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#### Immigrant overeducation across two generations: The role of gender and part-time work

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#### Abstract

A large body of literature shows that first-generation immigrants born in developing countries experience a higher likelihood of being overeducated than natives (i.e. immigrant overeducation). However, evidence is remarkably scarce when it comes to the overeducation of second-generation immigrants. Using a matched employer-employee database for Belgium over the period 1999-2016 and generalized ordered logit regressions, we contribute to the literature with one of the first studies on the intergenerational nexus between overeducation and origin among tertiary-educated workers. We show that immigrant overeducation disappears across two generations when workers work full-time. However, immigrant overeducation is a persistent intergenerational phenomenon when workers work part-time. Our gender-interacted estimates endorse these findings for female and male immigrants.

Keywords: Immigrants, intergenerational studies, labour market integration, overeducation, generalized ordered logit, moderating factors.

JEL classification: I21, I22, J15, J24, J61, J62, J71.

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#### Data availability statement

The data used in this paper are available at Statistics Belgium (STATBEL). However, restrictions may apply to the availability of these data, as confidentiality agreements and licenses must be signed with STATBEL.

The STATA do-files containing the codes for data cleaning, descriptive statistics and regressions are available upon request.

#### **Declarations of interest**

None

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#### **1. Introduction**

According to Eurostat (2023), the employment rate of foreign-born people in the European Union (EU) was 65.9% in 2021, around 5% points lower than that of native-born people (70.7%).<sup>1</sup> Difficulties in accessing the labour market also extend to the descendants of immigrants, whose employment rate was 65.3%. Zooming in on the EU employed population, several studies show that the wages of immigrants are lower than those of natives, albeit the extent of this wage gap significantly depends on immigrants' origin (e.g. Athari et al., 2019 for France; Hammarstedt and Palme, 2012 for Sweden; Pineda-Hernández et al., 2022 for Belgium). More precisely, while immigrants from developed countries earn similar wages to natives across two generations, immigrants from developing countries face persistent intergenerational wage inequalities. However, regarding the employment conditions of immigrants, intergenerational studies are scarce (e.g. Belfi et al., 2022; Belzil and Poinas, 2010), especially in the context of overeducation (i.e. the condition of having a higher level of education than that required to perform a specific job).

Moreover, it is of general interest to investigate whether origin contributes to the likelihood of being overeducated for a worker (i.e. immigrant overeducation) and whether this phenomenon holds across generations, as overeducation can have negative micro and macroeconomic consequences. In essence, overeducation can lead to: i) wage penalties and lower job satisfaction for overeducated workers because they are not paid according to their level of education, ii) lower productivity for firms due to the underutilisation of overeducated workers' skills<sup>2</sup>, higher labour turnover and an increase in absenteeism, iii) higher inequality and poverty for societies because overeducated workers may replace less-qualified workers, pushing them into low-paying occupations or unemployment; and iv) lower economic growth for countries as a result of funding non-productive education and misallocation of human capital (Brunello and Wruuck, 2019; Davia et al., 2017; McGuiness, 2006; Nielsen, 2011; Nugent, 2022).

<sup>&</sup>lt;sup>1</sup> Unless mentioned otherwise, this paper henceforth uses the words i) 'first-generation immigrants' and 'foreignborn people' for people born abroad; ii) 'second-generation immigrants', 'children of immigrants' and 'descendants of immigrants' for people born in the host country with at least one foreign-born parent; iii) 'immigrants' for first- and second-generation immigrants; and iv) 'natives' for people born in the host country with both parents born in the host country.

 $<sup>^2</sup>$  Jacobs et al. (2022a) find that although firms employing more overeducated workers should expect to pay higher wages, these workers positively impact firms' productivity. However, it is still possible to assume that overeducated workers would be more productive and earn more if they were employed in occupations where their skills would be correctly used.

