain only the latter one for our analysis (as this more likely reflects an individual’s choice to forgo employment). We also performed this analysis on the educational outcomes for U.S. Adults (see Section 5.3), finding the female share in the reported study field to differ substantially between male and female respondents after controlling for MHT.

²⁵See the reported p -values of Wald tests in Table 7 and results from a model with female binary gender interacted with CGI in Appendix Table A36.

Table 7: CGI and Labor Market Outcomes

Measure	Income	Full-time Homemaker	Weekly Ave. Work Hours	Female Industry Share	Managerial Responsibilities	Performance Pay	Wage Negotiation	Work Flexibility
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A. Binary gender								
Female	-23.265*** (1.865)	0.099*** (0.008)	-6.644*** (0.451)	0.062*** (0.007)	-0.125*** (0.016)	-0.082*** (0.021)	-0.132*** (0.016)	0.141*** (0.045)
R ²	0.038	0.043	0.053	0.039	0.016	0.008	0.017	0.005
Panel B. Gender identity								
CGI (feminine)	-10.841*** (0.958)	0.049*** (0.004)	-3.141*** (0.222)	0.025*** (0.003)	-0.061*** (0.008)	-0.043*** (0.010)	-0.066*** (0.008)	0.060*** (0.022)
R ²	0.033	0.042	0.047	0.026	0.015	0.009	0.017	0.004
Panel C. Binary gender and gender identity								
Female	-16.723*** (2.990)	0.057*** (0.012)	-4.553*** (0.806)	0.060*** (0.011)	-0.078*** (0.027)	-0.038 (0.035)	-0.074** (0.028)	0.126 (0.077)
CGI (feminine)	-3.998*** (1.542)	0.026*** (0.006)	-1.278*** (0.397)	0.001 (0.006)	-0.029* (0.014)	-0.028 (0.017)	-0.035** (0.014)	0.009 (0.039)
R ²	0.040	0.047	0.055	0.039	0.018	0.009	0.019	0.005
Panel D. Sample split by binary gender								
Female: CGI (feminine)	2.416 (1.124)	0.025*** (0.007)	-0.155 (0.336)	-0.004 (0.005)	0.027 (0.011)	-0.012 (0.014)	0.002 (0.011)	-0.079 (0.033)
R ²	0.002	0.006	0.000	0.001	0.003	0.001	0.000	0.006
Male: CGI (feminine)	-6.735*** (1.356)	0.005* (0.002)	-1.281*** (0.312)	0.005 (0.005)	-0.058*** (0.011)	-0.020 (0.015)	-0.041*** (0.011)	0.085*** (0.032)
R ²	0.011	0.002	0.009	0.001	0.014	0.002	0.007	0.007
Test: male sample split CGI = female sample split CGI								
p-value	0.000	0.005	0.018	0.145	0.000	0.767	0.009	0.000
Observations	3,902	3,902	3,902	1,994	3,902	1,994	3,902	1,994
Mean of Dependent Variable	61.421	0.062	25.281	0.497	0.400	0.303	0.491	0

Notes: The table regresses the labor market outcomes on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. The sample of the U.S. Adults (age 30–60) is used. The estimates in each column and panel come from a separate regression. Income is a self-reported categorical measure in thousands of U.S. dollars. Full-time homemaker is a dummy for working full-time at home. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. Female industry share is the share of female employees in a given industry. Managerial responsibilities is a dummy for having managerial responsibilities at work. Performance pay is a dummy that is equal to 1 if the respondent’s current or most recent job has performance-related pay. Wage negotiation is a dummy that is equal to 1 if the respondent ever negotiated wage. Work flexibility is a categorical measure of flexibility in working hours where 0 is no flexibility, 0.5 is can adapt work hours and 1 is full flexibility, which is standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. Equality of sample split coefficients in Panel D are tested with a Wald test. *** p<0.005, ** p<0.01, * p<0.05.

men.

5.5. Robustness Checks

We test the robustness of the main results in Tables 3 to 7 in four different ways. We report the results from these analyses in detail in the Appendix.

First, we test whether our results are robust to the inclusion of controls. Appendix Tables A4 to A8 show that adding varying controls for the different samples—including age, educational attainment and, for the U.S. Adults sample, ethnicity and geographical fixed effects—to the regressions in Appendix Tables 3 through 7 yields no substantive changes to the results. For example, when we add controls to Table 3 (incentivized behavioral measures) we observe that the coefficients all retain the predicted signs but,

as with Table 3, none are statistically significant at $p < 0.05$ (Appendix Table A4). In Appendix Tables A5 through A8, which correspond to Tables 4 through 7, the statistical significance of some coefficients increases or decreases slightly when adding controls, but there are no substantive changes.

Second, we test for possible non-linearities in the relationships between CGI and the outcome measures reported in Tables 3 through 7. We do so by replacing the linear CGI measure in the regressions with binary indicators for each possible response on the CGI scale. We present the outcome of this analysis in Appendix Tables A9 to A13. For each outcome variable, we focus on an F-test of the joint significance of all the binary CGI indicators when controlling for binary gender in Panel C. These tests sometimes yield greater statistical significance than our primary analysis. For example, the non-linear estimation yields a relationship for the unincentivized risk for Swedish Youths and redistribution for Swiss Uni that are statistically significant at $p < 0.005$ (Appendix Tables A10 and A11). Otherwise, the relationships in which CGI has at least a marginally statistically significant relationship when accounting for binary gender are very similar across both types of analyses. Thus, the linear inclusion of CGI does not seem to mask important non-linearities.

Third, we investigate the potential role of measurement error in CGI, which might bias coefficient estimates for this variable downward. For the first dataset we collected—Swiss Uni—we use the second elicitation of CGI (obtained about two weeks after the main study) to implement the Obviously Related Instrumental Variable Approach (ORIV) by Gillen, Snowberg and Yariv (2019) to account for possible measurement error in the CGI variable.²⁶ As shown in Appendix Table A14, applying the ORIV method does not substantively change our results compared to the OLS regressions, though there are a few instances where the statistical significance of the estimated coefficients changes.²⁷

Fourth, for the sample of U.S. Adults, we perform the analysis for all outcome measures that we elicited in both waves independently for each wave of data collection. Appendix Tables A15 and A16 reveal typically very similar results for all the outcome variables considered in both waves, but collected one year apart and with reversed order in which CGI and outcomes were measured. The signs of the coefficients in Panel C always have the same sign when estimated independently for the two waves and for six of nine cases where a coefficient is statistically significant in one wave it is also statistically significant in the other wave, though the statistical significance sometimes varies across

²⁶This approach instruments the original CGI measure with its second slightly modified measurement from the follow-up survey two weeks later.

²⁷In particular, the coefficient for staircase risk is no longer statistically significant when using ORIV (though it was significant at $p < 0.05$ in Table 4), while the unincentivized competitiveness measure becomes highly statistically significant ($p < 0.005$) and the unincentivized altruism measure becomes significant at $p < 0.05$ when using ORIV (though neither is statistically significant in our primary analysis).

5.6. Analysis of Combined Results

Our main analysis provides 33 tests of the relationships between various economic outcomes and binary gender and CGI. The results in Tables 3 through 7 reveal that the relationships between CGI and the outcome measures, when accounting for binary gender, are sometimes highly statistically significant and other times not. In this section, we attempt to discern patterns of significance across these multiple comparisons, and to assess the overall value that a simple measure of gender identity has for understanding economic outcomes.

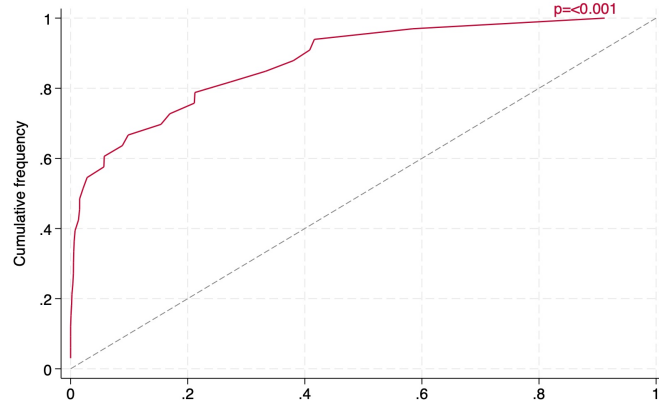
We first look at the full distribution of p -values obtained from our primary tests. Panel A of Figure 2 shows the cumulative distribution of p -values from all the primary tests reported in Subsections 5.2 through 5.4—that is, from all the coefficients for CGI in Panel C in Tables 3 through 7. Under the null hypothesis that CGI provides no additional information once controlling for binary gender, we would expect a uniform distribution of p -values. Our alternative hypothesis predicts a disproportionate prevalence of low p -values. The distribution clearly reveals stronger statistical relationships than one would expect by chance: a Kolmogorov-Smirnov test of the cumulative distribution strongly rejects the null hypothesis of uniformity ($p < 0.001$, one-tailed). Thus, viewing our analysis jointly, we reject that introducing a measure of continuous gender identity yields no significant relationships with the outcomes and behaviors we study, after controlling for binary gender.

Panel B of Figure 2 shows the cumulative distribution of p -values when splitting the sample by binary gender (the separate results from the regressions in Panel D of Tables 3 through 7). This graph clearly illustrates that most of the relationships between the outcome variables and CGI are driven by men. The distribution of p -values for men is skewed toward the left, and is highly statistically significantly different from the uniform distribution (KS test $p < 0.001$). For women, the cumulative distribution is more evenly distributed across all the possible values from 0 to 1, and we can reject the null hypothesis of uniformity only at the $p < 0.05$ level (KS test $p = 0.026$).

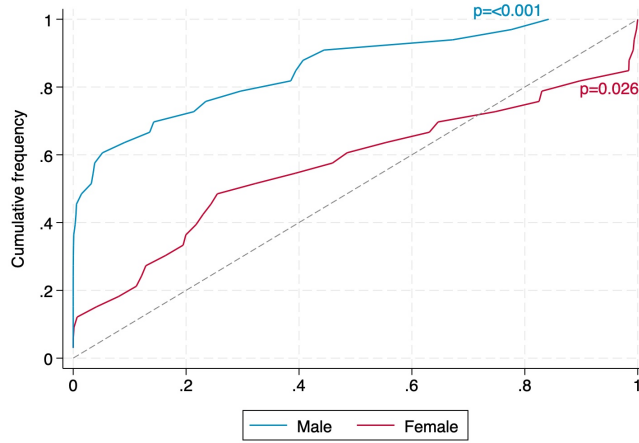
Next, we explore the change in the proportion of variance explained (R^2) in our outcome measures when we add CGI as an explanatory variable (in Panel C of the regressions in Tables 3 to 7), relative to when we only include the binary gender indicator as an explanatory variable (in Panel A of the same tables). Appendix Figure A3 plots this increase in R^2 for all 33 outcome variables. Consistent with our earlier observations,

²⁸For example, Wave 2 primarily drives the statistical significance of CGI for having managerial responsibilities in Panel C of Table 7 ($p < 0.05$). Conversely, the statistical significance for the CGI coefficients on financial risk in Panel C of Table 4 ($p < 0.05$) and attitudes to redistribution in Panel C of Table 5 ($p < 0.005$) are primarily driven by Wave 1.

Figure 2: Significance of CGI Across Regressions (Tables 3 through 7)



(a) Distribution of p -values of CGI when accounting for binary gender



(b) Distribution of p -values of CGI by gender

Notes: Panel A presents the cumulative distribution of p -values from all the coefficients for CGI in Panel C in Tables 3–7. Panel B shows the cumulative distribution of p values when splitting the sample by gender (p -values from all the coefficients for CGI in Panel D in Tables 3–7.). These p -values are computed with a one-tailed t -test in the direction of the coefficient estimate for female in panel A of Tables 3–7. The p -values on the plot region are obtained using a Kolmogorov-Smirnov test with the alternative hypothesis that the empirical cumulative distribution of p -values from the regression estimates are greater (smaller p -values) than the uniform cumulative distribution.

the increase is typically small in absolute terms. For example, the increases in R^2 are never larger than one percentage point. However, it is important to note that the original R^2 values in Panel A of Tables 3 to 7 also tend to be small (only four are above 0.1), meaning that the changes in relative terms are often not trivial.

5.7. Comparing Different Measures of Gender Identity

Finally, we investigate whether the broader conclusions that we draw about the relationships between economic outcomes, binary gender and continuous gender identity change depending on the exact measure of gender identity we use. For this, we systematically compare the predictive power of CGI to other gender identity measures for 10 outcomes

studied in Wave 2 of the U.S. Adult sample, where we also collected these alternative measures of gender identity.²⁹

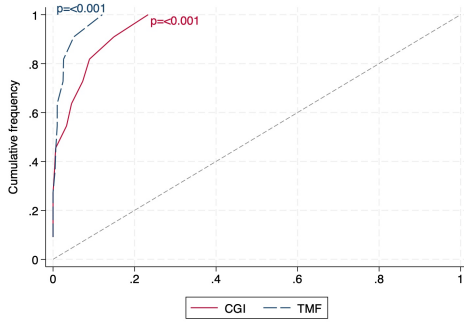
To focus on broader conclusions, rather than comparisons with respect to individual outcome measures, we consider the cumulative distributions of p -values obtained when using each measure of gender identity as an explanatory variable in regressions that also include binary gender—i.e., analogous to the regressions in Panel C of our main analysis. Figure 3 presents these cumulative distributions, while the underlying regressions are reported in greater detail in Appendix Tables A17 through A27. Each graph in Figure 3 shows the CDF of p -values for our measure of CGI (in red) together with an alternate gender identity scale (in blue). For the four alternate measures that elicit masculinity and femininity as two separate dimensions of gender identity, the distributions of p -values of the corresponding coefficients are shown separately. The cumulative distributions of our primary CGI measure differ somewhat across panels because we restrict the analysis in each graph to those sub-samples of respondents for which we also collected the alternate measure of gender identity.

Overall, we see little evidence that these alternate scales have substantially stronger correlations with our outcome measures than our single-item CGI scale. Using two-sided Kolmogorov-Smirnov tests to compare the equality of the distributions in each graph, we reject equality at $p < 0.05$ between the distributions of CGI and the other measures only in four cases—in two of which CGI tends to have lower p -values than the alternate scale (CGI vs Magliozzi femininity, Panel B, $p=0.021$; CGI vs OSRI femininity, Panel D, $p = 0.004$). Conversely, two scales measuring masculinity tend to have lower p -values than CGI (CGI vs. BSRI masculinity, Panel G, $p = 0.021$; CGI vs. CMNI, Panel F, $p = 0.021$). One reason why these two masculinity scales do well appears to be due to their inclusion of self-evaluations of the respondent’s risk-taking propensity. For instance, the BSRI elicits an evaluation of whether the respondent is “willing to take risks” as one of the 10 items indicating masculinity, while the CMNI includes three items measuring risk-taking (“I enjoy taking risks,” “I put myself in risky situations,” “I take risks”). Responses to these items correlate strongly with our incentivized and unincentivized risk measures and with stated competitiveness (all $p < 0.001$). Thus, we conclude that these alternate measures, despite their substantially lengthier nature, do not generally correlate more strongly than our CGI measure with our outcome measures, beyond a few exceptions in which the scales appear to directly measure a relevant behavioral trait.

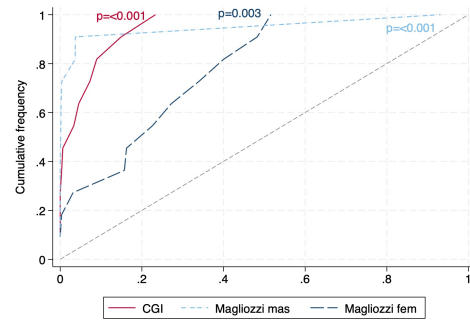
An interesting regularity in Figure 3 is that subscales measuring masculinity (rather than femininity) tend to have stronger relationships with the outcome measures, with distributions of p -values with greater mass on lower values. For the three femininity scales

²⁹The 10 outcomes comprise three risk measures (one incentivized and two unincentivized), stated competitiveness, female educational track share, income, full-time homemaker status, weekly average working hours, managerial responsibilities and wage negotiation.

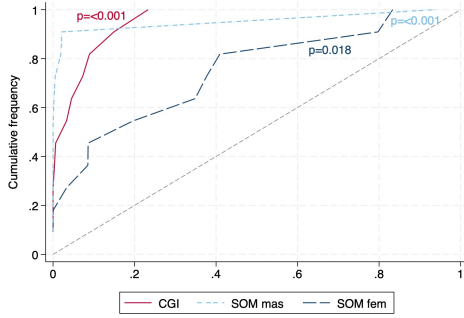
Figure 3: Significance of Different Gender Identity Measures Across Regressions



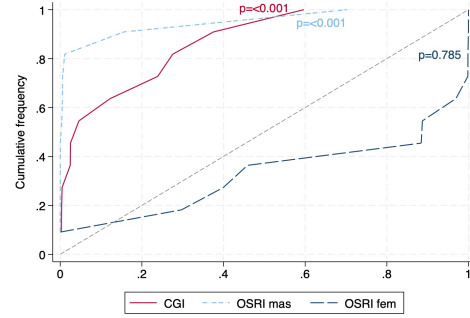
(a) TMF



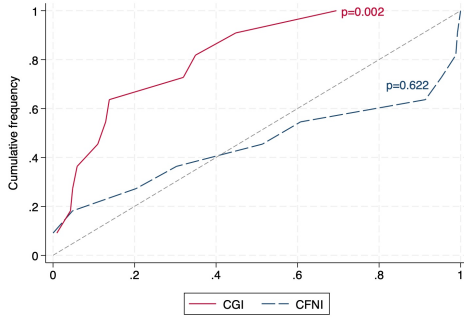
(b) Magliozzi



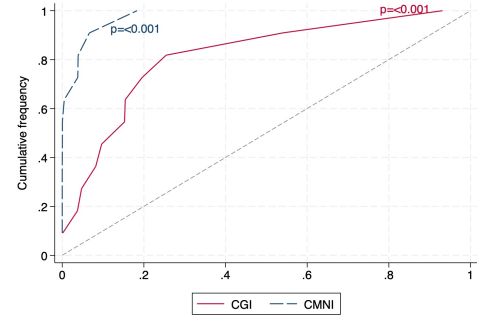
(c) SOM



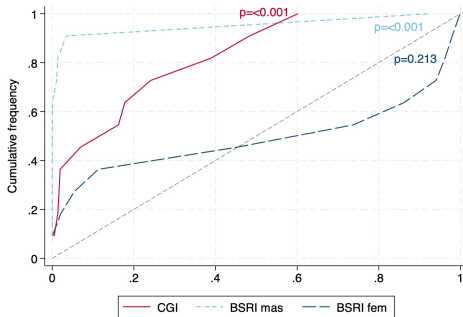
(d) OSRI



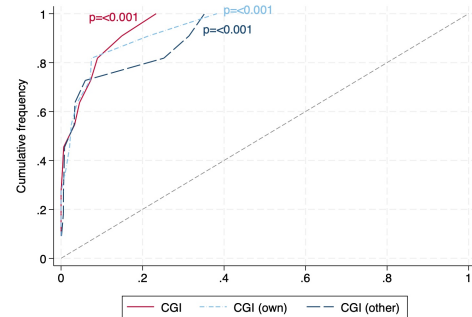
(e) CFNI



(f) CMNI



(g) BSRI



(h) Other CGI Measures

Notes: Each graph shows the cumulative distribution of p -values of gender measures (Panel C of Appendix Tables A17-A27). The p -values on the plot region are obtained using a Kolmogorov-Smirnov test with the alternative hypothesis that the empirical cumulative distribution of p -values from regression estimates are greater (smaller p -values) than the uniform cumulative distribution. The distribution of p -values for CGI varies slightly across panels since each panel uses the same sub-sample of respondents who also reported the alternate gender identity measure.

constructed from self-reported agreement with statements about female gender norms or gender-typical behavior—OSRI (Panel D), CFNI (Panel E), BSRI (Panel G)—we fail to reject that the CDF of p -values is uniform against the alternative hypothesis that it is less than uniform (KS tests $p > 0.213$). Moreover, when a scale contains both femininity and masculinity subscales, the latter tends to yield lower p -values as explanatory variables, and this difference is statistically significant in every case (Magliozzi, Panel B, $p = 0.021$; SOM, Panel C, $p = 0.004$; OSRI, Panel D, $p < 0.001$; BSRI, Panel G, $p = 0.004$). This pattern is generally consistent with our earlier observations that gender identity has stronger relationships with our outcome measures for men than for women.

As a final comparison, we consider the predictive power of our standard CGI measure to two alternate versions of our scale, in which we ask individuals to indicate how others see them on our scale from very masculine to very feminine—CGI (other)—and how they see themselves compared to others with the same binary gender, e.g., compared to other men when the respondent is a man—CGI (own). The CDFs of p -values look very similar (see Panel H of Figure 3) and do not differ significantly ($p > 0.833$ for all three comparisons), indicating that our broad conclusions do not change when eliciting CGI in slightly different ways.

6. Conclusion

We investigate the relationships between gender identity and preferences, behavior and outcomes across several samples of respondents. We focus on outcome measures that are particularly interesting for economists and for which previous research documents evidence of gaps between men and women. We present results from four distinct datasets in which we investigate the association between a validated measure of continuous gender identity (CGI) and economic preferences and outcomes, beyond the explanatory power of binary gender. The datasets comprise respondents across three countries and different age groups and elicit as outcome variables both incentivized and unincentivized measures of economic preferences and several dimensions of educational, labor market and workplace choices and outcomes.

For both genders and across all the samples we study, we find substantial variation in reported gender identity. In each sample, both men and women report varying degrees of masculinity and femininity, and the distributions of self-reported gender identity for men and women always overlap. This suggests that such broad and overlapping distributions of gender identity may provide an opportunity for identity to account for some of the variation in outcomes and behavioral tendencies among men and women.

Overall, we observe a pattern of moderate correlations between continuous gender identity and our outcome measures, once accounting for binary gender. Viewed jointly, the pattern of results across all our tests strongly rejects the null hypothesis of no cor-

relation between gender identity and economic outcome measures beyond binary gender. We observe the strongest statistical relationships for unincentivized measures of risk-seeking and competitiveness, the choice of gender-stereotypical educational tracks, and labor market outcomes like income, seeking work outside of the home, and number of work hours.

Interestingly, we find systematic differences when we consider men and women separately, thereby documenting patterns that have not previously been documented in gender economics. We find that our measure of gender identity is more strongly correlated with economic behaviors and outcomes for men than for women. One possible reason for this pattern could be that norms of appropriate economic behavior for women have relaxed over the last decades as women’s possibilities to pursue a career outside of the home, and thereby their economic agency, have increased. Norms about appropriate economic behavior for men may still rely primarily on their role in the labor market. We believe that these findings raise several interesting questions for future research. During the last decades, for instance, women have increasingly been entering STEM fields, while the entrance of men in traditionally female-dominated fields has remained low (Delfino, 2024). To what extent do societal constraints on men’s compared to women’s gender identity explain such patterns?

Broadly, we interpret our evidence as mixed in terms of the value of measures of self-reported gender identity for economic research. Our results clearly indicate that such measures correlate with economic behaviors and outcomes, even when controlling for binary gender, to a substantially greater degree than what is expected by chance. However, they tend to provide limited improvements in accounting for variation in the outcome measure conditional on binary gender. Our results identify those domains of interest to economists studying gender gaps which have the strongest and the weakest relationships between measures of gender identity and economic behaviors. A true advantage of nuanced measures of self-reported gender identity is that one can study the relationships between economic behaviors and gender identity separately for women and men. Given the novel and perhaps surprising patterns we find in this first systematic investigation of these relationships, we believe that it is a promising avenue for future research. We think that these relationships can point to domains of economic behavior in which societal views on gender-typical behavior may be particularly rigid.

We also provide evidence that our single-item measure of continuous gender identity generally correlates well with other measures used in gender studies. Thus, we find little reason to justify using richer measures, including ones that measure gender identity indirectly, use multiple items, or that elicit separate dimensions of masculinity and femininity. This is particularly true when considering the additional potential costs of including more items in surveys. Of course, we admit that it is entirely possible that such measures may have substantially stronger relationships with other economic behaviors and outcomes

that we do not study, which is a question for future research. We believe that all our findings together suggest that there is some value in utilizing our CGI measure in future research on gender gaps in economics, particularly when the cost is low.

Finally, we believe that our research highlights the value of a growing body of studies that investigate notions of gender and gender identity that depart from traditional classifications as “male” and “female.” The substantial variation in identity that we observe across samples suggests that such identification is a potentially important individual characteristic and that future research should further investigate domains in which such identity is both an important influence on outcomes and behavior and where these, in turn, affect individuals’ identities.

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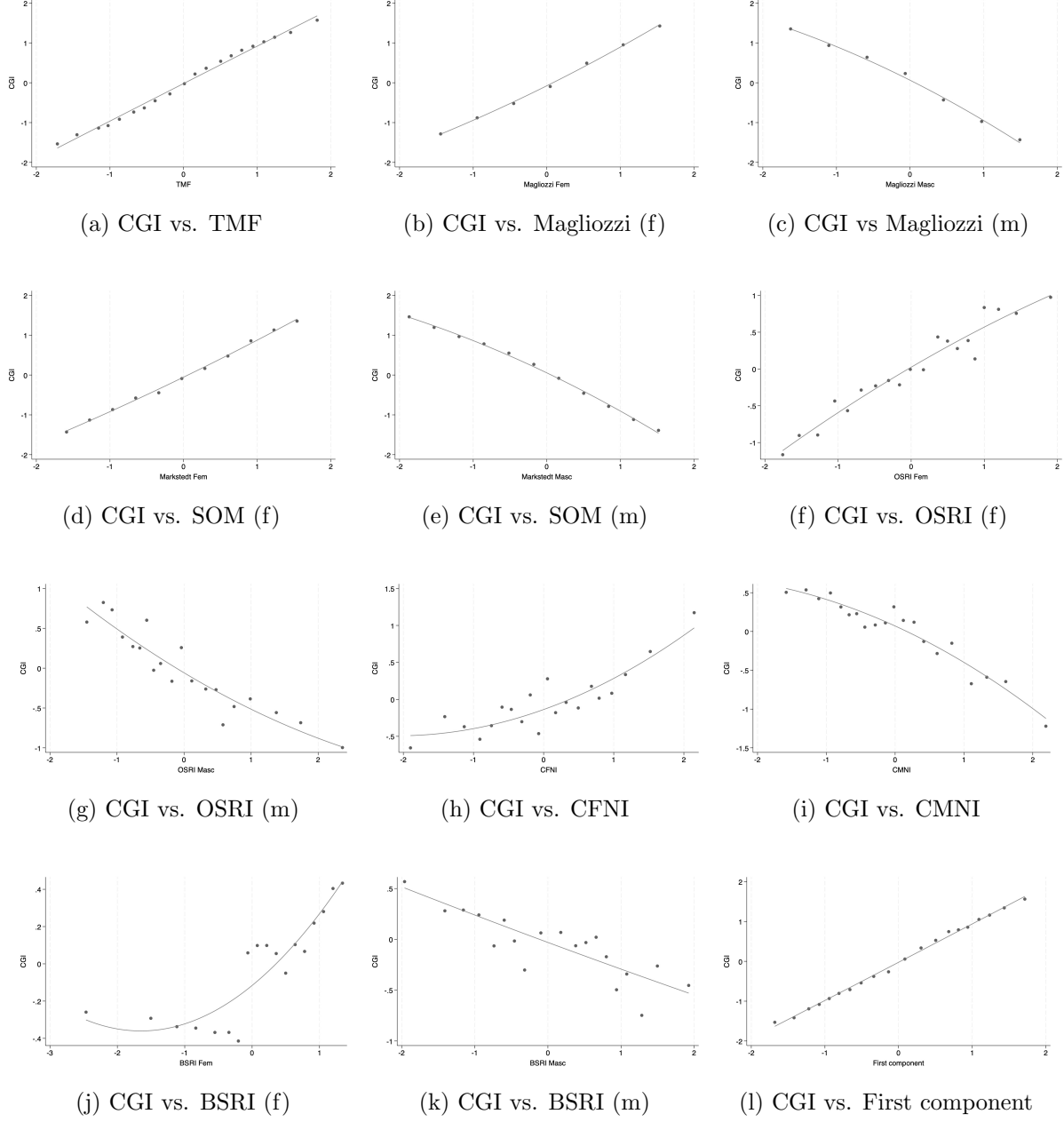
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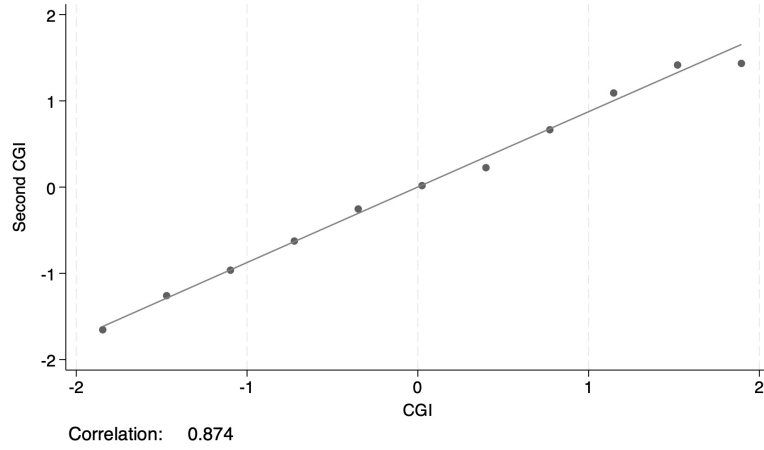
A. Appendix A: Tables and Figures

Figure A1: Binned Scatterplots of CGI vs. Alternative Gender Identity Measures



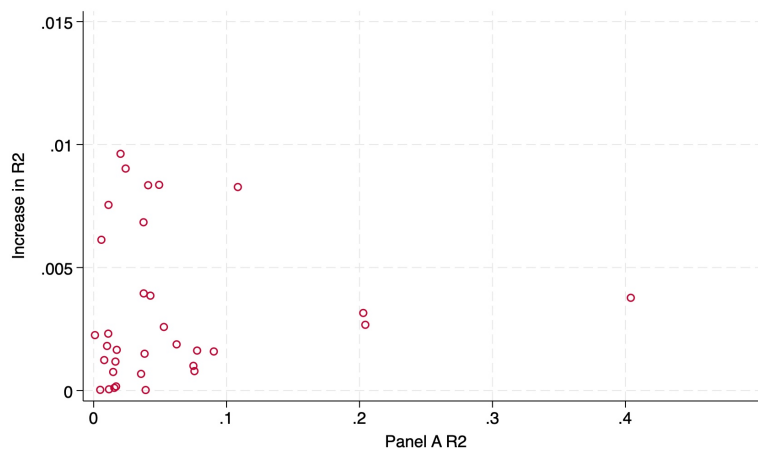
Notes: The figure presents the binscatter plots of CGI against alternative gender identity measures. U.S. Adults wave 2 (age 20–60) sample is used. CGI and TMF are unidimensional scales, with higher values indicating higher femininity. “f” refers to the femininity and “m” refers to the masculinity score of the two-dimensional scales BSRI, OSRI, Maglozzi and SOM. CFNI and CMNI are the conformity to feminine and masculine norm inventories. *First comp* combines the first factor from a principal component analysis of all gender identity scales (excluding CGI). N=2,659 for CGI, TMF, Maglozzi and SOM measures; N=662 for OSRI measures; N=659 for CFNI; N=662 for CMNI; N= 676 for BSRI measures. Quadratic line fit is added to the binscatter plot, where each scatter point represents the average of both gender identity measures in each bin.

Figure A2: Scatterplot of Separate Measurements of CGI



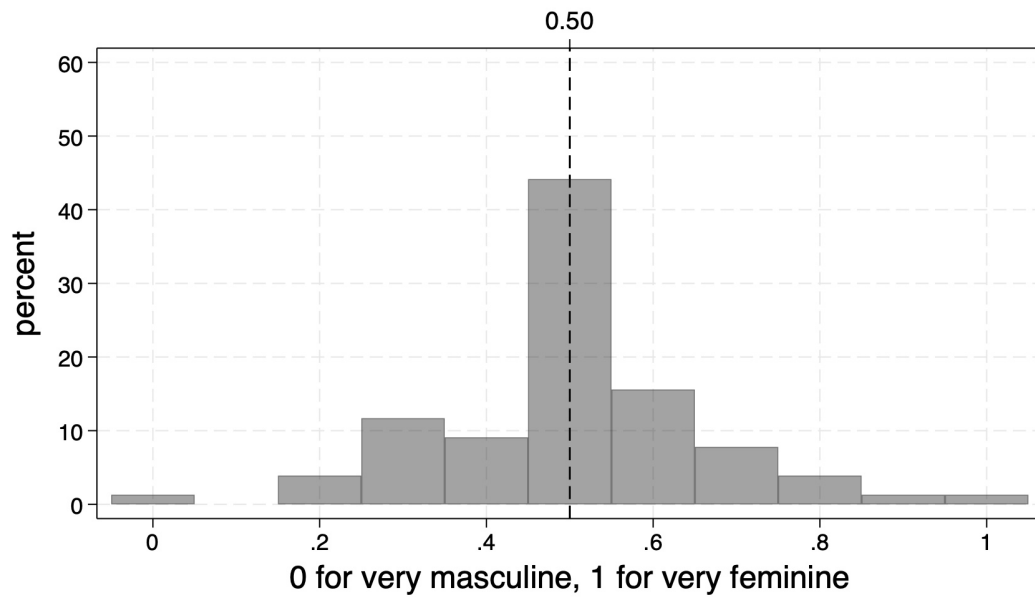
Notes: Swiss Uni sample (N=584) is used. CGI represents our first measurement, ranging from “very masculine” to “very feminine”. Second CGI represents the same question asked two weeks later, with a slightly modified scale and direction (“very feminine” to “very masculine”). Responses for second CGI are inverted and both measures are standardized (Mean=0, SD=1). Quadratic line fit is added to the binscatter plot, where each scatter point represents the average of both CGI measures in each bin.

Figure A3: Improvement in R^2



Notes: The figure plots R^2 obtained in a regression of an outcome variable on a binary gender indicator (Panel A) against the increase in R^2 when CGI is added as an explanatory variable (Panel C). Each dot corresponds to one of our 33 outcome measures presented in Tables 3 through 7.

