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Do Americans Favor Female or Male Politicians? Evidence from **Experimental Elections** 

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**Abstract** 

Women are severely underrepresented in American politics, especially among Republicans.

This underrepresentation can arise from women being less willing to run for office, from

voter bias against women, or from political structures that make it more difficult for women to

compete. Here we show to what extent support for female candidates varies by voters' party

affiliation and gender. We carried out hypothetical elections in which participants made vote

choices solely based on politicians' faces. When deciding between candidates of different

genders, Democrats, and particularly Democratic women, preferred female candidates, while

Republicans chose female and male candidates equally often. These patterns remained when

controlling for respondents' education, age, and political knowledge and for candidates' age,

attractiveness, and perceived conservativeness. Our results suggest that voter bias against

women cannot explain female underrepresentation. On the contrary, American voters appear

ready to further narrow the gender gap in politics.

Keywords: Gender; Elections; Gender discrimination; Political candidates

JEL Codes: D72; J16

#### 1 Introduction

Female politicians are more likely to implement policies that benefit women and children (Chattopadhyay & Duflo, 2004; Miller, 2008). Yet, women remain severely underrepresented in American politics. Though the percentage of elected women has steadily increased over the past four decades, women account for only 27% of House and Senate members in the 117th Congress, seated in January 2021, 31% of state legislatures, and 30% of statewide elective offices (Fig. 1A). Supply side explanations conclude that women are less willing to run for office, which could result from lack of encouragement or self-confidence (Fox & Lawless, 2004; Fox & Lawless, 2010; Kanthak & Woon, 2015). Demand side explanations suggest voter bias against women (Baltrunaite et al., 2019; Baskaran & Hessami, 2018), who need to be more qualified than men to be able to compete at all (Pearson & McGhee, 2013). Female disadvantage could also follow from women being held back by party leaders (Besley et al., 2017; Casas-Arce & Saiz, 2015). Furthermore, incumbency advantage (Gelman & King, 1990) hurts women given their historically low representation.

Yet, these forces do not seem to equally affect Democrats and Republicans, as a major gap has opened with respect to female representation between the two parties (Fig. 1B). In the 117th Congress, the percentage of women in the Democratic caucus (40%) is nearly three times that of the Republican caucus (14%), despite the spike in newly elected female Republicans (Fig. 1C). Research offers various explanations for this partisan gap. On the supply side, compared to their Democratic counterparts, Republican women are less likely to run in districts that are favorable to the party (Pearson & McGhee, 2013), and have a disadvantage in raising campaign funds (Bucchianeri, 2018; Thomsen & Swers, 2017). On the demand side, female politicians may simply appeal more to the Democratic electorate (Dolan, 2004; Sanbonmatsu, 2002), perhaps because Americans perceive women as more liberal than men of the same party (McDermott, 1997; Sanbonmatsu & Dolan, 2009). Another explanation is that voters tend to prefer candidates of their own gender (Sanbonmatsu, 2002). This should help female politicians among Democrats but harm them among Republicans, given that women identify more as Democrats than Republicans, while the reverse is true for men (Edlund & Pande, 2002; Gillion et al., 2020). A meta-analysis of vote choice experiments found that, on average, being described as a woman increased a candidate's vote share by four percentage points among Democrats, but reduced it by one point among Republicans (Schwarz & Coppock, 2021). However, the estimated effects of gender vary widely and some studies find that gender has no effect (Hainmueller et al., 2014) or that women have a small disadvantage (Ono & Burden, 2019).

To disentangle potentially interacting effects of voter gender and partisanship, we carried out hypothetical elections in which participants made vote choices solely based on politicians' faces. American respondents (N=729), recruited using Amazon Mechanical Turk (MTurk), were shown 99 pairs of randomly selected headshot photographs of politicians they did not know, and were asked each time for whom they would vote as their representative (see Figs. A1 and A2). Respondents were informed in advance that the task involved voting, but they were not informed that the study goal was to test the influence of candidate gender on vote choice. To ensure that respondents would not guess the study goal, each hypothetical election depicted with equal probability two females, two males, a female on the left and a male on the right, or a male on the left and a female on the right. Study I (N=293) took place in April 2016 during the presidential primaries and study II (N=436) in October 2020, just before the general election. The same photographs, depicting 736 elected Members of the European Parliament (MEPs), were used in both surveys. Experimental vote tasks were followed by background questions, one of which was the respondent's preferred candidate in that year's presidential election. Answers to that question were used to classify Democratic and Republican voters. While Democrats were overrepresented in study I, respondents' presidential preference for Biden (51.4%) and Trump (46.3%) in study II almost perfectly matched the final election outcome (Biden 51.3% vs. Trump 46.9%). In study II, we also asked respondents about their party identification, and we also present results on gender preference using that measure.

The rest of the paper is organized as follows. Section 2 presents a theoretical framework that distinguishes the role of candidate gender and other two thin slices of information, candidate attractiveness and perceived conservativeness, in a low-information election. We use this framework to derive testable predictions on how support for female candidates differs between female and male Democrats and Republicans, and how difference in candidate attractiveness and perceived conservativeness is related to their support. Section 3 presents our data, and section 4 empirical results. Section 5 concludes.

#### 2 Theoretical framework

We analyze voting in a low-information setting in which voters and candidates differ in their gender, ideology, and non-ideological characteristics. Voters belong to one of two political parties, Democrats on the left and Republicans on the right. When choosing whom to vote for, Democrats would prefer to choose a liberal candidate and Republicans a

conservative candidate, but they do not observe candidates' true ideology. Instead, voters derive cues on ideology from candidate photographs, particularly from candidate's gender, candidate's attractiveness, and how conservative the candidate looks. Voters may also value attractive looks either due to taste-based discrimination, in line with beauty premium in the labor market (Hamermesh & Biddle, 1994), or due to a halo effect, whereby good looks are associated with, say, competence or intelligence (Langlois et al., 2000).<sup>1</sup>

We denote gender by g, obtaining value f for females and m for males. An indicator variable  $I_K$  for candidate K's gender obtains value 0 for males and 1 for females. Party is denoted by P, which can also take two values, D for Democrats and R for Republicans. Voters' gender preference may depend on their party preference and own gender. More specifically, we denote valuation of the candidate being female by  $\varphi_{P,g}$  for supporters of party P of gender g. For voters who are indifferent to the politician's gender,  $\varphi_{P,g}$  would be equal to zero and any advantage of female or male candidates would arise through differences in their attractiveness or perceived conservativeness. If voters have an inherent gender preference,  $\varphi_{P,g} < 0$  for groups favoring men and  $\varphi_{P,g} > 0$  for groups favoring women. Attractiveness of candidate K is denoted by  $a_K$ . We denote the valuation of attractiveness by supporters of party P by  $\beta_P$ , with our prior being that this beauty premium is positive for both parties (i.e., voters generally prefer better looking candidates). Perceived conservativeness of candidate K is denoted by  $C_K$ . Its valuation by supporters of party P is denoted by  $K_P$ , with  $K_D < 0 < K_R$ .

We analyze probabilistic voting with two candidates, X and Y. As in Lindbeck and Weibull (1987, 1993) and Dixit and Londregan (1996), voters differ in a continuously distributed term whose ex-ante distribution is known but whose realization is unknown to parties or candidates. We analyze voting by individual i who belongs to group J, in which J is defined by a combination of party P,  $P \in \{D, R\}$ , and gender g,  $g \in \{f, g\}$ . As Persson and Tabellini (2002), we model the random component as a voter-specific popularity parameter  $\gamma_{iJ}$  that measures voter i's who belongs to group J idiosyncratic preference for candidate X, relative to candidate Y. The expected utility of voter i who belongs to party P and is of gender g from voting for candidate X is

$$EU_X^i = \varphi_{P,g}I_X + \beta_P a_X + \kappa_P c_X + \gamma_{iJ}.$$

<sup>&</sup>lt;sup>1</sup> Berggren et al. (2010) show that political candidates with more attractive looks are generally perceived also to be more competent, intelligent, likable, and trustworthy in ratings based on photographs only.

<sup>&</sup>lt;sup>2</sup> Theoretical predictions would remain qualitatively similar if we would assume that there is a separate random term associated with each candidate.

The expected utility from voting for candidate Y is

$$EU_Y^i = \varphi_{P,a}I_Y + \beta_P\alpha_Y + \kappa_Pc_Y.$$

Voter i maximizes his or her expected utility, and votes for X if  $EU_X^i > EU_Y^i$ . We introduce notation  $\tilde{a} = a_X - a_Y$  for the extent to which candidate X looks more attractive than candidate Y, with negative values indicating the extent to which Y looks more attractive, and  $\tilde{c} = c_X - c_Y$  for the extent to which candidate X looks more conservative than candidate Y, with negative values indicating the extent to which Y looks more conservative.  $EU_X^i > EU_Y^i$  is equivalent to

$$\gamma_{i,I} > \varphi_{P,a}(I_Y - I_X) - \beta_P \tilde{a} - \kappa_P \tilde{c} = \hat{\gamma}_{i,I}(I_X, I_Y, \tilde{a}, \tilde{c}).$$

The ex-ante probability that voter i votes for candidate X, measured before the realization of  $\gamma_{iJ}$  is revealed, is denoted by  $q_{iJ}$ . As Persson and Tabellini (2002), we assume that it follows a symmetric uniform distribution around zero, which is sufficiently wide so that  $0 < q_J < 1$ . This assumption rules out corner solutions and allows deriving closed-form solutions. Denoting the range of the distribution by  $[-\frac{1}{2\Gamma},\frac{1}{2\Gamma}]$  with  $\Gamma$  being the density function for  $\gamma_{iJ}$ , the ex-ante probability that voter i in group J votes for candidate X is given by

$$(1) \ q_{iJ}(I_X,I_Y,\tilde{\alpha},\tilde{c}) = \int_{\hat{\gamma}_{iJ}}^{\frac{1}{2\Gamma}} \Gamma d\gamma_{iJ} = \frac{1}{2} + \Gamma \left[ \varphi_{P,g}(I_X - I_Y) + \beta_P \tilde{\alpha} + \kappa_P \tilde{c} \right].$$

We can now derive

**Proposition 1.**  $\forall I_x, I_y, \tilde{\alpha}, \tilde{c}$ :

$$(\mathrm{i}) \ \ q_{iJ}(1,I_Y,\tilde{\alpha},\tilde{c}) - q_{iJ}(0,I_Y,\tilde{\alpha},\tilde{c}) = q_{iJ}(I_X,0,\tilde{\alpha},\tilde{c}) - q_{iJ}(I_X,1,\tilde{\alpha},\tilde{c}) = \Gamma \varphi_{P,g};$$

(ii) 
$$\frac{\partial q_{ij}(I_X,I_Y,\tilde{\alpha},\tilde{c})}{\partial \tilde{\alpha}} = \Gamma \beta_P;$$

(iii) 
$$\frac{\partial q_{ij}(I_X,I_Y,\tilde{\alpha},\tilde{c})}{\partial \tilde{c}} = \Gamma \kappa_P$$
.