The nexus between origin and overeducation can be established through different underlying mechanisms. First, human capital theory specifies that first-generation (F-G) immigrants face a disadvantageous position in the host country's labour market, as their foreign education and experience are unlikely to be perfectly transferred across borders (e.g. educational qualifications acquired in developing countries are often non-recognized in developed countries) (Basilio et al., 2017). Second, screening theory underlines the poor signal that a foreign diploma may send to employers (i.e. employers tend to undervalue schooling and language capabilities acquired in developing countries) (Chiswick and Miller, 2009; Zwysen and Demireva, 2018). Third, job search theory states that F-G immigrants may remain clustered in low-paying occupations (i.e. jobs that require a low level of education) due to their insufficient knowledge of the functioning of the host country's labour market (Akkaymak, 2017). Several empirical studies accord with these theories. In sum, they show that: i) F-G immigrants are more likely to be overeducated than natives (i.e. evidence of immigrant overeducation) (e.g. Jacobs et al., 2021; Lindley, 2009; Nielsen, 2011; Wen and Maani, 2018); ii) F-G immigrants experience a higher probability of state dependence in overeducation than natives<sup>3</sup> (e.g. Joona et al., 2014; Kalfa and Piracha, 2017); and iii) the negative effect of overeducation on earnings is more substantial for F-G immigrants than for natives (e.g. Jacobs et al., 2022b; Maani and Wen, 2021; Nielsen, 2011).

However, these underlying mechanisms are unlikely to explain the overeducation of secondgeneration (S-G) immigrants. Indeed, classical assimilation theory stipulates that since S-G immigrants are born, educated and socialised from childhood to adulthood in the host country, their likelihood of being overeducated should be lower than that of their F-G peers and similar to that of natives (Alba et al., 2011; Park and Myers, 2010). In other words, immigrant overeducation should disappear across two generations, as S-G immigrants possess human and social capital linked to the host country's labour market. However, segmented-assimilation theory sees this as a very optimistic assumption and instead suggests that S-G immigrants may still have to deal with marginalisation and discrimination due to the parental transmission of ethnic traits (e.g. religiosity, skin colour and patronymic) and social characteristics (e.g. lowincome families and concentration in immigrant-dense neighbourhoods) (Blau et al., 2013; OECD, 2017b; Phalet and Heath, 2010). OECD (2020c) further highlights that S-G immigrants may not wholly benefit from the host country's education due to school segregation and less

<sup>&</sup>lt;sup>3</sup> By state dependence in overeducation, we mean that being overeducated for a worker in the previous period increases his/her likelihood of being overeducated in the current period.

parental support in their learning process.<sup>4</sup> Therefore, differences in employment conditions between natives and S-G immigrants cannot be completely ruled out.

Furthermore, immigrant overeducation could be a persistent intergenerational phenomenon because: i) employers can make recruitment decisions based on imperfect information and ethnic stereotypes (i.e. statistical discrimination), ii) employers can have ethnic preferences for certain occupations, regardless of observed abilities and qualifications (i.e. taste-based discrimination), and iii) employers can take advantage of the barriers that immigrants face in accessing the primary labour market to hire them for jobs that do not match their level of education (e.g. monopsonistic discrimination<sup>5</sup>) (Becker, 1957; Zschirnt and Ruedin, 2016).

As far as we know, only a few empirical papers have explicitly considered S-G immigrants in the relationship between overeducation and origin. This is understandable because databases that examine this research topic rarely contain information on the country of birth of workers' parents. Using population data for Sweden in 2007 and logit regressions, Dahlstedt (2015) shows that the likelihood of being overeducated for S-G female immigrants is similar to that of female natives. In contrast, there is still an overeducation gap between S-G male immigrants and male natives. Using register data for Norway in 2014 and OLS regressions, Larsen et al. (2018) find that S-G female and male immigrants are just as likely to be overeducated as samegender natives, thus performing better than their F-G same-gender counterparts. Using the Labour Force Survey for Spain in the years 2008 and 2014 and logit regressions, Fernández-Reino et al. (2018) find that while immigrant overeducation significantly reduces across two generations among male workers, S-G female immigrants reverse the overeducation gap that their F-G same-gender peers experience. Using survey data for the Netherlands over the period 2006-2014 and multinomial logit regressions, Falcke et al. (2020) show that S-G female and male immigrants from non-Western countries (i.e. emerging and developing economies) are more likely to be overeducated than same-gender natives.