Figure A4: CGI of Non-Binary Participants



Notes: Histogram of CGI of participants identifying as non-binary. Swiss Uni (N=8), US Adults wave 2 (N=58) and Swedish Youths (N=12) samples are used. CGI represents our single question, ranging from “very masculine” to “very feminine”. CGI is rescaled from 0 to 1 where 1 is very feminine. The dashed line marks the sample mean.

Table A1: Detailed Overview of Main Outcome Variables

Variable	Description
<i>Independent variables</i>	
Sex	Binary classification of the sexes taking the value 1 for women.
Continuous gender identity (CGI)	The participant's self-reported gender identity on a scale from 0 (very masculine) to 10 (very feminine).
<i>Incentivized preferences</i>	
Risk	Measured through the Gneezy and Potters (1997) investment task. Our measure of risk taking is the amount a participant chooses to invest in the risky investment. Standardized based on an investment ranging from 0 to 100.
Competitiveness	Measured using the Niederle and Vesterlund (2007) approach. Binary variable taking the value 1 if the participant chose to compete.
Efficiency	Measured using the Fisman, Kariv and Markovits (2007) approach. We use the value of ρ split by decile as our measure, with lower values indicating stronger equality versus efficiency focus.
Overconfidence	Measured as relative overplacement following Gillen, Snowberg and Yariv (2019). The variable we use is a standardized summary index of three overplacement measures (participants rank guess with respect to relative performance on a series of matrix reasoning items, relative performance on the competitive task (in session and in group)).
<i>Non-Incentivized preferences</i>	
Staircase risk	Risk preferences measured as the certainty equivalence arising from the hypothetical risk module proposed by Falk et al. (2023). Standardized.
Risk	Self-reported general willingness to take risks based on the question validated in Dohmen et al. (2011). Standardized based on an answer scale from 1-10, where higher numbers indicate higher risk tolerance.
Financial risk	Binary variable taking the value of 1 if the respondent actively trades in securities.
Competitiveness	Self-reported general willingness to compete based on the question validated in Buser, Niederle and Oosterbeek (2021). Standardized based on an answer scale from 1-10, where higher numbers indicate higher competitiveness.
Redistribution	Self-reported preference for redistribution in society. Standardized based on an answer scale from 1-10, where higher numbers indicate preference for more redistribution.
Altruism	2 measures depending on dataset. 1. Self-reported amount donated from windfall gain. Standardized measure based on a donation ranging from 0-2400 (US Adults, Swiss Uni, Swiss Teens). 2. Self-reported willingness to donate to charity on a scale from 0-10. Standardized (Swedish Youth).
<i>Educational/Ocupational outcomes</i>	
Female educational track share	For U.S. Adults: share of women graduating with a bachelor's degree in a chosen college major field of study in 2020. For Swiss Teens: share of female apprentice graduates within the specific profession from 2019-2021. For Swedish Youths: share of women admitted in previous year (2022) to chosen educational field.
Math/language skill requirement	Math and language skills requirements with respect to occupations corresponding to the apprenticeships chosen by Swiss Teens, based on expert evaluation of job content. Standardized.
Income	Categorical variable based on self-reported income during the last 12 months (Categories 0-5000, 5001-10.000, 10.001-25.000, 25.001-50.000, 50.001-100.000, 100.001-250.000, >250.001)
Full-time homemaker	Binary variable taking the value of 1 if the participant reports being a full-time homemaker.
Weekly average work hours	Proxy measure constructed from self-reported number of weeks worked and average weekly hours.
Female industry share	Share of women working in the respondent's reported industry.
Managerial responsibilities	Binary variable taking the value of 1 if the survey respondent has managerial responsibilities at current/most recent employment.
Performance pay	Binary variable taking the value of 1 if the survey respondent has performance-related pay at current/most recent employment.
Wage negotiation	Binary variable taking the value of 1 if the respondent ever negotiated the wage.
Work flexibility	Categorical measure of flexibility in working hours where 0=No flexibility, 0.5=Can adapt work hours or Choose fix work hours, 1=Full flexibility. Standardized.

Notes: This table gives a brief description of each variable used for the analysis presented in tables 3 through 7. Please see Appendix B for more details. All standardized variables have a mean=0 and SD=1.

Table A2: Sample Selection

	Swiss Uni	U.S wave 1 (30-60)	U.S wave 2 (20-29)	U.S wave 2 (30-60)	Swiss Teens	Swedish Youth
Initial sample	597	2,002	800	2000	1,755	1,053
Non-binary gender	8	-	30	28	-	12
Inconsistent binary gender	5	8	12	25	15	-
Attention check fail	-	0	7	39	-	-
Final sample	584	1,994	751	1,908	1,740	1,041

Notes: This table provides an overview of how many observations were not included in the final sample for which reasons. *Inconsistent binary gender* means that there was any inconsistency across several measures of binary gender or binary sex we had from a respondent, for example when we compared self-reports in our survey to records obtained from Prolific Academic. *Non-binary gender* means that a person reported a non-binary gender category, for example “other”. *Attention check fail* means that a respondent failed at least one of two attention checks.

Table A3: Descriptive Statistics: Education and Labor Market

	Mean		Difference
	Male	Female	
	(1)	(2)	(3)
Education			
No Schooling Completed	0.00	0.01	0.01
Regular High School Diploma	0.09	0.08	-0.01
GED or Alternative Credential	0.02	0.02	0.00
Some College Credit (less than 1 year)	0.06	0.07	0.01
1 or more Years of College Credit, No Degree	0.11	0.10	-0.00
Associate's Degree	0.08	0.11	0.03
Bachelor's Degree	0.42	0.38	-0.04
Master's Degree	0.16	0.18	0.02
Doctorate Degree	0.03	0.02	-0.01
Professional Degree Beyond a Bachelor's Degree	0.02	0.02	-0.00
Female Educational Track Share	0.49	0.61	0.11***
Labor Market			
<i>Employment Status (respondents selected one response category)</i>			
Employed	0.73	0.59	-0.14**
Self-employed	0.14	0.15	0.01
Relative Assisting on a Farm or Business	0.00	0.00	-0.00
In Full Time Education (at school, university, etc.)	0.01	0.01	-0.00
Full Time Homemaker	0.01	0.11	0.10**
Currently on Child-care Leave or other Leave	0.00	0.00	0.00
Unable to Work due to Long Term Illness or Disability	0.03	0.04	0.01
Retired	0.01	0.02	0.01
Unemployed	0.06	0.07	0.01
Other	0.00	0.01	0.00
<i>Sector and Industry</i>			
Private Sector Employee	0.83	0.82	-0.01
Female Industry	0.47	0.53	0.06**
<i>Work Conditions</i>			
Income (in thousand US dollars)	73.83	52.25	-21.58**
Weekly Ave. Work Hours	29.16	21.55	-7.61**
Work flexibility (0 No flexibility, 1 Full flexibility)	0.45	0.50	0.05*
Changes to Working Arrangements (0 No, 1 Yes)	0.32	0.29	-0.03
Take off an Hour for Personal Matters (0 Very difficult, 4 Very easy)	3.14	3.07	-0.07
Managerial Responsibilities (0 No, 1 Yes)	0.47	0.34	-0.14**
Perfomance Pay (0 No, 1 Yes)	0.34	0.26	-0.08**
Wage Negotiation (0 No, 1 Yes)	0.48	0.36	-0.11**
Observations	999	995	

Notes: Multiple hypothesis test correction was done separately for the family of variables Labor Market and Education.
US Sample - Wave 1 is used. Significance levels: *** p < 0.005, ** p < 0.01, * p < 0.05

Table A4: CGI and Incentivized Behavioral Measures (with controls)

Measure	Risk		Competitiveness	Efficiency		Overconfidence
Sample	Swiss Uni (1)	US Adults (2)	Swiss Uni (3)	Swiss Uni (4)	Swedish Youths (5)	Swiss Uni (6)
Panel A. Binary gender						
Female	-0.383*** (0.082)	-0.211*** (0.032)	-0.135*** (0.041)	-0.147*** (0.023)	-0.149*** (0.017)	0.006 (0.083)
R ²	0.038	0.022	0.028	0.066	0.159	0.010
Panel B. Gender identity						
CGI (feminine)	-0.169*** (0.040)	-0.090*** (0.016)	-0.057*** (0.021)	-0.067*** (0.012)	-0.058*** (0.009)	
R ²	0.030	0.019	0.023	0.055	0.136	
Panel C. Binary gender and gender identity						
Female	-0.321* (0.153)	-0.197*** (0.057)	-0.127* (0.071)	-0.113*** (0.043)	-0.136*** (0.026)	
CGI (feminine)	-0.038 (0.075)	-0.009 (0.029)	-0.005 (0.036)	-0.021 (0.021)	-0.008 (0.013)	
R ²	0.039	0.022	0.028	0.068	0.159	
Panel D. Sample split by binary gender						
Female: CGI (feminine)	-0.020 (0.060)	0.000 (0.024)	-0.022 (0.029)	-0.042** (0.017)	-0.006 (0.012)	
R ²	0.022	0.013	0.043	0.033	0.133	
Male: CGI (feminine)	-0.016 (0.063)	-0.020 (0.024)	0.015 (0.030)	0.013 (0.017)	0.004 (0.000)	
R ²	0.009	0.014	0.005	0.004	0.145	
Test: male sample split CGI = female sample split CGI						
p-value	0.964	0.518	0.358	0.014	0.551	
Observations	584	3,902	584	584	1,041	584
Mean of Dependent Variable	0	0	0.413	0.538	0.550	0

Notes: The table presents results from regressing the incentivized measures of risk, competitiveness, equality versus efficiency and overconfidence on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), and Swedish Youths samples are used. The controls are as follows: quadratic age and education level for Swiss Uni; quadratic age, race, census divisions and wave for U.S. Adults; school for Swedish Youths. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of participants' investment decisions (mean=0, SD=1). Competitiveness is a dummy that takes the value 1 for those who chose to compete in the competitive task. Our measure of preferences for equality versus efficiency is measured in deciles, with increasing numbers indicating higher priority for efficiency. Overconfidence is measured as relative overplacement and standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A5: CGI and Unincentivized Risk and Competitiveness Measures (with controls)

Measure	Staircase Risk	Risk				Financial Risk	Competitiveness			
Sample	Swiss Uni (1)	Swiss Uni (2)	U.S Adults (3)	Swiss Teens (4)	Swedish Youths (5)	US Adults (6)	Swiss Uni (7)	U.S Adults (8)	Swiss Teens (9)	Swedish Youths (10)
Panel A. Binary gender										
Female	-0.206* (0.083)	-0.436*** (0.081)	-0.389*** (0.031)	-0.287*** (0.071)	-0.236*** (0.065)	-0.267*** (0.015)	-0.396*** (0.082)	-0.407*** (0.031)	-0.311*** (0.071)	-0.143* (0.063)
R ²	0.015	0.053	0.053	0.026	0.062	0.087	0.044	0.053	0.024	0.072
Panel B. Gender identity										
CGI (feminine)	-0.137*** (0.044)	-0.232*** (0.040)	-0.203*** (0.016)	-0.168*** (0.036)	-0.104*** (0.033)	-0.118*** (0.008)	-0.197*** (0.041)	-0.216*** (0.016)	-0.181*** (0.036)	-0.096*** (0.030)
R ²	0.023	0.059	0.056	0.034	0.059	0.070	0.044	0.058	0.033	0.076
Panel C. Binary gender and gender identity										
Female	0.051 (0.149)	-0.169 (0.148)	-0.171*** (0.057)	0.090 (0.168)	-0.177* (0.106)	-0.224*** (0.027)	-0.220 (0.139)	-0.161*** (0.057)	0.094 (0.170)	0.000 (0.106)
CGI (feminine)	-0.158* (0.078)	-0.163* (0.073)	-0.133*** (0.029)	-0.208** (0.086)	-0.040 (0.053)	-0.026* (0.013)	-0.108 (0.069)	-0.150*** (0.029)	-0.224*** (0.086)	-0.096* (0.050)
R ²	0.023	0.062	0.059	0.034	0.062	0.088	0.048	0.060	0.033	0.076
Panel D. Sample split by binary gender										
Female: CGI (feminine)	-0.064 (0.065)	-0.088 (0.059)	-0.026 (0.025)	-0.048 (0.060)	0.036 (0.051)	0.001 (0.010)	-0.011 (0.055)	0.062 (0.024)	0.016 (0.056)	0.006 (0.049)
R ²	0.029	0.011	0.015	0.003	0.057	0.019	0.012	0.016	0.001	0.098
Male: CGI (feminine)	-0.125* (0.069)	-0.108* (0.065)	-0.130*** (0.025)	-0.142*** (0.045)	-0.083 (0.000)	-0.034*** (0.012)	-0.111* (0.062)	-0.246*** (0.025)	-0.237*** (0.048)	-0.155 (0.000)
R ²	0.017	0.016	0.036	0.036	0.103	0.015	0.015	0.074	0.057	0.129
Test: male sample split CGI = female sample split CGI										
p-value	0.656	0.916	0.003	0.090	0.102	0.027	0.278	0.000	0.000	0.040
Observations	584	584	3,902	786	1,041	3,902	584	3,902	792	1,041
Mean of Dependent Variable	0	0	0	0	0	0.401	0	0	0	0

Notes: The table presents results from regressing unincentivized measures of risk attitudes and competitiveness on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens, and Swedish Youths samples are used. The controls are as follows: quadratic age, race, census divisions and wave for U.S Adults; quadratic age and education level for Swiss Uni; grade level and survey round for Swiss Teens; school for Swedish Youth. The estimates in each column and panel come from a separate regression. Staircase risk is a categorical certainty equivalence measure of risk-taking based on a series of hypothetical allocation decisions. Risk is a self-reported measure of risk-taking. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk taking or competitiveness. All measures except financial market risk are standardized (mean 0, SD 1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A6: CGI and Unincentivized Distributional Preferences (with controls)

Measure	Redistribution		Altruism			
Sample	Swiss Uni (1)	U.S. Adults (2)	Swiss Uni (3)	U.S. Adults (4)	Swiss Teens (5)	Swedish Youths (6)
Panel A. Binary gender						
Female	0.209* (0.082)	0.067* (0.032)	0.250*** (0.084)	-0.003 (0.032)	0.205*** (0.070)	0.540*** (0.063)
R ²	0.027	0.026	0.020	0.014	0.033	0.127
Panel B. Gender identity						
CGI (feminine)	0.111*** (0.040)	0.055*** (0.017)	0.097** (0.041)		0.067* (0.035)	0.222*** (0.034)
R ²	0.028	0.027	0.014		0.027	0.106
Panel C. Binary gender and gender identity						
Female	0.085 (0.156)	-0.072 (0.055)	0.274 (0.172)		0.459*** (0.166)	0.462*** (0.102)
CGI (feminine)	0.076 (0.076)	0.085*** (0.029)	-0.015 (0.086)		-0.140 (0.083)	0.052 (0.053)
R ²	0.029	0.028	0.020		0.036	0.129
Panel D. Sample split by binary gender						
Female: CGI (feminine)	-0.149 (0.062)	-0.105 (0.023)	-0.058 (0.073)		-0.073 (0.054)	0.056 (0.049)
R ²	0.044	0.050	0.029		0.036	0.074
Male: CGI (feminine)	0.198*** (0.062)	0.190*** (0.024)	0.023 (0.070)		-0.052 (0.046)	0.001 (0.000)
R ²	0.061	0.053	0.014		0.012	0.090
Test: male sample split CGI = female sample split CGI						
p-value	0.000	0.000	0.398		0.928	0.402
Observations	584	3,902	584	3,902	798	1,041
Mean of Dependent Variable	0	0	0	0	0	0

Notes: The table presents results from regressing unincentivized distributional measures on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens and Swedish Youth samples are used. The controls are as follows: quadratic age, race, census divisions and wave for U.S adults; quadratic age and education level for Swiss Uni; grade level and survey round for Swiss Teens; school for Swedish Youths. The estimates in each column and panel come from a separate regression. The table presents results from regressing self-reported preferences on redistribution and altruism on our standardized measure of CGI. Redistribution is a measure of how much economic redistribution one wants in society. Altruism is a measure of how much one would donate out of a windfall gain. A higher value means the greater willingness to redistribute or donate. All outcome measures are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A7: CGI and Educational Choice (with controls)

Measure	Female Educational Track Share			Skill Requirements	
				Math	Language
Sample	U.S Adults (1)	Swiss Teens (2)	Swedish Youths (3)	Swiss Teens (4)	Swiss Teens (5)
Panel A. Binary gender					
Female	0.126*** (0.007)	0.384*** (0.012)	0.090*** (0.010)	-0.917*** (0.047)	0.915*** (0.046)
R ²	0.120	0.411	0.154	0.220	0.212
Panel B. Gender identity					
CGI (feminine)	0.061*** (0.004)	0.168*** (0.006)	0.037*** (0.005)	-0.408*** (0.023)	0.403*** (0.023)
R ²	0.114	0.323	0.129	0.182	0.171
Panel C. Binary gender and gender identity					
Female	0.078*** (0.014)	0.328*** (0.023)	0.077*** (0.014)	-0.737*** (0.083)	0.760*** (0.083)
CGI (feminine)	0.029*** (0.007)	0.034*** (0.011)	0.009 (0.007)	-0.107*** (0.041)	0.092* (0.040)
R ²	0.127	0.415	0.155	0.223	0.215
Panel D. Sample split by binary gender					
Female: CGI (feminine)	0.010* (0.005)	0.026*** (0.010)	0.008 (0.007)	-0.045 (0.041)	0.115*** (0.041)
R ²	0.016	0.017	0.076	0.009	0.023
Male: CGI (feminine)	0.022*** (0.006)	0.014 (0.009)	0.007 (0.000)	-0.098*** (0.035)	0.019 (0.036)
R ²	0.038	0.027	0.126	0.042	0.014
Test: male sample split CGI = female sample split CGI					
p-value	0.172	0.721	0.715	0.110	0.230
Observations	2,279	1,448	1,041	1,433	1,433
Mean of Dependent Variable	0.541	0.442	0.521	0	0

Notes: The table presents results from regressing the educational track measures on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. U.S. Adults (age 30–60), Swiss Teens and Swedish Youths samples are used. The controls are as follows: quadratic age, race, census divisions and wave for U.S adults; grade level and survey round for Swiss Teens; school for Swedish Youths. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor's degree in the chosen field of study in 2020 for U.S. Adults; the share of women from past cohorts of graduates from the apprenticeship for Swiss Teens; the share of women accepted for undergraduate studies in that field the year before our sample made their educational choices for Swedish Youths.. Skills requirements are a standardized (mean=0, SD=1) measure based on expert evaluation of the job content in occupations chosen by Swiss Teens to start apprenticeships. Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A8: CGI and Labor Market Outcomes (with controls)

Measure	Income	Full-time Homemaker	Weekly Ave. Work Hours	Female Industry Share	Managerial Responsibilities	Performance Pay	Wage Negotiation	Work Flexibility
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A. Binary gender								
Female	-23.153*** (1.863)	0.100*** (0.008)	-6.607*** (0.452)	0.062*** (0.007)	-0.126*** (0.016)	-0.084*** (0.021)	-0.133*** (0.016)	0.142*** (0.045)
R ²	0.058	0.050	0.061	0.050	0.030	0.014	0.048	0.014
Panel B. Gender identity								
CGI (feminine)	-10.754*** (0.964)	0.050*** (0.004)	-3.127*** (0.223)	0.026*** (0.004)	-0.061*** (0.008)	-0.043*** (0.010)	-0.066*** (0.008)	0.062*** (0.023)
R ²	0.052	0.049	0.055	0.037	0.029	0.015	0.028	0.013
Panel C. Binary gender and gender identity								
Female	-16.584*** (2.974)	0.058*** (0.012)	-4.500*** (0.811)	0.059*** (0.012)	-0.081*** (0.027)	-0.041 (0.035)	-0.086*** (0.027)	0.119 (0.077)
CGI (feminine)	-4.017*** (1.547)	0.026*** (0.006)	-1.289*** (0.399)	0.002 (0.006)	-0.028* (0.014)	-0.026 (0.018)	-0.028* (0.014)	0.014 (0.039)
R ²	0.059	0.054	0.063	0.050	0.031	0.015	0.049	0.014
Panel D. Sample split by binary gender								
Female: CGI (feminine)	2.591 (1.146)	0.028*** (0.007)	-0.117 (0.339)	-0.004 (0.005)	0.023 (0.011)	-0.012 (0.014)	-0.006 (0.011)	-0.084 (0.033)
R ²	0.023	0.022	0.011	0.013	0.015	0.016	0.029	0.025
Male: CGI (feminine)	-6.751*** (1.382)	0.006* (0.003)	-1.410*** (0.316)	0.005 (0.005)	-0.051*** (0.011)	-0.015 (0.015)	-0.029** (0.011)	0.087*** (0.032)
R ²	0.037	0.010	0.027	0.020	0.034	0.016	0.043	0.021
Test: male sample split CGI = female sample split CGI								
p-value	0.000	0.004	0.006	0.209	0.000	0.942	0.157	0.000
Observations	3,902	3,902	3,902	1,994	3,902	1,994	3,902	1,994
Mean of Dependent Variable	61.421	0.062	25.281	0.497	0.400	0.303	0.491	0

Notes: The table presents results from regressing the labor market outcomes on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. The sample of the U.S. Adults (age 30-60) is used. The controls are as follows: quadratic age, race, census divisions and wave. The estimates in each column and panel come from a separate regression. Income is a self-reported categorical measure in thousand U.S. dollars. Full-time homemaker is a dummy for working full-time at home. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. Female industry share is the share of female employees in a given industry. Managerial responsibilities is a dummy for having managerial responsibilities at work. Performance pay is a dummy that is equal to 1 if the respondent's current or most recent job has performance-related pay. Wage negotiation is a dummy that is equal to 1 if the respondent ever negotiated wage. Work flexibility is a categorical measure of flexibility in working hours where 0 is no flexibility, 0.5 is can adapt work hours and 1 is full flexibility, which is standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A9: CGI and Incentivized Behavioral Measures (with CGI dummies)

Measure	Risk		Competitiveness	Efficiency		Overconfidence
Sample	Swiss Uni (1)	US Adults (2)	Swiss Uni (3)	Swiss Uni (4)	Swedish Youth (5)	Swiss Uni (6)
Panel A. Binary gender						
F-stat (overall)	21.513	45.689	9.963	38.626	86.394	0.001
p-value	0.000	0.000	0.002	0.000	0.000	0.978
Panel B. Gender identity						
F-stat (overall)	2.551	5.937	2.650	4.285	8.186	
p-value	0.005	0.000	0.004	0.000	0.000	
Panel C. Binary gender and gender identity						
F-stat (overall)	2.890	5.940	2.549	4.919	9.557	
p-value	0.001	0.000	0.004	0.000	0.000	
Test: Binary sex=0						
F-test	4.226	4.122	1.553	7.910	19.186	
p-value	0.040	0.042	0.213	0.005	0.000	
Test: All CGI dummies=0						
F-test	1.113	1.716	1.620	1.198	1.428	
p-value	0.350	0.071	0.097	0.289	0.163	
Observations	584	3,902	584	584	1,041	584

Notes: The table presents results from regressing the incentivized measures of risk, competitiveness, equality versus efficiency and overconfidence on binary indicators for each possible response on our measure of CGI. Swiss Uni, U.S. Adults (age 30-60), and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of participants' investment decisions standardized (mean=0, SD=1). Competitiveness is a dummy that takes the value 1 for those who chose to compete in the competitive task. Our measure of preferences for equality versus efficiency is measured in deciles, with increasing numbers indicating higher priority for efficiency. Overconfidence is measured as relative overplacement and standardized (mean=0, SD=1).

Table A10: CGI and Unincentivized Risk and Competitiveness Measures (with CGI dummies)

Measure	Staircase Risk	Risk				Financial Risk	Competitiveness			
Sample	Swiss Uni (1)	Swiss Uni (2)	US Adults (3)	Swiss Teens (4)	Swedish Youth (5)	US Adults (6)	Swiss Uni (7)	US Adults (8)	Swiss Teens (9)	Swedish Youth (10)
Panel A. Binary gender										
F-stat (overall)	6.549	30.068	152.059	16.247	15.415	316.622	22.681	166.561	19.221	6.109
p-value	0.011	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.014
Panel B. Gender identity										
F-stat (overall)	2.093	4.106	19.268	3.247	6.596	27.665	3.740	35.439	4.928	6.490
p-value	0.023	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
Panel C. Binary gender and gender identity										
F-stat (overall)	1.909	3.827	18.843	3.267	6.312	30.430	3.544	32.958	4.933	5.907
p-value	0.036	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
Test: Binary sex=0										
F-test	0.183	0.794	10.288	2.692	1.528	48.612	1.307	6.219	1.836	0.020
p-value	0.669	0.373	0.001	0.101	0.217	0.000	0.253	0.013	0.176	0.887
Test: All CGI dummies=0										
F-test	1.399	1.170	5.717	2.149	5.233	1.347	1.546	17.385	3.913	5.468
p-value	0.177	0.308	0.000	0.019	0.000	0.199	0.119	0.000	0.000	0.000
Observations	584	584	3,902	786	1,041	3,902	584	3,902	792	1,041

Notes: The table presents results from regressing unincentivized measures of risk attitudes and competitiveness on binary indicators for each possible response on our measure of CGI. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens, and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Staircase risk is a categorical certainty equivalence measure of risk-taking based on a series of hypothetical allocation decisions. Risk is a self-reported measure of risk-taking. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk-taking or competitiveness. All measures except financial market risk are standardized (mean 0, SD 1).

Table A11: CGI and Unincentivized Distributional Preferences (with CGI dummies)

Measure	Redistribution		Altruism			
Sample	Swiss Uni (1)	U.S. Adults (2)	Swiss Uni (3)	U.S. Adults (4)	Swiss Teens (5)	Swedish Youth (6)
Panel A. Binary gender						
F-stat (overall)	5.910	3.888	9.189	0.002	8.786	83.852
p-value	0.015	0.049	0.003	0.966	0.003	0.000
Panel B. Gender identity						
F-stat (overall)	3.165	14.081	1.809		1.904	6.768
p-value	0.001	0.000	0.056		0.036	0.000
Panel C. Binary gender and gender identity						
F-stat (overall)	2.990	12.840	2.428		1.840	8.994
p-value	0.001	0.000	0.006		0.039	0.000
Test: Binary sex=0						
F-test	0.454	0.710	2.225		1.598	18.738
p-value	0.501	0.400	0.136		0.207	0.000
Test: All CGI dummies=0						
F-test	2.416	13.809	1.374		1.316	0.706
p-value	0.008	0.000	0.189		0.217	0.720
Observations	584	3,902	584	3,902	798	1,041

Notes. The table presents results from regressing unincentivized distributional measures on binary indicators for each possible response on our measure of CGI. Swiss Uni, U.S. Adults (age 30–60), Swiss Teens and Swedish Youth samples are used. The estimates in each column and panel come from a separate regression. The table presents results from regressing self-reported preferences on redistribution and altruism on our standardized measure of CGI. Redistribution is a measure of how much economic redistribution one wants in society. Altruism is a measure of how much one would donate out of a windfall gain. A higher value means the greater willingness to redistribute or donate. All outcome measures are standardized (mean=0, SD=1).

Table A12: CGI and educational choices (with CGI dummies)

Measure	Female Educational Track Share			Skill Requirements	
				Math	Language
Sample	US Adults (1)	Swiss Teens (2)	Swedish Youth (3)	Swiss Teens (4)	Swiss Teens (5)
Panel A. Binary gender					
F-stat (overall)	280.099	970.728	102.972	381.072	382.547
p-value	0.000	0.000	0.000	0.000	0.000
Panel B. Gender identity					
F-stat (overall)	28.648	78.492	8.327	32.268	34.834
p-value	0.000	0.000	0.000	0.000	0.000
Panel C. Binary gender and gender identity					
F-stat (overall)	29.282	90.441	10.000	34.407	37.707
p-value	0.000	0.000	0.000	0.000	0.000
Test: Binary sex=0					
F-test	21.977	134.059	24.566	46.668	63.649
p-value	0.000	0.000	0.000	0.000	0.000
Test: All CGI dummies=0					
F-test	3.064	2.806	0.780	2.208	2.744
p-value	0.001	0.002	0.648	0.015	0.002
Observations	2,279	1,448	1,041	1,433	1,433

Notes: The table presents results from regressing the educational track measures on binary indicators for each possible response on our measure of CGI. U.S. Adults (age 30–60), Swiss Teens and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor's degree in the chosen field of study in 2020 for U.S. Adults; the share of women from past cohorts of graduates from the apprenticeship for Swiss Teens; the share of women accepted for undergraduate studies in that field the year before our sample made their educational choices for Swedish Youths. Skills requirements are a standardized (mean=0, SD=1) measure based on expert evaluation of the job content in occupations chosen by Swiss Teens to start apprenticeships.

Table A13: CGI and Labor Market Outcomes (with CGI dummies)

Measure	Income	Full-time Homemaker	Weekly Ave. Work Hours	Female Industry Share	Managerial Responsibilities	Performance Pay	Wage Negotiation	Work Flexibility
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A. Binary gender								
F-stat (overall)	155.642	172.645	217.216	81.057	65.005	15.972	68.948	9.929
p-value	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.002
Panel B. Gender identity								
F-stat (overall)	16.270	19.780	23.098	6.684	10.460	2.299	9.929	2.095
p-value	0.000	0.000	0.000	0.000	0.000	0.011	0.000	0.022
Panel C. Binary gender and gender identity								
F-stat (overall)	17.923	18.351	23.383	8.230	9.944	2.112	9.309	2.074
p-value	0.000	0.000	0.000	0.000	0.000	0.017	0.000	0.019
Test: Binary sex=0								
F-test	33.204	29.015	24.915	22.699	4.074	0.244	2.898	1.780
p-value	0.000	0.000	0.000	0.000	0.044	0.621	0.089	0.182
Test: All CGI dummies=0								
F-test	4.571	2.853	3.299	1.085	4.202	0.757	3.172	1.274
p-value	0.000	0.002	0.000	0.370	0.000	0.671	0.000	0.240
Observations	3,902	3,902	3,902	1,994	3,902	1,994	3,902	1,994

Notes: The table presents results from regressing the labor market outcomes on binary indicators for each possible response on our measure of CGI. The sample of the U.S. Adults (age 30-60) is used. The estimates in each column and panel come from a separate regression. Income is a self-reported categorical measure in thousand U.S. dollars. Full-time homemaker is a dummy for working full-time at home. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. Female industry share is the share of female employees in a given industry. Managerial responsibilities is a dummy for having managerial responsibilities at work. Performance pay is a dummy that is equal to 1 if the respondent's current or most recent job has performance-related pay. Wage negotiation is a dummy that is equal to 1 if the respondent ever negotiated wage. Work flexibility is a categorical measure of flexibility in working hours where 0 is no flexibility, 0.5 is can adapt work hours and 1 is full flexibility, which is standardized (mean=0, SD=1).