**Proof.** (i) follows by inserting equation (1) with different values of the gender indices and (ii) and (iii) follow by differentiating equation (1).

Proposition 1 shows how voters of a given type respond to changes in candidate characteristics. Part (i) establishes how preference for a candidate being female among voters of gender g among supporters of party P translates into difference in the probability of supporting a given candidate, once controlling for candidates' perceived attractiveness and

conservativeness. Part (ii) derives how party-specific attractiveness premium translates into votes, and part (iii) how party-specific preference for or against conservative-looking candidates translates into votes.

Previous research has concluded that female politicians appeal more to the Democratic electorate (McDermott, 1997; Sanbonmatsu & Dolan, 2009), suggesting that  $\varphi_{D,0} > \varphi_{R,0}$  and  $\varphi_{D,1} > \varphi_{R,1}$ . Also, some research has concluded that voters tend to favor candidates of the same gender (Sanbonmatsu, 2002), suggesting  $\varphi_{D,1} > \varphi_{D,0}$  and  $\varphi_{R,1} > \varphi_{R,0}$ . Therefore, we expect Democrats of either gender to be more likely to select a female candidate in a mixed-gender race than Republicans of the same gender. Furthermore, we expect Democratic women to choose female candidates more often than Democratic men and Republican women to choose female candidates more often than Republican men. Importantly, our model leaves it open whether a given voter group is more likely to select female or male candidates, and presents predictions only on how support for female candidates differs by voter gender and partisanship. Evaluating the overall gender preference is left to the empirical analysis. Our main testable hypotheses are<sup>3</sup>:

**Hypothesis 1.** Democrats of either gender are more likely to vote for women than Republicans of the same gender.

**Hypothesis 2.** Men tend to support male candidates more often than women.

**Hypothesis 3.** Women tend to support female candidates more often than men.

Previous research also found that both conservative and liberal voters favor more attractive-looking candidates but that beauty premium is larger for candidates on the right in both low-information real elections and in experimental elections (Berggren et al., 2017). Therefore, our prior was that  $\beta_R > \beta_D > 0$ . However, we do not assume that in our model, and leave the sign and the magnitude of  $\beta_D$  and  $\beta_R$  to be estimated in the econometric analysis. As Olivola et al. (2018) found that Republicans tended to support candidates who

<sup>&</sup>lt;sup>3</sup> We pre-registered these hypotheses concerning mixed-gender elections in AEA RCT Registry, with RCT ID AEARCTR-0006653. In the pre-registration plan, we operationalized Hypothesis 1 in two ways: with the current wording that relies on party identification and in terms of presidential vote preference as "those supporting Biden of either gender are more likely to vote for a female candidate than those supporting Trump of the same gender)". We present in the empirical results section evidence on both formulations.

looked more conservative, we further expect that Republicans are more likely and Democrats less likely to support candidates who appear conservative. Therefore, our prior is that  $\kappa_D < 0 < \kappa_R$ .

The most suitable application of our model to US politics are primary elections, in which candidates do not differ in their partisan labels. In general elections, voters would, plausibly, also use party membership as an additional queue on ideology which we do not include in our model. Given that most House districts typically elect repeatedly either a Democrat or a Republican, it may well be that the gender gap in American politics arises primarily from gender preference in intra-party competition for nominations, most notably in primary elections. At the cost of making the model more complicated, it would be possible to extend our model to also include party labels. In that case, strong partisans would likely choose always or almost always a candidate with their preferred party label, while independents and voters with only weak party preference would consider candidates from different parties, possibly tilting towards supporting candidates from one party, unless a candidate from the other party would be sufficiently preferred in terms of the combination of valuations related to gender, attractiveness, and perceived conservativeness to compensate for the partisan disadvantage.

## 3 Data

Our analysis is based on two online surveys conducted in April 2016 and October 2020, respectively. The Ethics Commission of the Department of Economics at the University of Munich approved the research, with decision numbers 2015-03 (Study I) and 2020-09 (Study II). Study II was pre-registered at the American Economic Association's registry for randomized trials on 22 October 2020 (AEARCTR-0006653). Respondents in both studies gave their informed consent prior to participation.

#### 3.1 Stimulus material

Both studies used the same database of photographs, which consisted of official headshots of all 736 elected Members of the European Parliament (MEPs) from the 2009–2014 parliamentary term, collected from the web page of the European Parliament. Thereby, 256 (35%) were female and 480 were male politicians. This database had already been used in prior research (Berggren et al., 2017), which obtained ratings of the MEPs' perceived attractiveness from 296, and ratings of perceived conservativeness from 292, American

respondents using MTurk. Table 1 presents descriptive statistics of the MEPs in the photographs. On average, female MEPs are evaluated as more attractive than male MEPs, and male MEPs are perceived as more conservative. The average age of male MEPs is 52 years and of female MEPs 50 years.

In study I, respondents were asked whether they recognized any of the persons shown in the photographs. As in prior research (Berggren et al., 2017), the likelihood that American respondents would recognize any of the MEPs was virtually zero. Only one respondent recognized any of the politicians (and only one). We therefore did not ask this question in study II. One study II respondent nevertheless wrote in the write-in comments that one person looked familiar and guessed correctly that the photographs could depict members of the European Parliament.

#### 3.2 Questionnaire

The questionnaires in both studies were programmed using SoSci Survey software. After a short introduction, respondents were asked to complete the experimental voting task. Respondents in both studies saw 99 pairs of candidates, randomly selected from all 736 photographs. Thereby, each pair was equally likely to depict two males, two females, a female on the left and a male on the right, and a male on the left and a female on the right. For each pair, respondents were asked for which person they would vote as a member of the House of Representatives if they would have to decide based on photographs only. In study I (Fig. A1), respondents were asked if they would vote for the person on the left or on the right. In study II (Fig. A2), respondents answered the question using a scale from 1 ("definitely the person on the left") to 6 ("definitely the person on the right"). In both studies, respondents could opt to abstain or prefer not to answer.

After making 99 vote choices, respondents were asked three policy questions, followed by which presidential candidate they would prefer to see elected as President. Then, respondents answered four political knowledge questions, and were asked to predict the presidential candidates' national vote shares. Finally, respondents were asked for gender, age, nationality, and education. In study II, respondents were also asked whether they think of themselves as a Republican, a Democrat, or an Independent, if they are registered to vote in the 2020 U.S. presidential election and, if so, in which state. Respondents could also provide comments on the survey or request a summary of the results before they received their MTurk completion code. The full questionnaires are available in the Online Appendix.

# 3.3 Participant recruiting

Participants in both studies were recruited through the Amazon Mechanical Turk (MTurk) platform.

Study I. We obtained 293 questionnaires 2-11 April 2016. Respondents had to be located in the U.S., had to have at least 1,000 MTurk tasks approved, as well as an approval rate of higher than 95%. Participants were paid \$2.14 for completing the questionnaire, which took 15 minutes (SD = 5 minutes) on average. This would translate to an hourly wage of \$8.7, which was 20% above the minimum wage in the U.S. at the time of the survey. Including fees paid to MTurk, we paid \$3.07 per respondent.

Study II. Our plan was to collect 360 questionnaires. Respondents had to be located in the U.S. and needed an approval rate of 90% or higher for previously completed MTurk tasks. We paid for an additional qualification to make sure that our sample contained 120 respondents who registered their political affiliation at MTurk as conservative, as well as another 120 respondents who registered their political affiliation at MTurk as liberal. We rejected approval from one respondent who failed to provide a completion code, six respondents who provided an incorrect code, and 19 respondents due to speeding. Speeders were identified based on a score calculated directly by the Sosci Survey software. This score, called DEG\_TIME, is normed in such a way that values higher than 100 indicate low-quality data, although the software recommends lower scores for stricter filtering. We rejected respondents with a score above 100. Another 12 (38) respondents did not meet our sampling criteria, because MTurk listed them as liberal (conservative) while they answered that they identified as Republican (Democrat). We initially decided to replace these respondents, due to concerns about reliability of their reported political affiliation. However, as we had not prespecified respondent exclusion criteria in our pre-registration, we later decided to keep all respondents in the analysis. This led to oversampling of 76 questionnaires, with the total being 436 questionnaires obtained 23-27 October 2020. Our online simulation tool, described below in subsection 4.2, allows readers to verify that the estimated effects change only marginally if implementing any combination of these possible exclusion criteria. Participants were paid \$2.1 for completing the questionnaire, which took 11 minutes (SD = 4 minutes) on average. This would translate to an hourly wage of \$11.5, which was 59% above the

minimum wage in the U.S. at the time of the survey. Including the fees paid to MTurk, we paid \$2.65 per respondent.