<sup>&</sup>lt;sup>4</sup> F-G immigrants tend to be less educated, less proficient in the host country's language and less informed about how the school system works, which reduces the degree of support in their children's learning.

<sup>&</sup>lt;sup>5</sup> Monopsonistic discrimination stipulates that in the context of a single buyer of the labour force, the more inelastic the labour supply, the lower wages relative to productivity (Barth and Dale-Olsen, 2009). This assumption can also be extrapolated to employment conditions. For instance, if the immigrant labour supply is more inelastic than the native labour supply, immigrants would be more likely than natives to accept low-paying jobs that do not correspond to their level of education.

Although these studies are the first to bring light to the overeducation of S-G immigrants, we believe there is room for significant improvement in the empirical strategy and data used for this research topic. First, most evidence is based on small samples or unbalanced data. Second, most studies use econometric regressions that may lead to biased estimations. For instance, binary non-linear models (e.g. logit and probit regressions) are unsuited for analysing educational mismatches, as a worker can be undereducated<sup>6</sup>, adequately educated or overeducated. Similarly, results based on OLS regressions may face under- or over-estimation issues, as these regressions impose a linear relationship between origin and overeducation and do not guarantee that every predicted probability stands between 0 and 1. Therefore, multiple-choice non-linear models (e.g. ordered or multinomial logit regressions) must be used to avoid misspecification issues. Third, none of the existing studies uses interactions between origin and gender in the regression analysis to explore a potential double penalty for female immigrants across generations (i.e. being penalised due to gender and foreign background).

Before delving into the details of our research, it is worth noting that we focus exclusively on tertiary-educated workers because they are, *per se*, individuals who can be reasonably at risk of being overeducated. Indeed, considering all individuals in a study on overeducation could lead to misleading conclusions, as workers holding at most a secondary diploma are much less likely to be overeducated. Several papers follow a similar strategy (e.g. Nielsen, 2011; Nugent, 2022; Shi et al., 2022). In addition, we also devote particular attention to immigrants from developing countries<sup>7</sup>, as a large body of literature shows that F-G immigrants born in developed countries and their S-G counterparts appear not to represent a significant integration concern for Western societies (e.g. Abramitzky et al., 2021; Algan et al., 2010; Fays et al., 2021).

Using a granular, matched employer-employee database for Belgium between 1999 and 2016, containing around 400,000 tertiary-educated workers, we aim to contribute to the literature with one of the first empirical studies on the intergenerational interplay between overeducation and origin. The novelty of our database is that it contains information on workers' country of birth and that of their parents, which allows us to identify F-G immigrants and their S-G peers

<sup>&</sup>lt;sup>6</sup> Workers are considered undereducated if their educational attainment is lower than that required for their job. Undereducation may notably result from periods of labour shortages (i.e. bottleneck vacancies) and technologically induced changes in job content and complexity.

<sup>&</sup>lt;sup>7</sup> By 'developing countries', we mean either transition and developing countries listed in the United Nations' (2020) classification and/or emerging market and developing economies listed in the IMF's (2020) classification. See Appendix 1.

correctly.<sup>8</sup> The rich detail of our database also allows us to identify overeducation cases across more than 13,500 occupation-age-sector cells using a realized matches approach. Moreover, we also have access to the 2021 Labour Force Survey for Belgium and its ad-hoc module on 'Migration and labour market'. This complementary data allows us to improve the quality of our paper using more detailed statistics. From an econometric viewpoint, as far as we know, our paper is the first to employ generalized ordered logit (GOLOGIT) regressions in the context of educational mismatches for the immigrant population. The main advantage of this econometric technique is that it relaxes the parallel regression assumption (i.e. the primary condition of an ordered logit regression) in the coefficients that do not meet it. The estimates of GOLOGIT regressions are also more parsimonious and interpretable than those of multinomial logit regressions (Williams, 2016; Williams and Quiroz, 2020).