Table A14: ORIV Analysis - Swiss Uni Sample

Measure	Incentivized Measures			Unincentivized Measures				
	Risk	Competitiveness	Efficiency	Staircase Risk	Risk	Competitiveness	Redistribution	Altruism
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A. Binary gender								
Female	-0.378*** (0.081)	-0.128*** (0.041)	-0.145*** (0.023)	-0.211* (0.083)	-0.444*** (0.081)	-0.388*** (0.081)	0.201* (0.083)	0.249*** (0.082)
R ²	0.036	0.017	0.062	0.011	0.049	0.038	0.010	0.016
Panel B. Gender identity								
ORIV CGI (feminine)	-0.139*** (0.046)	-0.069*** (0.023)	-0.070*** (0.013)	-0.111** (0.047)	-0.254*** (0.046)	-0.252*** (0.046)	0.129*** (0.046)	0.168*** (0.048)
Panel C. Binary gender and gender identity								
Female	-0.553** (0.199)	-0.055 (0.097)	-0.116* (0.057)	-0.111 (0.200)	-0.110 (0.189)	0.077 (0.193)	-0.034 (0.208)	-0.086 (0.198)
ORIV CGI (feminine)	0.108 (0.110)	-0.045 (0.054)	-0.018 (0.031)	-0.062 (0.111)	-0.205* (0.106)	-0.286*** (0.107)	0.144 (0.113)	0.206* (0.112)
Panel D. Sample split by binary gender								
Female: ORIV CGI (feminine)	0.086 (0.095)	-0.075 (0.046)	-0.035 (0.026)	-0.020 (0.095)	-0.102 (0.094)	-0.126 (0.091)	-0.121 (0.094)	0.108 (0.096)
Male: ORIV CGI (feminine)	0.045 (0.090)	0.024 (0.041)	0.015 (0.023)	-0.055 (0.088)	-0.142* (0.085)	-0.217** (0.090)	0.287*** (0.088)	0.133 (0.089)
Observations	584	584	584	584	584	584	584	584
Mean of Dependent Variable	0	0.413	0.538	0	0	0	0	0

Notes: The table presents results from Obviously Related Instrumental Variable Approach (ORIV) approach by Gillen, Snowberg and Yariv (2019). For the estimates presented in Panels B, C and D our standardized single-item CGI measure (10-point scale) was instrumented with a similar question from the follow-up survey (11-point scale). Swiss Uni sample is used. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of participants' investment decisions (mean=0, SD=1). Competitiveness is a dummy that takes the value 1 for those who chose to compete in the competitive task. Our measure of preferences for equality versus efficiency is measured in deciles, with increasing numbers indicating higher priority for efficiency. Staircase risk is a categorical certainty equivalence measure of risk-taking based on a series of hypothetical allocation decisions. Risk is a self-reported measure of risk-taking. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk-taking or competitiveness. All measures except incentivized competitiveness and efficiency are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A15: CGI and Preference Measures by U.S Waves

	Incentivized Risk		Unincentivized Risk		Financial Risk		Competitiveness		Redistribution		Altruism	
	Wave 1 (1)	Wave 2 (2)	Wave 1 (3)	Wave 2 (4)	Wave 1 (5)	Wave 2 (6)	Wave 1 (7)	Wave 2 (8)	Wave 1 (9)	Wave 2 (10)	Wave 1 (11)	Wave 2 (12)
Panel A. Binary gender												
Female	-0.210*** (0.045)	-0.220*** (0.046)	-0.345*** (0.044)	-0.434*** (0.045)	-0.260*** (0.021)	-0.277*** (0.021)	-0.351*** (0.044)	-0.461*** (0.045)	0.038 (0.045)	0.091* (0.046)	-0.004 (0.045)	0.007 (0.046)
R ²	0.011	0.012	0.030	0.047	0.069	0.082	0.031	0.053	0.000	0.002	0.000	0.000
Panel B. Gender identity												
CGI (feminine)	-0.088*** (0.022)	-0.095*** (0.023)	-0.165*** (0.023)	-0.245*** (0.023)	-0.118*** (0.011)	-0.121*** (0.011)	-0.189*** (0.023)	-0.245*** (0.022)	0.055** (0.023)	0.051* (0.024)		
R ²	0.008	0.009	0.027	0.060	0.057	0.062	0.036	0.060	0.003	0.003		
Panel C. Binary gender and gender identity												
Female	-0.195* (0.079)	-0.200* (0.081)	-0.225*** (0.079)	-0.091 (0.082)	-0.201*** (0.037)	-0.244*** (0.038)	-0.132 (0.078)	-0.177* (0.083)	-0.150 (0.077)	0.022 (0.079)		
CGI (feminine)	-0.010 (0.040)	-0.012 (0.041)	-0.074* (0.041)	-0.207*** (0.042)	-0.036* (0.019)	-0.020 (0.019)	-0.136*** (0.040)	-0.171*** (0.042)	0.116*** (0.041)	0.041 (0.042)		
R ²	0.011	0.012	0.032	0.060	0.071	0.082	0.037	0.062	0.005	0.003		
Panel D. Sample split by binary gender												
Female: CGI (feminine)	0.004 (0.033)	-0.006 (0.033)	-0.019 (0.037)	-0.019 (0.034)	-0.015 (0.015)	0.013 (0.014)	0.072 (0.033)	0.065 (0.035)	-0.115 (0.034)	-0.115 (0.031)		
R ²	0.000	0.000	0.000	0.000	0.001	0.001	0.005	0.004	0.013	0.013		
Male: CGI (feminine)	-0.013 (0.033)	-0.007 (0.032)	-0.066* (0.034)	-0.215*** (0.035)	-0.027* (0.016)	-0.035* (0.016)	-0.222*** (0.034)	-0.273*** (0.034)	0.226*** (0.034)	0.160*** (0.034)		
R ²	0.000	0.000	0.004	0.046	0.003	0.005	0.049	0.075	0.051	0.025		
Observations	1,994	1,908	1,994	1,908	1,994	1,908	1,994	1,908	1,994	1,908	1,994	1,908
Mean of Dependent Variable	0	0	0	0	0.420	0.381	0	0	0	0	0	0

Notes: The table presents results from regressing preference measures separately by waves for the U.S Adults (age 30–60) sample. The estimates in each column and panel come from a separate regression. Incentivized risk preference is a measure of participants' investment decisions (mean=0, SD=1). Unincentivized risk is a self-reported measure of risk-taking. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk-taking or competitiveness. Redistribution is a measure of how much economic redistribution one wants in society. Altruism is a measure of how much one would donate out of a windfall gain. A higher value means greater willingness to redistribute or donate. All measures except financial risk are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A16: CGI and Educational and Labor Market Outcomes by U.S Waves

	Female Educational Track Share		Income		Full-time Homemaker		Weekly Average Work Hours		Managerial Responsibilities		Wage Negotiation	
	Wave 1 (1)	Wave 2 (2)	Wave 1 (3)	Wave 2 (4)	Wave 1 (5)	Wave 2 (6)	Wave 1 (7)	Wave 2 (8)	Wave 1 (9)	Wave 2 (10)	Wave 1 (11)	Wave 2 (12)
Panel A. Binary gender												
Female	0.113*** (0.010)	0.139*** (0.011)	-21.580*** (2.707)	-25.045*** (2.557)	0.103*** (0.011)	0.096*** (0.011)	-7.614*** (0.640)	-5.632*** (0.634)	-0.137*** (0.022)	-0.114*** (0.022)	-0.114*** (0.022)	-0.150*** (0.022)
R ²	0.092	0.129	0.031	0.048	0.046	0.040	0.066	0.040	0.019	0.014	0.013	0.023
Panel B. Gender identity												
CGI (feminine)	0.053*** (0.005)	0.071*** (0.006)	-10.223*** (1.349)	-11.579*** (1.360)	0.051*** (0.006)	0.048*** (0.006)	-3.670*** (0.315)	-2.639*** (0.313)	-0.060*** (0.011)	-0.063*** (0.011)	-0.055*** (0.011)	-0.071*** (0.011)
R ²	0.080	0.133	0.028	0.041	0.045	0.040	0.062	0.035	0.015	0.016	0.012	0.021
Panel C. Binary gender and gender identity												
Female	0.081*** (0.019)	0.071*** (0.022)	-14.596*** (4.291)	-18.710*** (4.134)	0.060*** (0.015)	0.053** (0.019)	-4.854*** (1.121)	-4.023*** (1.164)	-0.116*** (0.036)	-0.031 (0.040)	-0.071 (0.037)	-0.101* (0.041)
CGI (feminine)	0.019* (0.009)	0.042*** (0.011)	-4.318* (2.137)	-3.822* (2.224)	0.026*** (0.008)	0.026*** (0.010)	-1.706*** (0.552)	-0.970* (0.573)	-0.013 (0.018)	-0.050** (0.020)	-0.026 (0.019)	-0.029 (0.020)
R ²	0.095	0.143	0.033	0.049	0.050	0.043	0.071	0.041	0.020	0.017	0.014	0.024
Panel D. Sample split by binary gender												
Female: CGI (feminine)	0.003 (0.007)	0.018* (0.009)	1.792 (1.583)	3.101 (1.604)	0.032*** (0.010)	0.019* (0.010)	-0.524 (0.477)	0.191 (0.476)	0.041 (0.015)	0.014 (0.016)	0.006 (0.015)	-0.005 (0.016)
R ²	0.000	0.010	0.001	0.004	0.010	0.004	0.001	0.000	0.007	0.001	0.000	0.000
Male: CGI (feminine)	0.019* (0.008)	0.028*** (0.009)	-6.443*** (1.927)	-7.359*** (1.896)	0.001 (0.002)	0.010* (0.005)	-1.438*** (0.446)	-1.275*** (0.434)	-0.051*** (0.015)	-0.070*** (0.016)	-0.035* (0.016)	-0.028* (0.016)
R ²	0.009	0.022	0.010	0.014	0.000	0.007	0.012	0.009	0.010	0.019	0.005	0.003
Observations	1,225	1,054	1,994	1,908	1,994	1,908	1,994	1,908	1,994	1,908	1,994	1,908
Mean of Dependent Variable	0.548	0.532	63.065	59.703	0.061	0.062	25.363	25.195	0.405	0.395	0.421	0.564

Notes: The table presents results from regressing educational and labor market outcomes separately by waves for the U.S Adults (age 30 –60) sample. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor's degree in the chosen field of study in 2020. Income is a self-reported categorical measure in thousand U.S. dollars. Full-time homemaker is a dummy for working full-time at home. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. Female industry share is the share of female employees in a given industry. Managerial responsibilities is a dummy for having managerial responsibilities at work. Performance pay is a dummy that is equal to 1 if the respondent's current or most recent job has performance-related pay. Wage negotiation is a dummy that is equal to 1 if the respondent ever negotiated wage. Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A17: Other Gender Identity Measures and Incentivized Risk

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-0.235*** (0.039)	-0.235*** (0.039)	-0.235*** (0.039)	-0.235*** (0.039)	-0.235*** (0.039)	-0.235*** (0.039)	-0.205** (0.077)	-0.205** (0.077)	-0.317*** (0.077)	-0.141 (0.078)	-0.270*** (0.077)	-0.270*** (0.077)	-0.235*** (0.039)	-0.235*** (0.039)
R ²	0.014	0.014	0.014	0.014	0.014	0.014	0.011	0.011	0.025	0.005	0.018	0.018	0.014	0.014
Panel B. Gender identity														
GI	-0.104*** (0.019)	-0.108*** (0.019)	-0.101*** (0.019)	0.116*** (0.020)	-0.079*** (0.020)	0.115*** (0.020)	-0.064 (0.039)	0.139*** (0.041)	-0.001 (0.040)	0.089** (0.037)	0.003 (0.038)	0.185*** (0.038)	-0.103*** (0.019)	-0.079*** (0.019)
R ²	0.011	0.012	0.010	0.013	0.006	0.013	0.004	0.019	0.000	0.008	0.000	0.034	0.011	0.006
Panel C. Binary gender and gender identity														
Female	-0.194*** (0.067)	-0.172** (0.066)	-0.203*** (0.065)	-0.137* (0.066)	-0.280*** (0.060)	-0.140* (0.059)	-0.200* (0.091)	-0.103 (0.087)	-0.354*** (0.082)	-0.080 (0.085)	-0.278*** (0.078)	-0.201** (0.078)	-0.199*** (0.065)	-0.225*** (0.050)
GI	-0.025 (0.034)	-0.039 (0.033)	-0.020 (0.033)	0.060* (0.034)	0.029 (0.030)	0.062* (0.030)	-0.005 (0.046)	0.117** (0.046)	0.057 (0.041)	0.072* (0.041)	0.025 (0.039)	0.164*** (0.039)	-0.022 (0.033)	-0.008 (0.025)
R ²	0.014	0.014	0.014	0.015	0.014	0.015	0.011	0.022	0.028	0.009	0.019	0.044	0.014	0.014
Panel D. Sample split by binary gender														
Female: GI	0.013 (0.029)	-0.002 (0.029)	0.013 (0.029)	0.038 (0.029)	0.066 (0.029)	0.025 (0.029)	-0.016 (0.057)	0.039 (0.058)	0.073 (0.057)	0.026 (0.051)	0.029 (0.058)	0.226*** (0.052)	0.015 (0.028)	0.012 (0.028)
R ²	0.000	0.000	0.000	0.001	0.004	0.001	0.000	0.002	0.005	0.001	0.001	0.051	0.000	0.000
Male: GI	-0.038 (0.027)	-0.041 (0.027)	-0.033 (0.027)	0.033 (0.028)	-0.019 (0.026)	0.060* (0.028)	0.004 (0.050)	0.147** (0.057)	0.035 (0.056)	0.092* (0.052)	0.020 (0.053)	0.092 (0.059)	-0.039 (0.027)	-0.022 (0.027)
R ²	0.001	0.002	0.001	0.001	0.000	0.004	0.000	0.022	0.001	0.009	0.000	0.009	0.001	0.001
Observations	2,659	2,659	2,659	2,659	2,659	2,659	662	662	659	662	676	676	2,659	2,659
Mean of Dependent Variable	0	0	0	0	0	0	0	0	0	0	0	0	0	0

Notes: This table presents results from regressing the incentivized measure of risk on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 20–60) is used. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of participants' investment decisions (mean=0, SD=1). A higher value means higher risk-taking. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. ** p<0.005, * p<0.01, * p<0.05.

Table A18: Other Gender Identity Measures and Unincentivized Risk

Gender Measure	CGI main (1)	TMF (2)	Maglioz fem (3)	Maglioz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-0.418*** (0.038)	-0.418*** (0.038)	-0.418*** (0.038)	-0.418*** (0.038)	-0.418*** (0.038)	-0.418*** (0.038)	-0.425*** (0.076)	-0.425*** (0.076)	-0.484*** (0.076)	-0.335*** (0.077)	-0.422*** (0.075)	-0.422*** (0.075)	-0.418*** (0.038)	-0.418*** (0.038)
R ²	0.044	0.044	0.044	0.044	0.044	0.044	0.045	0.045	0.058	0.028	0.045	0.045	0.044	0.044
Panel B. Gender identity														
GI	-0.218*** (0.020)	-0.216*** (0.020)	-0.168*** (0.020)	0.251*** (0.019)	-0.167*** (0.020)	0.247*** (0.019)	0.038 (0.039)	0.335*** (0.035)	-0.111*** (0.041)	0.377*** (0.033)	0.034 (0.040)	0.465*** (0.033)	-0.199*** (0.020)	-0.191*** (0.020)
R ²	0.048	0.047	0.028	0.063	0.028	0.061	0.001	0.112	0.012	0.142	0.001	0.216	0.040	0.037
Panel C. Binary gender and gender identity														
Female	-0.185** (0.068)	-0.196*** (0.065)	-0.416*** (0.064)	-0.030 (0.065)	-0.399*** (0.060)	-0.106 (0.060)	-0.724*** (0.092)	-0.164 (0.084)	-0.461*** (0.079)	-0.022 (0.084)	-0.444*** (0.076)	-0.236*** (0.070)	-0.279*** (0.065)	-0.293*** (0.049)
GI	-0.142*** (0.035)	-0.137*** (0.034)	-0.001 (0.033)	0.239*** (0.034)	-0.012 (0.031)	0.207*** (0.031)	0.252 (0.047)	0.299*** (0.040)	-0.035 (0.042)	0.372*** (0.039)	0.069 (0.039)	0.440*** (0.034)	-0.086** (0.034)	-0.098*** (0.026)
R ²	0.050	0.050	0.044	0.063	0.044	0.062	0.086	0.118	0.060	0.142	0.049	0.230	0.046	0.049
Panel D. Sample split by binary gender														
Female: GI	0.024 (0.029)	0.011 (0.028)	0.057 (0.028)	0.076*** (0.029)	0.084 (0.028)	0.091*** (0.029)	0.216 (0.059)	0.270*** (0.056)	-0.131* (0.062)	0.351*** (0.050)	0.036 (0.053)	0.451*** (0.046)	0.028 (0.028)	0.060 (0.029)
R ²	0.001	0.000	0.003	0.006	0.007	0.008	0.047	0.073	0.017	0.123	0.001	0.203	0.001	0.004
Male: GI	-0.184*** (0.030)	-0.165*** (0.029)	-0.052* (0.030)	0.218*** (0.028)	-0.085*** (0.029)	0.197*** (0.030)	0.203 (0.052)	0.284*** (0.050)	0.064 (0.055)	0.342*** (0.051)	0.105 (0.059)	0.429*** (0.051)	-0.127*** (0.029)	-0.207*** (0.028)
R ²	0.034	0.027	0.003	0.047	0.007	0.039	0.041	0.081	0.004	0.117	0.011	0.184	0.016	0.043
Observations	2,659	2,659	2,659	2,659	2,659	2,659	662	662	659	662	676	676	2,659	2,659
Mean of Dependent Variable	0	0	0	0	0	0	0	0	0	0	0	0	0	0

Notes: This table presents results from regressing the unincentivized measure of risk on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 20–60) is used. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of self-reported willingness to take risks (mean=0, SD=1). A higher value means higher risk-taking. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Maglioz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Maglioz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. .** p<0.005, ** p<0.01, * p<0.05.

Table A19: Other Gender Identity Measures and Financial Risk

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-0.277*** (0.021)	-0.277*** (0.021)	-0.277*** (0.021)	-0.277*** (0.021)	-0.277*** (0.021)	-0.277*** (0.021)	-0.286*** (0.043)	-0.286*** (0.043)	-0.321*** (0.042)	-0.209*** (0.043)	-0.292*** (0.042)	-0.292*** (0.042)	-0.277*** (0.021)	-0.277*** (0.021)
R ²	0.082	0.082	0.082	0.082	0.082	0.082	0.086	0.086	0.107	0.047	0.091	0.091	0.082	0.082
Panel B. Gender identity														
GI	-0.121*** (0.011)	-0.124*** (0.011)	-0.114*** (0.011)	0.134*** (0.011)	-0.118*** (0.011)	0.131*** (0.011)	-0.061*** (0.023)	0.143*** (0.022)	-0.067*** (0.022)	0.169*** (0.019)	-0.060*** (0.022)	0.074*** (0.022)	-0.117*** (0.011)	-0.105*** (0.011)
R ²	0.062	0.065	0.055	0.077	0.060	0.073	0.015	0.086	0.018	0.124	0.015	0.024	0.058	0.047
Panel C. Binary gender and gender identity														
Female	-0.244*** (0.038)	-0.227*** (0.038)	-0.270*** (0.036)	-0.175*** (0.039)	-0.240*** (0.035)	-0.186*** (0.035)	-0.323*** (0.052)	-0.197*** (0.050)	-0.313*** (0.046)	-0.085 (0.046)	-0.280*** (0.043)	-0.273*** (0.043)	-0.262*** (0.038)	-0.249*** (0.029)
GI	-0.020 (0.019)	-0.031 (0.019)	-0.005 (0.018)	0.062*** (0.019)	-0.024 (0.017)	0.059*** (0.017)	0.032 (0.027)	0.098*** (0.025)	-0.012 (0.024)	0.152*** (0.021)	-0.036 (0.022)	0.047* (0.021)	-0.009 (0.019)	-0.021 (0.014)
R ²	0.082	0.083	0.082	0.087	0.083	0.087	0.089	0.118	0.108	0.130	0.097	0.101	0.082	0.083
Panel D. Sample split by binary gender														
Female: GI	0.013 (0.014)	-0.004 (0.014)	0.016 (0.014)	0.019 (0.015)	0.008 (0.014)	0.027* (0.015)	0.028 (0.029)	0.122*** (0.029)	-0.032 (0.030)	0.104*** (0.030)	-0.045 (0.029)	-0.001 (0.028)	0.007 (0.014)	0.005 (0.014)
R ²	0.001	0.000	0.001	0.002	0.000	0.004	0.004	0.081	0.006	0.056	0.012	0.000	0.000	0.000
Male: GI	-0.035* (0.016)	-0.031* (0.016)	-0.021 (0.016)	0.052*** (0.016)	-0.035* (0.016)	0.049*** (0.016)	0.025 (0.032)	0.064* (0.033)	0.011 (0.033)	0.167*** (0.026)	-0.025 (0.033)	0.097*** (0.031)	-0.018 (0.016)	-0.038** (0.016)
R ²	0.005	0.004	0.002	0.011	0.005	0.010	0.003	0.016	0.001	0.112	0.003	0.038	0.001	0.006
Observations	1,908	1,908	1,908	1,908	1,908	1,908	472	472	475	475	486	486	1,908	1,908
Mean of Dependent Variable	0.381	0.381	0.381	0.381	0.381	0.381	0.388	0.388	0.402	0.362	0.370	0.370	0.381	0.381

Notes: This table presents results from regressing an unincentivized measure of financial market risk on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 30–60) is used. The estimates in each column and panel come from a separate regression. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. *** p<0.005, ** p<0.01, * p<0.05.

Table A20: Other Gender Identity Measures and Competitiveness

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-0.455*** (0.038)	-0.455*** (0.038)	-0.455*** (0.038)	-0.455*** (0.038)	-0.455*** (0.038)	-0.455*** (0.038)	-0.378*** (0.077)	-0.378*** (0.077)	-0.513*** (0.076)	-0.469*** (0.076)	-0.464*** (0.075)	-0.464*** (0.075)	-0.455*** (0.038)	-0.455*** (0.038)
R ²	0.052	0.052	0.052	0.052	0.052	0.052	0.036	0.036	0.066	0.055	0.054	0.054	0.052	0.052
Panel B. Gender identity														
GI	-0.230*** (0.019)	-0.239*** (0.019)	-0.194*** (0.019)	0.268*** (0.019)	-0.180*** (0.019)	0.268*** (0.019)	-0.017 (0.039)	0.249*** (0.037)	-0.022 (0.040)	0.405*** (0.035)	0.040 (0.039)	0.500*** (0.033)	-0.220*** (0.019)	-0.223*** (0.019)
R ²	0.053	0.057	0.038	0.072	0.032	0.072	0.000	0.062	0.000	0.164	0.002	0.250	0.048	0.050
Panel C. Binary gender and gender identity														
Female	-0.237*** (0.069)	-0.198*** (0.066)	-0.402*** (0.067)	-0.059 (0.069)	-0.440*** (0.062)	-0.117* (0.060)	-0.550*** (0.095)	-0.199* (0.084)	-0.559*** (0.082)	-0.157 (0.081)	-0.489*** (0.075)	-0.265*** (0.068)	-0.287*** (0.067)	-0.289*** (0.050)
GI	-0.134*** (0.035)	-0.159*** (0.033)	-0.033 (0.034)	0.244*** (0.034)	-0.010 (0.032)	0.224*** (0.030)	0.146 (0.048)	0.206*** (0.041)	0.070 (0.042)	0.372*** (0.040)	0.078 (0.038)	0.472*** (0.034)	-0.104*** (0.034)	-0.130*** (0.026)
R ²	0.058	0.061	0.052	0.072	0.052	0.073	0.049	0.070	0.070	0.169	0.060	0.267	0.056	0.062
Panel D. Sample split by binary gender														
Female: GI	0.062 (0.030)	0.034 (0.029)	0.100 (0.030)	0.047 (0.029)	0.132 (0.029)	0.049* (0.029)	0.178 (0.058)	0.259*** (0.049)	0.053 (0.061)	0.326*** (0.049)	0.026 (0.055)	0.511*** (0.044)	0.051 (0.029)	0.056 (0.029)
R ²	0.004	0.001	0.010	0.002	0.017	0.002	0.032	0.067	0.003	0.106	0.001	0.261	0.003	0.003
Male: GI	-0.226*** (0.030)	-0.225*** (0.029)	-0.141*** (0.029)	0.270*** (0.029)	-0.136*** (0.030)	0.284*** (0.028)	0.073 (0.055)	0.146*** (0.055)	0.087 (0.054)	0.376*** (0.056)	0.145 (0.056)	0.429*** (0.054)	-0.185*** (0.029)	-0.273*** (0.028)
R ²	0.051	0.051	0.020	0.073	0.019	0.081	0.005	0.021	0.008	0.141	0.021	0.184	0.034	0.075
Observations	2,659	2,659	2,659	2,659	2,659	2,659	662	662	659	662	676	676	2,659	2,659
Mean of Dependent Variable	0	0	0	0	0	0	0	0	0	0	0	0	0	0

Notes: This table presents results from regressing the unincentivized measure of competitiveness on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 20–60) is used. The estimates in each column and panel come from a separate regression. Competitiveness is a self-reported measure of competitiveness. A higher value mean higher competitiveness. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. ** p<0.005, * p<0.01, * p<0.05.

Table A21: Other Gender Identity Measures and Redistribution

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	0.088* (0.039)	0.088* (0.039)	0.088* (0.039)	0.088* (0.039)	0.088* (0.039)	0.088* (0.039)	0.126 (0.078)	0.126 (0.078)	-0.029 (0.078)	0.110 (0.078)	0.147 (0.077)	0.147 (0.077)	0.088* (0.039)	0.088* (0.039)
R ²	0.002	0.002	0.002	0.002	0.002	0.002	0.004	0.004	0.000	0.003	0.005	0.005	0.002	0.002
Panel B. Gender identity														
GI	0.051** (0.021)	0.063*** (0.021)	0.067*** (0.020)	-0.018 (0.020)	0.074*** (0.021)	-0.012 (0.021)	0.143*** (0.040)	-0.121*** (0.041)	-0.170*** (0.041)	-0.145*** (0.039)	0.139*** (0.038)	-0.086* (0.040)	0.040* (0.021)	0.040* (0.021)
R ²	0.003	0.004	0.005	0.000	0.005	0.000	0.020	0.015	0.029	0.021	0.019	0.007	0.002	0.002
Panel C. Binary gender and gender identity														
Female	0.011 (0.065)	-0.041 (0.064)	-0.057 (0.063)	0.171** (0.065)	-0.066 (0.059)	0.160*** (0.057)	-0.067 (0.094)	0.025 (0.087)	0.092 (0.081)	-0.014 (0.084)	0.106 (0.077)	0.116 (0.078)	0.066 (0.065)	0.061 (0.049)
GI	0.047 (0.035)	0.080* (0.034)	0.090*** (0.033)	0.051 (0.034)	0.099*** (0.032)	0.048 (0.031)	0.162*** (0.049)	-0.116** (0.046)	-0.185*** (0.043)	-0.148*** (0.043)	0.130*** (0.039)	-0.073* (0.040)	0.013 (0.035)	0.021 (0.027)
R ²	0.003	0.004	0.005	0.003	0.006	0.003	0.021	0.015	0.031	0.021	0.022	0.011	0.002	0.002
Panel D. Sample split by binary gender														
Female: GI	-0.111 (0.028)	-0.133 (0.029)	-0.101 (0.028)	0.155 (0.027)	-0.116 (0.029)	0.169 (0.029)	0.008 (0.059)	-0.046 (0.061)	-0.245*** (0.059)	-0.034 (0.058)	0.054 (0.052)	-0.021 (0.055)	-0.103 (0.029)	-0.120 (0.029)
R ²	0.012	0.018	0.010	0.024	0.013	0.029	0.000	0.002	0.060	0.001	0.003	0.000	0.011	0.014
Male: GI	0.158*** (0.029)	0.209*** (0.028)	0.196*** (0.027)	-0.113*** (0.029)	0.216*** (0.026)	-0.130*** (0.028)	0.220*** (0.052)	-0.141** (0.056)	-0.094* (0.057)	-0.213*** (0.053)	0.207*** (0.055)	-0.127* (0.058)	0.117*** (0.028)	0.149*** (0.028)
R ²	0.025	0.044	0.038	0.013	0.047	0.017	0.048	0.020	0.009	0.045	0.043	0.016	0.014	0.022
Observations	2,659	2,659	2,659	2,659	2,659	2,659	662	662	659	662	676	676	2,659	2,659
Mean of Dependent Variable	0	0	0	0	0	0	0	0	0	0	0	0	0	0

Notes: This table presents results from regressing the unincentivized measure of redistribution on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 20–60) is used. The estimates in each column and panel come from a separate regression. Redistribution is a measure of how much economic redistribution one wants in society. A higher value means a greater willingness to redistribute. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. .** p<0.005, ** p<0.01, * p<0.05.

Table A22: Other Gender Identity Measures and Female Educational Track Share

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	0.131*** (0.009)	0.131*** (0.009)	0.131*** (0.009)	0.131*** (0.009)	0.131*** (0.009)	0.131*** (0.009)	0.128*** (0.018)	0.128*** (0.018)	0.129*** (0.019)	0.143*** (0.018)	0.122*** (0.019)	0.122*** (0.019)	0.131*** (0.009)	0.131*** (0.009)
R ²	0.117	0.117	0.117	0.117	0.117	0.117	0.117	0.117	0.115	0.141	0.097	0.097	0.117	0.117
Panel B. Gender identity														
GI	0.067*** (0.005)	0.065*** (0.005)	0.065*** (0.005)	-0.066*** (0.005)	0.063*** (0.005)	-0.065*** (0.005)	0.039*** (0.009)	-0.048*** (0.009)	0.029*** (0.009)	-0.067*** (0.009)	0.028*** (0.010)	0.000 (0.010)	0.060*** (0.005)	0.057*** (0.005)
R ²	0.124	0.117	0.115	0.118	0.108	0.115	0.044	0.067	0.024	0.122	0.020	0.000	0.099	0.089
Panel C. Binary gender and gender identity														
Female	0.062*** (0.017)	0.072*** (0.017)	0.075*** (0.016)	0.070*** (0.016)	0.083*** (0.014)	0.076*** (0.015)	0.121*** (0.022)	0.106*** (0.020)	0.122*** (0.019)	0.105*** (0.020)	0.117*** (0.020)	0.128*** (0.019)	0.097*** (0.016)	0.097*** (0.012)
GI	0.042*** (0.009)	0.036*** (0.008)	0.035*** (0.008)	-0.037*** (0.008)	0.031*** (0.007)	-0.036*** (0.007)	0.006 (0.011)	-0.024* (0.010)	0.014* (0.009)	-0.044*** (0.010)	0.021* (0.010)	0.014 (0.010)	0.021** (0.008)	0.027*** (0.006)
R ²	0.133	0.130	0.128	0.129	0.128	0.132	0.118	0.130	0.121	0.183	0.108	0.101	0.121	0.129
Panel D. Sample split by binary gender														
Female: GI	0.016** (0.007)	0.014* (0.007)	0.016** (0.006)	-0.016** (0.006)	0.010* (0.006)	-0.020*** (0.006)	0.015 (0.011)	-0.009 (0.013)	0.024* (0.010)	-0.030** (0.013)	0.036** (0.015)	0.024 (0.013)	0.003 (0.006)	0.012* (0.006)
R ²	0.010	0.007	0.009	0.009	0.004	0.014	0.009	0.003	0.022	0.036	0.039	0.017	0.000	0.005
Male: GI	0.031*** (0.008)	0.028*** (0.008)	0.025*** (0.007)	-0.027*** (0.007)	0.029*** (0.007)	-0.028*** (0.007)	-0.003 (0.014)	-0.029* (0.013)	0.003 (0.013)	-0.048*** (0.013)	0.005 (0.014)	0.003 (0.014)	0.021*** (0.007)	0.031*** (0.007)
R ²	0.027	0.022	0.018	0.020	0.023	0.021	0.000	0.024	0.000	0.061	0.001	0.000	0.012	0.026
Observations	1,531	1,531	1,531	1,531	1,531	1,531	395	395	369	387	380	380	1,531	1,531
Mean of Dependent Variable	0.532	0.532	0.532	0.532	0.532	0.532	0.535	0.535	0.523	0.539	0.531	0.531	0.532	0.532

Notes: This table presents results from regressing female educational track share on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 20–60) is used. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor's degree in the chosen field of study in 2020. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. .** p<0.005, ** p<0.01, * p<0.05.

Table A23: Other Gender Identity Measures and Income

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-20.376*** (2.099)	-20.376*** (2.099)	-20.376*** (2.099)	-20.376*** (2.099)	-20.376*** (2.099)	-20.376*** (2.099)	-13.953*** (4.010)	-13.953*** (4.010)	-26.463*** (4.127)	-17.840*** (4.165)	-23.488*** (4.503)	-23.488*** (4.503)	-20.376*** (2.099)	-20.376*** (2.099)
R ²	0.034	0.034	0.034	0.034	0.034	0.034	0.018	0.018	0.058	0.027	0.040	0.040	0.034	0.034
Panel B. Gender identity														
GI	-9.401*** (1.112)	-9.439*** (1.110)	-8.770*** (1.088)	9.819*** (1.082)	-8.424*** (1.095)	9.081*** (1.084)	-2.135 (1.936)	9.227*** (2.019)	4.753 (1.958)	6.432*** (1.958)	1.414 (2.157)	10.259*** (2.315)	-9.214*** (1.116)	-8.142*** (1.117)
R ²	0.029	0.029	0.025	0.032	0.023	0.027	0.002	0.031	0.007	0.014	0.001	0.030	0.028	0.022
Panel C. Binary gender and gender identity														
Female	-15.093*** (3.281)	-14.756*** (3.366)	-17.640*** (3.270)	-13.016*** (3.262)	-18.290*** (3.005)	-15.495*** (3.088)	-17.576*** (5.280)	-7.305 (4.284)	-33.137*** (4.393)	-15.105*** (4.612)	-24.538*** (4.562)	-20.053*** (4.350)	-15.860*** (3.333)	-16.849*** (2.529)
GI	-3.232* (1.759)	-3.480* (1.788)	-1.709 (1.699)	4.528*** (1.694)	-1.349 (1.579)	3.236* (1.595)	3.059 (2.556)	7.641*** (2.166)	10.185 (2.035)	3.257 (2.166)	3.342 (2.142)	8.144*** (2.267)	-2.787 (1.787)	-2.762* (1.357)
R ²	0.035	0.035	0.034	0.036	0.034	0.036	0.020	0.035	0.088	0.030	0.043	0.058	0.035	0.036
Panel D. Sample split by binary gender														
Female: GI	3.734 (1.300)	2.099 (1.398)	3.180 (1.258)	-1.878 (1.185)	4.211 (1.194)	-1.857 (1.396)	-2.424 (3.080)	0.327 (2.400)	3.635 (2.496)	2.599 (2.385)	1.393 (2.703)	2.961 (3.091)	2.895 (1.335)	3.359 (1.379)
R ²	0.006	0.002	0.005	0.002	0.008	0.002	0.003	0.000	0.007	0.003	0.001	0.003	0.004	0.005
Male: GI	-7.265*** (1.541)	-5.829*** (1.583)	-4.921*** (1.580)	7.892*** (1.617)	-5.248*** (1.530)	6.920*** (1.594)	6.214 (2.757)	11.263*** (2.863)	16.269 (2.944)	3.279 (2.998)	5.380 (3.323)	13.698*** (3.212)	-6.148*** (1.622)	-7.559*** (1.546)
R ²	0.014	0.009	0.007	0.017	0.008	0.013	0.013	0.042	0.069	0.003	0.007	0.045	0.010	0.016
Observations	2,659	2,659	2,659	2,659	2,659	2,659	662	662	659	662	676	676	2,659	2,659
Mean of Dependent Variable	54.914	54.914	54.914	54.914	54.914	54.914	53.074	53.074	56.153	54.728	55.692	55.692	54.914	54.914

Notes: This table presents results from regressing income on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 20–60) is used. The estimates in each column and panel come from a separate regression. Income is a self-reported categorical measure in thousand U.S. dollars. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. .** p<0.005, ** p<0.01, * p<0.05.