Six respondents reported a nationality other than American, but we did not exclude them as they could be naturalized citizens who feel more attached to another country. A handful of people guessed in the comments that we are interested in how gender, age and/or appearance affects vote choice, but none guessed that we are interested in how respondents' political views affect their choice between men and women. We did not exclude any respondents based on write-in comments.

## 3.4 Participants

Table 2 presents summary statistics of all respondents in the two surveys. Of the 293 respondents in *study I*, 49% (N=144) were female. Respondents' average age was 40 years (SD = 12.6). A total of 62% said that they would vote for a Democratic candidate (either Hillary Clinton or Bernie Sanders) in the 2016 presidential election, while 33% said that they would vote for one of three Republican candidates (Donald Trump, Ted Cruz, or John Kasich). The share of Democratic respondents in our survey was somewhat higher compared to the general population. For example, a Gallup poll conducted April 6-10, 2016, around the same time as our survey, found that 49% of respondents identified as Democratis/Democratic leaners and 41% as Republicans/Republican leaners (Gallup Organization, 2016).

Of the 436 respondents in *study II*, 44% (N=193) were female. Respondents' average age was 39 years (SD = 12.3). A total of 51.4% said that they would vote for Joe Biden in the 2020 presidential election, while 46.3% said that they would vote for Donald Trump. This almost perfectly matched the final election outcome (Biden 51.3% vs. Trump 46.9%). With respect to political affiliation, 51% identified as Democrats (including Independent leaners), 38% as Republicans (including Independent leaners), and 9% as Independents. In comparison, a Gallup poll conducted around the same time of our survey found that 49% of respondents identified as Democrats/Democratic leaners and 45% as Republicans/Republican leaners (Gallup Organization, 2020).

#### 4 Results

## 4.1 Descriptive evidence on gender preference

We start by presenting descriptive evidence on female vote by male and female Republicans and Democrats. As respondents in each group were presented large numbers of randomly chosen pairs of photographs using the same randomization rule and set of photographs, the share of female vote should not vary systematically if the groups would not differ in their gender preference. Therefore, any systematic differences would be indicative of group-level differences in tendency to choose a female candidate in experimental elections.

As shown in Fig. 2, Democrats favored female candidates in mixed-gender races, and the effect was particularly strong for female Democrats. Republicans, regardless of their gender, chose female candidates about half of the time, thereby not favoring either gender. Average support for female candidates remained stable across male and female voters of both parties from 2016 to 2020, ranging from 42% (95%-confidence: 35–49%) among Republican men in 2016 to 75% (95%-confidence: 71–79%) among Democratic women in 2020. The picture is similar when comparing respondents based on their stated party identification (Fig. A3) as well as when accounting for the strength of the reported choice (Fig. A4), each of which was collected in 2020 only. Taken together, these figures confirm Hypothesis 1 that Democrats of either gender are more likely to vote for women than Republicans of the same gender, both when using presidential voting intentions and when using stated party identification.

Fig. 3 compares female voting by respondent gender, without controlling for respondent partisanship. Female respondents chose female candidates almost twice as often as male candidates, while male respondents chose female and male candidates about equally often. This figure confirms Hypothesis 2 that men tend to support male candidates more often than women and Hypothesis 3 that women tend to support female candidates more often than men. The gap in support for female candidates between female and male respondents is similar in size to the gap in support for female candidates between Democratic and Republican men, but clearly smaller than the gap in support for female candidates between Democratic and Republican women. Taken together, our descriptive evidence suggests that women and Democrats strongly favor female candidates in low-information experimental elections, while men and Republicans do not exhibit strong gender preference in low-information experimental elections, on average.

Prior research suggests that voters tend to vote for more attractive candidates (Ahler et al., 2017; Berggren et al., 2010, 2017), and use candidate appearance as a cue to ideology (Berggren et al., 2017; Olivola & Todorov, 2010; Rule & Ambady, 2010). To address

potential concerns that Democrats may prefer female candidates simply because women may look better, or more liberal, we took into account evaluations of each MEP's attractiveness and perceived conservativeness, obtained in prior research (Berggren et al., 2017). The partisan gender gap prevailed when analyzing separately hypothetical elections in which the female candidate looked more attractive (Fig. A5) or less attractive (Fig. A6). Similarly, partisan gender gap remains when analyzing separately hypothetical elections in which the female (Fig. A7) or the male (Fig. A8) candidate looked more conservative. In each case, female Democrats clearly favored female candidates.

#### 4.2 Econometric evidence on gender preference

We used ordinary least squares (OLS) regression to analyze how vote choice relates to respondent and candidate characteristics, as recommended when estimating treatment effects on binary outcomes (Gomila, 2020). All variables and their coding are described in Table A1. Our dependent variable *vote choice* is a dummy variable, coded as 1 if the respondent chose the female candidate in hypothetical elections with a male and a female candidate, and 0 otherwise. Hypothetical elections in which the respondent abstained were excluded. Although the assumption that errors are normally distributed is violated because of a binary outcome variable, this should not be a large problem given our sample size (Hellevik, 2009). To account for heteroscedasticity and correlation of the outcome variable within respondents, we used robust standard errors clustered at the respondent level (N= 684).

Table 3 presents the results for the support for female candidate in the mixed-gender elections, using four different specifications. In model 1, we used only respondent characteristics related to gender, partisanship defined as Democrat for those supporting Democratic presidential candidate and Republican for those supporting Republican presidential candidate in the upcoming presidential election, age, level of education, and time dummy for those who responded in 2016. In model 2, we added a measure of the two candidates' attractiveness difference, measured as average attractiveness score of the female candidate minus the average attractiveness score of the male candidate, and its interaction with Democrat. In model 3, we added a measure of the two candidates' difference in perceived conservativeness, measured as average perceived conservativeness of the female candidate minus the average perceived conservativeness of the male candidate, as well as the age difference between the two candidates, defined as the age of the female candidate minus the age of the male candidate, and the interactions of both measures with the respondent being

Democrat. In the model 4, we added a measure of respondent's political knowledge and its interaction with respondent being Democrat. Regressions were run using STATA/MP 16.0. Pre-registered conjecture that Democrats of either gender are more likely to vote for women than Republicans of the same gender was tested against one-sided null hypothesis that this is not the case with an F-test, separately for females and males.

The conjecture that Democrats of either gender are more likely to support female candidates than Republicans of the same gender received strong support in all models. According to model 1, Democratic men were 15 percentage points more likely than Republican men to vote for female candidates (estimated change in probability 0.146; P <0.001; F(1.683) = 42.27) and Democratic women were 22 percentage points more likely than Republican women to vote for female candidates (estimated change in probability 0.222; P < 0.001; F(1,683) = 73.84). Column 2 shows that the results remain when controlling for the difference in candidate attractiveness. Respondent education has no statistically significant effects. Older respondents are somewhat likelier to support female candidates than younger respondents, although the magnitude of this effect is small. Moving from column 1 to column 2 confirms findings from previous research: respondents were more likely to select more attractive candidates (Ahler et al., 2017; Berggren et al., 2010, 2017). The regressions also confirm that Republicans tended to vote for women about as often as for men, as already suggested by Fig. 2. The reference person (male Republican with bachelor's degree in 2020) voted for female candidate in 46% of mixed-gender races in column 1, and female Republicans in 52% of mixed-gender races.

Column 3 adds perceived conservativeness and age difference between candidates. Republicans tended to support candidates who looked more conservative as in (Olivola et al., 2018), while Democrats tended to support candidates who looked more liberal. Candidate age difference has no effect on average vote shares. Controlling for perceived conservativeness somewhat reduces the effect of gender, but most of the gender gap remains: male Democrats are 11 percentage point more likely to support female candidate than male Republicans, and female Democrats 18 percentage point more likely than female Republicans. Finally, column 4 shows that the effects of political knowledge on gender preference varied by party. While political knowledge did not affect gender preference of Republicans, Democrats who were more knowledgeable about US politics were more likely to support female candidates.

In our pre-registration, we did not list hypotheses with respect to perceived conservativeness. Therefore, the estimates in columns 1 and 2 are the best test of our preregistered hypotheses. Importantly, controlling for attractiveness leaves the estimated effect of gender almost unchanged. This is in line with Proposition 1 deriving the effects of gender and attractiveness that are independent of each other. Comparing columns 2 and 3 allows evaluating how our testable hypotheses can be expected to change if voters have access to only two thin slices of information, candidate gender and attractiveness. In that case, we would expect that both could be used as a cue for candidate ideology. The empirical results confirm this clearly for gender: if perceived conservativeness is not controlled for, the estimate for differential support for female candidates among Democrats compared with among Republicans is somewhat higher. The difference in gender preference between male Democrats and male Republicans in our model,  $\Gamma(\varphi_{D,0} - \varphi_{R,0})$ , would correspond to the estimated coefficient for Democrat in Table 3, and the difference in gender preference between female Democrats and female Republicans in our model,  $\Gamma(\varphi_{D,1} - \varphi_{R,1})$ , would correspond to the sum of the estimated coefficients for Democrat and Democrat x Female in Table 3. Our theoretical model corresponds most closely to column 3 that includes perceived conservativeness, and has value 0.11 for Democrat (gap among men) and 0.18 for Democrat + Democrat x Female (gap among women). Leaving out perceived conservativeness corresponds to a move to column 2, and increases partisan gap among men to 0.15 and among women to 0.22. Therefore, not including conservativeness as an additional thin slice of information increases gender preference for female candidates by about four percentage points among both male Democrats compared with male Republicans and among female Democrats compared with female Republicans. This can be explained by women being generally perceived as less conservative, as shown in Table 1 for evaluations of MEPs in our photographs, and (McDermott, 1997; Sanbonmatsu & Dolan, 2009) for perceptions of female and male politicians in the United States.