Our empirical strategy starts with a GOLOGIT regression where, conditioning on a wide range of covariates (e.g. worker, employment and firm characteristics), we estimate the likelihood of being overeducated for immigrants from developing countries. Using a more fine-grained classification, we further explore how geographical origins<sup>9</sup> (e.g. the Maghreb, Sub-Saharan Africa and the Near and Middle East) shape the likelihood of being overeducated for immigrants from developing countries. We follow this approach as source-country characteristics at the individual level (e.g. patronymic, physical appearance, religion and cultural manners) and the macro level (e.g. economic and political stability, the quality of the education system and reasons for migration in the source country) can shape the integration of immigrants (Fleischmann and Dronkers, 2010; Levels and Dronkers, 2008). Several studies on immigrants' employment and earnings outcomes accord with this premise (e.g. Athari et al., 2019; Lindley, 2009; Pineda-Hernández et al., 2022a; Piton and Rycx, 2021).

The contributions of our paper also extend to the analysis of two moderating variables (i.e. gender and part-time work) in the intergenerational relationship between overeducation and origin. There are theoretical and empirical justifications for this additional analysis. Regarding

<sup>&</sup>lt;sup>8</sup> Our empirical analysis is consistent with that used in the vast majority of existing (also more recent) studies documenting the labour market performance of immigrants from an intergenerational perspective (e.g., Algan et al., 2010; Athari et al., 2019; Duncan and Trejo, 2018; Gueye and Ceci-Renaud, 2022; Piton and Rycx, 2021). Indeed, we identify F-G immigrants and their S-G counterparts on the basis of workers' country of birth and that of their parents. In other words, like most previous studies, our analysis does not focus on clearly identified parent-child pairs.

<sup>&</sup>lt;sup>9</sup> For the sake of accuracy in correctly classifying immigrants by geographical origin and economic development level, we constructed our geographical classification of countries based on the United Nations' (2020) classification and the IMF's (2020) classification. See Appendix 2.

the role of gender, when female workers become mothers, they may revisit their labour market trajectories or employment conditions to deal with motherhood and childcare (i.e. accepting jobs that do not often match their level of education but allow them to spend more time with their children) (Kifle et al., 2014; Petrongolo, 2019).<sup>10</sup> The European Commission (2019) further reveals that in the EU, tertiary-educated women are still underrepresented in professional or managerial occupations (i.e. potential overeducation issues). Recent papers underpin these statements as they find evidence of a gender overeducation gap in Europe (e.g. Nugent, 2022; Santiago-Vela and Mergener, 2022).

Moreover, traditional values and ethnic behaviour (e.g. family hierarchy and fertility and partner choices) can affect the working time decisions of F-G female immigrants, as they tend to be more involved in informal childcare and household production than natives (Baudin and Kondi, 2023; Ferrer and Mascella, 2022). Put differently, F-G female immigrants may find more difficulties than female natives in balancing their professional careers and ethnic identity (Blau et al., 2013; Fernández and Fogli, 2009), leading to a higher likelihood of being overeducated. There is also evidence that the influence of traditional values and ethnic norms extends to S-G female immigrants (Biegel et al., 2016; Maes et al., 2023; Noghanibehambari et al., 2022; Ng, 2023). Indeed, Jacobs et al. (2022c) point out that S-G female immigrants still face discrimination and prejudices in the workplace due to the construction of professional identities that match their migration background. However, these statements are not fully supported by the few papers that analyse the moderating role of gender in the relationship between origin and overeducation (e.g. Falckle et al., 2020; Jacobs et al., 2021; McGuinness and Byrne, 2015).

Regarding the role of part-time work, Davia et al. (2017) and Wen and Maani (2018) state that the level of education required for part-time jobs rarely corresponds to that of workers, as they may have chosen them for family reasons or personal preferences rather than career aspirations. Moreover, it could be argued that F-G immigrants may struggle to find full-time jobs due to poor human capital, imperfect transferability of qualifications or discrimination (Zschirnt and Ruedin, 2016), leading them to accept part-time jobs in a disadvantageous position (i.e. involuntary part-time work). Similarly, if employers present ethnic stereotypes or preferences for natives when recruiting full-time workers, an employment selection could also be applied

<sup>&</sup>lt;sup>10</sup> It is also very likely that some women leave the labour market because they cannot find a job that allows them to reconcile family and work.

to S-G immigrants. The findings of Fernández-Reino et al. (2018) and Green et al. (2007) go in this direction, suggesting that people with a migration background are more likely to be in a part-time job than natives. Therefore, since immigrants are more likely to be in part-time jobs, it is interesting to investigate whether immigrants working part-time also experience more educational mismatches than natives working part-time.