Table A24: Other Gender Identity Measures and Full-time Homemaker

Gender Measure	CGI main (1)	TMF (2)	Maglioz fem (3)	Maglioz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	0.096*** (0.011)	0.096*** (0.011)	0.096*** (0.011)	0.096*** (0.011)	0.096*** (0.011)	0.096*** (0.011)	0.103*** (0.022)	0.103*** (0.022)	0.123*** (0.024)	0.080*** (0.020)	0.080*** (0.021)	0.080*** (0.021)	0.096*** (0.011)	0.096*** (0.011)
R ²	0.040	0.040	0.040	0.040	0.040	0.040	0.044	0.044	0.057	0.032	0.028	0.028	0.040	0.040
Panel B. Gender identity														
GI	0.048*** (0.006)	0.047*** (0.006)	0.045*** (0.006)	-0.046*** (0.006)	0.043*** (0.005)	-0.048*** (0.006)	0.031*** (0.009)	-0.019* (0.009)	0.021* (0.013)	-0.029*** (0.009)	0.017* (0.009)	-0.033*** (0.011)	0.047*** (0.006)	0.041*** (0.006)
R ²	0.040	0.039	0.035	0.036	0.031	0.039	0.016	0.006	0.007	0.017	0.005	0.018	0.038	0.028
Panel C. Binary gender and gender identity														
Female	0.053** (0.019)	0.056*** (0.019)	0.068*** (0.018)	0.064*** (0.021)	0.077*** (0.018)	0.056*** (0.017)	0.100*** (0.025)	0.108*** (0.025)	0.123*** (0.024)	0.068*** (0.021)	0.077*** (0.021)	0.070*** (0.021)	0.059*** (0.018)	0.076*** (0.015)
GI	0.026*** (0.010)	0.024** (0.010)	0.017* (0.009)	-0.019* (0.011)	0.012 (0.009)	-0.026*** (0.009)	0.003 (0.010)	0.005 (0.010)	-0.000 (0.013)	-0.015* (0.009)	0.011 (0.009)	-0.026* (0.011)	0.023** (0.009)	0.015* (0.008)
R ²	0.043	0.043	0.041	0.042	0.041	0.044	0.044	0.045	0.057	0.036	0.030	0.039	0.042	0.042
Panel D. Sample split by binary gender														
Female: GI	0.019* (0.010)	0.020* (0.011)	0.014 (0.010)	-0.012 (0.010)	0.008 (0.011)	-0.022* (0.010)	0.003 (0.018)	0.014 (0.021)	0.012 (0.022)	-0.022 (0.016)	0.023 (0.016)	-0.041* (0.019)	0.022* (0.010)	0.012 (0.010)
R ²	0.004	0.004	0.002	0.002	0.001	0.005	0.000	0.002	0.001	0.006	0.006	0.018	0.005	0.002
Male: GI	0.010* (0.005)	0.008* (0.005)	0.006* (0.003)	-0.009* (0.005)	0.008* (0.004)	-0.009* (0.005)	0.001 (0.002)	-0.002 (0.004)	-0.014 (0.008)	-0.008 (0.007)	-0.003 (0.005)	-0.009 (0.010)	0.003 (0.004)	0.010* (0.005)
R ²	0.007	0.005	0.003	0.006	0.004	0.006	0.000	0.000	0.016	0.005	0.000	0.004	0.001	0.007
Observations	1,908	1,908	1,908	1,908	1,908	1,908	472	472	475	475	486	486	1,908	1,908
Mean of Dependent Variable	0.062	0.062	0.062	0.062	0.062	0.062	0.064	0.064	0.072	0.053	0.062	0.062	0.062	0.062

Notes: This table presents results from regressing full-time homemaker on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 30–60) is used. The estimates in each column and panel come from a separate regression. Full-time homemaker is a dummy for working full-time at home. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Maglioz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Maglioz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. .** p<0.005, ** p<0.01, * p<0.05.

Table A25: Other Gender Identity Measures and Weekly Average Work Hours

Gender Measure	CGI main (1)	TMF (2)	Maglioz fem (3)	Maglioz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-5.632*** (0.634)	-5.632*** (0.634)	-5.632*** (0.634)	-5.632*** (0.634)	-5.632*** (0.634)	-5.632*** (0.634)	-4.437*** (1.242)	-4.437*** (1.242)	-7.655*** (1.303)	-4.815*** (1.268)	-5.573*** (1.266)	-5.573*** (1.266)	-5.632*** (0.634)	-5.632*** (0.634)
R ²	0.040	0.040	0.040	0.040	0.040	0.040	0.027	0.027	0.069	0.030	0.039	0.039	0.040	0.040
Panel B. Gender identity														
GI	-2.639*** (0.313)	-2.765*** (0.311)	-2.424*** (0.317)	2.860*** (0.312)	-2.573*** (0.313)	2.703*** (0.315)	-0.367 (0.629)	1.563*** (0.597)	0.048 (0.668)	1.507** (0.633)	-0.553 (0.635)	2.609*** (0.643)	-2.757*** (0.314)	-2.432*** (0.314)
R ²	0.035	0.038	0.029	0.041	0.033	0.037	0.001	0.013	0.000	0.012	0.002	0.034	0.038	0.030
Panel C. Binary gender and gender identity														
Female	-4.023*** (1.164)	-3.354*** (1.102)	-4.963*** (1.099)	-2.841* (1.184)	-4.168*** (1.028)	-3.614*** (1.022)	-6.004*** (1.456)	-3.804** (1.429)	-8.748*** (1.361)	-4.305*** (1.383)	-5.548*** (1.284)	-4.724*** (1.279)	-3.395*** (1.150)	-4.309*** (0.884)
GI	-0.970* (0.573)	-1.391** (0.541)	-0.413 (0.548)	1.677*** (0.583)	-0.928* (0.506)	1.298** (0.508)	1.362 (0.727)	0.697 (0.688)	1.570 (0.691)	0.624 (0.690)	-0.074 (0.642)	2.139*** (0.655)	-1.356** (0.569)	-0.986* (0.437)
R ²	0.041	0.043	0.040	0.044	0.041	0.043	0.033	0.029	0.079	0.031	0.039	0.060	0.043	0.042
Panel D. Sample split by binary gender														
Female: GI	0.191 (0.476)	-0.255 (0.468)	0.426 (0.495)	0.080 (0.468)	0.303 (0.470)	0.427 (0.460)	1.529 (0.896)	-0.171 (0.986)	0.136 (0.909)	0.834 (0.894)	-1.633* (0.872)	2.831*** (0.887)	-0.344 (0.466)	0.021 (0.479)
R ²	0.000	0.000	0.001	0.000	0.000	0.001	0.012	0.000	0.000	0.003	0.012	0.037	0.001	0.000
Male: GI	-1.275*** (0.434)	-1.320*** (0.414)	-0.886* (0.418)	1.955*** (0.438)	-1.356*** (0.414)	1.293*** (0.445)	0.787 (0.815)	1.212 (0.812)	2.932 (0.914)	0.380 (0.893)	1.521 (0.899)	1.307 (0.929)	-1.207*** (0.446)	-1.529*** (0.433)
R ²	0.009	0.010	0.004	0.022	0.010	0.010	0.004	0.009	0.047	0.001	0.013	0.010	0.008	0.013
Observations	1,908	1,908	1,908	1,908	1,908	1,908	472	472	475	475	486	486	1,908	1,908
Mean of Dependent Variable	25.195	25.195	25.195	25.195	25.195	25.195	26.160	26.160	25.585	24.177	24.873	24.873	25.195	25.195

Notes: This table presents results from regressing weekly average work hours on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 30–60) is used. The estimates in each column and panel come from a separate regression. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Maglioz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Maglioz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. ** p<0.005, * p<0.01, * p<0.05.

Table A26: Other Gender Identity Measures and Managerial Responsibilities

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-0.114*** (0.022)	-0.114*** (0.022)	-0.114*** (0.022)	-0.114*** (0.022)	-0.114*** (0.022)	-0.114*** (0.022)	-0.049 (0.044)	-0.049 (0.044)	-0.178*** (0.044)	-0.078 (0.045)	-0.149*** (0.044)	-0.149*** (0.044)	-0.114*** (0.022)	-0.114*** (0.022)
R ²	0.014	0.014	0.014	0.014	0.014	0.014	0.003	0.003	0.033	0.006	0.023	0.023	0.014	0.014
Panel B. Gender identity														
GI	-0.063*** (0.011)	-0.059*** (0.011)	-0.046*** (0.011)	0.066*** (0.011)	-0.046*** (0.011)	0.073*** (0.011)	0.051 (0.022)	0.065*** (0.023)	0.020 (0.022)	0.086*** (0.022)	0.050 (0.020)	0.144*** (0.020)	-0.059*** (0.011)	-0.050*** (0.011)
R ²	0.016	0.015	0.009	0.018	0.009	0.022	0.011	0.018	0.002	0.030	0.010	0.086	0.014	0.011
Panel C. Binary gender and gender identity														
Female	-0.031 (0.040)	-0.050 (0.039)	-0.115*** (0.038)	-0.015 (0.040)	-0.107*** (0.036)	-0.001 (0.035)	-0.160*** (0.054)	0.013 (0.050)	-0.219*** (0.047)	-0.010 (0.049)	-0.172*** (0.045)	-0.096* (0.044)	-0.053 (0.039)	-0.084*** (0.030)
GI	-0.050** (0.020)	-0.039* (0.020)	0.001 (0.019)	0.059*** (0.020)	-0.004 (0.018)	0.072*** (0.018)	0.097 (0.027)	0.068*** (0.026)	0.058 (0.024)	0.084*** (0.024)	0.065 (0.021)	0.134*** (0.021)	-0.037* (0.020)	-0.022 (0.015)
R ²	0.017	0.016	0.014	0.018	0.014	0.022	0.030	0.019	0.046	0.030	0.040	0.095	0.015	0.015
Panel D. Sample split by binary gender														
Female: GI	0.014 (0.016)	0.006 (0.015)	0.018 (0.016)	0.004 (0.016)	0.027 (0.015)	0.024 (0.016)	0.026 (0.032)	0.074* (0.033)	0.010 (0.032)	0.006 (0.034)	0.019 (0.029)	0.117*** (0.028)	0.014 (0.016)	0.031 (0.016)
R ²	0.001	0.000	0.001	0.000	0.003	0.003	0.003	0.025	0.000	0.000	0.002	0.061	0.001	0.004
Male: GI	-0.070*** (0.016)	-0.049*** (0.017)	-0.017 (0.017)	0.067*** (0.016)	-0.030* (0.017)	0.072*** (0.016)	0.124 (0.030)	0.052 (0.032)	0.103 (0.032)	0.130*** (0.031)	0.111 (0.028)	0.148*** (0.030)	-0.057*** (0.016)	-0.067*** (0.016)
R ²	0.019	0.010	0.001	0.018	0.004	0.021	0.064	0.011	0.042	0.068	0.049	0.088	0.013	0.018
Observations	1,908	1,908	1,908	1,908	1,908	1,908	472	472	475	475	486	486	1,908	1,908
Mean of Dependent Variable	0.395	0.395	0.395	0.395	0.395	0.395	0.362	0.362	0.394	0.415	0.409	0.409	0.395	0.395

Notes: This table presents results from regressing managerial responsibilities on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 30–60) is used. The estimates in each column and panel come from a separate regression. Managerial responsibilities is a dummy for having managerial responsibilities at work. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. .** p<0.005, ** p<0.01, * p<0.05.

Table A27: Other Gender Identity Measures and Wage Negotiation

Gender Measure	CGI main (1)	TMF (2)	Magloz fem (3)	Magloz mas (4)	SOM fem (5)	SOM mas (6)	OSRI fem (7)	OSRI mas (8)	CFNI (9)	CMNI (10)	BSRI fem (11)	BSRI mas (12)	CGI other (13)	CGI own (14)
Panel A. Binary gender														
Female	-0.150*** (0.022)	-0.150*** (0.022)	-0.150*** (0.022)	-0.150*** (0.022)	-0.150*** (0.022)	-0.150*** (0.022)	-0.118** (0.046)	-0.118** (0.046)	-0.189*** (0.045)	-0.069 (0.045)	-0.219*** (0.044)	-0.219*** (0.044)	-0.150*** (0.022)	-0.150*** (0.022)
R ²	0.023	0.023	0.023	0.023	0.023	0.023	0.014	0.014	0.037	0.005	0.048	0.048	0.023	0.023
Panel B. Gender identity														
GI	-0.071*** (0.011)	-0.077*** (0.011)	-0.064*** (0.011)	0.089*** (0.011)	-0.054*** (0.011)	0.086*** (0.011)	0.020 (0.023)	0.098*** (0.022)	-0.027 (0.023)	0.065*** (0.022)	0.005 (0.023)	0.117*** (0.022)	-0.074*** (0.011)	-0.065*** (0.011)
R ²	0.021	0.024	0.016	0.032	0.012	0.030	0.002	0.039	0.003	0.017	0.000	0.055	0.022	0.017
Panel C. Binary gender and gender identity														
Female	-0.101* (0.041)	-0.075 (0.040)	-0.137*** (0.039)	-0.007 (0.041)	-0.174*** (0.036)	-0.041 (0.036)	-0.211*** (0.054)	-0.036 (0.051)	-0.194*** (0.047)	-0.019 (0.049)	-0.228*** (0.045)	-0.180*** (0.045)	-0.090* (0.040)	-0.114*** (0.031)
GI	-0.029 (0.020)	-0.046* (0.020)	-0.008 (0.019)	0.086*** (0.020)	0.015 (0.018)	0.070*** (0.018)	0.081 (0.027)	0.090*** (0.025)	0.007 (0.024)	0.061** (0.024)	0.025 (0.023)	0.099*** (0.023)	-0.036* (0.020)	-0.027* (0.015)
R ²	0.024	0.026	0.023	0.032	0.023	0.031	0.032	0.040	0.037	0.018	0.051	0.086	0.025	0.024
Panel D. Sample split by binary gender														
Female: GI	-0.005 (0.016)	-0.019 (0.016)	0.000 (0.016)	0.050*** (0.016)	0.013 (0.016)	0.043*** (0.016)	0.038 (0.033)	0.080** (0.032)	-0.025 (0.034)	0.103*** (0.029)	0.009 (0.032)	0.133*** (0.031)	-0.013 (0.016)	-0.010 (0.016)
R ²	0.000	0.001	0.000	0.010	0.001	0.007	0.006	0.026	0.002	0.043	0.000	0.072	0.001	0.000
Male: GI	-0.028* (0.016)	-0.034* (0.016)	-0.009 (0.016)	0.045*** (0.016)	0.006 (0.015)	0.047*** (0.016)	0.089 (0.031)	0.081** (0.032)	0.040 (0.031)	0.021 (0.031)	0.040 (0.032)	0.059* (0.032)	-0.029* (0.016)	-0.030* (0.016)
R ²	0.003	0.005	0.000	0.009	0.000	0.009	0.033	0.027	0.007	0.002	0.007	0.015	0.004	0.004
Observations	1,908	1,908	1,908	1,908	1,908	1,908	472	472	475	475	486	486	1,908	1,908
Mean of Dependent Variable	0.564	0.564	0.564	0.564	0.564	0.564	0.542	0.542	0.592	0.592	0.533	0.533	0.564	0.564

Notes: This table presents results from regressing wage negotiation on all 14 measures of gender identity collected in this sample. U.S. Adults Wave 2 sample (age 30–60) is used. The estimates in each column and panel come from a separate regression. Wage negotiation is a dummy that is equal to 1 if the respondent has ever engaged in wage negotiations. N differs across columns since each participant responded to only one of the four lengthy gender identity inventories (OSRI, CFNI, CMNI, BSRI). See Subsection 2.1 for a detailed description of the alternate gender identity measures. CGI other and CGI own indicate how respondents think that others see them and how they see themselves compared to others with the same binary gender on a scale from 0-very masculine to 10-very feminine. For CGI, TMF, Magloz fem, SOM fem, CFNI, BSRI fem, CGI other, CGI own higher values indicate greater femininity. For these measures significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Magloz mas, SOM mas, OSRI mas, CMNI, BSRI mas higher values indicate greater masculinity. For these measures, significance levels in Panels B, C and D are determined by a one-tailed test in the opposite direction of the coefficient estimate of female in Panel A. Significance levels in Panel A are determined by a two-tailed test for all measures. All gender identity measures are standardized (mean=0, SD=1) within the relevant sample. ** $p < 0.005$, * $p < 0.01$, * $p < 0.05$.

Table A28: CGI and U.S. Adults Preference Measures with and without Non-binary Participants

	Incentivized Risk		Unincentivized Risk		Financial Risk		Competitiveness		Redistribution		Altruism	
	exclude nonbinary (1)	include nonbinary (2)	exclude nonbinary (3)	include nonbinary (4)	exclude nonbinary (5)	include nonbinary (6)	exclude nonbinary (7)	include nonbinary (8)	exclude nonbinary (9)	include nonbinary (10)	exclude nonbinary (11)	include nonbinary (12)
Panel A. Binary gender												
Female	-0.220*** (0.046)	-0.215*** (0.045)	-0.434*** (0.045)	-0.431*** (0.044)	-0.277*** (0.021)	-0.280*** (0.021)	-0.461*** (0.045)	-0.464*** (0.044)	0.091* (0.046)	0.106* (0.045)	0.007 (0.046)	0.002 (0.045)
R ²	0.012	0.012	0.047	0.046	0.082	0.083	0.053	0.054	0.002	0.003	0.000	0.000
Panel B. Gender identity												
CGI (feminine)	-0.095*** (0.023)	-0.097*** (0.023)	-0.245*** (0.023)	-0.243*** (0.023)	-0.121*** (0.011)	-0.120*** (0.011)	-0.245*** (0.022)	-0.243*** (0.022)	0.051* (0.024)	0.051* (0.024)		
R ²	0.009	0.009	0.060	0.059	0.062	0.061	0.060	0.059	0.003	0.003		
Panel C. Binary gender and gender identity												
Female	-0.200* (0.081)	-0.173* (0.078)	-0.091 (0.082)	-0.096 (0.079)	-0.244*** (0.038)	-0.252*** (0.037)	-0.177* (0.083)	-0.203* (0.081)	0.022 (0.079)	0.067 (0.077)		
CGI (feminine)	-0.012 (0.041)	-0.026 (0.040)	-0.207*** (0.042)	-0.204*** (0.041)	-0.020 (0.019)	-0.017 (0.019)	-0.171*** (0.042)	-0.159*** (0.040)	0.041 (0.042)	0.024 (0.041)		
R ²	0.012	0.012	0.060	0.060	0.082	0.084	0.062	0.062	0.003	0.003		
Panel D. Sample split by binary gender												
Female: CGI (feminine)	-0.006 (0.033)	-0.010 (0.033)	-0.019 (0.034)	-0.030 (0.034)	0.013 (0.014)	0.015 (0.014)	0.065 (0.035)	0.077 (0.034)	-0.115 (0.031)	-0.144 (0.031)		
R ²	0.000	0.000	0.000	0.001	0.001	0.001	0.004	0.006	0.013	0.021		
Male: CGI (feminine)	-0.007 (0.032)	-0.020 (0.032)	-0.215*** (0.035)	-0.207*** (0.035)	-0.035* (0.016)	-0.034* (0.016)	-0.273*** (0.034)	-0.278*** (0.033)	0.160*** (0.034)	0.170*** (0.034)		
R ²	0.000	0.000	0.046	0.043	0.005	0.005	0.075	0.077	0.025	0.029		
Observations	1,908	1,936	1,908	1,936	1,908	1,936	1,908	1,936	1,908	1,936	1,908	1,936
Mean of Dependent Variable	0	0	0	0	0.381	0.379	0	0	0	0	0	0

Notes: The table presents the results from regressing incentivized and unincentivized preference measures on our standardized CGI measure (mean=0, SD=1). CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. U.S. Adults (age 30-60) sample is used. In the first column for each outcome measure, we exclude individuals identifying as non-binary and in the second column, we include individuals identifying as non-binary. Binary gender is measured with the question “What is your sex?” with answer options “female” and “male” which every participant had to answer. Non-binary gender is measured with the question “What gender are you currently?” with answer options “Man (including Trans Male/Trans Man)”, “Woman (including Trans Female/Trans Woman)”, “Non-binary” and “Would rather not say”. The estimates in each column and panel come from a separate regression. Incentivized risk preference is a measure of participants’ investment decisions. Unincentivized risk is a self-reported measure of risk-taking. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk taking or competitiveness. All measures except financial risk are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A29: CGI and U.S. Adults Educational and Labor Market Outcomes with and without Non-binary Participants

	Female Educational Track Share		Income		Full-time Homemaker		Weekly Average Work Hours		Managerial Responsibilities		Wage Negotiation	
	exclude nonbinary (1)	include nonbinary (2)	exclude nonbinary (3)	include nonbinary (4)	exclude nonbinary (5)	include nonbinary (6)	exclude nonbinary (7)	include nonbinary (8)	exclude nonbinary (9)	include nonbinary (10)	exclude nonbinary (11)	include nonbinary (12)
Panel A. Binary gender												
Female	0.139*** (0.011)	0.139*** (0.011)	-25.045*** (2.557)	-24.971*** (2.535)	0.096*** (0.011)	0.094*** (0.011)	-5.632*** (0.634)	-5.468*** (0.631)	-0.114*** (0.022)	-0.114*** (0.022)	-0.150*** (0.022)	-0.150*** (0.022)
R ²	0.129	0.128	0.048	0.048	0.040	0.039	0.040	0.037	0.014	0.014	0.023	0.023
Panel B. Gender identity												
CGI (feminine)	0.071*** (0.006)	0.071*** (0.006)	-11.579*** (1.360)	-11.552*** (1.349)	0.048*** (0.006)	0.048*** (0.006)	-2.639*** (0.313)	-2.646*** (0.310)	-0.063*** (0.011)	-0.062*** (0.011)	-0.071*** (0.011)	-0.072*** (0.011)
R ²	0.133	0.131	0.041	0.041	0.040	0.039	0.035	0.035	0.016	0.016	0.021	0.021
Panel C. Binary gender and gender identity												
Female	0.071*** (0.022)	0.072*** (0.022)	-18.710*** (4.134)	-18.398*** (3.970)	0.053** (0.019)	0.050** (0.018)	-4.023*** (1.164)	-3.446*** (1.136)	-0.031 (0.040)	-0.040 (0.038)	-0.101* (0.041)	-0.101* (0.039)
CGI (feminine)	0.042*** (0.011)	0.041*** (0.011)	-3.822* (2.224)	-4.011* (2.141)	0.026*** (0.010)	0.027*** (0.009)	-0.970* (0.573)	-1.234* (0.559)	-0.050** (0.020)	-0.045** (0.019)	-0.029 (0.020)	-0.030 (0.020)
R ²	0.143	0.142	0.049	0.049	0.043	0.043	0.041	0.040	0.017	0.017	0.024	0.024
Panel D. Sample split by binary gender												
Female: CGI (feminine)	0.018* (0.009)	0.017* (0.009)	3.101 (1.604)	3.452 (1.581)	0.019* (0.010)	0.022* (0.010)	0.191 (0.476)	0.167 (0.470)	0.014 (0.016)	0.013 (0.016)	-0.005 (0.016)	-0.003 (0.016)
R ²	0.010	0.009	0.004	0.005	0.004	0.005	0.000	0.000	0.001	0.001	0.000	0.000
Male: CGI (feminine)	0.028*** (0.009)	0.029*** (0.009)	-7.359*** (1.896)	-8.058*** (1.860)	0.010* (0.005)	0.010* (0.004)	-1.275*** (0.434)	-1.583*** (0.437)	-0.070*** (0.016)	-0.066*** (0.016)	-0.028* (0.016)	-0.031* (0.016)
R ²	0.022	0.023	0.014	0.016	0.007	0.006	0.009	0.014	0.019	0.017	0.003	0.004
Observations	1,054	1,064	1,908	1,936	1,908	1,936	1,908	1,936	1,908	1,936	1,908	1,936
Mean of Dependent Variable	0.532	0.533	59.703	59.278	0.062	0.061	25.195	25.101	0.395	0.394	0.564	0.562

Notes: The table presents the results from regressing educational and labor market outcomes on our standardized CGI measure (mean=0, SD=1). CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Adults (age 30-60) sample is used. In the first column for each outcome measure, we exclude individuals identifying as non-binary and in the second column, we include individuals identifying as non-binary. U.S. Binary gender is measured with the question “What is your sex?” with answer options “female” and “male” which every participant had to answer. Non-binary gender is measured with the question “What gender are you currently?” with answer options “Man (including Trans Male/Trans Man)”, “Woman (including Trans Female/Trans Woman)”, “Non-binary” and “Would rather not say”. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor’s degree in the chosen field of study in 2020. Income is a self-reported categorical measure in thousands of U.S. dollars. Full-time homemaker is a dummy for working full-time at home. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. Managerial responsibilities is a dummy for having managerial responsibilities at work. Wage negotiation is a dummy that is equal to 1 if the respondent ever negotiated wage. Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A30: CGI and Swiss Uni Incentivized Behavioral Measures with and without Non-binary Participants

	Risk		Competitiveness		Efficiency		Overconfidence	
	exclude nonbinary (1)	include nonbinary (2)	exclude nonbinary (3)	include nonbinary (4)	exclude nonbinary (5)	include nonbinary (6)	exclude nonbinary (7)	include nonbinary (8)
Panel A. Binary gender								
Female	-0.378*** (0.081)	-0.380*** (0.081)	-0.128*** (0.041)	-0.130*** (0.040)	-0.145*** (0.023)	-0.144*** (0.023)	-0.002 (0.083)	0.005 (0.083)
R ²	0.036	0.036	0.017	0.017	0.062	0.061	0.000	0.000
Panel B. Gender identity								
CGI (feminine)	-0.169*** (0.040)	-0.168*** (0.039)	-0.056*** (0.020)	-0.056*** (0.020)	-0.066*** (0.012)	-0.067*** (0.011)		
R ²	0.028	0.028	0.013	0.013	0.052	0.053		
Panel C. Binary gender and gender identity								
Female	-0.305* (0.153)	-0.317* (0.151)	-0.110 (0.072)	-0.115 (0.071)	-0.110* (0.043)	-0.103* (0.043)		
CGI (feminine)	-0.045 (0.074)	-0.039 (0.073)	-0.011 (0.036)	-0.009 (0.035)	-0.022 (0.021)	-0.025 (0.021)		
R ²	0.036	0.037	0.017	0.018	0.064	0.063		
Panel D. Sample split by binary gender								
Female: CGI (feminine)	-0.039 (0.060)	-0.043 (0.060)	-0.029 (0.030)	-0.031 (0.029)	-0.042** (0.017)	-0.044*** (0.017)		
R ²	0.002	0.002	0.004	0.004	0.023	0.025		
Male: CGI (feminine)	-0.017 (0.063)	-0.008 (0.062)	0.013 (0.030)	0.017 (0.029)	0.013 (0.017)	0.011 (0.017)		
R ²	0.000	0.000	0.001	0.001	0.002	0.002		
Observations	584	590	584	590	584	590	584	590
Mean of Dependent Variable	0	0	0.413	0.414	0.538	0.537	0	0

Notes: The table presents results from regressing incentivized behavioral measures on our standardized CGI measure (mean=0, SD=1). CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni sample is used. In the first column for each outcome measure, we exclude individuals identifying as non-binary and in the second column, we include individuals identifying as non-binary. Binary gender is measured with the question “Are you male or female?” with answer options “female” and “male” which every participant had to answer. Non-binary gender is measured with the question “What is your current gender?” with answer options “Woman”, “Man”, “Transgender”, “Non-binary”, “Prefer not to answer” and “Other”. The estimates in each column and panel come from a separate regression. Risk preference is a standardized measure of participants’ investment decisions (mean=0, SD=1). Competitiveness is a dummy that takes the value 1 for those who chose to compete in the competitive task. Our measure of preferences for equality versus efficiency is measured in deciles, with increasing numbers indicating higher priority for efficiency. Overconfidence is measured as relative overplacement and is standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. .*** p<0.005, ** p<0.01, * p<0.05.

Table A31: CGI and Swiss Uni Unincentivized Behavioral Measures with and without Non-binary Participants

	Staircase Risk		Risk		Competitiveness		Redistribution		Altruism	
	exclude nonbinary (1)	include nonbinary (2)	exclude nonbinary (3)	include nonbinary (4)	exclude nonbinary (5)	include nonbinary (6)	exclude nonbinary (7)	include nonbinary (8)	exclude nonbinary (9)	include nonbinary (10)
Panel A. Binary gender										
Female	-0.211* (0.083)	-0.217** (0.082)	-0.444*** (0.081)	-0.441*** (0.081)	-0.388*** (0.081)	-0.390*** (0.081)	0.201* (0.083)	0.196* (0.082)	0.249*** (0.082)	0.257*** (0.082)
R ²	0.011	0.012	0.049	0.049	0.038	0.038	0.010	0.010	0.016	0.017
Panel B. Gender identity										
CGI (feminine)	-0.137*** (0.044)	-0.140*** (0.043)	-0.234*** (0.040)	-0.233*** (0.039)	-0.194*** (0.040)	-0.195*** (0.040)	0.106*** (0.041)	0.101** (0.040)	0.095* (0.041)	0.099** (0.041)
R ²	0.019	0.020	0.055	0.054	0.038	0.038	0.011	0.010	0.009	0.010
Panel C. Binary gender and gender identity										
Female	0.031 (0.148)	0.028 (0.146)	-0.188 (0.148)	-0.187 (0.147)	-0.212 (0.138)	-0.214 (0.137)	0.082 (0.155)	0.092 (0.154)	0.279 (0.166)	0.283 (0.164)
CGI (feminine)	-0.149* (0.078)	-0.151* (0.077)	-0.157* (0.073)	-0.156* (0.072)	-0.108 (0.068)	-0.108 (0.068)	0.073 (0.076)	0.064 (0.076)	-0.018 (0.084)	-0.016 (0.083)
R ²	0.019	0.020	0.058	0.057	0.042	0.042	0.012	0.011	0.016	0.017
Panel D. Sample split by binary gender										
Female: CGI (feminine)	-0.050 (0.064)	-0.056 (0.064)	-0.081 (0.057)	-0.092 (0.057)	-0.015 (0.054)	-0.007 (0.054)	-0.136 (0.063)	-0.130 (0.062)	-0.047 (0.071)	-0.039 (0.070)
R ²	0.003	0.003	0.007	0.009	0.000	0.000	0.019	0.017	0.002	0.002
Male: CGI (feminine)	-0.123* (0.068)	-0.121* (0.067)	-0.105 (0.064)	-0.095 (0.064)	-0.110* (0.062)	-0.118* (0.062)	0.193*** (0.062)	0.177*** (0.062)	0.020 (0.068)	0.015 (0.068)
R ²	0.015	0.015	0.011	0.009	0.012	0.014	0.037	0.031	0.000	0.000
Observations	584	590	584	590	584	590	584	590	584	590
Mean of Dependent Variable	0	0	0	0	0	0	0	0	0	0

Notes: The table presents results from regressing unincentivized behavioral measures on our standardized CGI measure (mean=0, SD=1). CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni sample is used. In the first column for each outcome measure, we exclude individuals identifying as non-binary and in the second column, we include individuals identifying as non-binary. Binary gender is measured with the question “Are you male or female?” with answer options “female” and “male” which every participant had to answer. Non-binary gender is measured with the question “What is your current gender?” with answer options “Woman”, “Man”, “Transgender”, “Non-binary”, “Prefer not to answer” and “Other”. The estimates in each column and panel come from a separate regression. Staircase risk is a categorical certainty equivalence measure of risk-taking based on a series of hypothetical allocation decisions. Risk is a self-reported measure of risk-taking. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk taking or competitiveness. Redistribution is a measure of how much economic redistribution one wants in society. Altruism is a measure of how much one would donate out of a windfall gain. A higher value means a greater willingness to redistribute or donate. All measures are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. *** p<0.005, ** p<0.01, * p<0.05.

Table A32: CGI and Incentivized Behavioral Measures (with Binary Gender and CGI Interaction)

Measure	Risk		Competitiveness	Efficiency		Overconfidence
Sample	Swiss Uni (1)	US Adults (2)	Swiss Uni (3)	Swiss Uni (4)	Swedish Youths (5)	Swiss Uni (6)
Panel A. Binary gender						
Female	-0.378*** (0.081)	-0.215*** (0.032)	-0.128*** (0.041)	-0.145*** (0.023)	-0.161*** (0.017)	-0.002 (0.083)
R ²	0.036	0.012	0.017	0.062	0.078	0.000
Panel B. Gender identity						
CGI (feminine)	-0.169*** (0.040)	-0.092*** (0.016)	-0.056*** (0.020)	-0.066*** (0.012)	-0.068*** (0.009)	
R ²	0.028	0.008	0.013	0.052	0.055	
Panel C. Binary gender and gender identity						
Female	-0.305* (0.153)	-0.195*** (0.057)	-0.110 (0.072)	-0.110** (0.043)	-0.135*** (0.026)	
CGI (feminine)	-0.045 (0.074)	-0.012 (0.029)	-0.011 (0.036)	-0.022 (0.021)	-0.017 (0.013)	
R ²	0.036	0.012	0.017	0.064	0.079	
Panel D. Sample split by binary gender						
Female: CGI (feminine)	-0.039 (0.060)	-0.001 (0.023)	-0.029 (0.030)	-0.042** (0.017)	-0.014 (0.011)	
R ²	0.002	0.000	0.004	0.023	0.003	
Male: CGI (feminine)	-0.017 (0.063)	-0.012 (0.023)	0.013 (0.030)	0.013 (0.017)	-0.010 (0.014)	
R ²	0.000	0.000	0.001	0.002	0.001	
Panel E. Binary gender and CGI interaction						
Female	-0.301* (0.152)	-0.196*** (0.057)	-0.101 (0.072)	-0.098* (0.043)	-0.134*** (0.026)	
CGI (feminine)	-0.029 (0.106)	-0.022 (0.041)	0.021 (0.048)	0.021 (0.027)	-0.013 (0.018)	
Female*CGI	-0.037 (0.146)	0.021 (0.057)	-0.075 (0.073)	-0.098* (0.042)	-0.010 (0.027)	
R ²	0.036	0.012	0.017	0.064	0.079	
Observations	584	3,902	584	584	1,041	584
Mean of Dependent Variable	0	0	0.413	0.538	0.550	0

Notes: The table presents results from regressing the incentivized measures of risk, competitiveness, equality versus efficiency and overconfidence on our standardized (mean=0, SD=1) measure of CGI. Swiss Uni, U.S. Adults (age 30-60), and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Risk preference is a standardized measure of participants' investment decisions (mean=0, SD=1). Competitiveness is a dummy that takes the value 1 for those who chose to compete in the competitive task. Our measure of preferences for equality versus efficiency is measured in deciles, with increasing numbers indicating higher priority for efficiency. Overconfidence is measured as relative overplacement and standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Panel E, significance levels are determined using a one-tailed test in the direction of the coefficient estimate of female in Panel A, except for the interaction term, which is determined using a two-tailed test. *** p<0.005, ** p<0.01, * p<0.05.