As Berggren et al. (2017) showed that more attractive-looking candidates are perceived as more conservative, we also expected that not controlling for perceived conservativeness could increase the estimated rewards for attractiveness among Republicans, but it among Democrats. This conjecture was not supported, as the estimated returns to attractiveness are similar in columns 2 and 3, and remain similar for Democrats and Republicans independently of whether perceived conservativeness is controlled for.

The effects of respondent gender and partisanship are robust with respect to researcher decisions. In Table A2, we present corresponding analysis based on self-reported party

identification, available only in 2020. If anything, the effects of respondent ideology are even stronger, at least for women. In columns 1 and 2, female Democrats are 28 percentage points more likely to vote for female candidate in a hypothetical election than female Republicans. In columns 3 and 4, the gap is still about 25 percentage points. While female Democrats are 13 percentage points more likely to support a female candidate than male Democrats, there is no statistically significant difference in support for female candidates between female and male Republicans.

In an effort to provide full transparency, we created an online simulation tool that allows readers to interact with our complete data (tinyurl.com/femalevote). For example, readers can estimate the likelihood of a female vote based on voter and candidate characteristics, and test the effects of introducing data cleaning rules to reject respondents who were identified as speeders or using different regression model specifications.

# 4.3 Insights from same-gender elections

Although our main focus is on mixed-gender elections, we also present analyses on same-gender elections, to shed further light on the use of attractiveness and perceived conservativeness as voting cues. Table A3 analyzes the probability of voting for the more attractive-looking candidate in hypothetical elections between two male candidates and table A4 in hypothetical elections between two female candidates, when partisanship is defined based on presidential voting preference. In column 1, constant presents the baseline probability that the reference person (who is male Republican with bachelor's degree who responded in 2020) votes for the more attractive-looking candidate. In column 2, we include as an additional explanatory variable the marginal effect of attractiveness gap between the more and the less attractive-looking candidates. Although the constant drops then below 0.5, more attractive-looking candidates have still clear electoral advantage, captured now partly by Attractiveness advantage. The third column again adds difference in perceived conservativeness and the fourth column political knowledge.

In our pre-registration, we listed as additional conjectures that Republicans are, on average, more likely to choose better-looking candidates than Democrats, and that Trump supporters are, on average, more likely to choose better-looking candidate than Biden supporters, at least when it comes to choices between two candidates of the same gender. The empirical results refuted these two conjectures. Instead, we find that more attractive-looking candidates have a similar advantage among Democrats and Republicans. This differs from the

finding in Berggren et al. (2017) that politicians on the right gained more from good looks in hypothetical elections in both Europe and the United States, whether using MTurk or student respondents, and also in low-information real elections, using performance in terms of the number of personal votes within party list. As Berggren et al. (2017) was written and uses data collected before Trump's presidency, which has changed the Republican Party profoundly in many ways, it is left for future research to evaluate whether the attractiveness premium has converged between Democrats and Republicans more generally since the Trump presidency.

As in mixed-gender elections, Republicans are more likely and Democrats less likely to vote for the candidate with more conservative looks. This supports the underlying idea of our theoretical model that voters use both attractiveness and perceived conservativeness as cues in their voting choices. These results are confirmed also if defining party based on self-identification, available only in 2020 (Tables A5 and A6).

Comparing Tables 3, A2, A3, A4, A5, and A6 shows that the effects of attractiveness and perceived conservativeness are quite similar in mixed-gender elections and in samegender elections, independently of other controls included. The marginal returns to attractiveness advantage in columns 2 to 4 in these six tables ranges between 0.050 and 0.076, and interaction with Democrat is always small and statistically insignificant. The marginal returns to more conservative look among Republicans range from 0.022 to 0.040, and among Democrats (after adding interaction term to the baseline estimate for Republicans) from - 0.010 to -0.033. The finding that the effects of attractiveness and perceived conservativeness are rather similar independently of candidate gender is in line with our theoretical model that treats attractiveness and perceived conservativeness independently of candidate gender.

# 4.4 Gender voting and ideological positions

We also carried out exploratory analyses on gender preference according to ideological positions in three central dimensions of contemporary American politics: support for redistribution, environmental policies, and social issues. We asked respondents whether they support increasing taxes on the rich and redistributing money to those with low incomes, whether they support increasing taxes on gasoline and using the money to protect the environment, as well as whether they support same-sex marriage or registered partnership. In each issue, Republicans were more conservative than Democrats (Fig. A9), and respondents who took liberal positions were more likely to vote for female candidates in hypothetical

elections (Fig. A10). These patterns suggest that the partisan gap in support for female candidates coincides with a gap between conservatives and liberals in economic, environmental, and social issues.

4.5 Are Republicans more confident in voting choices based on candidate looks?

Previous research has suggested that conservatives tend to rely more heavily on stereotypes than liberals (Olivola et al., 2012; Olivola et al., 2018). To study whether this is the case also in our setting, we use the feature that we collected vote choice in experimental elections in 2020 using a six-point scale, from voting definitely for the candidate on the left to voting definitely for the candidate on the right. In our pre-registration, we listed two hypotheses related to the strength of respondent choices in hypothetical elections: first that Republicans are, on average, more certain in their choice than Democrats, and second that Trump supporters are, on average, more certain in their choice than Biden supporters, at least when it comes to choices between two candidates of the same gender. In Figs A11 and A12, we present the evidence on these hypotheses, separately for mixed-gender elections, elections between two males, and elections between two females. In each case, a considerably higher share of Republicans than of Democrats makes the choice to vote definitely for the candidate on the left or vote definitely for the candidate on the right, whether Democrats and Republicans are identified based on self-reported party identification or based on their intention to vote for Biden or Trump. Therefore, also these two additional pre-registered hypotheses are confirmed.

#### **5** Conclusion

Major gender gaps have opened in American politics in recent decades. Women are more likely than men to support Democrats (Gillion et al., 2020), Democratic voters are more likely than Republicans to support female candidates (Schwarz & Coppock, 2021), and the female share of congressional Democrats is almost three times that of congressional Republicans (Fig. 1B). We carried out hypothetical elections in 2016 and 2020 to disentangle how voter gender and partisanship interact in support for female candidates. Our results show that Democrats generally favored female candidates, and that preference for female candidates was particularly strong among Democratic women. In our 2020 survey, Democratic women chose the female candidate three times as often as the male candidate. Republican respondents, instead, chose female and male candidates about equally often. Our findings suggest that voter bias against women cannot explain female underrepresentation in

American politics, even among Republicans. If anything, voters, on average, prefer women over men.

Our approach to study gender discrimination in voting complements vignette and conjoint survey experiments, which have become an established practice in political science research (Hainmueller et al., 2015; Hainmueller et al., 2014). In these studies, respondents state their preferences based on short, standardized descriptions of hypothetical candidates. Vignette and conjoint survey experiments allow studying simultaneously the effects of different cues, like gender, age, and reported experience. However, this comes at the cost that researchers define the characteristics that are presented to respondents, and how these are presented. Our approach of asking respondents to make vote choices based on candidate photographs does not require researchers to specify what textual cues are provided to respondents and in which order. Instead, we collected vote choices for hypothetical elections among all 736 Members of the European Parliament. One advantage of using MEPs was that they are real and elected politicians. Hence, the photographs likely incorporate cues that are relevant in politics, which may not be the case when using stock photographs. Another advantage of using MEPs was that American respondents are unlikely to recognize the candidates, which could have introduced bias. Finally, previous research has shown that evaluations of politicians' photographs help to predict election outcomes around the world, providing external validity for using photographs (Antonakis & Dalgas, 2009; Ballew & Todorov, 2007; Berggren et al., 2010; Lawson et al., 2010; Todorov et al., 2005).

A major concern in all surveys is that subjects might change their behavior due to cues about what constitutes appropriate behavior (Zizzo, 2010). In our setting, the concern is that respondents would find supporting female candidates in hypothetical elections the appropriate choice, even if they would not vote for the female candidate in a real election. Our study design alleviates these concerns by randomizing gender combinations in hypothetical elections. We also did not refer to gender – but only to voting under very little information – in our task description. Furthermore, recent research has found that experimenter demand effects are small in online surveys even when respondents are provided a hint on the hypothesis that researchers are testing (de Quidt et al., 2018; Mummolo & Peterson, 2019). Comparing conjoint and vignette experiments with real referendums in Switzerland also suggests that estimates from survey experiments perform remarkably well in predicting actual voting outcomes (Hainmueller et al., 2014).

Our results emphasize the critical role of supply side factors as remaining barriers to closing the gender gap in political representation, such as women's reluctance to enter politics and discrimination by party elites and donors, as well as the weight of historical female underrepresentation through incumbency advantage. Given that voters with prior exposure to female leaders are more likely to vote for women (Baskaran & Hessami, 2018; Beaman et al., 2009; Bhavnani, 2009), recent increases in the share of elected female politicians, and the election of Kamala Harris as the first female Vice President of the United States, could foreshadow a narrowing gender gap in years to come.

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**Data and materials availability:** All data, code, and materials used in the analysis will be made available via Harvard Dataverse.