Last but not least, several papers show that women are more likely to work part-time than men (e.g. de Quinto et al., 2020; McIntosh et al., 2012; Piton, 2022). According to Fernández-Kranz et al. (2013) and Kifle et al. (2014), this overrepresentation of women in part-time jobs is mainly driven by mothers. Furthermore, Piton (2022) highlights that the size effect of motherhood on the likelihood of working part-time significantly depends on women's foreign background. Based on these premises, we also investigate how part-time work affects the likelihood of being overeducated for female and male immigrants from developing countries across two generations. To achieve this goal, we re-estimate our GOLOGIT regression using a three-way interaction (gender, part-time work and origin).

Our paper shows that, unlike their F-G peers, S-G immigrants from developing countries experience the same likelihood of being overeducated as natives (i.e. immigrant overeducation disappears across two generations). This finding applies to all geographical groups, with the exception of S-G immigrants from the Maghreb, who are somewhat more likely to be overeducated than natives. Our gender-interacted estimates suggest that F-G female immigrants born in developing countries face a double penalty in their likelihood of being overeducated due to their gender and migration background. In contrast, S-G female immigrants from developing countries perform on par with female natives (i.e. only evidence of a gender penalty in their likelihood of being overeducated). Moreover, part-time work is positively associated with immigrant overeducation, and this relationship largely persists across two generations. Finally, we find that immigrant overeducation is a persistent intergenerational issue when workers work part-time, irrespective of gender.

#### 2. Belgian context

Belgium is one of the developed countries with the largest immigrant population. According to the 2021 Labour Force Survey, F-G immigrants accounted for 21.4% of the total population aged 20-64 in Belgium, while S-G immigrants represented 13.3%. Among this immigrant

population, 44.7% had an EU background, while 55.3% came from non-EU countries (primarily immigrants from developing countries).<sup>11</sup> This demographic context makes Belgium an interesting case study to assess the labour market integration of immigrants across generations.

In the developed world, there is a large consensus regarding the positive role of education in boosting employment and wages (OECD, 2022). However, holding a tertiary diploma in Belgium seems more profitable for natives than for people with a migration background.<sup>12</sup> Indeed, tertiary-educated people born in developing countries experienced a sizeable immigrant-native employment gap in 2018, ranging between 17% and 30% points according to their geographical origin (FPS Employment and Unia, 2022). Moreover, several papers show that this integration issue extends to S-G immigrants (e.g. Corluy et al. 2015; De Cuyper et al. 2018). Piton and Rycx (2021) also find that F-G female immigrants born in developing countries and their same-gender descendants experience a double penalty in accessing the Belgian labour market.

Once in employment, F-G immigrants born in developing countries earn less than natives and face wage discrimination (e.g. Fays et al. 2021; Kampelmann and Rycx 2016; Grinza et al. 2020). From an intergenerational perspective, Pineda-Hernández et al. (2022) show that the wages of S-G immigrants from developing countries are similar to those of their F-G peers, being substantially lower than those of natives. The authors also find evidence of a significant double penalty in the wages of female immigrants from developing countries across two generations. However, controlling for observables at the firm level, Pineda-Hernández et al. (2022) show that the adjusted wage gap (e.g. wage discrimination) for immigrants from developing countries disappears across two generations (from 2.7% to nil). In terms of employment conditions, Jacobs et al. (2021) show that F-G immigrants born in developing countries also suggest that F-G female immigrants born in developing countries face a similar likelihood of being

<sup>&</sup>lt;sup>11</sup> It should be noted that the 2021 Labour Force Survey only allows us to identify if a worker's parent was born in an EU or non-EU country. Nevertheless, immigrants from non-EU developed countries (e.g. North America, Norway, Japan or Australia) represent less than 5% of the total immigrant population from non-EU countries in Belgium (FPS Employment and Unia, 2022). Therefore, immigrants from non-EU countries can be mainly considered as immigrants from developing countries in our statistics.