Table A33: CGI and Unincentivized Risk and Competitiveness Measures (with Binary Gender and CGI Interaction)

Measure	Staircase Risk		Risk			Financial Risk	Competitiveness			
Sample	Swiss Uni	Swiss Uni	U.S Adults	Swiss Teens	Swedish Youths	US Adults	Swiss Uni	U.S Adults	Swiss Teens	Swedish Youths
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)
Panel A. Binary gender										
Female	-0.211*	-0.444***	-0.387***	-0.285***	-0.243***	-0.268***	-0.388***	-0.405***	-0.311***	-0.153*
	(0.083)	(0.081)	(0.031)	(0.071)	(0.062)	(0.015)	(0.081)	(0.031)	(0.071)	(0.062)
R ²	0.011	0.049	0.038	0.020	0.015	0.075	0.038	0.041	0.024	0.006
Panel B. Gender identity										
CGI (feminine)	-0.137***	-0.234***	-0.206***	-0.171***	-0.109***	-0.119***	-0.194***	-0.218***	-0.181***	-0.109***
	(0.044)	(0.040)	(0.016)	(0.036)	(0.031)	(0.008)	(0.040)	(0.016)	(0.036)	(0.029)
R ²	0.019	0.055	0.042	0.029	0.012	0.059	0.038	0.048	0.033	0.012
Panel C. Binary gender and gender identity										
Female	0.031	-0.188	-0.152***	0.124	-0.182*	-0.224***	-0.212	-0.145**	0.087	0.022
	(0.148)	(0.148)	(0.056)	(0.166)	(0.103)	(0.027)	(0.138)	(0.057)	(0.168)	(0.101)
CGI (feminine)	-0.149*	-0.157*	-0.144***	-0.227***	-0.041	-0.027*	-0.108	-0.159***	-0.220***	-0.117**
	(0.078)	(0.073)	(0.029)	(0.085)	(0.051)	(0.013)	(0.068)	(0.029)	(0.085)	(0.048)
R ²	0.019	0.058	0.044	0.030	0.015	0.076	0.042	0.049	0.033	0.012
Panel D. Sample split by binary gender										
Female: CGI (feminine)	-0.050	-0.081	-0.018	-0.051	0.058	-0.001	-0.015	0.068	0.021	0.015
	(0.064)	(0.057)	(0.025)	(0.060)	(0.046)	(0.010)	(0.054)	(0.024)	(0.055)	(0.044)
R ²	0.003	0.007	0.000	0.003	0.003	0.000	0.000	0.005	0.000	0.000
Male: CGI (feminine)	-0.123*	-0.105	-0.143***	-0.154***	-0.097*	-0.029***	-0.110*	-0.253***	-0.238***	-0.163***
	(0.068)	(0.064)	(0.024)	(0.045)	(0.052)	(0.011)	(0.062)	(0.024)	(0.047)	(0.050)
R ²	0.015	0.011	0.020	0.024	0.009	0.003	0.012	0.064	0.056	0.027
Panel E. Binary gender and CGI interaction										
Female	0.020	-0.191	-0.163***	0.156	-0.209*	-0.227***	-0.229	-0.171***	0.150	-0.007
	(0.152)	(0.147)	(0.057)	(0.165)	(0.105)	(0.027)	(0.140)	(0.057)	(0.165)	(0.104)
CGI (feminine)	-0.190*	-0.167	-0.245***	-0.411***	-0.131*	-0.049***	-0.169*	-0.410***	-0.605***	-0.210***
	(0.105)	(0.103)	(0.041)	(0.120)	(0.071)	(0.019)	(0.095)	(0.039)	(0.120)	(0.064)
Female*CGI	0.095	0.024	0.213***	0.311	0.228*	0.048	0.142	0.533***	0.648***	0.236*
	(0.161)	(0.145)	(0.059)	(0.168)	(0.104)	(0.027)	(0.138)	(0.058)	(0.167)	(0.100)
R ²	0.019	0.058	0.044	0.030	0.015	0.076	0.042	0.049	0.033	0.012
Observations	584	584	3,902	786	1,041	3,902	584	3,902	792	1,041
Mean of Dependent Variable	0	0	0	0	0	0.401	0	0	0	0

Notes: The table presents results from regressing unincentivized measures of risk attitudes and competitiveness on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens, and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Staircase risk is a categorical certainty equivalence measure of risk-taking based on a series of hypothetical allocation decisions. Risk is a self-reported measure of risk-taking. Financial risk is a dummy that takes the value 1 for those who report to actively trade in securities. Competitiveness is a self-reported measure of competitiveness. A higher value means higher risk taking or competitiveness. All measures except financial market risk are standardized (mean 0, SD 1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Panel E, significance levels are determined using a one-tailed test in the direction of the coefficient estimate of female in Panel A, except for the interaction term, which is determined using a two-tailed test. *** p<0.005, ** p<0.01, * p<0.05.

Table A34: CGI and Unincentivized Distributional Preferences (with Binary Gender and CGI Interaction)

Measure	Redistribution		Altruism			
Sample	Swiss Uni (1)	U.S. Adults (2)	Swiss Uni (3)	U.S. Adults (4)	Swiss Teens (5)	Swedish Youths (6)
Panel A. Binary gender						
Female	0.201* (0.083)	0.063* (0.032)	0.249*** (0.082)	0.001 (0.032)	0.210*** (0.071)	0.551*** (0.060)
R ²	0.010	0.001	0.016	0.000	0.011	0.076
Panel B. Gender identity						
CGI (feminine)	0.106*** (0.041)	0.053*** (0.017)	0.095* (0.041)		0.074* (0.035)	0.224*** (0.032)
R ²	0.011	0.003	0.009		0.005	0.050
Panel C. Binary gender and gender identity						
Female	0.082 (0.155)	-0.072 (0.055)	0.279* (0.166)		0.412** (0.165)	0.489*** (0.099)
CGI (feminine)	0.073 (0.076)	0.083*** (0.029)	-0.018 (0.084)		-0.112 (0.082)	0.042 (0.052)
R ²	0.012	0.003	0.016		0.013	0.077
Panel D. Sample split by binary gender						
Female: CGI (feminine)	-0.136 (0.063)	-0.115 (0.023)	-0.047 (0.071)		-0.051 (0.054)	0.052 (0.046)
R ²	0.019	0.013	0.002		0.013	0.003
Male: CGI (feminine)	0.193*** (0.062)	0.196*** (0.024)	0.020 (0.068)		-0.046 (0.046)	0.013 (0.053)
R ²	0.037	0.038	0.000		0.002	0.000
Panel E. Binary gender and CGI interaction						
Female	0.148 (0.155)	-0.045 (0.055)	0.292* (0.167)		0.414** (0.164)	0.481*** (0.097)
CGI (feminine)	0.315*** (0.101)	0.337*** (0.041)	0.033 (0.112)		-0.122 (0.121)	0.017 (0.073)
Female*CGI	-0.562*** (0.152)	-0.538*** (0.058)	-0.117 (0.169)		0.017 (0.165)	0.063 (0.102)
R ²	0.012	0.003	0.016		0.013	0.077
Observations	584	3,902	584	3,902	798	1,041
Mean of Dependent Variable	0	0	0	0	0	0

Notes: The table presents results from regressing unincentivized distributional measures on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. Swiss Uni, U.S. Adults (age 30-60), Swiss Teens and Swedish Youth samples are used. The estimates in each column and panel come from a separate regression. The table presents results from regressing self-reported preferences on redistribution and altruism on our standardized measure of CGI. Redistribution is a measure of how much economic redistribution one wants in society. Altruism is a measure of how much one would donate out of a windfall gain. A higher value means the greater willingness to redistribute or donate. All outcome measures are standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Panel E, significance levels are determined using a one-tailed test in the direction of the coefficient estimate of female in Panel A, except for the interaction term, which is determined using a two-tailed test. *** p<0.005, ** p<0.01, * p<0.05.

Table A35: CGI and Educational Choice (with Binary Gender and CGI Interaction)

Measure	Female Educational Track Share			Skill Requirements	
				Math	Language
Sample	U.S Adults (1)	Swiss Teens (2)	Swedish Youths (3)	Swiss Teens (4)	Swiss Teens (5)
Panel A. Binary gender					
Female	0.125*** (0.007)	0.383*** (0.012)	0.093*** (0.009)	-0.917*** (0.047)	0.913*** (0.047)
R ²	0.108	0.404	0.090	0.204	0.203
Panel B. Gender identity					
CGI (feminine)	0.061*** (0.004)	0.167*** (0.006)	0.039*** (0.005)	-0.403*** (0.024)	0.404*** (0.023)
R ²	0.104	0.315	0.063	0.163	0.163
Panel C. Binary gender and gender identity					
Female	0.075*** (0.014)	0.329*** (0.023)	0.079*** (0.014)	-0.762*** (0.084)	0.745*** (0.083)
CGI (feminine)	0.031*** (0.007)	0.033*** (0.011)	0.009 (0.007)	-0.092* (0.042)	0.100** (0.040)
R ²	0.117	0.408	0.092	0.207	0.206
Panel D. Sample split by binary gender					
Female: CGI (feminine)	0.009* (0.005)	0.028*** (0.009)	0.007 (0.006)	-0.028 (0.040)	0.135*** (0.038)
R ²	0.003	0.014	0.002	0.001	0.018
Male: CGI (feminine)	0.024*** (0.006)	0.009 (0.009)	0.005 (0.007)	-0.087** (0.035)	0.005 (0.035)
R ²	0.016	0.002	0.001	0.008	0.000
Panel E. Binary gender and CGI interaction					
Female	0.078*** (0.014)	0.328*** (0.023)	0.079*** (0.014)	-0.763*** (0.082)	0.744*** (0.084)
CGI (feminine)	0.041*** (0.010)	0.020 (0.018)	0.007 (0.009)	-0.171** (0.067)	0.010 (0.070)
Female*CGI	-0.024 (0.014)	0.022 (0.023)	0.006 (0.014)	0.136 (0.084)	0.156 (0.084)
R ²	0.117	0.408	0.092	0.207	0.206
Observations	2,279	1,448	1,041	1,433	1,433
Mean of Dependent Variable	0.541	0.442	0.521	0	0

Notes: The table presents results from regressing the educational track measures on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. U.S. Adults (age 30–60), Swiss Teens and Swedish Youths samples are used. The estimates in each column and panel come from a separate regression. Female educational track share is the share of women graduating with a bachelor's degree in the chosen field of study in 2020 for U.S. Adults; the share of women from past cohorts of graduates from the apprenticeship for Swiss Teens; the share of women accepted for undergraduate studies in that field the year before our sample made their educational choices for Swedish Youths.. Skills requirements are a standardized (mean=0, SD=1) measure based on expert evaluation of the job content in occupations chosen by Swiss Teens to start apprenticeships. Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Panel E, significance levels are determined using a one-tailed test in the direction of the coefficient estimate of female in Panel A, except for the interaction term, which is determined using a two-tailed test. *** p<0.005, ** p<0.01, * p<0.05.

Table A36: CGI and Labor Market Outcomes (with Binary Gender and CGI Interaction)

Measure	Income	Full-time Homemaker	Weekly Ave. Work Hours	Female Industry Share	Managerial Responsibilities	Performance Pay	Wage Negotiation	Work Flexibility
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
Panel A. Binary gender								
Female	-23.265*** (1.865)	0.099*** (0.008)	-6.644*** (0.451)	0.062*** (0.007)	-0.125*** (0.016)	-0.082*** (0.021)	-0.132*** (0.016)	0.141*** (0.045)
R ²	0.038	0.043	0.053	0.039	0.016	0.008	0.017	0.005
Panel B. Gender identity								
CGI (feminine)	-10.841*** (0.958)	0.049*** (0.004)	-3.141*** (0.222)	0.025*** (0.003)	-0.061*** (0.008)	-0.043*** (0.010)	-0.066*** (0.008)	0.060*** (0.022)
R ²	0.033	0.042	0.047	0.026	0.015	0.009	0.017	0.004
Panel C. Binary gender and gender identity								
Female	-16.723*** (2.990)	0.057*** (0.012)	-4.553*** (0.806)	0.060*** (0.011)	-0.078*** (0.027)	-0.038 (0.035)	-0.074** (0.028)	0.126 (0.077)
CGI (feminine)	-3.998*** (1.542)	0.026*** (0.006)	-1.278*** (0.397)	0.001 (0.006)	-0.029* (0.014)	-0.028 (0.017)	-0.035** (0.014)	0.009 (0.039)
R ²	0.040	0.047	0.055	0.039	0.018	0.009	0.019	0.005
Panel D. Sample split by binary gender								
Female: CGI (feminine)	2.416 (1.124)	0.025*** (0.007)	-0.155 (0.336)	-0.004 (0.005)	0.027 (0.011)	-0.012 (0.014)	0.002 (0.011)	-0.079 (0.033)
R ²	0.002	0.006	0.000	0.001	0.003	0.001	0.000	0.006
Male: CGI (feminine)	-6.735*** (1.356)	0.005* (0.002)	-1.281*** (0.312)	0.005 (0.005)	-0.058*** (0.011)	-0.020 (0.015)	-0.041*** (0.011)	0.085*** (0.032)
R ²	0.011	0.002	0.009	0.001	0.014	0.002	0.007	0.007
Panel E. Binary gender and CGI Interaction								
Female	-17.508*** (2.960)	0.055*** (0.012)	-4.647*** (0.809)	0.061*** (0.011)	-0.085*** (0.027)	-0.038 (0.035)	-0.078*** (0.028)	0.150* (0.077)
CGI (feminine)	-11.425*** (2.300)	0.009* (0.004)	-2.173*** (0.529)	0.009 (0.008)	-0.098*** (0.019)	-0.032 (0.024)	-0.069*** (0.019)	0.134*** (0.049)
Female*CGI	15.740*** (3.053)	0.036** (0.013)	1.896* (0.800)	-0.017 (0.012)	0.146*** (0.027)	0.010 (0.035)	0.072** (0.028)	-0.278*** (0.078)
R ²	0.040	0.047	0.055	0.039	0.018	0.009	0.019	0.005
Observations	3,902	3,902	3,902	1,994	3,902	1,994	3,902	1,994
Mean of Dependent Variable	61.421	0.062	25.281	0.497	0.400	0.303	0.491	0

Notes: The table presents results from regressing the labor market outcomes on our standardized (mean=0, SD=1) measure of CGI. CGI is standardized (mean=0, SD=1) separately for the male and female samples in Panel D. The sample of the U.S. Adults (age 30-60) is used. The estimates in each column and panel come from a separate regression. Income is a self-reported categorical measure in thousand U.S. dollars. Full-time homemaker is a dummy for working full-time at home. Weekly average work hours is a proxy measure constructed from weeks worked and hours worked, using the product of the two categorical measures and dividing by 52. Female industry share is the share of female employees in a given industry. Managerial responsibilities is a dummy for having managerial responsibilities at work. Performance pay is a dummy that is equal to 1 if the respondent's current or most recent job has performance-related pay. Wage negotiation is a dummy that is equal to 1 if the respondent ever negotiated wage. Work flexibility is a categorical measure of flexibility in working hours where 0 is no flexibility, 0.5 is can adapt work hours and 1 is full flexibility, which is standardized (mean=0, SD=1). Robust standard errors are reported in parentheses. Significance levels in Panel A are determined by a two-tailed test. Significance levels in Panels B, C and D are determined by a one-tailed test in the direction of the coefficient estimate of female in Panel A. For Panel E, significance levels are determined using a one-tailed test in the direction of the coefficient estimate of female in Panel A, except for the interaction term, which is determined using a two-tailed test. *** p<0.005, ** p<0.01, * p<0.05.

B. Appendix B

B.1. Online experiment with Swiss students—*Swiss Uni*

B.1.1 *Participants and Procedures*

Our first data collection focused primarily on the explanatory power of CGI for gender gaps in economic preferences. This experiment was pre-registered (<https://osf.io/phyt6/>). Changes were implemented compared to the pre-registration mainly because we adapted and intended laboratory experiment to an online format due to Covid-19 restrictions. Most importantly, we increased the sample size based on an expectation of additional noise in the online setting, a few secondary measures were excluded for a shorter experiment more suitable to the online setting, and the competitive task was exchanged to prevent cheating, or any beliefs thereof. A detailed description of the departure from the initial pre-registration is available in the document “Updates to Pre-registration Final.pdf” available at <https://osf.io/phyt6/>.

The online experiment was conducted in September and October 2021. The experiment was implemented in English using o-Tree (Chen, Schonger and Wickens, 2016). The 597 participants were students recruited from a subject pool consisting mainly of students at the University of Zurich (UZH) and the Swiss Federal Institute of Technology (ETH), using the software h-root (Bock, Baetge and Nicklisch, 2014). Participants received a variable amount based on the outcome of their decisions in the incentivized tasks. During the experiment, payoffs were measured in Experimental Currency Units (ECU) with an exchange rate of 20 ECU to 1 CHF. A follow-up survey programmed with Qualtrics was conducted two weeks after the main experiment. Participation in the follow-up survey was strongly incentivized—participants received a fixed payment of 50 CHF only for successfully completing both the laboratory session and the online survey. Thus, the dropout rate was extremely small (less than 1%). The main purpose of the follow-up survey was to duplicate the measure of CGI using a slightly modified scale to account for possible measurement error, following Gillen, Snowberg and Yariv (2019). We considered this approach because self-reported gender may be susceptible to measurement error. In the experiment and follow-up survey, we collected several incentivized and un-incentivized preference measures. We list our main measures below.³⁰ All the

³⁰Our main measure of CGI and the incentivized preference measures are relevant for our primary pre-registered analysis. For secondary analysis, we additionally elicited a two-dimensional measure (measuring masculinity and femininity on separate dimensions) in first and third person—how others see a respondent—following Magliozzi, Saperstein and Westbrook (2016). The questionnaire also included an unincentivized measure of overconfidence following Gillen, Snowberg and Yariv (2019), an unincentivized measure of willingness to engage in self-promotion following the design of Exley and Kessler (2022), questions about field of study, family and sibling structure, parental division of household work, parental education, occupation, a short version of the Big-5 personality inventory, and perceived gender identity (male, female, transgender, other), sex at birth, sexual orientation, and relationship status.

study material (experimental decision screens and the follow-up survey) are reproduced in Appendix C.

B.1.2 *Elicited Measures*

CGI. In the main experiment, CGI was elicited in a final survey after completion of the incentivized parts of the study. The question came right after demographic questions (age, binary gender, nationality and mother tongue) on a new screen. The precise question was “In general, how do you see yourself? Where would you put yourself on this scale from “0-Very masculine” to “10-Very feminine”?”. On the same screen, we also asked how most people see a respondent on this scale (“In general, how do most people see you? Where would most people put you on this scale from “0-Very masculine” to “11-Very feminine”?”). In the follow-up survey, CGI was elicited as the first question right after initial instructions. It was asked in the following way “In general, how do you see yourself? Where would you put yourself on this scale from “0-Very feminine” to “11-Very masculine”?”.

Incentivized preference measures. Our main outcome variables were elicited at the onset of the experiment and comprise incentivized preference measures of attitudes to risk, competitiveness, preference for equality versus efficiency, and overconfidence that are frequently used in the literature, and for which earlier studies have documented a gender gap.

Risk preferences were elicited through a one-shot investment task (Gneezy and Potters, 1997), in which participants allocate between 0 and 100 ECU to a risky investment. The investment has a 50% probability of success, in which case it returns 2.5 times the invested amount. If the investment fails, the invested ECU are lost. Our outcome variable for risk seeking is the amount invested. Based on earlier research, we expect women to invest less than men.

We elicited *Competitiveness* following the design introduced by Niederle and Vesterlund (2007), although with a different competitive task. The experimental task chosen for the online implementation is the matrix task used by Buser, Niederle and Oosterbeek (2021), in which participants are asked to identify the two numbers in a 3x3 matrix that sum to a target number. We used this task since participants could easily solve the arithmetic task from the original design with the aid of a calculator. Participants were incentivized to solve as many tasks as possible across three rounds of three minutes each. In Round 1, participants received a piece-rate payment of 10 ECU per correct exercise. In Round 2, participants were compensated under a tournament scheme—they were randomly assigned to groups of four and the participant who solved the most exercises within a group earned 40 ECU per correct calculation, with the other three group members earning nothing. Ties were randomly broken. Participants did not find out

how they performed relative to other group members until the end of the experiment. In Round 3, participants chose between the piece-rate or tournament payment schemes. If they choose the tournament scheme, their performance was compared to their group members' previous performance in Round 2. One round was randomly selected for payout. Our outcome variable for competitiveness is the binary choice in Round 3, with the choice of the tournament indicating a preference for competition. Based on prior research, we expect women to compete less than men.

We elicited preferences for *Equality versus efficiency* by implementing 15 graphical budget sets, similarly to Fisman, Kariv and Markovits (2007). In each choice, a participant distributes ECU between him- or herself and another randomly assigned participant with the relative price of giving varying across choices. Our main estimate of interest is the parameter ρ , which measures the equality-efficiency tradeoff from a CES utility function. To determine ρ , we use the maximum likelihood (ML) estimation method provided by Fisman et al adapted to our setting with 15 predetermined budget sets instead of 50 randomly chosen ones. Positive values of ρ , specifically $0 < \rho \leq 1$, indicate distributional preferences that are weighted towards efficiency (increasing total income), while negative values of ρ indicate weighting toward equality (reducing differences in income). Our outcome variable for this preference is the decile to which a participant's estimated ρ belongs, following the approach proposed in Fisman, Jakiela and Kariv (2017) to deal with outliers. Lower deciles indicate a relatively lower weight placed on efficiency relative to equity. Prior research documents that women often prioritize equality to a greater extent than men (Fisman, Jakiela and Kariv, 2017). Hence, we expect women to have lower scores than men.

We elicit *Overconfidence* using three measures of relative overplacement (Moore and Healy, 2008). First, following Gillen, Snowberg and Yariv (2019), participants solve a selection of eight matrix reasoning items (similar to Raven's matrices) provided by Condon and Revelle (2014). Participants have 45 seconds to solve each puzzle. One of the eight puzzles is randomly chosen for payout, which is 50 ECU if it is solved correctly and 0 if it is not. Participants then guess their performance rank within a randomly chosen reference group of 26 study participants, including themselves, receiving an additional 20 ECU if they guess within two ranks of their actual rank.³¹ Second, we elicited two additional measures of overplacement during the measurement of competitive preferences, during which we asked participants to guess their relative rank, first within the group of four contestants and then within a randomly chosen reference group of 26 study participants. For a correct rank-out-of-four-guess, and a rank-out-of-26-guess within two ranks of their actual rank, participants receive 20 ECU, respectively. We construct our measure of overconfidence as a standardized summary index of these three

³¹They also guess how many of the eight tasks they solved correctly (overestimation) and, if this guess is correct, they again receive 20 ECU. As our focus is on overplacement, we do not use this measure.

overplacement measures and reverse the sign such that a higher score indicates greater overconfidence. Based on prior research, we expect women to be less overconfident than men.

Unincentivized preference measures. In the follow-up survey we also elicited non-incentivized preference measures such as self-reported measures of risk seeking, competitiveness, overprecision, preference for redistribution, and altruism. As hypothetical measures of risk we use i) a question from Dohmen et al. (2011) asking participants to report their general willingness to take risks on a scale from 0 to 10, and ii) the staircase measure of hypothetical choices between a lottery and a safe value, varying the amount of money in the safe option across choices, as proposed in Falk et al. (2023). The measure of competitiveness follows a similar logic as the Dohmen et al. (2011) general risk measure (Buser, Niederle and Oosterbeek, 2021). As a hypothetical measure of altruism participants stated how much of an unexpected CHF 2400 windfall they would donate to a good cause (Dohmen et al., 2011). Finally, we asked participants to state how much redistribution they want in society on a scale ranging from 1 (no redistribution) to 10 (full redistribution).

B.2. Prolific Survey with U.S. Adults—*U.S. Adults*

B.2.1 Participants and Procedures

To study whether CGI explains variation in economic choices and outcomes in a broader sample and across a more varied set of outcomes, we also administered an online survey to U.S. residents of working age in two waves. The respondents were recruited and paid through Prolific Academic. We limited recruitment to participants reporting English as their primary language and with an approval rate for prior studies above 98 percent. Only participants who passed two attention checks were allowed to complete the study in wave 1, and those who failed attention checks were later excluded from the sample in wave 2. Recruitment was stratified on gender and age. In wave 1, conducted in March 2022, we recruited 800 respondents in the age brackets 30-39 and 40-49 years, and 400 respondents in the age bracket 50-60 years, giving slightly lower weight to older participants who are closer to exiting the labor market. Participants received \$1.25 for survey completion and could earn up to \$2.5 in an incentivized risky investment task. In March 2023, we recruited a new sample of 2,802 participant from the Prolific subject pool again stratified on gender and age, excluding previous participants. In wave 2, we also recruited adults in the age bracket 20-29. We collected 800 participants from each of the age groups, 20-29, 30-39 and 40-49, and 400 participants aged 50-60, due to the lower number of available respondents in the latter age group.

B.2.2 *Elicited Measures*

CGI. In wave 1, CGI was elicited at the beginning of the survey right after demographic questions (age, U.S state of residence, ZIP code, sex) and on a separate screen. In addition to self-view, we also asked where most people would put a respondent on that scale from 0-“very masculine” to 10-“very feminine” (*CGI (other)*). Both measures were elicited prior to any of the outcome measures and on the same screen. Additionally, at the very end of the survey, we asked respondents to place themselves on a scale from 0-“very masculine” to 10-“very feminine” relative to 1) women and 2) men. The placement relative to others with the same gender is the *CGI (own)* measure reported in the main text. In wave 2, CGI was elicited after all of the outcome measures. It was part of the second block of the survey in which different gender identity measures were administered in random order. All four versions of CGI (standard, as seen by most people, relative to men, relative to women) were asked together on the same screen.

Other measures. All other survey items fall into four domains: Demographics, Education, Family, Employment and Work, and Preferences. We briefly list all the variables collected here. The exact wording of all questions and answer options is available in Appendix C. Moreover, Table A1 describes all the main variables that are used for the analysis presented in the Results section of the paper.

The set of *Demographics* include age, sex, race/ethnicity, current state of residence and zip code. The question “What is your sex?” with answer options “male” and “female” was taken from the U.S. census questionnaire. The set of *Family and Education* variables consists of relationship status, sexual orientation, children, division of housework, level of education (highest degree completed) and major field of study when applicable. *Employment and Work* includes employment status, job search behavior (if applicable), sector of employment, industry, income, weeks worked last year, usual hours worked, flexibility of working hours, changes to working arrangements, the ability to take off an hour for personal matters, managerial responsibilities, performance pay, and experience with wage negotiation. The female share in industry and educational track share were constructed from publicly available datasets for the year 2020 (U.S. Bureau of Labor Statistics, National Center for Education Statistics). The educational track share is the share of women who graduated with a bachelor’s degree in a field of study in 2020.

We also collected three types of preference measures. First, we asked the same incentivized risk task as in the Swiss Uni Sample with an investment endowment of 1 USD. Second, as a proxy for risk-taking, we asked whether a respondent actively trades in securities. Third, we elicited the same unincentivized measures of willingness to take risk (the general questions from Dohmen et al. (2011)), competitiveness, attitude towards redistribution and altruism as in our Swiss Uni sample (see section B.1.2 **Unincentivized**

Preference Measures for a detailed description).

B.3. Online survey with Swiss Adolescents—*Swiss Teens*

B.3.1 *Participants and Procedures*

Our third dataset is based on two surveys comprising 1,755 Swiss adolescents. We recruited these respondents through two newsletters sent from the largest Swiss online platform for apprenticeship search. This platform covers around 90 percent of all online postings for apprenticeships lasting two to four years, which form the most important part of the vocational education and training system in the German-speaking part of Switzerland. About two-thirds of a birth cohort do an apprenticeship after compulsory education, which allows them to combine vocational school with on-the-job training. Most respondents (91 percent) were 8th and 9th grade students (with an average age of 14.8 years) who were planning to do an apprenticeship after compulsory schooling (9th grade). At the time of our surveys (December 2021 and March 2022), the 8th graders had just started considering which apprenticeship they would like to do in the future. They would subsequently apply for trial apprenticeships, which typically last 1-5 days and allow the student to experience a specific apprenticeship at a specific firm. The 9th graders were further along—52 percent of them had already signed a contract for their apprenticeships with a specific company.

B.3.2 *Elicited Measures*

CGI. In the first survey, we elicited CGI on the first page after the consent form; the CGI question was asked after the traditional binary gender classification (Are you male or female?). In the second survey, we asked our CGI question at the very end of the survey. The CGI question was in both surveys asked on a scale from 0-“Very feminine” to 11-“Very masculine”; for the analysis we reverse the scale.

Other measures. In the first survey, after the CGI question, we elicited the same unincentivized measures of willingness to take risk (the general questions from Dohmen et al. (2011)), competitiveness, attitude towards redistribution and altruism as in our Swiss Uni and U.S. Adults samples (see section B.1.2 **Unincentivized Preference Measures** for a detailed description). Moreover, for these respondents we were also able to merge their survey responses to their administrative user profile data from the platform. From this administrative data, we observe all applications submitted for trial apprenticeships and apprenticeships by a respondent through the platform. Thus, as a complement to the measures elicited in our survey, we observe relevant real-life behaviors. While the platform covers around 90 percent of all online apprenticeship postings, it only covers around

a quarter of all trial apprenticeships, as these are much more common to organize informally through informal networks. Given this and because trial apprenticeships mainly take place in the spring, our administrative data on occupational search predominantly captures 9th graders. Therefore, we conducted our second survey to elicit all trial apprenticeships and one-day information events students had attended among 8th and 9th graders by March 2022. Attendance at these trial apprenticeships and information events (we refer to these combined as experiences henceforth) count as excused school absences and represent an important part of students’ occupational choice process.

For each application (observed in the administrative data set) and experience (reported in the second survey), we know the exact classification for the “apprenticeship profession” and merge this to characteristics about the specific apprenticeship. For the analysis of the relationship between CGI and occupational choices, we focus on three main outcome variables that characterize the student’s occupational preferences in terms of gender composition and skill requirements. Since a student may apply to multiple (trial) apprenticeships, we take the average of each characteristic across all considered (trial) apprenticeships. First, we consider the female share of apprentice graduates from 2019–2021 (i.e., the gender composition of past apprentice cohorts, using the administrative LABB (*Längsschnittanalysen im Bildungsbereich*) data from the Swiss Federal Statistical Office). Second, we consider the math and school language skill requirements, based on expert evaluations of the academic requirements associated with each apprenticeship; these data come from <https://anforderungsprofile.ch>. We standardize the latter two measures to have a mean of zero and a standard deviation of one.

B.4. Experiment with Swedish Secondary School Students—*Swedish Youths*

B.4.1 *Participants and Procedures*

We had the opportunity to insert our measure of CGI into an experiment with Swedish students (age 18-19 years). The experiment was implemented in several schools and all 1,063 respondents were students in the final year of the natural science track in Swedish secondary education. It was implemented either in the respondents’ school or online and took place in the beginning of 2022, in the weeks before the students applied for their preferred educational field for subsequent university studies.

B.4.2 *Elicited Measures*

CGI. CGI was elicited in a final survey after completion of the incentivized parts of the study. The question came right after demographic questions (age and a discrete measure of gender identity “Do you identify as a woman or man?” with answer options “Woman”, “Man” and “Other”).

Incentivized preferences. Preferences for *Equality versus efficiency* were elicited at the end of the main study through the implementation of 20 graphical budget sets, similarly to Fisman, Kariv and Markovits (2007). In each choice, a participant distributes ECU between him- or herself and another randomly assigned participant with the relative price of giving varying across choices. Our main estimate of interest is the parameter ρ , which measures the equality-efficiency tradeoff from a CES utility function. To determine ρ , we use the maximum likelihood (ML) estimation method provided by Fisman et al adapted to our setting with 20 predetermined budget sets instead of 50 randomly chosen ones. Positive values of ρ , specifically $0 < \rho \leq 1$, indicate distributional preferences that are weighted towards efficiency (increasing total income), while negative values of ρ indicate weighting toward equality (reducing differences in income). Our outcome variable for this preference is the decile to which a participant’s estimated ρ belongs, following the approach proposed in Fisman, Jakiela and Kariv (2017) to deal with outliers. Lower deciles indicate a relatively lower weight placed on efficiency relative to equity.

Unincentivized preferences. The end survey also included unincentivized preferences for risk (general willingness to take risks from Dohmen et al. (2011)) and competitiveness (Buser, Niederle and Oosterbeek, 2021), as well as a measure asking participants to report their willingness to donate to charity on a scale from 0-10. Finally, the data set also comprises information on the participant’s intended field of undergraduate studies. We use register data from Statistics Sweden on the actual gender composition of these study fields in 2022 to generate a variable measuring the gender composition of the respondent’s intended field of study.

C. Appendix C

This Appendix reproduces the original full instructions and survey questions that we used for our primary data collection.

The data collection using Swedish Youths was implemented as part of another ongoing study by one of the authors of this paper (Ranehill). The full instructions for this data collection will be made available when that main study is completed.

Order in which Study Materials are Presented

1. Swiss Uni: Experiment
2. Swiss Uni: Follow-Up Survey
3. U.S. Adults: Wave 1
4. U.S. Adults: Wave 2
5. Swiss Teens Survey 1: English Translation
6. Swiss Teens Survey 1: German Original
7. Swiss Teens Survey 2: English Translation
8. Swiss Teens Survey 2: German Original

Swiss Uni: Experiment

These screenshots reproduce the instructions as shown to participants.

Acknowledgement of Informed Consent

You are taking part in a study designed by researchers at the University of Zurich, the University of Gothenburg and the Humboldt University of Berlin. The study is part of research aimed at better understanding how people make economic decisions. The following is intended to provide you with information on the procedures involved with the study in which you will participate today.

- You will work on a computer while making decisions that will be described to you before the study starts and over the course of the study.
- You will be paid for your participation. Your exact payment will depend on the choices that you and others make during the study. Your payment will be paid by bank transfer in the coming weeks. For this, the ETH Decision Science Laboratory will collect information on a bank account to which the payment should be transferred. At the end of today's session, you will be directed to a secure online form through which you can submit your bank account details. Note that this information on a bank account will only be used to transfer your payment by personnel of the ETH Decision Science Laboratory. The researchers in charge of this study will not have access to this information, and no record of this information will be kept after your payment has been made. Importantly, that means that no personally identifying information will be stored with the data on the choices that you have made in today's study.
- There is no risk associated with participating in this study, beyond any risks you would encounter in performing activities on a computer in everyday life.
- Your participation in this study is voluntary, and you may withdraw your participation or your data at any time without any penalty. Withdrawing from the study may involve a loss of payment from the study.
- Your data will be kept completely confidential by not recording your name in combination with your data. All records will be identified only by a random ID number.

If you have any question regarding this study, you can contact the ETH Decision Science Laboratory (see contact details below).