# **Tables and Figures**

**Table 1.** Members of European Parliament (MEPs): descriptive statistics.

	Male (N=480)	Female (N=256)
Attractiveness	-0.17	0.33
	(0.90)	(1.10)
Perceived conservativeness	0.38	-0.71
	(0.92)	(0.71)
Age	52	50
	(11)	(10)

Table shows average standardized ratings for attractiveness and perceived conservativeness as well as age (standard deviation in parentheses) per MEP gender.

**Table 2.** Summary statistics for respondents

	Study	Study
	1	II
	2016	2020
N of respondents	293	436
Gender		
Female	49.1%	44.3%
Male	50.9%	55.7%
Age	40.3	38.5
	(sd=12.6)	(sd=12.3)
Presidential preference		
Democrat	61.8%	51.4%
Republican	33.1%	46.3%
Other	5.1%	2.3%
Party identification		
Democrat	na	50.7%
Republican	na	38.3%
In dependent	na	9.2%
Other	na	1.8%
Education		
High school orless	15.7%	7.6%
Associate's degree or some	33.4%	20.4%
college		
Bachelor's degree	36.9%	53.4%
Master's degree or higher	14.0%	18.6%
Political knowledge	1.8	2
	(sd=1.2)	(sd=1.2)

Table displays the distribution of respondents along demographic and political variables in studies I and II. Distributions of gender, presidential preference, party identification, and education are presented as column percentages, and age and political knowledge as average (standard deviation in parentheses). Distributions of presidential preference and party identification are shown excluding abstentions.

**Table 3.** OLS regression analysis on the determinants of voting for the female candidate in mixed-gender elections (Party classification based on Presidential voting intention)

	1	2	3	4
Democrat	0.146***	0.149***	0.108***	0.100***
	(0.023)	(0.023)	(0.023)	(0.022)
Female	0.064*	0.066*	0.066*	0.067*
	(0.026)	(0.026)	(0.026)	(0.026)
Democratx Female	0.076*	0.073*	0.073*	0.080*
	(0.034)	(0.034)	(0.034)	(0.033)
Age	0.002**	0.002**	0.002**	0.001
	(0.001)	(0.001)	(0.001)	(0.001)
High school or less	-0.023	-0.026	-0.026	-0.015
	(0.028)	(0.028)	(0.028)	(0.028)
Associate degree or some college	-0.011	-0.012	-0.012	0.000
	(0.022)	(0.022)	(0.022)	(0.022)
Master's degree or higher	-0.001	-0.002	-0.001	0.002
	(0.021)	(0.021)	(0.021)	(0.021)
Responded in 2016 (Study I)	-0.027	-0.027	-0.026	-0.018
	(0.018)	(0.018)	(0.018)	(0.018)
Female more attractive		0.061***	0.062***	0.063***
		(0.005)	(0.005)	(0.005)
Female more attractive x Democrat		-0.005	-0.006	-0.006
		(0.006)	(0.006)	(0.006)
Female looks more conservative			0.025***	0.025***
			(0.004)	(0.004)
Female looks more conservative x Democrat			-0.037***	-0.038***
			(0.006)	(0.006)
Female looks older			-0.000	-0.000
			(0.000)	(0.000)
Female looks older x Democrat			0.000	0.000
			(0.001)	(0.001)
Political knowledge				-0.007
				(0.011)
Political knowledge x Democrat				0.047**
				(0.014)
Constant	0.461***	0.429***	0.456***	0.449***
	(0.018)	(0.018)	(0.018)	(0.018)
R squared	0.05	0.08	0.08	80.0
Number of respondents	683	683	683	683
Number of hypothetical elections	30633	30633	30633	30633

Each observation is a hypothetical election between a female and a male candidate, excluding hypothetical elections in which the respondent abstained. Dependent variable equals 1 if the respondent chose the female candidate and 0 if the respondent chose the male candidate. Robust standard errors are clustered at the respondent level. Respondent reference categories (not shown) are Republican, Male, Bachelor's degree and Responded in 2020 (study II).

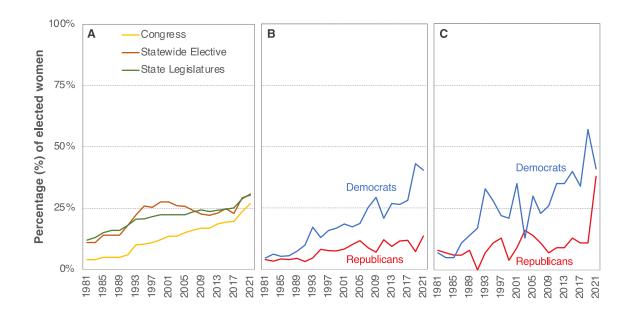


Fig. 1. Percentage of elected female politicians in the United States.

(A) In Congress, statewide elective offices (e.g., governor or lieutenant governor, attorney general, and secretary of state), and state legislatures. (B) Among Democrats and Republicans in Congress. (C) Among newly elected Democrats and Republicans in Congress. Data sources: Center for American Women and Politics, Eagleton Institute of Politics, Rutgers University; congress.gov; house.gov; senate.gov.

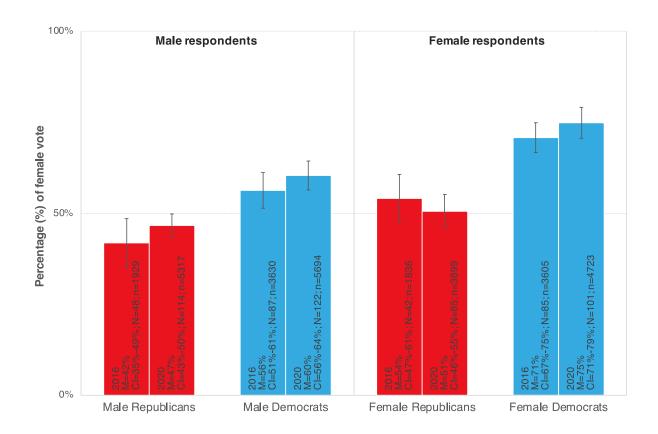


Fig. 2. Percentage of female vote by respondent gender and partisanship.

Female vote shown separately for studies I and II. Partisanship measured based on presidential preference. Results include only hypothetical elections with one female and one male candidate, excluding abstentions. N: number of respondents; n: number of hypothetical elections. Error bars depict 95% CI of the mean.

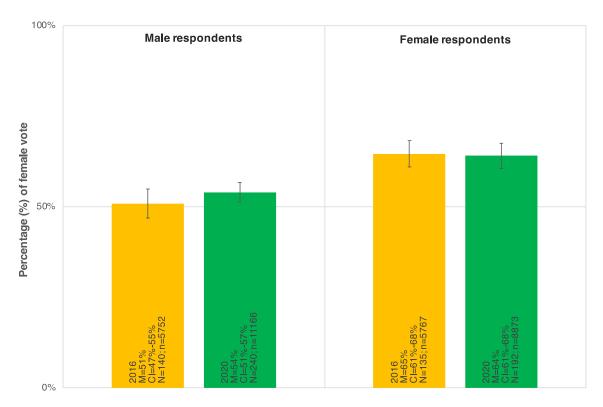


Fig. 3. Percentage of female vote by respondent gender.

Female vote shown separately for studies I and II. Results include only hypothetical elections with one female and one male candidate, excluding abstentions. N: number of respondents; n: number of hypothetical elections. Error bars depict 95% CI of the mean.

# Appendix

Table A1. Variables and their coding

MEASURE	STUDY	QUESTION(S)	ANSWER OPTIONS	CODING IN ANALYSIS
Party	II	Generally	a strong Democrat	1
dentification		speaking, do you usually	not a strong Democrat	1
		think of yourself as a	an Independent, leaning Democratic	1
		Republican, a Democrat, an	an Independent	0
		Independent, or what?	an Independent, leaning Republican	-1
			not a strong Republican	-1
			a strong Republican	-1
			Other	-9
			I am not interested in politics	-9
Democrat	I	Whom would	Hillary Clinton (Democrat)	1
		you like to see elected as the	Bernie Sanders (Democrat)	1
		next President of the United	Donald Trump (Republican)	0
		States?	Ted Cruz (Republican)	0
		Ctates:		
			John Kasich (Republican)	0
			Don't know / prefer not to answer	-1
			(Candidates rotated)	
	II	Whom would	Joe Biden (Democrat)	1
		you like to see elected as the	Donald Trump (Republican)	0
		next President of the United	Don't know / prefer not to answer	-1
		States?	(Candidates rotated)	
Female	1811	What is your	Male	0
		gender?	Female	1
			Other/Prefer not to answer	-1
Age	1&11	What is your	Free text (4-digit number), converted to	•
Age	ian	year of birth?	age and mean-centered	
Education	1&11	What is the	No schooling completed	1
Ludcation	IXII	highest degree or level of	Nursery school to 8th grade	1
				1
		school you have completed?	Some high school, no diploma	!
			High school graduate, diploma or the	1
			equivalent	2
			Some college credit, no degree	
			Trade/technical/vocational training	2
			Associate degree	2
			Bachelor's degree	3
			Master's degree	4
			Professional degree	4
			Doctorate degree	4
Knowledge	1&11	Index from 0 to 4	calculated as the number of correct answers to the four k	nowledge questions
		KQ1 to KQ4		
KQ1	I	How many	None	0
		female justices are there	One	0
		currently on the Supreme	Two	0
		Court of the United States?	Three	1
			Four	0
			Don't know	0
	11	How many male	3	0
		justices are there currently	4	0
		on the Supreme Court of the	5	0
		United States?	6	1
			7	0
		i de la companya de		
			Don't know	0
KQ2	1&11	What is the	Don't know 100	0