<sup>&</sup>lt;sup>12</sup> For people who studied abroad, two possible scenarios exist for getting their diplomas recognised in Belgium. First, if the diploma was awarded in a European Economic Area (EEA) country after the implementation of the Bologna process in 1999, it can be directly processed by the Belgian higher education system (i.e. a fast-track procedure). Second, if the diploma was awarded in a non-EEA country, it must be analysed by an equivalency commission (i.e. a lengthy procedure).

overeducated as their F-G male peers (i.e. no evidence of a double penalty). Nevertheless, as far as we know, intergenerational migration issues related to overeducation remain unaddressed in Belgium.

#### 3. Methodology

#### **3.1. Realized matches approach**

Given the structure of our data (i.e. a matched employer-employee database), we follow a realized matches (RM) approach (also known as a statistical approach) to measure overeducation.<sup>13</sup> The RM approach uses workers' distribution of levels of education (ISCED: 7 categories) to calculate the modal value within each occupational group (i.e. workers' educational attainment that repeats the most in an occupation) (Kiker et al., 1997; Sellami et al., 2018; Verdugo and Verdugo, 1989).<sup>14</sup> Then, the modal value is used as a reference to identify educational mismatches (i.e. undereducation or overeducation).<sup>15</sup> For instance, a taxi driver is overeducated if she holds a bachelor's degree, whereas most taxi drivers only hold a secondary diploma.

However, it should be noted that the level of education required in an occupation may vary over workers' careers (i.e. age cohort effects) (Lindley, 2009). Put differently, as young workers' skills increase over time, the probability of being overeducated for old workers mechanically increases. In this regard, Voets (2022) shows that in the EU, while overeducation decreases over time for the youngest age groups, an opposite path is observed for the oldest ones. Moreover, the possibility of occupational skill upgrading or downgrading across sectors should also be considered (i.e. sector cohort effects). For instance, while the required level of education

<sup>&</sup>lt;sup>13</sup> Two other approaches are also available in the literature to measure overeducation: i) a job analysis approach that gauges overeducation by occupation based on analysts' criteria and ii) a worker self-assessment approach that uses surveys to ask workers if they consider or not themselves in a situation of overeducation given their current educational attainment (Pérez Rodriguez et al., 2020; Turmo-Garuz et al., 2019). However, our data do not contain the necessary information to implement them.

<sup>&</sup>lt;sup>14</sup> Information on workers' educational attainment, available in 7 categories in our dataset, has been reported by firms' human resources departments (based on their registers). We converted that information into years of education, applying the following rule: i) primary education: 6 years of education; ii) lower secondary education: 9 years; iii) general, technical, and artistic upper secondary education: 12 years; iv) higher non-university education, short type (i.e. bachelor): 15 years; v) university and non-university education, long type (i.e. master): 17 years; vi) postgraduate education (i.e. advanced master): 18 years; and vii) doctoral education (i.e. PhD): 21 years. Since firms' human resources departments have provided information on workers' educational attainment, this might be somewhat underestimated for F-G immigrants who obtained their educational qualifications abroad. Therefore, the likelihood of being overeducated for F-G immigrants should be considered as a lower bound.

<sup>&</sup>lt;sup>15</sup> In an RM approach, the mean by occupation can also be used as a reference. However, the mean is very likely to be influenced by outliers.

for a consultant in the real estate sector is a bachelor's degree, a master's degree is needed for the same occupation in the banking sector. Therefore, in order to minimize a potential bias related to these cohort effects, we sort workers by occupation (ISCO classification at a threedigit level), age group (five categories)<sup>16</sup> and sector (NACE classification at a two-digit level). Then, we identify educational mismatches using the modal value of the level of education in each occupation-age-sector cell.<sup>17</sup> Our granular employer-employee database allows us to identify 13,628 occupation-age-sector cells.<sup>18</sup> Therefore, the realized matches approach of our paper is notably more precise than those implemented in previous studies on immigrant overeducation.