Contact: ETH Zurich, Decision Science Laboratory, descil@ethz.ch

I confirm that I understand the above information and that I give my consent to participate in the study and allow the researchers to use the data obtained from my participation.

- ☐ **Yes**
- ☐ **No**

Please click "Next" now.

Next

Initial Instructions

The experiment comprises four main parts. We will provide you with detailed instructions before each part. At the end of the experiment, you will also be asked to fill out a survey.

In the four main parts of this experiment, you have the opportunity to earn money. *You will receive any money that you earn in these four parts. This money will be paid via bank transfer in the coming weeks.*

As part of this study, we also ask you to take part in an online follow-up survey, for which you will receive a link directly by e-mail in approximately 2 weeks. The follow-up survey takes approximately 20 minutes to complete, and you can fill in the survey from anywhere, as long as you have access to a computer or a smartphone. Once you receive the link you will have 72 hours (3 days) to complete the survey. Once you have completed the survey, your participation payment of 40 CHF will be transferred to your bank account.

Please note that you will only receive your participation payment of 40 CHF if you complete the follow-up survey in addition to participating in today's experiment. The participation payment of 40 CHF combines today's show-up fee and a survey completion payment.

To sum up: if you take part in the online follow-up survey, you will receive the 40 CHF participation payment in addition to the money that you earn in today's experiment. If you do not take part in the online follow-up survey, you will only receive the money that you earn in today's experiment. Either way, your earnings from today's study will be paid via bank transfer in the coming weeks.

During the study you are not allowed to use electronic devices or to communicate with other participants. While you participate in this study, you are allowed to only use the programs and functions intended for the study.

If your internet connection were to break down during the study, you could pick up the study where you have left it by accessing the link again that was e-mailed to you shortly before the beginning of today's session. Note that some screens have timers on them. When they are up, the program will automatically advance to the next page. It is therefore important that you remain seated at your computer throughout the study in order to follow its progress attentively.

Next

Additional Earnings

In addition to the 40 CHF payment that you receive for your participation, you will be paid an additional amount of money that you accumulate from decision tasks in the four parts of the study. The exact amount you receive will be determined during the experiment, and will depend on your decisions, and the decisions of others. Please note that the decisions you make in any part of the experiment will have no effect on what happens in other parts.

All monetary amounts in this experiment will be denominated in experimental currency units (ECU). At the end of the experiment, your earnings in ECU will be exchanged into CHF at a rate of

$$20 \text{ ECU} = 1 \text{ CHF}$$

You will be paid these additional earnings via bank transfer in the coming weeks.

Note that all your interactions in the study are anonymous. This means that you will not know the identity of any other participant with whom you interact, and no other participants will know your identity.

Please click "Next" now to see the instructions for Part 1.

Next

Instructions to Part 1

In Part 1, you will make a simple economic decision. You will receive detailed instructions before the decision.

Please note that the decision you make in this part is independent and does not influence the future course of the study. You will be informed about the outcome of your decision and the earnings that you will receive in a few weeks via e-mail. This e-mail will be sent after the time window for completing the online follow-up survey has closed.

Please press the "Next" button now to see the instructions for the decision.

Next

Part 1, Investment Decision

In the task for part 1, we ask you to make an investment decision. This decision is for real money; the result of your decision will be added to your earnings and paid to you via bank transfer in the coming weeks.

You start the investment task with a balance of 100 ECU. You choose how much of this amount (from 0 ECU to 100 ECU) you wish to allocate to the investment.

The ECU that you choose not to invest will be yours to keep and cannot be lost. You will receive these ECU when you are paid for this study.

The value of the ECU you choose to invest depends on the success or failure of the investment.

The success or failure of the investment will be determined by a computerized random draw, similar to a coin flip. There are two possible outcomes:

- **With 50% probability the investment fails**, and you lose the amount invested.
- **With 50% probability the investment succeeds**, and you receive 2.5 times the amount invested.

So, for any amount A that you invest, you will keep $100 - A$, regardless of what happens with the investment.

If the investment fails, which happens with 50% probability, your total earnings from the investment decision will be $100 - A$, since you lose the amount that you invested

If the investment succeeds, which also happens with 50% probability, your total earnings from the investment decision will be $100 - A + 2.5 \cdot A = 100 + 1.5 \cdot A$.

Your Decision:

Please enter the amount of money (from ECU 0 to ECU 100) you wish to invest.

Risky investment:

Once you enter a number, please submit your investment by clicking "Next".

Next

Instructions to Part 2

In Part 2, you will be asked to complete three different tasks. You will receive detailed instructions before each task.

Please note that each of the decisions you make in this part is independent and does not influence the future course of the study. After today's experiment, the computer program will randomly select one of the three tasks in Part 2. You will be paid based on your performance in that task. The method we use to determine your earnings varies across the three tasks. Before each task we will describe in detail how your payment is determined.

Please press the "Next" button now to see the instructions for Task 1 of Part 2.

Next

Part 2, Task 1 - Piece Rate

For Task 1 you will be asked to find the two numbers out of nine that sum to a target number. You will be given 3 minutes to work on a series of these problems in which you have to find the correct two numbers that add up to a target number. Below, you can see what your screen will look like. When completing the task, you will enter each number in a separate box:

Example Screen:

example

In the example above, the correct answer is the two numbers 48 and 34. Each number was entered in a separate box. It does not matter into which box you enter the two numbers, but you must enter the correct two numbers.

Once you have entered the two numbers that solve a problem into the boxes, you can submit your answer by clicking "Submit answer".

If Task 1 is the one randomly selected for payment for Part 2, then you receive **10 ECU per problem you solve correctly** in the 3 minutes during which you work on the task. Your payment does not decrease if you provide an incorrect answer to a problem. We refer to this payment as the **piece rate** payment.

Before proceeding, please answer the following question on the procedure described above:

Suppose you solve 2 problems correctly and 2 problems incorrectly in this part in Task 1 of Part 2. How many ECU will you receive if this part is randomly selected to count for your payment?

Once you are ready to start Task 1 of Part 2, press the "Next" button. If you have answered the above question correctly, you will then proceed to Task 1. You will then have exactly 3 minutes to work on the task.

Next

Part 2, Task 1 - Piece Rate

Which two of the following nine numbers sum to:

Correct: 0

Incorrect: 0

88

66	59	71
47	36	63
41	62	69

Number 1:

Number 2:

Once you have entered the two numbers that solve a problem into the boxes, you can submit your answer by clicking "Next".

Next

Part 2, Task 2 - Tournament

As in Task 1 of Part 2, you will again be given 3 minutes to find the two numbers out of nine that sum to a target number. However, for this task your payment depends on your performance relative to that of a group of other participants. Each group consists of four people. The three other members of your group are randomly selected members who are also participating in this study now. You will not know who is in your group.

If Task 2 is the one randomly selected for payment, then **if you solve the largest number of problems correctly in the group** you receive 40 ECU per problem you solved correctly. If you do not solve the largest number of problems correctly in the group, then you receive 0 ECU. We refer to this as the **tournament** payment. If there are ties the winner will be randomly determined among those solving the most problems. You will not be informed of how you did in the tournament until in a few weeks when you receive feedback on your earnings from today's study via e-mail.

Once you are ready to start Task 2, press the "Next" button. You will then have exactly 3 minutes to work on the task.

Next

Part 2, Task 2 - Tournament

Which two of the following nine numbers sum to:

Correct: 0
Incorrect: 0

64

29	42	33
12	36	13
49	15	38

Number 1:

Number 2:

Once you have entered the two numbers that solve a problem into the boxes, you can submit your answer by clicking "Next".

Next

Part 2, Task 3 – Choice

As in the previous two tasks of Part 2, you will be given 3 minutes to find the two numbers out of nine that add to a target number. However, **you will now get to choose whether you want to be paid according to the piece rate or the tournament.**

If Task 3 is the one randomly selected for payment, then your earnings for this task are determined as follows.

If you choose the **piece rate** you receive 10 ECU per problem you solve correctly.

If you choose the **tournament** your performance will be compared to the performance of the other three participants of your group in Task 2. Task 2 is the one you just completed. If you correctly solve more problems than they did in Task 2, then you receive four times the payment from the piece rate, which is 40 ECU per correct problem. You will receive no earnings for this task if you choose the tournament and do not solve more problems correctly now, than the others in your group did in Task 2.

As your performance in this task is compared to the group members' performance in Task 2, your decision and performance in this task **do not impact their outcomes in any way**. For any participant who chooses the tournament pay, their current performance will also be compared to the group's performance in Task 2.

You will not be informed of how you did in the tournament until a few weeks from now when you receive feedback on your earnings from today's study via e-mail. If there are ties, the winner will be randomly selected.

Please indicate below which payment scheme you choose: **piece rate** or **tournament**.

Make your choice:

- ☐ Piece Rate
☐ Tournament

Once you have made your choice and are ready to start Task 3, press the "Next" button. You will then have exactly 3 minutes to work on the task.

Next

Part 2, Task 3

Which two of the following nine numbers sum to:

Correct: 0

Incorrect: 0

77

22	46	55
26	48	12
64	32	53

Number 1:

Number 2:

Once you have entered the two numbers that solve a problem into the boxes, you can submit your answer by clicking "Next".

Next

Part 2, Questionnaire

Before Part 2 ends, we would like you to answer two additional questions. You can earn additional money based on your answers. Please fill in your answer to the questions below on the screen.

You confirm your entry by clicking the "Next" button. Once you have submitted your answers, Part 3 will begin.

1. This question is about your performance in Task 2 (the tournament task) relative to others in your group. What do you think was your rank within the group in terms of number of problems solved correctly? A problem consisted of correctly finding the two numbers out of nine that sum to a target number. Please choose a number from 1 (meaning that you were the best in your group of four) to 4 (meaning that you were the 4th in your group of four). If your guess is correct, you receive 20 ECU.

My estimated rank in the group:

☐ 1 ☐ 2 ☐ 3 ☐ 4

2. This question is about your performance in Task 2 (the tournament task) relative to 25 randomly selected other participants who are currently also participating in the study. What do you think was your rank in terms of sums solved correctly within 26 participants who are now taking the study? If, for example, you think that you solved the most problems correctly and that no other participant solved as many, you enter the rank of 1. If you think that 10 participants performed better than you, you enter the rank 11. If you think you solved the fewest puzzles correctly, and that all other participants solved more than you, you enter the rank 26. You submit your guessed rank by clicking the "Next" button.

Please consider your guess carefully. You will earn an additional 20 ECU if your answer lies within 2 ranks of your actual rank. That is, if your actual rank is only 0, 1, or 2 away from the rank that you guess, you will receive an extra payment of 20 ECU at the end of the study.

My estimated rank among 26 participants who are participating in this study:

Next

Instructions to Part 3

In Part 3, you will participate repeatedly in 15 independent decision problems that share a common form. You will receive detailed instructions and see a practice example before the task begins.

Please note that each of the decisions you make in this part is independent and does not influence the future course of the study. After today's experiment, the computer program will randomly select one of the decisions and implement it. More details about the implementation of this payment will be provided on the next screen.

Please press the "Next" button now to see the instructions for the allocation decisions in Part 3.

Next

Part 3, Allocation Decisions

You will make 15 decisions in which you allocate ECU between yourself and one other person who is currently participating in this study. The payoff from this task will be added to your earnings and paid to you at the end of the experiment.

For each of the 15 decisions, you will see a range of possible allocations for you and for one other randomly selected participant. Your task in each decision is to choose your preferred allocation among all the alternatives.

After today's experiment, the computer will randomly form pairs from all the participants who are taking the study now. So pairs will be formed after you have made your 15 allocation decisions. You will not find out about the identity of the person with whom you are paired and this person will not find out about your identity. For each pair, the computer will randomly select one person to have their decisions count and will select one of the 15 allocation decisions made by this person. Each allocation decision is equally likely to be selected. Only this choice will be paid out for that pair. You will not know the outcome of these allocation decisions until in a few weeks when you receive feedback on your earnings from today's study via e-mail.

This means that you should treat each decision as if it is the one that will determine your earnings and the earnings of one other person, because this could be the case.

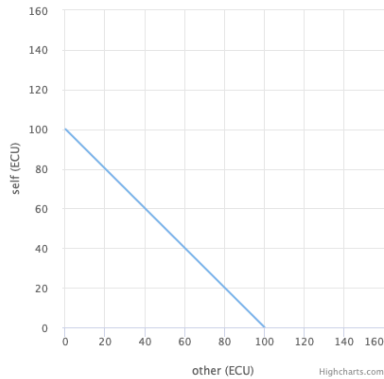
Please press the "Next" button now to see a sample allocation decision.

Next

Part 3, Example of an Allocation Decision

Each decision will involve choosing a point on a graph representing possible ECU allocations. Below, you find an example of the kind of graph you might face. The numbers in this example are used only to illustrate the task and do not correspond to the numbers in the actual decisions.

Each decision problem will start by displaying a blue line on which you can allocate ECU between yourself and one other participant. A pointer will appear once you move your mouse inside the graph area. To choose an allocation, use the mouse to move the pointer on the computer screen to the allocation that you desire. Please try this now.

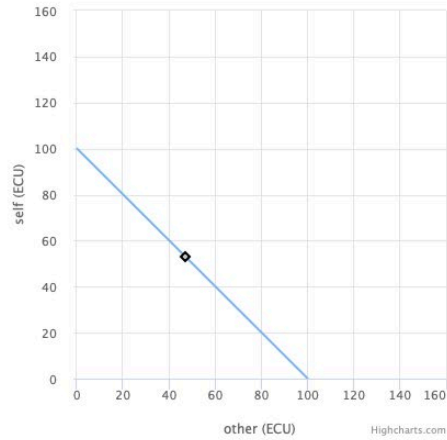


At any point, the computer displays the corresponding payments for you ("self") and for the other ("other"). Note that you can choose only ECU allocations that are on the blue line. On each screen, you will make one choice, corresponding to how you decide to allocate ECU between yourself and the other participant. You are free to select any point you want on the blue line.

Once you have selected a point on the blue line, you have to enter the corresponding allocation. You can do this by entering the ECU that you want to allocate to yourself in the box "Self". And you enter the corresponding ECU that you want to allocate to the other participant in the box "Other". Once you have entered an allocation in the two boxes, you can click anywhere on the screen to display your chosen allocation on the blue line. To submit your chosen allocation, press the "Next"-button or hit "enter".

Please answer the following question on the procedure describe above.

Below you see an example allocation. What are the corresponding ECU that go to "Self" and "Other" in this example? Please enter the two numbers in the box "Self" and the box "Other":



Self:

Other:

Click "Next" to submit your answers.

Next

Part 3, Allocation Decisions - Submitting Your Choices

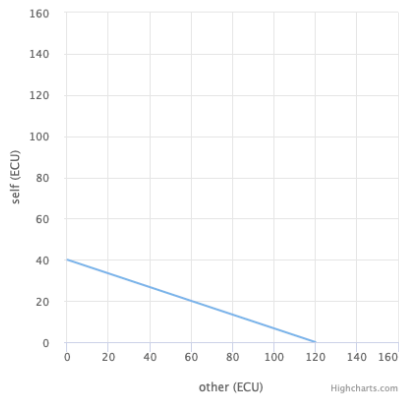
When you are ready to make your decision, click "Next" to submit your chosen allocation. The computer will then check whether the allocation you have chosen is feasible, that is, whether it corresponds to a point on the blue line.

After that, you will be asked to make an allocation in another decision. This process will be repeated until all the 15 decisions are completed.

Once you are ready, please start the task by clicking "Next".

Next

Part 3, Decision 1



Self:

Other:

Click "Next" to submit your answers.

Next

Please enter your decision in the two boxes. Then click anywhere on the screen to display the corresponding point on the line.

The Participant made a total of 15 allocation decisions on varying budget lines

Instructions to Part 4

In Part 4, you will be asked to solve 8 logical puzzles. You will be given 45 seconds to complete each puzzle. You will receive detailed instructions and see an example of a puzzle before the task begins.

Please note that each of the decisions you make in this part is independent and does not influence the future course of the study. After today's experiment, the computer program will randomly select one of the puzzles and pay you 50 ECU if it is solved correctly (and nothing if it is not solved correctly).

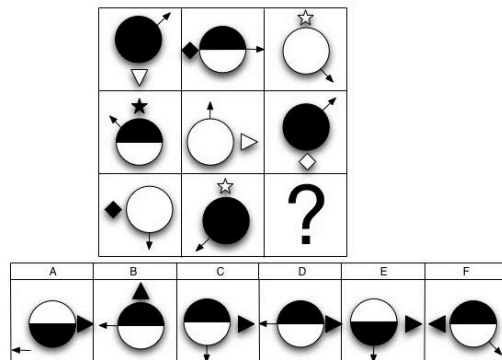
Please press the "Next" button now to see the instructions for Part 4.

Next

Part 4, Logical Puzzles

Below, you see an example of a logical puzzle. You will be asked to decide which of the pictures (A, B, C, D, E, or F) is the correct one to replace the question mark.

Please take 45 seconds now to practice solving this puzzle.



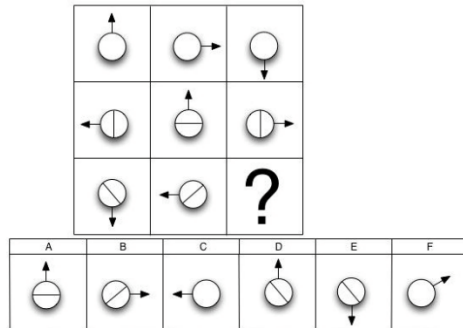
The correct answer to the example question above is option D. You will be able to select the letter using radio buttons (not pictured here) that will be below the puzzle. Once you select the letter from the options below the puzzle, please submit your choice by clicking "Next". **Please note that if you do not click the "Next" button before the 45 seconds run out, your answer will not be recorded.**

Once you are ready to start solving the puzzles, press the "Next" button. You will then have exactly 45 seconds for each puzzle.

Next

Part 4, Puzzle 1

Time left to complete this page: 0:41



Which of the pictures (A, B, C, D, E or F) is the correct one to replace the question mark?

☐ A ☐ B ☐ C ☐ D ☐ E ☐ F ☐ None of these ☐ I don't know

Next

***The Participant solved a total of
8 Puzzles like Puzzle 1***

Part 4, Questionnaire

Before Part 4 ends, we would like you to answer a few additional questions. You can earn additional money based on your answers. Please fill in your answers to the questions on the screen.

You confirm your entries by clicking the "Next" button. Once you have answered these questions, the experimental part of this study will be over and a survey will begin. This is a survey that you are asked to complete in addition to the online follow-up survey that you will complete 2-3 weeks from today.

1. How many of the 8 puzzles do you think you solved correctly? If your guess is correct, you receive 20 ECU.

Number of problems I believe I solved correctly:

2. We also ask you to assess your performance (how many puzzles you solved correctly) in comparison to the performance of 25 participants, selected randomly from all the participants who are taking the study now. Please enter your guess of your performance ranking in the box below. For this comparison, the computer will randomly select 25 other participants who are currently participating in the study. If, for example, you think that you solved the most problems correctly and that no other participant solved as many, you enter the rank of 1. If you think that 6 participants performed better than you, you enter the rank 7. If you think you solved the fewest puzzles correctly, and that all other participants solved more than you, you enter the rank 26. You submit your guessed rank by clicking the "Next" button.

Please enter your guess below.

My estimated rank:

Next

Questionnaire

You have now finished the part of the study in which you could earn additional money. You will see your earnings from today in an e-mail that will be sent to you once the time window for completing the online follow-up survey has closed.

The next part of the study consists of a questionnaire. You are asked to complete this short questionnaire now in addition to the online follow-up survey that will be sent to you in 2 weeks. Part of your participation payment of 40 CHF is for completing this questionnaire. We will provide you with detailed instructions before the questionnaire begins.

Please note that your answers to these questions are independent of the preceding part of the experiment and do not influence your payment for the study.

This is the final task for today. After you have completed these questions, you will be prompted to fill in a secure online form. This form is administered by the Decision Science Laboratory of the ETH and collects data from you so that the Lab can process your payment for today's study. The payment will be processed once the time window for completing the online follow-up survey has closed.

Please press the "Next" button to begin the questionnaire.

Next

Questionnaire

Please fill in your answers to the questions on the screen. Keep in mind that there are no right or wrong answers.

Please take your time to read through the questions carefully and consider which answer is right for you.

Please follow the instructions on the screen. You confirm your entries by clicking the "Arrow" button.



How old are you (in years)?

Are you male or female?

- ☐ Male
☐ Female

What is your nationality? You can select more than one.

- ☐ Swiss
☐ Other

If you selected the answer option "Other" in the previous question, please list all nationalities that you have.

What is your native language? You can select more than one.

- ☐ (Swiss) German
☐ English
☐ French
☐ Italian
☐ Other

If you selected the answer option "Other" in the previous question, please list your native language(s).

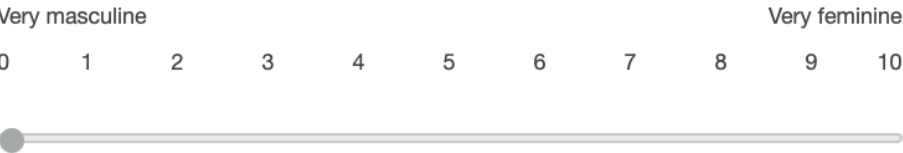


In the following, you will be asked to reply to a series of questions by positioning a slider on the screen. Note that you have to move the slider for your answer to be recorded, even if you want to move the slider back to its initial position.



In general, how do you see yourself? Where would you put yourself on this scale from “0-Very masculine” to “10-Very feminine”?

Please indicate your response below.



In general, how do you see yourself?

Please answer on both scales below from “0-Not at all” to “10-Very”.

Not at all

Very

012345678910

Feminine

Masculine



In general, how do **most people** see you? Where would **most people** put you on this scale from “0-Very masculine” to “11-Very feminine”?

Please indicate your response below.

Very masculine

Very feminine

01234567891011



In general, how do **most people** see you?

Please answer on both scales below from “0-Not at all” to “11-Very”.

Not at all

Very

01234567891011

Feminine



Masculine



What is your highest educational attainment?

- ☐ High school
 - ☐ Bachelor degree
 - ☐ Master degree
 - ☐ PhD degree
 - ☐ Other
-

At which of the following institutions have you completed most of your studies until now?

- ☐ UZH
 - ☐ ETZH or EPFL
 - ☐ University of Applied Sciences
 - ☐ Other
-

What is or was your major?

Which of the following categories best fits the major you reported above?

- ☐ Business and Economics
 - ☐ Engineering and Computer Science
 - ☐ Humanities and Other Social Sciences
 - ☐ Natural Sciences and Math
-

What is your height (in cm)?



What is the highest educational level your mother obtained?

- ☐ Secondary school or less
- ☐ Some additional education after secondary school but no university degree
- ☐ A bachelor's or master's degree
- ☐ An additional post-graduate degree (e.g., a law degree, PhD or medical degree)

What is your mother's occupation? If she is not working at the moment but worked previously, please indicate her most recent primary occupation. Please choose the most appropriate category from the list and write her occupation in the box below.

In which industry is your mother working or was her most recent primary occupation?

When you were in primary school, what percentage of a full workweek did your mother work on average?

- | | | | | | | | | | | | | |
|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| 0 % | 10 % | 20 % | 30 % | 40 % | 50 % | 60 % | 70 % | 80 % | 90 % | 100 % | can't
tell | not
applicable |
| <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> | <input type="radio"/> |



What is the highest educational level your father obtained?

- ☐ Secondary school or less
- ☐ Some additional education after secondary school but no university degree
- ☐ A bachelor's or master's degree
- ☐ An additional post-graduate degree (e.g., a law degree, PhD or medical degree)

What is your father's occupation? If he is not working at the moment but worked previously, please indicate his most recent primary occupation. Please choose the most appropriate category from the list and write his occupation in the box below.

In which industry is your father working or was his most recent primary occupation?

When you were in primary school, what percentage of a full workweek did your father work on average?

0 %	10 %	20 %	30 %	40 %	50 %	60 %	70 %	80 %	90 %	100 %	can't tell	not applicable
<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>



In what country were you born?

- ☐ Switzerland
☐ Other

If you answered the previous question with "Other", please enter the name of the country you were born in in this box.

In what country was your mother born?

- ☐ Switzerland
☐ Other

If you answered the previous question with "Other", please enter the name of the country your mother was born in in this box.

In what country was your father born?

- ☐ Switzerland
☐ Other

If you answered the previous question with "Other", please enter the name of the country your father was born in in this box.

What is your relationship status?



Comments

Were any of the questions in the survey unclear? If so, please specify which ones.



You have almost completed the survey.

Do you have any further comments for the researchers?



Please answer the following question truthfully. Remember that your responses are entirely anonymous.

- ☐ My answers should be treated as if they are reliable; I understood the questions and tried to answer as honestly as possible.
- ☐ My answers should not be treated as if they are reliable; I either did not understand many of the questions or did not take the survey very seriously.



Swiss Uni: Follow-Up Survey

WELCOME

Thank you for participating in our follow-up survey. The survey comprises three main parts. We will provide you with detailed instructions before each part.

Part of your compensation in the participation payment is for carefully completing this survey. You will receive a combined payment of 40 CHF for your completion of this survey and participation in the experiment via bank transfer.

Please note that if you do not complete this survey fully, you will not receive your combined show-up fee and survey completion payment of 40 CHF.

All your answers will be anonymized and stored confidentially. All data collected will be analyzed in anonymous, aggregate form. You have already provided details on where to transfer your earnings through an online form that was administered at the end of the online experiment in which you participated a few weeks ago.

The ETH Decision Science Laboratory will be able to link the total amount that you have earned with the information on your bank account that you provided at the end of the online experiment and information on whether or not you have completed this follow-up survey. The ETH Decision Science Laboratory can make this link based on an anonymous token that was generated by a computer when you participated in the online experiment. The researchers who analyze the data do not have access to the information on your bank account. The ETH Decision Science Laboratory will destroy this information once the transfer has been made.

Please keep in mind that there are no right or wrong answers and that the results will be fully anonymous. Please read each question carefully and do not talk to anyone else while completing the survey. It is important that you complete the study alone and in one sitting (and always use the same computer and web browser). Also, please do not use any (other) websites during the study. Note that you cannot return to an earlier page after you have moved on. Please do not use your browser navigation button to go back.

Once you are ready, click the “Arrow” button on the bottom right of this page to start the survey.

Instructions to Part 1

In Part 1, you will be asked to evaluate a number of characteristics and statements in terms of how well they apply to you. Keep in mind that there are no right or wrong answers. Please take your time to read through the questions carefully and consider which answer is right for you.

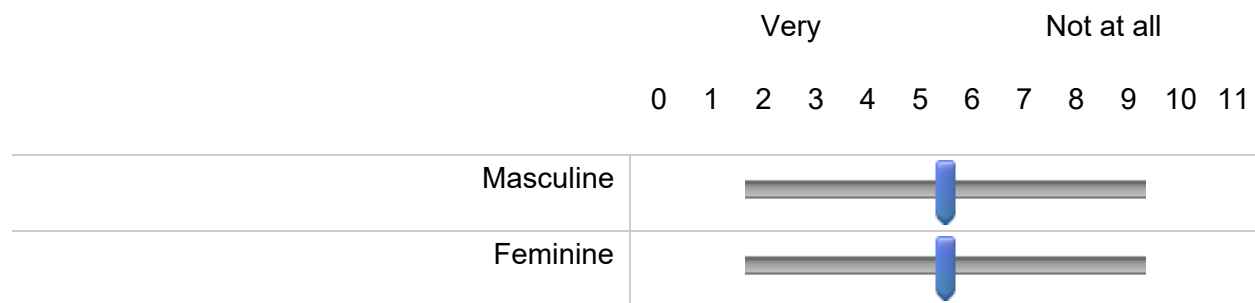
You answer some of the questions by positioning a slider on the screen. Note that you have to move the slider for your answer to be recorded, even if you want to move the slider back to its initial position.

Please click the “Arrow” button now to begin Part 1.

In general, how do you see yourself? Where would you put yourself on this scale from “0-Very feminine” to “11-Very masculine”? Please indicate your response below.

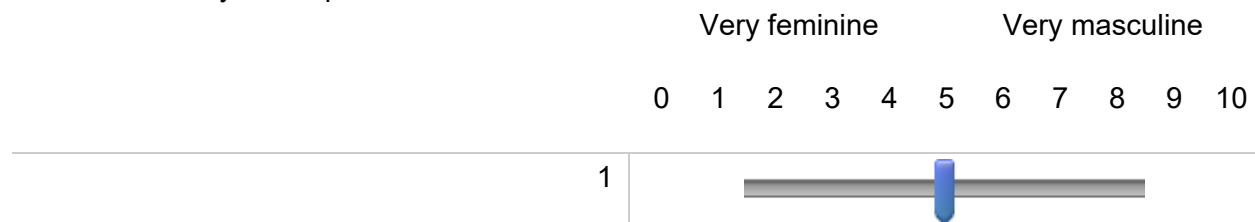


In general, how do you see yourself? Please answer on both scales below from “0-Very” to “11-Not at all”.





In general, how do **most people** see you? Where would **most people** put you on this scale from “0-Very feminine” to “10-Very masculine”?

Please indicate your response below.



In general, how do **most people** see you?

Please answer on both scales below from “0-Very ” to “10-Not at all”.

		Very	Not at all										
		0	1	2	3	4	5	6	7	8	9	10	
	Masculine												
	Feminine												

How important is each of the following to you in selecting a job or career?

You can rank each statement by dragging it to change the order in which it appears. If you place it at the top of the list (rank 1), this means it is the most important one to you in selecting a job or career, while if you place it at the bottom (rank 4) it is the least important. The order in which you place the statements determines your ranking.

Please rank the following statements in order from most (1) to least (4) important in selecting a job or career.

- _____ Making a lot of money
- _____ Opportunities to be helpful to others or useful to society
- _____ The chance to be a leader
- _____ Opportunities to work with people rather than things

How important is each of the following to you in your life?

You can rank each statement by dragging it to change the order in which it appears. If you place it at the top of the list (rank 1), this means it is the most important one to you in your life, while if you place it at the bottom (rank 5) it is the least important. The order in which you place the statements determines your ranking.

- _____ Being successful in work
- _____ Having lots of money
- _____ Being able to give children better opportunities
- _____ Living close to parents and relatives
- _____ Helping other people in my community

Please imagine your work situation at age 40.

Likert scale: 0-“Very unlikely“ to 10-“Very likely“



How likely is it that you will work in the public sector?

How likely is it that you will work in the private sector?

How likely is it that you will work in the non-profit sector?

How likely is it that you will be self-employed?

Please imagine your work situation at age 40.

What workload would you ideally like to have? 0-100%	
What workload do you realistically think you that you will have? 0-100%	

Please imagine your family situation at age 40.

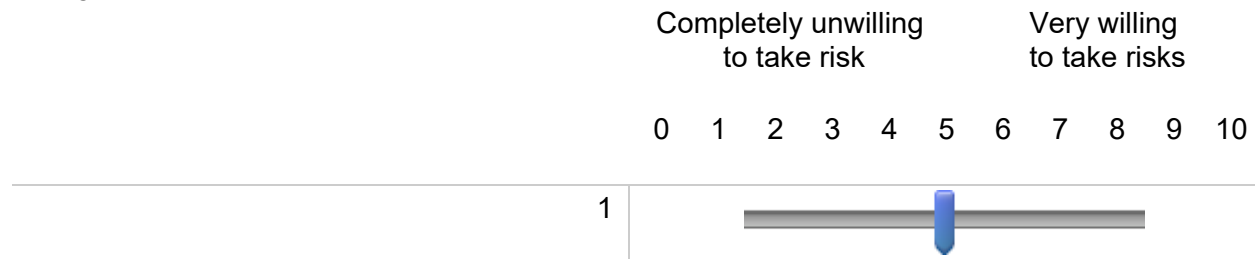
How many children would you ideally want to have?	▼ 0 ... 5+
By age 40, how many children do you realistically think you will have?	▼ 0 ... 5+

Instructions to Part 2

In Part 2, you will be asked to answer some questions about yourself, make some hypothetical decisions and give some estimates. You will receive detailed information for each question.

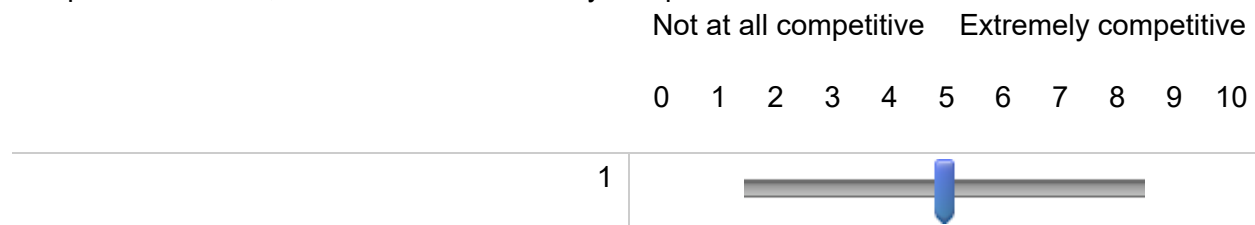
Please click the “Arrow” button now to begin Part 2.

Please state, **in general, how willing or unwilling you are to take risks**, using a scale from 0 to 10, where 0 means you are “completely unwilling to take risks” and 10 means you are “very willing to take risks”.



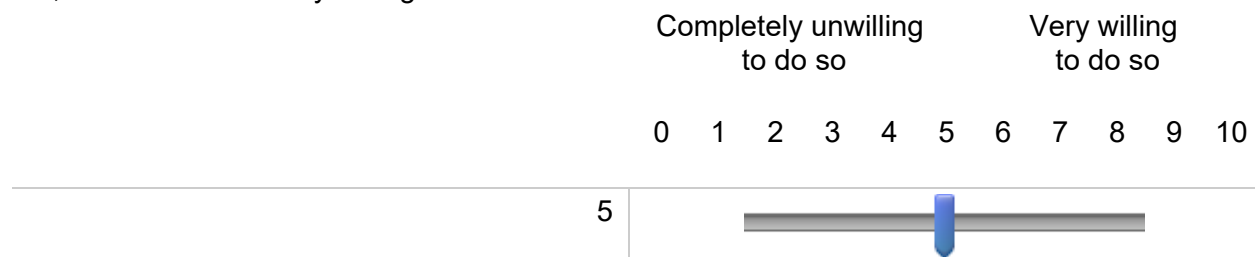
How competitive do you consider yourself to be?