MEASURE	STUDY	QUESTION(S)	ANSWER OPTIONS	CODING IN ANALYSIS
		with full voting rights in the	435	1
		United States House of	501	0
		Representatives?	603	0
		'	Don't know	0
KQ3	1811	Which of the	Currently, the United States Serate has a	1 in
i i i	ian	following statements is	Republican and the House has a Democratic majority.	study II, 0 in study I
		correct? (answer options	Currently, the United States Senate and	1 in
		rotated)	**	study I, 0 in study II
		Totaled)	House both have Republican majorities.	, ,
			Currently, the United States Senate and	0
			House both have Democratic majorities.	
			Currently, the United States Senate has a	0
			Democratic and the House has a Republican majority.	
			Currently, the United States Senate is	0
			evenly divided between Democrats and Republicans,	
			while the House has a Republican majority.	
			Don't know	0
KQ4	I	Who is the	Dick Cheney	0
	·	current Secretary of	Ashton Carter	1
		Defense? (answer options	Chuck Hagel	0
		rotated)	Leon Panetta	0
				0
			John Kerry Don't know	0
		Who in the		
	II	Who is the	Mark Esper	1
		current Secretary of	Ashton Carter	0
		Defense? (answer options	Chuck Hagel	0
		rotated)	Leon Panetta	0
			John Kerry	0
			Don't know	0
Vote		For which	Person on the left	0
choice		person would you vote?	Person on the right	1
(recoded so			Abstain from voting	-9
that left refers to the			Prefer not to answer	-9
male candidate and	П	Which person	Definitely the person on the left	0
right to the female		would you vote for?	The person on the left	0
candidate)			Most probably the person on the left	0
			Most probably the person on the right	1
			The person on the right	1
			Definitely the person on the right	1
			Abstain from voting	-9
			Prefer not to answer	-9
Strength of	II	Which person	Definitely the person on the left	1
vote choice	"	would you vote for?	The person on the left	2
(recoded so			Most probably the person on the left	3
hat left refers to the			, , ,	3
male candidate and			Most probably the person on the right	-
right to the female			The person on the right	5
candidate)			Definitely the person on the right	6
oandidate)			Abstain from voting	-9
		<u> </u>	Prefer not to answer	-9
Female	Data		verage attractiveness rating of the female candidate minus	•
more attractive	obtained from prior	=	ale candidate. Individual attractiveness ratings were origin	·
	research (22)	, ,	o (very handsome or beautiful). We standardized those rational and a second state of the second state of t	•
			andardized each individual respondent's ratings to a mean	
		•	ed scores per MEP were then obtained by averaging stand	
		·	politician, and standardizing those averages. See Table S´	ror descriptive
		statistics.		
Female			verage perceived conservativeness of the female candidate	=
ooks more		perceived conservativeness	of the male candidate. Perceived conservativeness ratings	were originally
conservative			farthest to the left) to 10 (farthest to the right). We standar	

MEASURE	STUDY	QUESTION(S)	ANSWER OPTIONS	CODING IN ANALYSIS		
		follows: After excluding abster	ntions, we standardized each individual respondent	's ratings to a mean of 0 and		
		a standard deviation of unity. Standardzed scores per MEP were then obtained by averaging standardized				
		individual respondents' ratings of each politician, and standardizing those averages. See Table S1 for				
		descriptive statistics.				
Female		Defined as the ag	e of the female candidate minus the age of the mal	e candidate. Candidate age		
older		information obtained from the European Parliament's web site. See Table S1 for descriptive statistics.				

**Table A2.** OLS regression analysis on the determinants of voting for the female candidate in mixed-gender elections in 2020 (Party classification based on self-identification)

	1	2	3	4
Democrat	0.149***	0.151***	0.115***	0.115***
	(0.029)	(0.029)	(0.029)	(0.029)
Female	0.003	0.004	0.004	0.004
	(0.037)	(0.037)	(0.037)	(0.037)
Democrat x Female	0.131**	0.130**	0.130**	0.130**
	(0.046)	(0.046)	(0.046)	(0.046)
Age	0.002*	0.002*	0.002*	0.002*
	(0.001)	(0.001)	(0.001)	(0.001)
High school or less	-0.023	-0.025	-0.025	-0.025
	(0.039)	(0.038)	(0.038)	(0.038)
Associate degree or some college	-0.010	-0.011	-0.011	-0.011
	(0.032)	(0.032)	(0.032)	(0.032)
Master's degree or higher	-0.005	-0.006	-0.006	-0.006
	(0.028)	(0.028)	(0.028)	(0.028)
Female more attractive		0.050***	0.050***	0.050***
		(0.006)	(0.007)	(0.007)
Female more attractive x Democrat		-0.007	-0.006	-0.005
		(0.007)	(800.0)	(800.0)
Female looks more conservative			0.024***	0.024***
			(0.006)	(0.006)
Female looks more conservative x Democrat			-0.034***	-0.034***
			(800.0)	(800.0)
Female looks older			-0.000	-0.000
			(0.001)	(0.001)
Female looks older x Democrat			0.001	0.001
			(0.001)	(0.001)
Political knowledge				-0.004
				(0.005)
Political knowledge x Democrat				0.003
				(0.006)
Constant	0.467***	0.441***	0.467***	0.467***
	(0.022)	(0.022)	(0.022)	(0.022)
R squared	0.06	0.08	0.08	0.08
Number of respondents	383	383	383	383
Number of hypothetical elections	17880	17880	17880	17880

Each observation is a hypothetical election between a female and a male candidate in 2020, excluding hypothetical elections in which the respondent abstained. Dependent variable equals 1 if the respondent chose the female candidate and 0 if the respondent chose the male candidate. Robust standard errors are clustered at the respondent level. Respondent reference categories (not shown) are Republican, Male, and Bachelor's degree. Party identification is based on self-identification. Respondents who do not identify as Democrats or Republicans or at least leaning towards one of these parties are excluded.

**Table A3.** OLS regression analysis on the determinants of voting for the more attractive candidate in elections between two males (Party classification based on Presidential voting intention)

	1	2	3	4
Democrat	0.021	0.006	0.004	0.003
	(0.013)	(0.016)	(0.016)	(0.016)
Female	0.007	0.005	0.006	0.007
	(0.015)	(0.015)	(0.015)	(0.015)
Democrat x Female	0.002	0.004	0.002	0.001
	(0.019)	(0.019)	(0.019)	(0.019)
Age	0.001	0.001	0.001	0.001
	(0.000)	(0.000)	(0.000)	(0.000)
High school or less	0.011	0.010	0.009	0.012
	(0.018)	(0.018)	(0.018)	(0.018)
Associate degree or some college	0.015	0.018	0.017	0.019
	(0.012)	(0.012)	(0.012)	(0.012)
Master's degree or higher	0.005	0.007	0.008	0.008
	(0.013)	(0.013)	(0.013)	(0.012)
Responded in 2016 (Study I)	0.042***	0.041***	0.041***	0.041***
	(0.010)	(0.010)	(0.010)	(0.010)
Attractiveness advantage		0.065***	0.075***	0.076***
		(800.0)	(0.009)	(0.009)
Attractiveness advantage x Democrat		0.015	0.004	0.003
		(0.011)	(0.011)	(0.011)
More attractive looks more conservative			0.040***	0.040***
			(0.006)	(0.006)
More attractive looks more conservative x Democrat			-0.069***	-0.069***
			(800.0)	(800.0)
More attractive looks more older			0.001*	0.001*
			(0.000)	(0.000)
More attractive looks more older x Democrat			-0.001	-0.001
			(0.001)	(0.001)
Political knowledge				0.011*
				(0.006)
Political knowledge x Democrat				-0.011
				(800.0)
Constant	0.552***	0.486***	0.488***	0.487***
	(0.011)	(0.013)	(0.013)	(0.013)
R squared	0.00	0.02	0.03	0.03
Number of respondents	675	675	675	675
Number of hypothetical elections	14414	14414	14414	14414

Each observation is a hypothetical election between two male candidates, excluding hypothetical elections in which the respondent abstained. Dependent variable equals 1 if the respondent chose the candidate who looks more attractive and 0 if the respondent chose the candidate who looks less attractive. Robust standard errors are clustered at the respondent level. Respondent reference categories (not shown) are Republican, Bachelor's degree and Responded in 2020 (study II).

**Table A4.** OLS regression analysis on the determinants of voting for the more attractive candidate in elections between two females (Party classification based on Presidential voting intention)

	1	2	3	4
Democrat	0.000	-0.005	-0.005	-0.006
	(0.014)	(0.016)	(0.016)	(0.016)
Female	-0.038*	-0.037*	-0.037*	-0.033*
	(0.016)	(0.016)	(0.016)	(0.016)
Democrat x Female	-0.009	-0.008	-0.008	-0.011
	(0.021)	(0.021)	(0.021)	(0.021)
Age	0.000	0.000	0.000	0.000
	(0.000)	(0.000)	(0.000)	(0.000)
High school or less	0.011	0.010	0.009	0.018
	(0.020)	(0.019)	(0.019)	(0.019)
Associate degree or some college	0.024	0.025*	0.024*	0.029*
	(0.013)	(0.012)	(0.012)	(0.012)
Master's degree or higher	-0.009	-0.008	-0.008	-0.008
	(0.014)	(0.014)	(0.014)	(0.014)
Responded in 2016 (Study I)	0.049***	0.048***	0.048***	0.048***
	(0.011)	(0.011)	(0.011)	(0.011)
Attractiveness advantage		0.058***	0.058***	0.058***
		(0.007)	(800.0)	(0.008)
Attractiveness advantage x Democrat		0.004	0.004	0.003
		(0.009)	(0.010)	(0.010)
More attractive looks more conservative			0.022**	0.022**
			(0.007)	(0.007)
More attractive looks more conservative x Democrat			-0.044***	-0.044***
			(0.010)	(0.010)
More attractive looks more older			-0.000	-0.000
			(0.001)	(0.001)
More attractive looks more older x Democrat			-0.000	-0.000
			(0.001)	(0.001)
Political knowledge				0.027***
				(0.006)
Political knowledge x Democrat				-0.026**
				(800.0)
Constant				0.524***
				(0.013)
R squared	0.01	0.02	0.02	0.02
Number of respondents	681	681	681	681
Number of hypothetical elections	15379	15379	15379	15379

Each observation is a hypothetical election between two female candidates, excluding hypothetical elections in which the respondent abstained. Dependent variable equals 1 if the respondent chose the candidate who looks more attractive and 0 if the respondent chose the candidate who looks less attractive. Robust standard errors are clustered at the respondent level. Respondent reference categories (not shown) are Republican, Bachelor's degree and Responded in 2020 (study II).