#### **3.2. Generalized ordered logit regressions**

Ordered logit (OLOGIT) regressions and multinomial logit (MLOGIT) regressions are the two main econometric methods used in the literature to estimate educational mismatches (e.g. the probability of being under- or over-educated for a job). The choice between these two models depends on the possibility of establishing a ranking for the ordinal dependent variable. Initially, assuming an unequivocal order for educational mismatches (1 – undereducated, 2 – adequately educated, and 3 – overeducated), OLOGIT regressions are the right econometric choice. However, OLOGIT regressions depend on the parallel regression assumption (i.e. the effect of any explanatory variable is consistent or proportional across the different categories), which is often violated (Williams and Quiroz, 2020).<sup>19</sup> An alternative solution is to leverage MLOGIT regression assumption. However, MLOGIT regressions depend on the assumption of independence of irrelevant alternatives (IIA) (i.e. the characteristics of one particular category do not impact the relative probabilities of choosing other categories), which is unlikely to hold in the analysis of educational mismatches.<sup>20</sup>

<sup>&</sup>lt;sup>16</sup> We consider the following age groups: 20-24, 25-29, 30-39, 40-49 and > 50.

<sup>&</sup>lt;sup>17</sup> When several modes are identified within an occupation-age-sector cell, the minimum mode is used as the unique reference.

<sup>&</sup>lt;sup>18</sup> The distribution of occupation-age-sector cells by size is as follows: 7.3% have less than ten observations, 38.7% have between 10 and 100 observations, 45.2% have between 101 and 1000 observations, and 8.8% have more than 1000 observations.

<sup>&</sup>lt;sup>19</sup> The null hypothesis of the Brant test for parallel regression assumption specifies that the coefficients should be the same in each cumulative logistic regression. If the null hypothesis of this test is rejected, OLOGIT estimates can be highly misleading. Using our database, we reject the null hypothesis of the Brant test. The results of the test can be obtained on request.

<sup>&</sup>lt;sup>20</sup> The null hypothesis of the Hausman test for the IIA assumption states that there is no systematic change in the coefficients if a category of the dependent variable is excluded from the model. If the null hypothesis of this test is rejected, the disturbances of the categories are not independent. Using our database, we reject the null hypothesis of the IIA assumption. The results of the test can be obtained on request.

Against this background, Williams (2016) states that "generalized ordered logit (GOLOGIT) regressions can fit estimates that are less restrictive than OLOGIT regressions, whose parallel regression assumption is often violated, but more parsimonious and interpretable that those fitted by MLOGIT regressions".<sup>21</sup> Moreover, GOLOGIT regressions present an additional advantage compared to other non-linear models. They allow the parallel regression assumption to be relaxed for estimates that do not satisfy it while imposing it on those that do (i.e. estimating partial proportional odds) (Williams, 2006). Therefore, we use GOLOGIT regressions to assess the intergenerational nexus between overeducation and origin among tertiary-educated workers in Belgium. Our benchmark GOLOGIT regression is written as follows:

$$P(Y_{it} > j) = g(\beta_j X) = \frac{\exp(\alpha_j + \sum_{k=1}^4 \beta_{jk} Origin_{itk} + \underline{z}_{it}^T \underline{\vartheta}_j + \underline{g}_{it}^T \underline{\lambda}_j + \underline{f}_{it}^T \underline{\xi}_j + \delta_{jt})}{1 + \left\{ \exp(\alpha_j + \sum_{k=1}^4 \beta_{jk} Origin_{itk} + \underline{z}_{it}^T \underline{\vartheta}_j + \underline{g}_{it}^T \underline{\lambda}_j + \underline{f}_{it}^T \underline{\xi}_j + \delta_{jt}) \right\}}$$
(1)

From equation (1), it can be established that the coefficients and fixed effects (summarised with the letter  $\phi$ ) associated with our regressors (*X*) vary for each ordered category *j*.<sup>22</sup> Therefore, the probability that a worker *i* at time *t* within an occupation-age-sector cell will be under-, adequately or over-educated is respectively equal to:

$$P(Y_{it} = 1 = undereducated) = 1 - g(\phi_1 X)$$

$$P(Y_{it} = 2 = adequately educated) = g(\phi_1 X) - g(\phi_2 X)$$

 $P(Y_{it} = 3 = overeducated) = g(\phi_2 X)$ 

As we are interested in the overeducation of workers, GOLOGIT regressions estimate the probability of being in category 3 (i.e. overeducated) compared to being in a lower category (i.e. under- or adequately educated). It should also be noted that we use the average marginal effects of GOLOGIT regressions to facilitate the interpretation of non-linear estimations.