You can express this using the scale below, which runs from 0 – meaning, "not at all competitive" – to 10, which means "extremely competitive".



How willing are you to give to good causes without expecting anything in return?

Please indicate your answer on a scale from 0 to 10, where 0 means “completely unwilling to do so”, and 10 means “very willing to do so”.



Please imagine the following situation: You can choose between a sure payment of a particular amount of money, OR a draw, where you would have an equal chance of getting 750 CHF or getting nothing. We will present to you five different situations.

What would you prefer:

- ☐ A draw with a 50-percent chance of receiving 750 CHF and the same 50-percent chance of receiving nothing,
- ☐ OR the amount of 400 CHF as a sure payment?

Sequence of questions depending on respondent's answers to determine a certainty equivalent.

Below are different qualities a person can have. You will probably find that some apply to you perfectly and that some do not apply to you at all. With others, you may be somewhere in between.

Please answer according to the following scale: 0 means "Does not apply to me at all"; 6 means "Applies to me perfectly".

I see myself as someone who...

- ...does a thorough job
- ...is communicative, talkative
- ...is sometimes somewhat rude to others
- ...is original, comes up with new ideas
- ...worries a lot
- ...has a forgiving nature
- ...tends to be lazy
- ...is outgoing, sociable
- ...values artistic, aesthetic experiences
- ...gets nervous easily
- ...does things effectively and efficiently
- ...is reserved
- ...is considerate and kind to others
- ...has an active imagination
- ...is relaxed, handles stress well

You will now be shown a picture of a jar of jellybeans. Please give us your best guess as to the number of jellybeans in the jar.

Please enter the number of jellybeans you think are in this jar (between 1 and 1000).
My estimate:

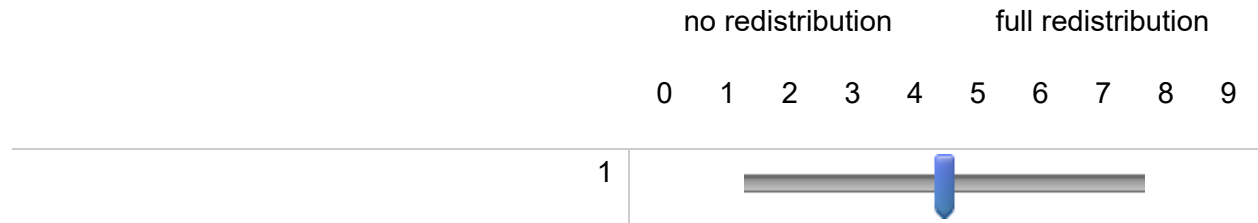
How confident are you of your answer to this question?

- ☐ Not confident at all
- ☐ Not very confident
- ☐ Somewhat unconfident
- ☐ Somewhat confident
- ☐ Very confident
- ☐ Certain

How much economic redistribution do you want in society?

“No redistribution” means the state not influencing the income distribution at all, and “full redistribution” means everyone earning the same amount after taxes and subsidies.

Please indicate your response on the scale below from 0 to 9, where 0 means you want “no redistribution” and 9 means you want “full redistribution”.



Please state your agreement or disagreement with each of the following statements, using a scale from “Strongly disagree” to “Strongly agree”.

Likert scale: 1 “Strongly disagree” -5 “Strongly agree”

I can always manage to solve difficult problems if I try hard enough.

If someone opposes me, I can find the means and ways to get what I want.

It is easy for me to stick to my aims and accomplish my goals.

I am confident that I could deal efficiently with unexpected events.

Thanks to my resourcefulness, I know how to handle unforeseen situations.

I can solve most problems if I invest the necessary effort.

I can remain calm when facing difficulties because I can rely on my coping abilities.

When I am confronted with a problem, I can usually find several solutions.

If I am in trouble, I can usually think of a solution.

I can usually handle whatever comes my way.

Next, you will be asked to answer 10 questions from a Vocational Aptitude Battery (ASVAB) that has been developed for the Armed Services in the U.S.

Each question will test your aptitude in one of the five categories: Arithmetic Reasoning, Math Knowledge, Mechanical Comprehension, and Assembling Objects. In addition to being used by the U.S. military to determine which jobs armed service members are qualified for, performance on the ASVAB is often used as a measure of cognitive ability by academic researchers.

You will be presented with each of the 10 questions on separate pages. You will be given up to 30 seconds to answer each question, although you may push the arrow at the bottom of the page to answer a question before the 30 seconds are up.

Please try to answer each question as best as you can.

Question 1 out of 10

ARITHEMTICE REASONING: Three people go to a restaurant. Their bill comes to \$78.00. They decide to split the cost. One person pays \$4.50; the next person pays 3 times that amount. How much will the third person have to pay?

- ☐ 60 \$
- ☐ 66 \$
- ☐ 70.50 \$
- ☐ 73.50 \$

Question 2 out of 10

MATH KNOWLEDGE: A cafeteria requires 3 workers for every 24 students. How many workers will be needed for a school with 136 students?

- ☐ 24
- ☐ 13
- ☐ 8
- ☐ 17

Question 3 out of 10

GENERAL SCIENCE: Waves bend when they move around obstructions or pass through narrow openings. This type of bending is called ____.

- ☐ Diffraction
- ☐ Reflection
- ☐ Refraction
- ☐ Intererence

Question 4 out of 10

ASSEMBLING OBJECTS: Choose the figure that shows the shapes assembled into an object.

- ☐ A
- ☐ D
- ☐ B
- ☐ C

Question 5 out of 10

MECHANICAL COMPREHENSION: Which of the following is the basis for Pascal's law?

- ☐ The amplification of force in a hydraulic system.
- ☐ The definition of pressure as pounds per square inch.
- ☐ The manner in which liquids conform to their container.
- ☐ The relationship between force and volume of a liquid.

Question 6 out of 10

ARITHMETIC REASONING: Which of the following is not a factor of 90?

- ☐ 5
- ☐ 6
- ☐ 15
- ☐ 12

Question 7 out of 10

GENERAL SCIENCE: Which layer of the atmosphere is closest to the earth?

- ☐ Stratosphere
- ☐ Mesosphere
- ☐ Troposphere
- ☐ Thermosphere

Question 8 out of 10

MECHANICAL COMPREHENSION: Tension can best be called ____

- ☐ A movement
- ☐ A force
- ☐ A direction
- ☐ A weight

Question 9 out of 10

ASSEMBLING OBJECTS: Choose the figure that shows the shapes assembled into an object.

- ☐ A
- ☐ B
- ☐ C
- ☐ D

Question 10 out of 10

MATH KNOWLEDGE: A classroom has 15 boys and 13 girls. If 10 more girls join the class, what is the ratio of girls to boys?

- ☐ 15:23
- ☐ 13:15
- ☐ 10:15
- ☐ 23:15

On the test, you answered **X** questions correctly out of 10 questions.

To confirm that you have read the prior sentence, please answer the following question.

Out of the 10 questions on the test, how many questions did you answer correctly?

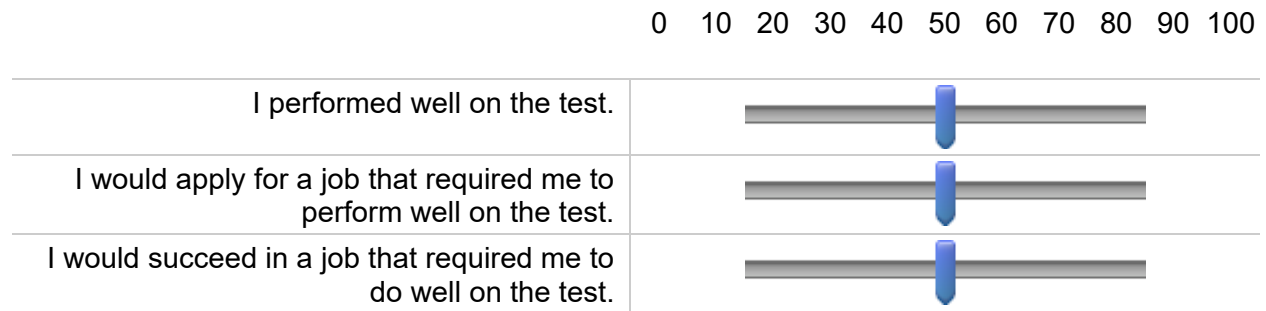
Please answer the following questions.

Please describe how well you think you performed on the test and why.

Please indicate how well you think you performed on the test.

- ☐ terrible
- ☐ very poor
- ☐ neutral
- ☐ good
- ☐ very good
- ☐ excellent

On a scale from 0 (entirely disagree) to 100 (entirely agree), please indicate the extent to which you agree with the following statement:



Instructions to Part 3

In Part 3, you will be asked to answer some more questions.

You will receive detailed information for each question. Please click the “Arrow” button now to begin Part 3.

What was your parents' marital status when you were born?

- ☐ Married
- ☐ Unmarried

What was your family structure at age 12?

- ☐ I lived with both biological parents
- ☐ I lived with one biological parent and a step-parent
- ☐ I lived with a single parent
- ☐ Other

How many children were in your household when you were growing up (e.g. including yourself, and any step, adopted, or foster children)?

How many *biological* siblings do you have (i.e. you share at least one parent)?

For example, if you have a brother and a sister (excluding you), please enter “2”.

This question differed depending on the number of siblings a participant indicated to have.

What is the birth order of you and your biological sibling?

What is your and your sibling's current age?

Please enter the information below. For example, if you have one sister and you are the oldest child, then tick the "Me" box for first-born, "Sister" for second-born. Please enter your and your sibling's current age below the labels "First-born" and "Second-born".

	Me	Brother	Sister
First-born (current age)	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>
Second-born (current age)	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>

Who was your primary caregiver as a child?

- ☐ Mother
- ☐ Father
- ☐ Both parents equally
- ☐ Grandparent(s)
- ☐ Sibling(s)
- ☐ Other/Non-relative

Think back to your childhood when you lived with both your parents.

Who did the following things...?

Please tick one box on each line.

Likert scale: 1 "Always my mother" – 7 "Always my father" | option 8 "I grew up in a single parent household"

- ... Did the laundry
- ... Made small repairs around the house
- ... Cared for sick family members
- ... Shopped for groceries
- ... Did the household cleaning
- ... Prepared the meals
- ... Brought me (and my siblings) to school or spare time activities

In general, how do you see your mother? Where would you put her on this scale from "Very masculine" to "Very feminine"? Please indicate your response below.



In general, how do you see your father? Where would you put him on this scale from "Very masculine" to "Very feminine"? Please indicate your response below.



Please state your agreement or disagreement with each of the following statements, using a scale from “Strongly disagree” to “Strongly agree”.

Likert scale: 1 “Strongly disagree” – 5 “Strongly agree”

A woman has to have children in order to be fulfilled.

A working mother can establish just as warm and secure a relationship with her children as a mother who does not work.

Being a housewife is just as fulfilling as working for pay.

Both husband and wife should contribute to household income.

A job is all right but what most women really want is a home and children.

Having a job is the best way for a woman to be an independent person.

When jobs are scarce, men have more right to a job than women.

Whats was your assigned sex at birth? (For example on your birth certificate)

- ☐ Male
- ☐ Female
- ☐ Intersex

Whats is your current gender?

- ☐ Woman
- ☐ Man
- ☐ Transgender
- ☐ Non-binary
- ☐ Prefer not to answer
- ☐ Other

What is your sexual orientation?

- ☐ Straight (heterosexual)
- ☐ Gay (homosexual)
- ☐ Lesbian (homosexual)
- ☐ Bisexual
- ☐ Prefer not to answer
- ☐ Other

Below are some questions about your gender identity. If you, for example, identify as a woman, you respond in relation to your experience as a woman.

For each of the following statements, please indicate on the scale how much you agree with it.

Likert scale: 1 “Strongly disagree” – 5 “Strongly agree”

My gender identity has very little to do with how I feel about myself.

My gender identity is an important part of my self-image.

My gender identity is an important reflection of who I am.

My gender identity has no significance for my sense of what kind of person I am.

Please list 3 characteristics that you associate with being “very masculine”.

- ☐ Characteristic 1 _____
- ☐ Characteristic 2 _____
- ☐ Characteristic 3 _____

Please list 3 characteristics that you associate with being “very feminine”.

- ☐ Characteristic 1 _____
- ☐ Characteristic 2 _____
- ☐ Characteristic 3 _____

Comments

Were any of the questions in the survey unclear? If so, please specify which ones.

You have almost completed the follow-up survey.

Do you have any further comments for the researchers?

Please answer the following question truthfully. Remember that your responses are entirely anonymous.

- ☐ My answers should be treated as if they are reliable; I understood the questions and tried to answer as honestly as possible.
- ☐ My answers should not be treated as if they are reliable; I either did not understand many of the questions or did not take the survey very seriously.

U.S. Adults: Wave 1

Welcome to this Research Study

Overview of the study: You are participating in a research study. It consists of a survey and a final decision-making task. Your bonus payment depends on your decision in this task. Therefore, it is important that you read all descriptions carefully. Your answers and choices are *anonymous*.

Payment: You will receive a participation payment of 1.25 USD. In addition, you can earn an additional bonus payment of up to 2.5 USD. You must complete all parts of the study to receive payment, which will be administered within the next two working days.

General Rules of Conduct: Completing this study will take around 10 minutes. We ask for your full attention during the study. Please find a quiet space to complete the study. During the study, please do not use other devices, have conversations with other people, use social media, etc. Please remain solely in this browser tab for the entire time of the study.

Verification: During the survey, you will encounter two attention screening questions. If you fail both, you will not be paid.

Consent: I have read and understood the information above. I agree to comply with the rules of conduct stated above and choose to participate in this study.

☐ I agree

☐ I do not agree

What is your **Prolific ID**? *Please note that this response should auto-fill with the correct ID*

Start of Block: Demographics and Education

Please **take your time** to read through each of the questions carefully and consider which answer is **right** for you.

What is your **age**?

▼ 30 (1) ... 60 (31)

In which **state** do you currently reside?

▼ Alabama (1) ... Wyoming (52)

What is your **ZIP code**?

What is your **sex**?

☐ Male

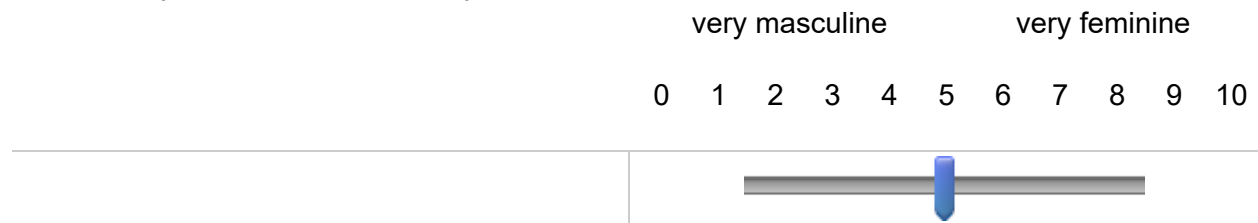
☐ Female

In the following, you will reply to a series of questions by positioning a slider on the screen. Note that you have to **move the slider** for your answer to be recorded, even if you want to move the slider back to its initial position.

In general, how do **you** see yourself? Where would **you** put yourself on this scale from “0-Very masculine” to “10-Very feminine”?



In general, how do **most people** see you? Where would **most people** put you on this scale from “0-Very masculine” to “10-Very feminine”?



Which of the following best describes your **race** or **ethnicity**?

- ☐ American Indian or Alaska Native
- ☐ Asian
- ☐ Black or African American
- ☐ Native Hawaiian or Other Pacific Islander
- ☐ White
- ☐ Hispanic or Latino or Spanish Origin
- ☐ Other

What is the **highest degree** or **level of education** you have completed? *(If currently enrolled, mark the previous grade or highest degree already received)*

- ☐ No schooling completed
- ☐ Regular high school diploma
- ☐ GED or alternative credential
- ☐ Some college credit (< 1 year)
- ☐ 1 or more years of college credit, no degree
- ☐ Associate's degree (for example AA, AS)
- ☐ Bachelor's degree (for example BA, BS)
- ☐ Master's degree (for example MA, MS, MEng, MEd, MSW, MBA)
- ☐ Professional degree beyond a bachelor's degree (for example MD, DDS, DVM, LLB, JD)
- ☐ Doctorate degree (for example PhD, EdD)

Display This Question:

If education_level = Bachelor's degree (for example BA, BS)

Or education_level = Master's degree (for example MA, MS, MEng, MEd, MSW, MBA)

Or education_level = Professional degree beyond a bachelor's degree (for example MD, DDS, DVM, LLB, JD)

Or education_level = Doctorate degree (for example PhD, EdD)

Which of the following best describes your **field of study**?

- ☐ Education
- ☐ Arts and Humanities
- ☐ Social Sciences, Journalism, and Information
- ☐ Business, Administration and Law
- ☐ Biological and Related Science / Environmental Science
- ☐ Physical Sciences / Mathematics and Statistics
- ☐ Information and Communication Technologies
- ☐ Engineering, Manufacturing and Construction
- ☐ Agriculture, Forestry, Fisheries and Veterinary
- ☐ Health and Welfare
- ☐ Other

Attention Screening: You have to select the number **17** from the following dropdown list.

▼ 1 (1) ... 20 (20)

End of Block: Demographics and Education

Start of Block: Labor Market and Economic Outcomes

Which of these **categories** describes your current situation?

- ☐ Employed by another individual or entity
- ☐ Self-employed
- ☐ Relative assisting on family farm or business
- ☐ Unemployed
- ☐ Unable to work due to long-term illness or disability
- ☐ Currently on child-care leave or other leave
- ☐ Retired
- ☐ Full time homemaker
- ☐ In full time education (at school, university, etc.)
- ☐ Other

Display This Question:

If employment_status = Unemployed

Or employment_status = Unable to work due to long-term illness or disability

Or employment_status = Currently on child-care leave or other leave

Or employment_status = Retired

Or employment_status = Full time homemaker

Or employment_status = In full time education (at school, university, etc.)

Or employment_status = Other

During the last **4 weeks**, have you been **actively** looking for work?

- ☐ Yes
- ☐ No

During the past **12 months**, how many **weeks** did you work? *Include weeks when you only worked for a few hours or took paid leave, for example, vacations.*

- ☐ 0 – 13 weeks (0 - 3 months)
- ☐ 14 - 26 weeks (4 - 6 months)
- ☐ 27 – 39 weeks (7 - 9 months)
- ☐ 40 – 52 weeks (10 - 12 months)

During the past **12 months**, in the weeks you worked, how many **hours** did you usually work **each week**?

- ☐ 0 – 10 hours
- ☐ 11 – 20 hours
- ☐ 21 – 30 hours
- ☐ 31 – 40 hours
- ☐ 41 – 50 hours
- ☐ More than 50 hours

Which of the following **sectors** best describes your current employment?

- ☐ Private sector employee
- ☐ Government sector employee

Which of the following **industries** best describes your current employment?

- ☐ Agriculture, Forestry and Fishing
- ☐ Mining and Quarrying, and Oil and Gas Extraction
- ☐ Construction
- ☐ Manufacturing
- ☐ Wholesale and Retail Trade
- ☐ Transportation and Utilities
- ☐ Information
- ☐ Financial Activities
- ☐ Professional and Business Services
- ☐ Education and Health Services
- ☐ Leisure and Hospitality
- ☐ Other Services
- ☐ Public Administration

What was your **total income** over the past 12 months? *Income includes wages, salary, commissions, bonuses, tips, interest and dividend payments, as well as self-employment income.*

- ☐ \$0 - \$5,000
- ☐ \$5,001 - \$10,000
- ☐ \$10,001 - \$25,000
- ☐ \$25,001 - \$50,000
- ☐ \$50,001 - \$100,000
- ☐ \$100,001 - \$250,000
- ☐ More than \$250,000

Do you **actively trade** in securities like stocks, cryptocurrencies, etc.?

- ☐ Yes
- ☐ No

End of Block: Labor Market and Economic Outcomes

Start of Block: Job Market Characteristics

Do you have **managerial responsibilities** at your current employment? *By managerial responsibilities we mean things like directly supervising others, the ability to hire or terminate other employees, etc.*

- ☐ Yes
- ☐ No

How are your **working time arrangements** set at your current employment?

- ☐ They are set by the company/organization with no possibility for changes
- ☐ I can choose between several fixed working schedules determined by the company/organization
- ☐ I can adapt my working hours within certain limits (e.g., flextime)
- ☐ My working hours are entirely or almost entirely determined by myself

Do **changes** to your working time arrangements occur regularly at your current employment? If yes, how long in **advance** are you informed about these changes?

- ☐ No
- ☐ Yes, the same day
- ☐ Yes, the day before
- ☐ Yes, several weeks in advance

Would you say that, for you, taking an hour or two off during working hours to take care of **personal** or **family matters** is ...

- ☐ Very easy
- ☐ Fairly easy
- ☐ Fairly difficult
- ☐ Very difficult

Does your **income** at your current employment include **performance-related pay** or **bonuses**?
If yes, how much of your total income does this amount to?

- ☐ No
- ☐ Yes, less than 1/3 of my total income
- ☐ Yes, more than 1/3 but less than 2/3 of my total income
- ☐ Yes, more than 2/3 of my total income

Have you ever tried to **negotiate** a higher wage or greater compensation in an employment relationship?

- ☐ No
- ☐ Yes, but only once
- ☐ Yes, more than once

Attention Screening: You have to select the number **14** from the following dropdown list.

▼ 1 (1) ... 20 (20)

End of Block: Job Market Characteristics

Start of Block: Family

What is your **relationship status**?

- ☐ Never married: single
- ☐ Never married: in a relationship
- ☐ Married
- ☐ Separated
- ☐ Divorced
- ☐ Widowed

Do you consider **yourself** to be ...

- ☐ Heterosexual or Straight
- ☐ Gay or Lesbian
- ☐ Bisexual
- ☐ Other
- ☐ Prefer not to answer

How many **children** do you have?

- ☐ None
- ☐ 1
- ☐ 2
- ☐ 3
- ☐ 4
- ☐ 5 or more

Think about your current or most recent experience of **living with someone** (e.g., spouse, partner, roommates). What **best** describes the typical **division of housework** in this situation?

Likert scale: 1 “Always me”, 2 “Usually me”, 3 “About equal or both together”, 4 “Usually the other person(s)”, 5 “Always the other person(s)”, 6 “Usually done by an outsider or not done at all”

Does the laundry

Makes small repairs around the house

Shops for groceries

Does the household cleaning


Prepares the meals

Start of Block: Preferences

Imagine the following situation: Suppose you unexpectedly received **\$2400** How much of this amount would you donate to a **good cause**? (*Values between 0 and 2400 are allowed*)

In the following, you will again be asked to reply to a series of questions by positioning a slider on the screen. Note that you have to **move the slider** for your answer to be recorded, even if you want to move the slider back to its initial position.

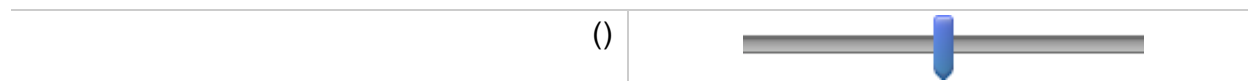
How **willing** or **unwilling** are you to take **risks**, in general? *Please indicate your response on this scale from 0-completely unwilling to take risks to 10-completely willing to take risks.*

	Completely unwilling to take risks		Very willing to take risks								
	0	1	2	3	4	5	6	7	8	9	10
()											

How **competitive** do you consider yourself to be? *Please indicate your response on this scale from 0-not at all competitive to 10-extremely competitive.*

Not at all competitive Extremely competitive

0 1 2 3 4 5 6 7 8 9 10

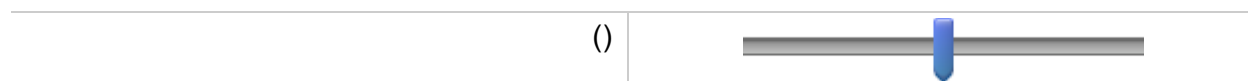


How much **economic redistribution** do you want in society? *“No redistribution” means the state does not influence the income distribution at all, and “full redistribution” means everyone earns the same amount after taxes and subsidies.*

No redistribution

Full redistribution

0 1 2 3 4 5 6 7 8 9 10



Relative to men in general, how do you see yourself? Where would you put yourself on this scale from "0-Very masculine" to "10-Very feminine"?



Relative to women in general, how do you see yourself? Where would you put yourself on this scale from "0-Very masculine" to "10-Very feminine"?



Investment Decision

In this part of the survey, you can earn an **additional bonus payment**. This is your final task for today.

We ask you to make an investment decision. For this, we endow you with an additional **100 cents = \$1 USD**. You choose how much of this amount (from 0 to 100 cents) you wish to invest.

The amount that you choose not to invest will be yours to keep and cannot be lost.

The amount that you chose to invest will either increase or decrease in value. This depends on the success or failure of the investment. The success or failure will be determined by a computerized coin flip after you have completed this survey.

If the virtual coin comes up tails, the **investment fails**, and you **lose** the amount invested.

If the virtual coin comes up heads, the **investment succeeds**, and you receive **2.5 times** the amount invested.

Summary bonus payment: Your additional bonus payment will equal **100 cents**, minus the amount that you invest, plus **2.5times** the amount that you invest if the investment succeeds.

Your Investment Decision:

Please move the slider to the **amount** that you wish to **invest**.

Cents

0 10 20 30 40 50 60 70 80 90 100

()

Thank you for participating in our study. You have reached the end.

You will receive **1.25 USD** for your participation in the study and an additional bonus of up to **2.5 USD** depending on your investment decision. You will receive the payment within two working days after completing this study. The payment will be processed by Prolific.

Please click the button below to be redirected back to Prolific and register your submission.

U.S. Adults: Wave 2

Welcome to this Research Study

Overview of the study: This study consists of a survey and a decision making task. Your answers and choices are *anonymous*.

Payment: You will receive a participation payment of \$2.5. In addition, you can earn an additional bonus payment of up to \$2.5 based on your decision in the task. Therefore, it is important that you read all descriptions carefully. You must complete all parts of the study to receive payment, which will be administered within the next three working days.

General Rules of Conduct: Completing this study takes around 15 minutes. We ask for your full attention during the study. Please find a quiet space to complete the study. During the study, please do not use other devices, have conversations with other people, use social media, etc. Please remain solely in this browser tab for the entire time of the study.

Verification: During the survey, you will encounter two attention screening questions. If you fail both, you will not be paid.

Consent: I have read and understood the information above. I agree to comply with the rules of conduct stated above and choose to participate in this study.

☐ I agree

☐ I do not agree

What is your **Prolific ID**? *Please note that this response should auto-fill with the correct ID*

Start of Block: Demographics and Education

Please answer the following questions about you.

What is your **age**?

▼ 30 ... 60

In which **state** do you currently reside?

▼ Alabama ... Wyoming

What is your **ZIP code**?

What is your **sex**?

- ☐ Male
- ☐ Female

Which of the following best describes your **race** or **ethnicity**?

- ☐ American Indian or Alaska Native
- ☐ Asian
- ☐ Black or African American
- ☐ Native Hawaiian or Other Pacific Islander
- ☐ White
- ☐ Hispanic or Latino or Spanish Origin
- ☐ Other

What is the **highest degree or level of education** you have completed? *(If currently enrolled, mark the previous grade or highest degree already received)*

- ☐ No schooling completed
- ☐ Regular high school diploma
- ☐ GED or alternative credential
- ☐ Some college credit (< 1 year)
- ☐ 1 or more years of college credit, no degree
- ☐ Associate's degree (for example AA, AS)
- ☐ Bachelor's degree (for example BA, BS)
- ☐ Master's degree (for example MA, MS, MEng, MEd, MSW, MBA)
- ☐ Professional degree beyond a bachelor's degree (for example MD, DDS, DVM, LLB, JD)
- ☐ Doctorate degree (for example PhD, EdD)

Display This Question:

If What is the highest degree or level of education you have completed? (If currently enrolled, mark... = Bachelor's degree (for example BA, BS)

Or What is the highest degree or level of education you have completed? (If currently enrolled, mark... = Master's degree (for example MA, MS, MEng, MEd, MSW, MBA)

Or What is the highest degree or level of education you have completed? (If currently enrolled, mark... = Professional degree beyond a bachelor's degree (for example MD, DDS, DVM, LLB, JD)

Or What is the highest degree or level of education you have completed? (If currently enrolled, mark... = Doctorate degree (for example PhD, EdD)

Which of the following best describes your **field of study**?

- ☐ Education
- ☐ Arts and Humanities
- ☐ Social Sciences, Journalism, and Information
- ☐ Business, Administration and Law
- ☐ Biological and Related Science / Environmental Science
- ☐ Physical Sciences / Mathematics and Statistics
- ☐ Information and Communication Technologies
- ☐ Engineering, Manufacturing and Construction
- ☐ Agriculture, Forestry, Fisheries and Veterinary
- ☐ Health and Welfare
- ☐ Other, please specify: _____

End of Block: Demographics and Education

Start of Block: Labor Market and Economic Outcomes

Which of these **categories** describes your current situation?

- ☐ Employed by another individual or entity
- ☐ Self-employed
- ☐ Relative assisting on family farm or business
- ☐ Unemployed
- ☐ Unable to work due to long-term illness or disability
- ☐ Currently on child-care leave or other leave
- ☐ Retired
- ☐ Full time homemaker
- ☐ In full time education (at school, university, etc.)
- ☐ Other

Display This Question:

If Which of these categories describes your current situation? = Unemployed

Or Which of these categories describes your current situation? = Currently on child-care leave or other leave

Or Which of these categories describes your current situation? = Full time homemaker

Or Which of these categories describes your current situation? = Retired

Or Which of these categories describes your current situation? = Unable to work due to long-term illness or disability

Or Which of these categories describes your current situation? = In full time education (at school, university, etc.)

Or Which of these categories describes your current situation? = Other

During the last **4 weeks**, have you been **actively** looking for work?

- ☐ Yes
- ☐ No

During the past **12 months**, how many **weeks** did you work? *Include weeks when you only worked for a few hours or took paid leave, for example, vacations.*

- ☐ 0 – 13 weeks (0 - 3 months)
- ☐ 14 - 26 weeks (4 - 6 months)
- ☐ 27 – 39 weeks (7 - 9 months)
- ☐ 40 – 52 weeks (10 - 12 months)

During the past **12 months**, in the weeks you worked, how many **hours** did you usually work **each week**?

- ☐ 0 – 10 hours
- ☐ 11 – 20 hours
- ☐ 21 – 30 hours
- ☐ 31 – 40 hours
- ☐ 41 – 50 hours
- ☐ More than 50 hours

What was your **total income** over the past 12 months? *Income includes wages, salary, commissions, bonuses, tips, interest and dividend payments, as well as self-employment income.*

- ☐ \$0 - \$5,000
- ☐ \$5,001 - \$10,000
- ☐ \$10,001 - \$25,000
- ☐ \$25,001 - \$50,000
- ☐ \$50,001 - \$100,000
- ☐ \$100,001 - \$250,000
- ☐ More than \$250,000

Do you **actively trade** in securities like stocks, cryptocurrencies, etc.?

- ☐ Yes
- ☐ No

Do you have **managerial responsibilities** at your current employment? *By managerial responsibilities we mean things like directly supervising others, the ability to hire or terminate other employees, etc.*

- ☐ Yes
- ☐ No

Have you ever tried to **negotiate** a higher wage or greater compensation in an employment relationship?

- ☐ No
- ☐ Yes, but only once
- ☐ Yes, more than once

You have to select the number **17** from the following dropdown list.

▼ 1 ... 20

End of Block: Labor Market and Economic Outcomes

Start of Block: Family

What is your **relationship status**?

- ☐ Never married: single
- ☐ Never married: in a relationship
- ☐ Married
- ☐ Separated
- ☐ Divorced
- ☐ Widowed

Do you consider **yourself** to be ...

- ☐ Heterosexual or Straight
- ☐ Gay or Lesbian
- ☐ Bisexual
- ☐ Other
- ☐ Prefer not to answer

How many **children** do you have?

- ☐ None
- ☐ 1
- ☐ 2
- ☐ 3
- ☐ 4
- ☐ 5 or more

End of Block: Family

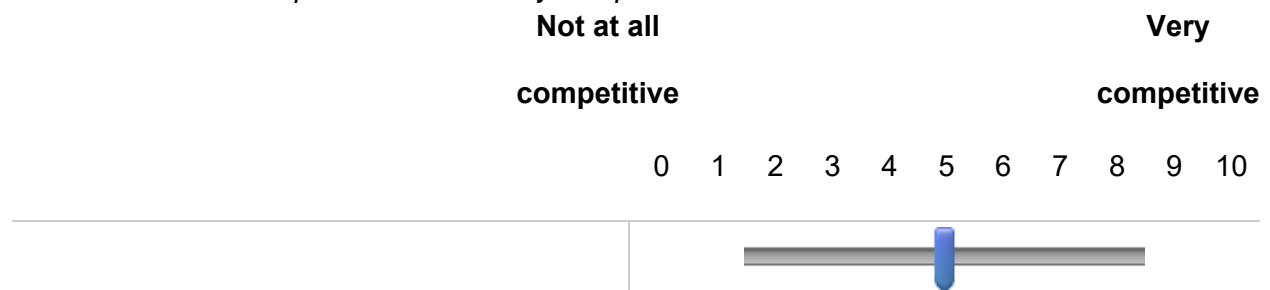
Start of Block: Preferences

In the following, you will be asked to reply to questions by positioning a slider on the screen. Note that you have to **move the slider** for your answer to be recorded, even if you want to move the slider back to its initial position.