**Table A5.** OLS regression analysis on the determinants of voting for the more attractive candidate in elections between two males in 2020 (Party classification based on self-identification)

	1	2	3	4
Democrat	0.011	-0.006	-0.007	-0.008
	(0.017)	(0.020)	(0.020)	(0.020)
Female	-0.001	-0.003	0.002	0.002
	(0.020)	(0.020)	(0.020)	(0.020)
Democratx Female	0.016	0.017	0.012	0.013
	(0.025)	(0.025)	(0.025)	(0.025)
Age	-0.000	-0.000	-0.000	-0.000
	(0.001)	(0.001)	(0.001)	(0.001)
High school or less	0.036	0.033	0.034	0.035
	(0.027)	(0.027)	(0.027)	(0.027)
Associate degree or some college	0.019	0.021	0.022	0.023
	(0.017)	(0.017)	(0.017)	(0.017)
Master's degree or higher	-0.011	-0.011	-0.009	-0.009
	(0.016)	(0.017)	(0.017)	(0.017)
Attractiveness advantage		0.055***	0.063***	0.063***
		(0.011)	(0.012)	(0.012)
Attractiveness advantage x Democrat		0.016	0.006	0.006
		(0.014)	(0.015)	(0.015)
More attractive looks more conservative			0.037***	0.037***
			(800.0)	(800.0)
More attractive looks more conservative x Democrat			-0.070***	-0.070***
			(0.010)	(0.010)
More attractive looks more older			0.001	0.001
			(0.001)	(0.001)
More attractive looks more older x Democrat			-0.001	-0.001
			(0.001)	(0.001)
Political knowledge				0.001
				(800.0)
Political knowledge x Democrat				0.000
				(0.011)
Constant	0.559***	0.503***	0.503***	0.503***
	(0.013)	(0.016)	(0.016)	(0.016)
R squared	0.00	0.01	0.02	0.02
Number of respondents	382	382	382	382
Number of hypothetical elections	8526	8526	8526	8526

Each observation is a hypothetical election between two male candidates in 2020, excluding hypothetical elections in which the respondent abstained. Dependent variable equals 1 if the respondent chose the candidate who looks more attractive and 0 if the respondent chose the candidate who looks less attractive. Robust standard errors are clustered at the respondent level. Respondent reference categories (not shown) are Republican and Bachelor's degree. Party identification is based on self-identification. Respondents who do not identify as Democrats or Republicans or at least leaning towards one of these parties are excluded.

**Table A6.** OLS regression analysis on the determinants of voting for the more attractive candidate in elections between two females in 2020 (Party classification based on self-identification)

	1	2	3	4
Democrat	0.012	0.005	0.004	0.006
	(0.017)	(0.021)	(0.021)	(0.020)
Female	-0.026	-0.026	-0.027	-0.026
	(0.024)	(0.024)	(0.024)	(0.023)
Democratx Female	-0.008	-0.007	-0.007	-0.008
	(0.029)	(0.029)	(0.029)	(0.028)
Age	0.000	0.000	0.000	0.000
	(0.001)	(0.001)	(0.001)	(0.001)
High school or less	0.033	0.031	0.030	0.041
	(0.031)	(0.030)	(0.030)	(0.030)
Associate degree or some college	0.039*	0.040*	0.040*	0.043*
	(0.017)	(0.017)	(0.017)	(0.017)
Master's degree or higher	-0.015	-0.014	-0.014	-0.017
	(0.019)	(0.019)	(0.020)	(0.019)
Attractiveness advantage		0.055***	0.051***	0.052***
		(0.010)	(0.010)	(0.010)
Attractiveness advantage x Democrat		0.005	0.004	0.003
		(0.012)	(0.013)	(0.013)
More attractive looks more conservative			0.035***	0.036***
			(0.009)	(0.009)
More attractive looks more conservative x Democrat			-0.064***	-0.065***
			(0.013)	(0.013)
More attractive looks more older			-0.001	-0.001
			(0.001)	(0.001)
More attractive looks more older x Democrat			-0.000	-0.000
			(0.001)	(0.001)
Political knowledge				0.029**
				(0.010)
Political knowledge x Democrat				-0.027*
				(0.012)
Constant				0.512***
				(0.016)
R squared	0.00	0.01	0.02	0.02
Number of respondents	383	383	383	383
Number of hypothetical elections	9003	9003	9003	9003

Each observation is a hypothetical election between two female candidates in 2020, excluding hypothetical elections in which the respondent abstained. Dependent variable equals 1 if the respondent chose the candidate who looks more attractive and 0 if the respondent chose the candidate who looks less attractive. Robust standard errors are clustered at the respondent level. Respondent reference categories (not shown) are Republican and Bachelor's degree. Party identification is based on self-identification. Respondents who do not identify as Democrats or Republicans or at least leaning towards one of these parties are excluded.



**Fig. A1. Experimental voting task in study I.** Participants made a binary choice between the candidate on the left and the candidate on the right. Candidates were randomly selected, and 50% of the pairs were mixed-gender races.



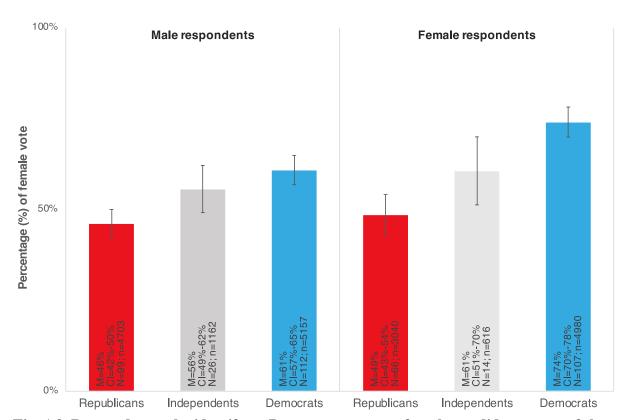


#### Which person would you vote for?

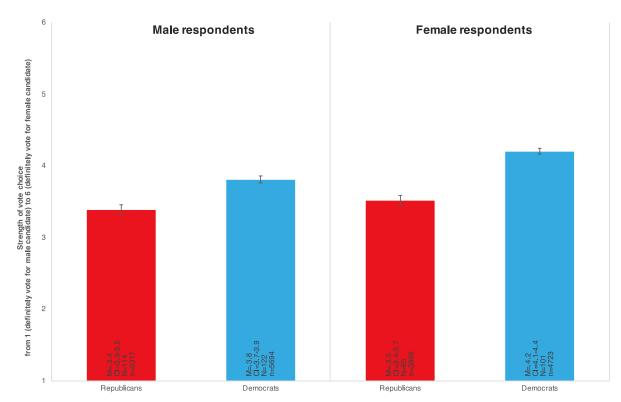
You can either click on your preferred option or use the keys on your keyboard.

[1]	[2]	[3]	[4]	[5]	[6]	[A]	[N]
Definitely	The person	Most probably	Most probably	The person	Definitely	Abstain from	Prefer not to
the person	on the left	the person	the person	on the right	the person	voting	answer
41 1-44		41 1-44	and the analysis		46		

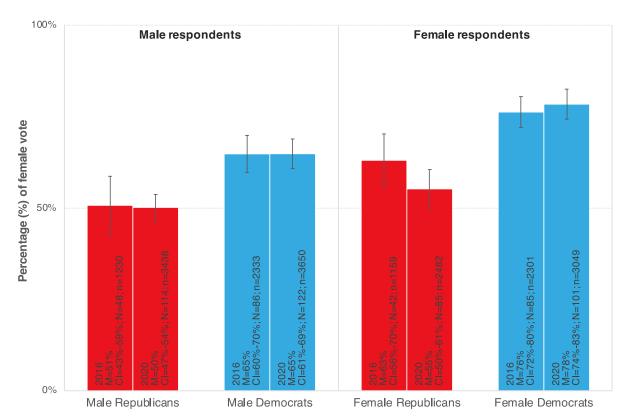
**Fig. A2. Experimental voting task in study II.** Participants chose between the candidate on the left and the candidate on the right using a 6-point scale from 1 ("definitely the person on the left") to 6 ("definitely the person on the right"). Candidates were randomly selected, and 50% of the pairs were mixed-gender races.



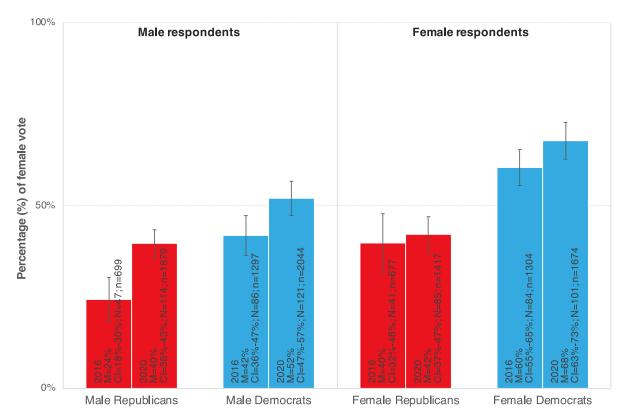
**Fig. A3.** Respondents who identify as Democrats support female candidates most of the time. The share of female candidates being elected in hypothetical elections by male and female respondents is shown according to their reported *party identification*. Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Only available in study II. Error bars depict 95% CI of the mean. N: number of respondents; n: number of hypothetical elections.