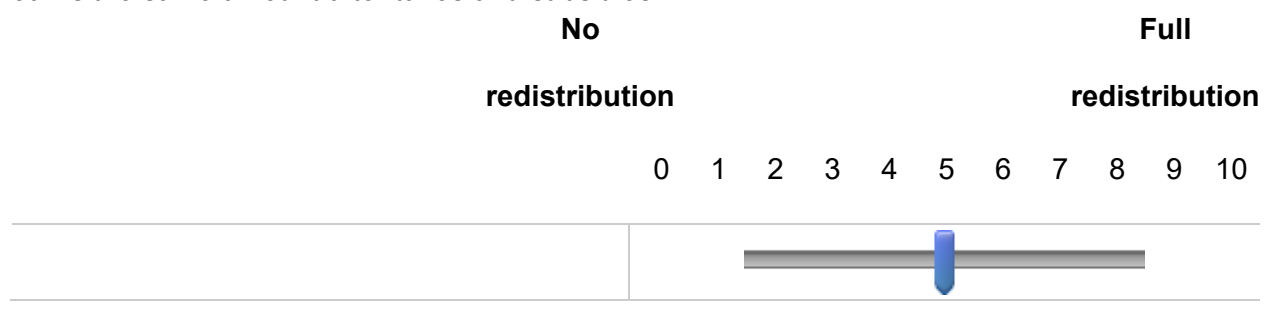
How **willing** or **unwilling** are you to take **risks**, in general? *Please indicate your response on this scale from 0-"Completely unwilling to take risks" to 10-"Very willing to take risks".*



How **competitive** do you consider yourself to be? *Please indicate your response on this scale from 0-"Not at all competitive" to 10-"Very competitive".*



How much **economic redistribution** do you want in society? *"No redistribution" means the state does not influence the income distribution at all, and "Full redistribution" means everyone earns the same amount after taxes and subsidies.*



Imagine the following situation: Suppose you unexpectedly received **\$2400** How much of this amount would you donate to a **good cause**? (*Values between 0 and \$2400 are allowed*)

Investment Decision

In this part of the survey, you can earn an **additional bonus payment**.

We ask you to make an investment decision. For this, we endow you with an additional **100 cents = \$1**. You choose how much of this amount (from 0 to 100 cents) you wish to invest.

The amount that you choose not to invest will be yours to keep and cannot be lost.

The amount that you chose to invest will either increase or decrease in value. This depends on the success or failure of the investment. The success or failure will be determined by a computerized 50/50 coin flip after you have completed this survey.

If the virtual coin comes up tails, the **investment fails**, and you **lose** the amount invested.

If the virtual coin comes up heads, the **investment succeeds**, and you receive **2.5 times** the amount invested.


Summary bonus payment: Your additional bonus payment will equal **100 cents**, minus the amount that you invest, plus **2.5 times** the amount that you invest if the investment succeeds.

Your Investment Decision:

Please move the slider to the **amount** that you wish to **invest**.

Cents

0 10 20 30 40 50 60 70 80 90 100



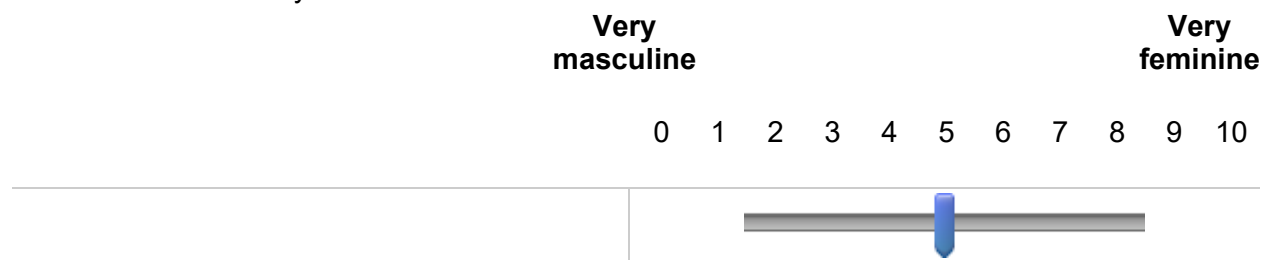
End of Block: Preferences

NOTE: Participants were randomly assigned to one of the long inventories (BSRI, OSRI, CMNI, CFNI). Additionally, every participant took all the shorter scales (CGI, TMF, SOM, Magliozzi). All gender identity measures were elicited in a random order (randomization at the respondent-level).

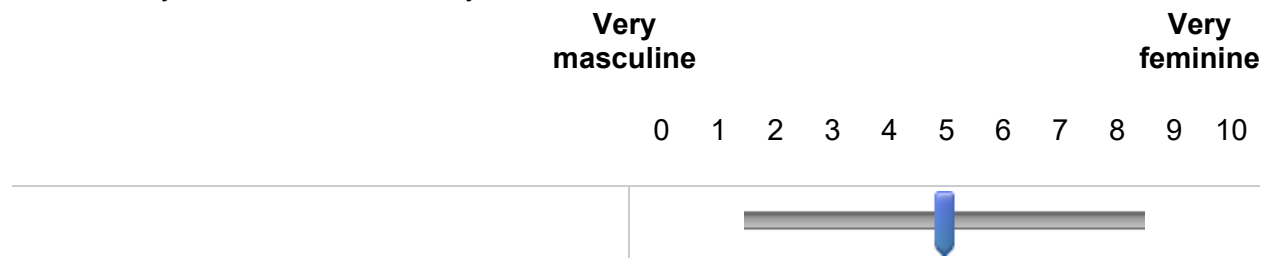
Start of Block: CGI

In the following, you will reply to four questions by positioning a slider on the screen. Note that you have to **move the slider** for your answer to be recorded, even if you want to move the slider back to its initial position.

In general, how do **you** see yourself? Where would **you** put yourself on this scale from 0-"Very masculine" to 10-"Very feminine"?



In general, how do **most people** see you? Where would **most people** put you on this scale from 0-"Very masculine" to 10-"Very feminine"?



Relative to men in general, how do you see yourself? Where would you put yourself on this scale from 0-"Very masculine" to 10-"Very feminine"?



Very masculine

0 1 2 3 4 5 6 7 8 9 10

Very feminine

Start of Block: TMF

Please rate the following statements with what you think is the right answer for you in the given 7-point scale. *The scale ranges from 0-"Very masculine" to 6-"Very feminine".*

Likert scale: 0-"Very masculine" to 6-"Very feminine".

I consider myself as...

Ideally, I would like to be...

Traditionally, my interests would be considered as...

Traditionally, my attitudes and beliefs would be considered as...

Traditionally, my behavior would be considered as...

Traditionally, my outer appearance would be considered as...

End of Block: TMF

Start of Block: BSRI

To which extent do you disagree or agree with each of the following statements?

Likert scale: 1 "Strongly disagree" - 5 "Strongly agree"

I defend my own beliefs.

I am affectionate.

I am conscientious.

I am independent.

I am sympathetic.

I am moody.

I am assertive.

I am sensitive to the needs of others.

I am reliable.

I have a strong personality.

I am understanding.

I am jealous.

I am forceful.

I am compassionate.

I am truthful.

I have leadership abilities.

I am eager to soothe hurt feelings.

I am secretive.

I am willing to take risks.

I am warm.

I am adaptable.

I am dominant.

I am tender.

I am conceited.

I am willing to take a stand.

I love children.

I am tactful.

I am aggressive.

I am gentle.

I am conventional.

End of Block: BSRI

Start of Block: attention_screening2

To demonstrate that you are reading this study carefully, just go ahead and select both "strongly agree" and "strongly disagree" among the alternatives below.

- ☐ Strongly agree
- ☐ Somewhat agree
- ☐ Neither agree nor disagree
- ☐ Somewhat disagree
- ☐ Strongly disagree

End of Block: attention_screening2

Start of Block: OSRI

To which extent do you disagree or agree with each of the following statements?

Likert scale: 1 “Strongly disagree” - 5 “Strongly agree”

- I have been very interested in historical wars.
- I like guns.
- I bake sweets just for myself sometimes.
- I give people handmade gifts.
- I have taken apart machines just to see how they work.
- I wear a blanket around the house.
- I have burned things up with a magnifying glass.
- I really like dancing.
- I leave nice notes for people now and then.
- I have kept a personal journal.
- I have set fuels, aerosols or other chemicals on fire, just for fun.
- I have thrown knives, axes or other sharp things.
- I think a natural disaster would be kind of exciting.
- I have considered joining the military.
- I decorate my things (e.g. stickers on laptop).
- I jump up and down in excitement sometimes.
- I take lots of pictures of my activities.
- I have studied how to win at gambling.
- I have day dreamed about saving someone from a burning building.
- I think horoscopes are fun.
- I have thought about dying my hair.
- I playfully insult my friends.

End of Block: OSRI

Start of Block: CMNI

To which extent do you disagree or agree with each of the following statements?

Likert scale: 1 “Strongly disagree” - 5 “Strongly agree”

I tend to share my feelings.

I would get angry if people thought I was gay.

I dislike any kind of violence.

It bothers me when I have to ask for help.

I bring up my feelings when talking to others.

Work comes first for me.

For me, the best feeling in the world comes from winning.

I enjoy taking risks.

I think that trying to be important is a waste of time.

The women in my life should obey me.

I would be furious if someone thought I was gay.

I would change sexual partners often if I could.

I like to talk about my feelings.

I would find it enjoyable to date more than one person at a time.

It's never ok for me to be violent.

In general I must get my way.

It would be awful if people thought I was gay.

Having status is not important to me.

I put myself in risky situations.

Things tend to be better when men are in charge.

I feel good when work is my first priority.

I would hate to be important.

I will do anything to win.

I think that violence is sometimes necessary.

I never ask for help.

I need to prioritize my work over other things.

I love it when men are in charge of women.

I am not ashamed to ask for help.

I would feel good if I had many sexual partners.

I take risks.

End of Block: CMNI

Start of Block: CFNI

To which extent do you disagree or agree with each of the following statements?

Likert scale; 1 “Strongly disagree” - 5 “Strongly agree”

I would be happier if I was thinner.

It is important to keep your living space clean.

I spend more than 30 minutes a day doing my hair and make-up.

I tell everyone about my accomplishments.

I clean my home on a regular basis.

I feel attractive without make-up.

I believe that my friendships should be maintained at all costs.

I find children annoying.

I would feel guilty if I had a one-night stand.

When I succeed, I tell my friends about it.

Having a romantic relationship is essential in life.

I enjoy spending time making my living space look nice.

Being nice to others is extremely important.

I regularly wear make-up.

I don't go out of my way to keep in touch with friends.

Most people enjoy children more than I do.

I would like to lose a few pounds.

It is not necessary to be in a committed relationship to have sex.

I hate telling people about my accomplishments.

I get ready in the morning without looking in the mirror very much.

I would feel burdened if I had to maintain a lot of friendships.

I would feel comfortable having casual sex.

I make it a point to get together with my friends regularly.

I always downplay my achievements.

Being in a romantic relationship is important.

I don't care if my living space looks messy.

I never wear make-up.

I always try to make people feel special.

I am not afraid to tell people about my achievements.

My life plans do not rely on my having a romantic relationship.

I am always trying to lose weight.

I would only have sex with the person I love.

When I have a romantic relationship, I enjoy focusing my energies on it.

There is no point to cleaning because things will get dirty again.

I am not afraid to hurt people's feelings to get what I want.

Taking care of children is extremely fulfilling.

I would be perfectly happy with myself even if I gained weight.

If I were single, my life would be complete without a partner.

I rarely go out of my way to act nice.

I actively avoid children.

I am terrified of gaining weight.

I would only have sex if I was in a committed relationship like marriage.

I like being around children.

I don't feel guilty if I lose contact with a friend.

I would be ashamed if someone thought I was mean.

End of Block: CFNI_45

Start of Block: SOM question

To what extent would you say that you yourself have **feminine** traits? *Please answer on this scale from 0-"I have few such traits" to 10-"I have many such traits".*

Likert scale 0-10

To what extent would you say that you yourself have **masculine** traits? *Please answer on this scale from 0-"I have few such traits" to 10-"I have many such traits".*

Likert scale 0-10

It is sometimes said that there are feminine and masculine traits and that people may have both feminine and masculine traits. What do you consider to be feminine and masculine traits, respectively?

Please write each trait in a separate box.

What do you consider to be **feminine** traits?

- ☐ _____
- ☐ _____
- ☐ _____

What do you consider to be **masculine** traits?

- ☐ _____
- ☐ _____
- ☐ _____

End of Block: SOM question

Start of Block: gender_questions

What was your assigned **sex** at birth? *(For example on your birth certificate.)*

- ☐ Male
- ☐ Female
- ☐ Intersex

What **gender** are you currently?

- ☐ Man (including Trans Male/Trans Man)
- ☐ Woman (including Trans Female/Trans Woman)
- ☐ Non-binary
- ☐ Would rather not say

What best describes your sibling situation? *Please think about siblings with whom you share at least one biological parent.*

- ☐ I have no siblings
- ☐ I have older siblings only
- ☐ I have younger siblings only
- ☐ I have both older and younger siblings

Display This Question:

*If What best describes your sibling situation? Please think about siblings with whom you share at le...
= I have younger siblings only*

*Or What best describes your sibling situation? Please think about siblings with whom you share at le...
= I have both older and younger siblings*

What is the **sex** of your younger sibling who is closest to you in age?

- ☐ Male
- ☐ Female

Do you have any **comments** you want to share with us?

Thank you for participating in our study! You have reached the end.

You invested X cents from your endowment of 100 cents. Your virtual coin turned out heads so you earn $2.5 \times X$ cents from this investment. Therefore, \$X will be paid out to you as a bonus, in addition to the \$2.5 participation fee.

Your **final payment** for this study is **\$X**

You will receive this payment within three working days after completing this study. The payment will be processed by Prolific. Please click the button below to be redirected back to Prolific and register your submission.

Please click the button below to be redirected back to Prolific and register your submission.

Swiss Teens Survey 1: English Translation

This is the English translation of the original survey, which was administered in German and French. The German original is also provided.

CAREER CHOICE STUDY

UNIVERSITY OF ZURICH

Department: Economics

Head: Prof. Anne Brenøe

We invite you to take part in the **CAREER CHOICE STUDY**, which deals with how students search for trial apprenticeships and apprenticeships. Your participation will help researchers better understand what influences students' career aspirations.

The survey takes about 3-5 minutes. Among all participants in this survey, we will raffle **4 vouchers from Galaxus worth CHF 250 each**.

You have the right to cancel your participation at any time. The confidentiality of your answers and data will be maintained throughout the study. We adhere to Swiss data security standards. With your consent, you allow us and Yousty to combine your answers from this survey with Yousty's data.

If you have any questions or concerns about this study, you can always contact our research team at berufswahl@econ.uzh.ch.

Please tick the box below to confirm that you have read and understood the above information and agree to participate in this study.

- ☐ **I have read and understood the above information and agree to participate in this study.**
- ☐ I do not want to participate in this study
- ☐ **I agree that my answers from this survey will be merged with Yousty's data.**
- ☐ I do not agree that my answers from this survey will be merged with Yousty's data

What is your current age?

What is the best way to describe your current situation?

☐ I attend the 8th grade.

☐ I attend the 9th grade.

☐ Other

What is your sex?

☐ Male

☐ Female

In general, how do **you** see yourself? Where would **you** put yourself on this scale from “0-Very feminine” to “11-Very masculine”?

Please indicate your answer below

1 “Very feminine” - 11 “Very masculine”

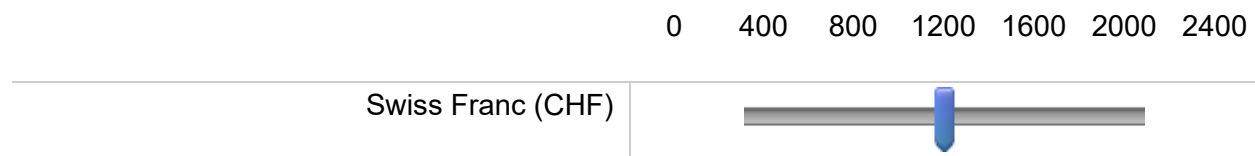
How **competitive** do you consider yourself to be? *Please indicate your response on this scale from 0-not at all competitive to 10-extremely competitive.*

0 “Not at all competitive” – 10 “Extremely competitive”

How **willing** or **unwilling** are you to take **risks**, in general? *Please indicate your response on this scale from 0-completely unwilling to take risks to 10-completely willing to take risks.*

0 “Completely unwilling to take risks” – 10 “Very willing to take risks”

Imagine the following situation: Suppose you unexpectedly received CHF **2400** How much of this amount would you donate to a **good cause**? (*Values between 0 and CHF 2400 are allowed*)



How much economic redistribution do you want in society?

“No redistribution” means the state not influencing the income distribution at all, and “full redistribution” means everyone earning the same amount after taxes and subsidies.

0 " No redistribution" – 10 " Full redistribution"

Thank you for taking the survey. You will automatically be entered into the draw for 4 Galaxus vouchers worth CHF 250 each.

We also have one last request for you: In order to gain further insights from this project for research, we would like to ask you to allow the researchers, Yousty and the Federal Statistical Office to merge your answers from this survey with administrative data (Central Equalisation Register and longitudinal analyses in the education sector). Note that the confidentiality of your answers and data will be maintained throughout the study and that we will adhere to Swiss data security standards.

It's up to you which option you choose, and your decision won't affect the raffle!

- ☐ I agree that my answers from this survey will be merged with administrative data.
- ☐ I do not agree to my answers from this survey being merged with administrative data.

Swiss Teens Survey 1: German Original

BERUFSWAHLSTUDIE

UNIVERSITÄT ZÜRICH

Abteilung: Volkswirtschaftslehre

Leitung: Prof. Anne Brenøe

Wir laden dich ein, an der **BERUFSWAHLSTUDIE** teilzunehmen, die sich mit der Suche von Schüler*innen nach Schnupperlehren und Lehrstellen befasst. Deine Teilnahme hilft Forscher*innen dabei, besser zu verstehen, was die Berufswünsche von Schüler*innen beeinflusst.

Die Umfrage dauert etwa 3-5 Minuten. Unter allen Teilnehmer*innen dieser Umfrage verlosen wir **4 Gutscheine von Galaxus im Wert von je CHF 250.**

Du hast jederzeit das Recht, deine Teilnahme abubrechen. Die Vertraulichkeit deiner Antworten und Daten wird während der gesamten Studie gewahrt. Wir halten uns dabei an die Schweizer Datensicherheitsstandards. Mit deinem Einverständnis erlaubst du uns und Yousty, deine Antworten aus dieser Umfrage mit den Daten von Yousty zusammenzuführen.

Wenn du Fragen oder Bedenken zu dieser Studie hast, kannst du unser Forschungsteam jederzeit unter berufswahl@econ.uzh.ch kontaktieren.

Bitte klick das nachstehende Kästchen an, um zu bestätigen, dass du die obigen Informationen gelesen und verstanden hast und mit der Teilnahme an dieser Studie einverstanden bist.

☐ Ich habe die obigen Informationen gelesen und verstanden und bin mit der Teilnahme an dieser Studie einverstanden. (1)

☐ Ich möchte nicht an dieser Studie teilnehmen. (2)

☐ Ich bin damit einverstanden, dass meine Antworten aus dieser Umfrage mit den Daten von Yousty zusammengeführt werden. (1)

☐ Ich bin **nicht** damit einverstanden, dass meine Antworten aus dieser Umfrage mit den Daten von Yousty zusammengeführt werden. (2)

Wie alt bist du derzeit?

Wie lässt sich deine derzeitige Situation am besten beschreiben?

☐ Ich besuche die 8. Klasse. (1)

☐ Ich besuche die 9. Klasse. (2)

☐ Sonstiges (3)

Bist du männlich oder weiblich?

☐ Männlich (1)

☐ Weiblich (2)

Wie siehst du dich im Allgemeinen? Wo würdest du dich auf dieser Skala von "0 - sehr weiblich" bis "11 - sehr männlich" einordnen?

Bitte gib deine Antwort unten an.

1 "Sehr weiblich" – 11-"Sehr männlich"

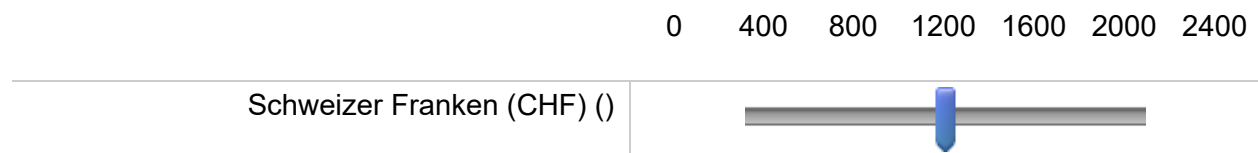
Für wie konkurrenzbetont hältst du dich im Allgemeinen?

0-"Überhaupt nicht konkurrenzbetont" 0 -"Extrem konkurrenzbetont" 10

Wie hoch ist deine Risikobereitschaft im Allgemeinen?

0-"Absolut keine Bereitschaft, Risiken einzugehen" – 10-"Sehr hohe Bereitschaft, Risiken einzugehen"

Stell dir die folgende Situation vor: Angenommen, du hast unerwartet CHF 2'400 erhalten. Wie viel von dieser Summe würdest du für einen guten Zweck spenden?



Wie viel wirtschaftliche Umverteilung wünschst du dir in der Gesellschaft?

"Keine Umverteilung" bedeutet, dass der Staat die Einkommensverteilung überhaupt nicht beeinflusst, und "volle Umverteilung" bedeutet, dass alle nach Steuern und Subventionen gleich viel verdienen.

0-"Keine Umverteilung" 0 – 10 "Volle Umverteilung" 10

Vielen Dank, dass du an der Umfrage teilgenommen hast. Du nimmst damit automatisch an der Verlosung der 4 Gutscheine von Galaxus im Wert von je 250 CHF teil.

Wir haben außerdem noch eine letzte Bitte an dich: Um aus diesem Projekt weitere Erkenntnisse für die Forschung zu gewinnen, würden wir dich gerne darum bitten, dass du es den Forscher*innen, Yousty und dem Statistische Bundesamt erlaubst, deine Antworten aus dieser Umfrage mit administrativen Daten (Zentrales Ausgleichsregister und Längsschnittanalysen im Bildungsbereich) zusammenzuführen. Beachte, dass die Vertraulichkeit deiner Antworten und Daten während der gesamten Studie gewahrt wird und wir uns dabei an die Schweizer Datensicherheitsstandards halten.

Es steht dir ganz frei, für welche Option du dich entscheidest und deine Entscheidung wird keinen Einfluss auf die Verlosung haben!

Ich bin damit einverstanden, dass meine Antworten aus dieser Umfrage mit administrativen Daten zusammengeführt werden.

Ich bin nicht damit einverstanden, dass meine Antworten aus dieser Umfrage mit administrativen Daten zusammengeführt werden.

Swiss Teens Survey 2: English Translation

This is the English translation of the original survey, which was administered in German and French. The German original is also provided.

CAREER CHOICE STUDY

UNIVERSITY OF ZURICH

Department: Economics

Head: Prof. Anne Brenøe

The University of Zurich (UZH) would like to learn more about career choices. This survey will take about 10 minutes to complete. We **will raffle off two vouchers from Digitec/Galaxus worth CHF 100 each among** all participants. Please try to answer honestly. This survey is intended for students in the 8th and 9th grades.

Check the box below to confirm that you have read and understood the information here and agree to participate in this study. If you have any questions or concerns about this study, you can always contact our research team at berufswahl@uzh.ch. You have the right to withdraw your consent and/or cancel your participation at any time, for any reason. UZH will collect your answers and email address using the German online survey tool Qualtrics – they will be stored accordingly in Germany. UZH will delete your answers and your e-mail address after five years. UZH will use an anonymized data set for research purposes only. UZH may share this anonymized data set with other universities for research purposes.

- ☐ I have read and understood the above information and agree to participate in this study.
- ☐ I do not want to participate in this study
- ☐ I agree that my answers from this survey will be merged with Yousty's data.
- ☐ I do not agree that my answers from this survey will be merged with Yousty's data

We will first ask some general questions about you.

What is the best way to describe your current situation?

☐ I attend the 8th grade.

☐ I attend the 9th grade.

☐ Other

What is your sex?

☐ Male

☐ Female

How many **trial apprenticeships (on-site in a company)** have you completed?

In which occupations did you do a trial apprenticeship?

Plant operator (74)
Automation technician (59)
Automotive specialist (31)
Automotive mechatronics technician (68)
Baker-confectioner-confectioner (63)
Hairdresser (37)
Dental assistant (8)
Retail trade specialist (2)
Electrician (13)
Electronics technician (60)
Care specialist (6)
Operational maintenance specialist (64)
Healthcare specialist (5)
Housekeeping specialist (66)
Forest warden (62)
Gardener (35)
Heating engineer (61)
Hotel specialist (67)
Computer scientist (11)
Businessman/woman (1)
Cook (28)
Design engineer (70)
Biology laboratory technician (57)
Chemistry laboratory technician (58)
Agricultural machinery mechanic (69)
Farmer (36)
Logistician (3)
Painter (53)
Bricklayer (54)
Mediamatician (73)
Medical practice assistant (51)
Metal worker (71)
Assembly electrician (14)
Pharmaceutical assistant (52)
Polymechanic (24)
Production mechanic (72)
Restoration specialist (65)
Sanitary fitter (55)
Carpenter (19)
Road builder (56)
Draftsman/draftswoman (18)
Carpenter (20)
Other. Please enter: (39)

How many **information events of companies (on site in a company)** have you attended?

▼ none (6) ... 10 or more (14)

In which occupation did you attend these information events?

Plant operator (74)
Automation technician (59)
Automotive specialist (31)
Automotive mechatronics technician (68)
Baker-confectioner-confectioner (63)
Hairdresser (37)
Dental assistant (8)
Retail trade specialist (2)
Electrician (13)
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Medical practice assistant (51)
Metal worker (71)
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Polymechanic (24)
Production mechanic (72)
Restoration specialist (65)
Sanitary fitter (55)
Carpenter (19)
Road builder (56)
Draftsman/draftswoman (18)
Carpenter (20)
Other. Please enter: (39) _____

How many **information events of companies (on-site in a company)** have you attended in each of these professions?

List of occupations results from the answer to the previous question. If more than three occupations were specified in the previous question, only the first three listed occupations were queried.

How many **online information events from companies** have you attended?

▼ None (6) ... 10 or more (14)

In which occupations have you attended these information events?

Plant operator (74)
Automation technician (59)
Automotive specialist (31)
Automotive mechatronics technician (68)
Baker-confectioner-confectioner (63)
Hairdresser (37)
Dental assistant (8)
Retail trade specialist (2)
Electrician (13)
Electronics technician (60)
Care specialist (6)
Operational maintenance specialist (64)
Healthcare specialist (5)
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Forest warden (62)
Gardener (35)
Heating engineer (61)
Hotel specialist (67)
Computer scientist (11)
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Metal worker (71)
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Pharmaceutical assistant (52)
Polymechanic (24)
Production mechanic (72)
Restoration specialist (65)
Sanitary fitter (55)
Carpenter (19)
Road builder (56)
Draftsman/draftswoman (18)
Carpenter (20)
Other. Please enter: (39) _____

How many **online information events of companies** have you attended in each of these professions?

List of occupations results from the answer to the previous question. If more than three occupations were specified in the previous question, only the first three listed occupations were queried.

In general, how do **you** see yourself? Where would **you** put yourself on this scale from “0-Very feminine” to “11-Very masculine”?

Please indicate your answer below

1 “Very feminine” - 11 “Very masculine”

Do you have any general comments for us here at the end of the survey?

Swiss Teens Survey 2: German Original

BERUFSWAHLSTUDIE
UNIVERSITÄT ZÜRICH

Abteilung: Volkswirtschaftslehre
Leitung: Prof. Anne Brenøe

Die Universität Zürich (UZH) möchte mehr über das Thema Berufswahl erfahren. Diese Umfrage dauert etwa 10 Minuten. Unter allen Teilnehmenden **verlosen wir zwei Gutscheine von Digitec/Galaxus im Wert von je CHF 100.** Bitte versuche, ehrlich zu antworten. Diese Umfrage ist für Schüler-innen der 8. und 9. Klasse bestimmt.

Klicke das untenstehende Kästchen an, um zu bestätigen, dass du die hier stehenden Informationen gelesen und verstanden hast und mit der Teilnahme an dieser Studie einverstanden bist. Wenn du Fragen oder Bedenken zu dieser Studie hast, kannst du unser Forschungsteam unter berufswahl@uzh.ch jederzeit kontaktieren. Du hast das Recht, deine Einwilligung jederzeit zu widerrufen und/oder die Teilnahme abubrechen, egal aus welchem Grund. Die UZH wird deine Antworten und deine E-Mail-Adresse mit dem Deutschen Online-Umfragetool Qualtrics erfassen – sie werden entsprechend in Deutschland gespeichert. Die UZH wird deine Antworten und deine E-Mail-Adresse nach fünf Jahren löschen. Die UZH wird einen anonymisierten Datensatz nur für Forschungszwecke verwenden. Die UZH kann diesen anonymisierten Datensatz zu Forschungszwecken mit anderen Universitäten teilen.

- ☐ **Ich habe die obigen Informationen gelesen und verstanden und bin mit der Teilnahme an dieser Studie einverstanden. (1)**
- ☐ Ich möchte nicht an dieser Studie teilnehmen. (2)
- ☐ **Ich bin damit einverstanden, dass meine Antworten aus dieser Umfrage mit den Daten von Yousty zusammengeführt werden. (1)**
- ☐ Ich bin nicht damit einverstanden, dass meine Antworten aus dieser Umfrage mit den Daten von Yousty zusammengeführt werden. (2)

Wir werden zunächst einige allgemeine Fragen zu deiner Person stellen.

Welche Klasse besuchst du gerade?

☐ 8. Klasse (1)

☐ 9. Klasse (2)

☐ Andere (3)

Bist du männlich oder weiblich?

☐ Männlich (1)

☐ Weiblich (2)

Wie viele **Schnupperlehren (vor Ort in einer Firma)** hast du absolviert?

In welchem Beruf hast du geschnuppert?

Anlagenführer/in (74)
Automatiker/in (59)
Automobil-Fachmann/-frau (31)
Automobil-Mechatroniker/in (68)
Bäcker/in-Konditor/in-Confiseur/in (63)
Coiffeur/-euse (37)
Dentalassistent/in (8)
Detailhandelsfachmann/-frau (2)
Elektroinstallateur/in (13)
Elektroniker/in (60)
Fachmann/-frau Betreuung (6)
Fachmann/-frau Betriebsunterhalt (64)
Fachmann/-frau Gesundheit (5)
Fachmann/-frau Hauswirtschaft (66)
Forstwart/in (62)
Gärtner/in (35)
Heizungsinstallateur/in (61)
Hotelfachmann/-frau (67)
Informatiker/in (11)
Kaufmann/-frau (1)
Koch/Köchin (28)
Konstrukteur/in (70)
Laborant/in Biologie (57)
Laborant/in Chemie (58)
Landmaschinenmechaniker/-in (69)
Landwirt/in (36)
Logistiker/in (3)
Maler/in (53)
Maurer/in (54)
Mediamatiker/in (73)
Medizinische/r Praxisassistent/in (51)
Metallbauer/in (71)
Montage-Elektriker/in (14)
Pharma-Assistent/in (52)
Polymechaniker/in (24)
Produktionsmechaniker/in (72)
Restaurationsfachmann/-frau (65)
Sanitärinstallateur/in (55)
Schreiner/in (19)
Strassenbauer (56)
Zeichner/in (18)
Zimmermann/Zimmerin (20)
Andere. Bitte eingeben: (39) _____

Wie viele **Informationsevents von Firmen (vor Ort in einer Firma)** hast du besucht?

▼ Keine (6) ... 10 oder mehr (14)

In welchen Berufen hast du diese informationsevents besucht?

Anlagenführer/in (74)
Automatiker/in (59)
Automobil-Fachmann/-frau (31)
Automobil-Mechatroniker/in (68)
Bäcker/in-Konditor/in-Confiseur/in (63)
Coiffeur/-euse (37)
Dentalassistent/in (8)
Detailhandelsfachmann/-frau (2)
Elektroinstallateur/in (13)
Elektroniker/in (60)
Fachmann/-frau Betreuung (6)
Fachmann/-frau Betriebsunterhalt (64)
Fachmann/-frau Gesundheit (5)
Fachmann/-frau Hauswirtschaft (66)
Forstwart/in (62)
Gärtner/in (35)
Heizungsinstallateur/in (61)
Hotelfachmann/-frau (67)
Informatiker/in (11)
Kaufmann/-frau (1)
Koch/Köchin (28)
Konstrukteur/in (70)
Laborant/in (57)
Landmaschinenmechaniker/-in (69)
Landwirt/in (36)
Logistiker/in (3)
Maler/in (53)
Maurer/in (54)
Mediamatiker/in (73)
Medizinische/r Praxisassistent/in (51)
Metallbauer/in (71)
Montage-Elektriker/in (14)
Pharma-Assistent/in (52)
Polymechaniker/in (24)
Produktionsmechaniker/in (72)
Restaurationsfachmann/-frau (65)
Sanitärinstallateur/in (55)
Schreiner/in (19)
Strassenbauer (56)
Zeichner/in (18)
Zimmermann/Zimmerin (20)
Andere. Bitte eingeben: (39) _____

Wie viele **Informationsevents von Firmen (vor Ort in einer Firma)** hast du in jedem dieser Berufe besucht?

Liste mit Berufen ergibt sich aus der Antwort zur vorherigen Frage. Falls in der vorherigen Frage mehr als drei Berufe angegeben wurden, wurden nur die ersten drei gelisteten Berufe abgefragt.

Wie viele **online Informationsevents von Firmen** hast du besucht?

▼ Keine (6) ... 10 oder mehr (14)

In welchen Berufen hast du diese informationsevents besucht?

Anlagenführer/in (74)
Automatiker/in (59)
Automobil-Fachmann/-frau (31)
Automobil-Mechatroniker/in (68)
Bäcker/in-Konditor/in-Confiseur/in (63)
Coiffeur/-euse (37)
Dentalassistent/in (8)
Detailhandelsfachmann/-frau (2)
Elektroinstallateur/in (13)
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Fachmann/-frau Betreuung (6)
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Sanitärinstallateur/in (55)
Schreiner/in (19)
Strassenbauer (56)
Zeichner/in (18)
Zimmermann/Zimmerin (20)
Andere. Bitte eingeben: (39) _____

Wie viele **online Informationsevents von Firmen** hast du in jedem dieser Berufe besucht?

Liste mit Berufen ergibt sich aus der Antwort zur vorherigen Frage. Falls in der vorherigen Frage mehr als drei Berufe angegeben wurden, wurden nur die ersten drei gelisteten Berufe abgefragt.

Wie siehst du dich im Allgemeinen? Wo würdest du dich auf dieser Skala von «0 - sehr weiblich» bis «11 - sehr männlich» einordnen?

Bitte gib deine Antwort unten an.

0 - sehr weiblich» bis «11 - sehr männlich»

Hast du irgendwelche generellen Kommentare für uns hier am Ende der Umfrage?