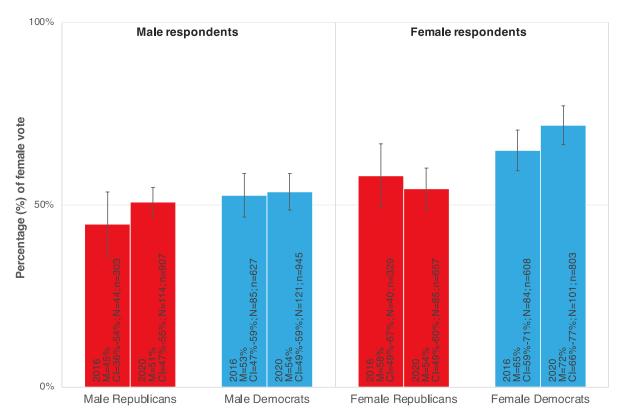
**Fig. A4. Democrats support female candidates more also when accounting for the intensity of support.** Average support for female candidates is measured using scale 1 to 6, in which 1 is maximum support for male candidate and 6 is maximum support for female candidates, and shown by respondent gender and partisanship. Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Only available in study II. Error bars depict 95% CI of the mean. N: number of respondents; n: number of hypothetical elections.



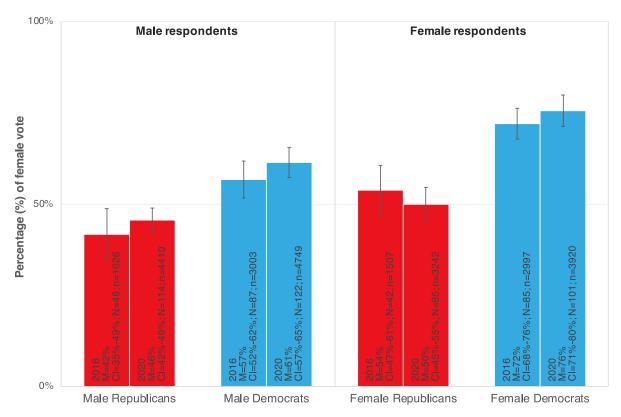
**Fig. A5. Percentage of female vote for hypothetical elections in which the female candidate looked more attractive than the male candidate**, by respondent gender and partisanship, shown separately for studies I and II. Partisanship measured based on presidential preference. Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Error bars depict 95% CI of the mean. N: number of respondents; n: number of hypothetical elections.



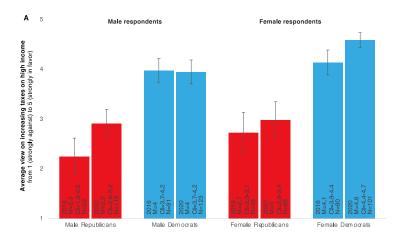
**Fig. A6.** Percentage of female vote for hypothetical elections in which the male candidate looked more attractive than the female candidate, by respondent gender and partisanship, shown separately for studies I and II. Partisanship measured based on presidential preference. Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Error bars depict 95% CI of the mean. N: number of respondents; n: number of hypothetical elections.

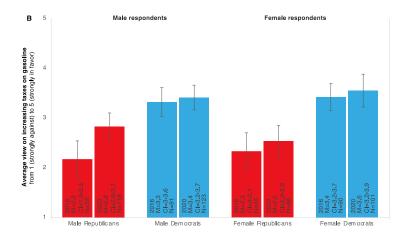


**Fig. A7. Percentage of female vote for hypothetical elections in which the female candidate looked more conservative than the male candidate**, by respondent gender and partisanship, shown separately for studies I and II. Partisanship measured based on presidential preference. Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Error bars depict 95% CI of the mean. N: number of respondents; n: number of hypothetical elections.



**Fig. A8.** Percentage of female vote for hypothetical elections in which the male candidate looked more conservative than the female candidate, by respondent gender and partisanship, shown separately for studies I and II. Partisanship measured based on presidential preference. Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Error bars depict 95% CI of the mean. N: number of respondents; n: number of hypothetical elections.





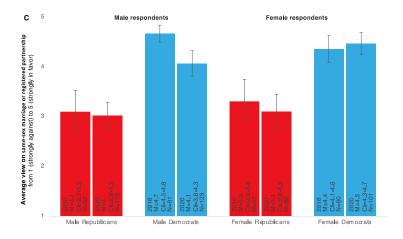


Fig. A9. Respondents' ideological views in three central dimensions of

contemporary American politics, by respondent gender and partisanship, shown separately for studies I and II. Respondents were asked whether they support (A) increasing taxes on the rich and redistributing money to those with low incomes, (B) increasing taxes on gasoline and using the money to protect the environment, as well as (C) support same-sex marriage or registered partnership. Answer scale from 1 (strongly against) to 5 (strongly in favor). Error bars depict 95% CI of the mean. N: number of respondents.

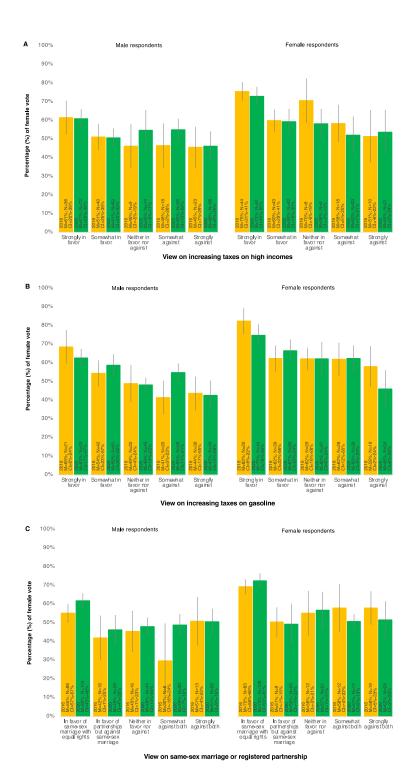
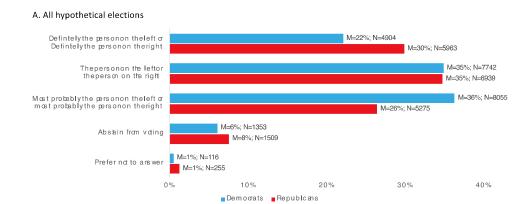
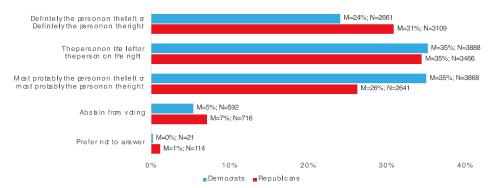


Fig. A10. Percentage of female vote depending on respondents' ideological views in three central dimensions of contemporary American politics: (A) support for redistribution, (B) environmental policies, and (C) social issues (see Fig. A11). Results shown separately for studies I (orange) and II (green). Only hypothetical elections with one female and one male candidate are included. Hypothetical elections with abstentions are excluded. Error bars depict 95% CI of the mean. N: number of respondents.

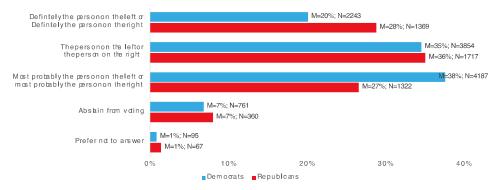
Fig. A11. Confidence in vote choice among Democrats and Republicans. (Party classification based on Presidential voting intention)



#### B. Hypothetical elections between a female and a male candidate



#### C. Hypothetical elections between two male candidates



# D. Hypothetical elections between two female candidates

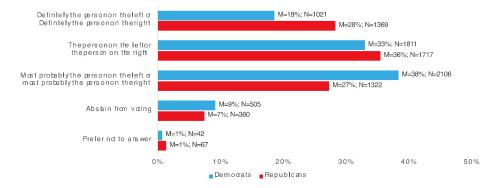
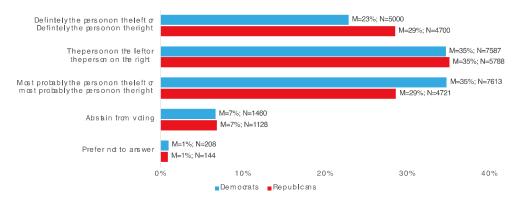
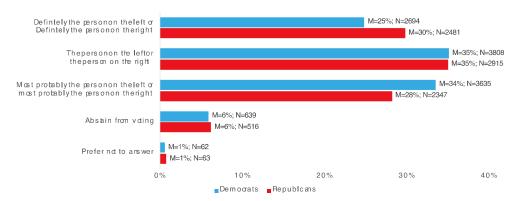


Fig. A12. Confidence in vote choice among Democrats and Republicans. (Party classification based on self-identification)

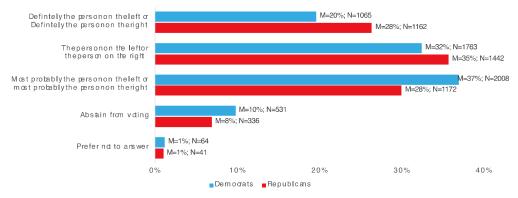
### A. All hypothetical elections



#### B. Hypothetical elections between a female and a male candidate



# C. Hypothetical elections between two male candidates



## D. Hypothetical elections between two female candidates