In equation (1), our main explanatory variable is ' $Origin_{itk}$ '<sup>23</sup>, which categorizes workers into five groups *k*: i) natives, i.e. workers born in Belgium with both parents born in Belgium (the

<sup>&</sup>lt;sup>21</sup> Case-specific multinomial probit (CSMP) regressions can also deal with the issues of OLOGIT regressions and MLOGIT regressions, as they relax both the IIA and the parallel regression assumptions. However, CSMP regressions are highly time-consuming, preventing them from generating simulated maximum likelihoods in granular and large databases. Indeed , the expected optimisation performance in CSMP models was never achieved using our database.

<sup>&</sup>lt;sup>22</sup> In equation (1),  $\alpha_j$  represents the threshold parameter for each of the *j* ordered categories.

<sup>&</sup>lt;sup>23</sup> We use parental country of birth instead of parents' nationality or ethnicity to avoid ethnic attrition. Relying on nationality or ethnicity may lead to substantial classification errors (Kone, 2018).

reference group), ii) F-G immigrants born in developing countries<sup>24</sup>, iii) S-G immigrants from developing countries (i.e. workers born in Belgium with at least one parent born in a developing country)<sup>25</sup>, iv) immigrants from developed countries, excluding Belgian natives<sup>26</sup>, and v) others (i.e. workers born abroad with both parents born in Belgium)<sup>27</sup> (see Appendix 1 for a chart of developed and developing countries).

We also introduce an extensive range of covariates in our benchmark GOLOGIT regression to reduce a potential omitted variable bias. In order to facilitate the presentation of our covariates and their corresponding coefficients in equation (1), they are written as vectors in the following manner:

$$\underline{\mathbf{z}}_{it} = (\mathbf{z}_{it1}, \dots, \mathbf{z}_{itM})^{\mathrm{T}} \qquad \underline{\boldsymbol{\vartheta}}_{j} = (\boldsymbol{\vartheta}_{j1}, \dots, \boldsymbol{\vartheta}_{jM})^{\mathrm{T}}$$
$$\underline{\mathbf{g}}_{it} = (\mathbf{g}_{it1}, \dots, \mathbf{g}_{itL})^{\mathrm{T}} \qquad \underline{\boldsymbol{\lambda}}_{j} = (\lambda_{j1}, \dots, \lambda_{jL})^{\mathrm{T}}$$
$$\underline{\mathbf{f}}_{it} = (\mathbf{f}_{it1}, \dots, \mathbf{f}_{itQ})^{\mathrm{T}} \qquad \underline{\boldsymbol{\xi}}_{j} = (\xi_{j1}, \dots, \xi_{jQ})^{\mathrm{T}}$$

where  $\underline{\mathbf{z}}_{it}$  is a M x 1 vector of observations on worker characteristics (i.e. gender, tenure, squared tenure, level of tertiary education and type of household);  $\underline{\mathbf{g}}_{it}$  is a L x 1 vector of observations on employment characteristics (i.e. type of contract and dummies for part-time and overtime work);  $\underline{\mathbf{f}}_{it}$  is a Q x 1 vector that contains observations on the firm where the worker is employed (i.e. size of the firm, region where the firm is located and dummies for the existence of a firm-level collective agreement and the type of economic and financial control); and  $\delta_{jt}$  represents year fixed effects.

<sup>&</sup>lt;sup>24</sup> See footnote 7.

<sup>&</sup>lt;sup>25</sup> Regarding S-G immigrants from developing countries, it should be noted that their origin has been firstly determined by the father's country of birth, except if the father was born in Belgium and the mother in a developing country. In that case, the mother's country of birth has been retained. This is a common approach in recent empirical studies covering the labour market outcomes of immigrants across generations (e.g. Corluy et al., 2015; Jacobs et al., 2022c; Piton and Rycx, 2021).

<sup>&</sup>lt;sup>26</sup> F-G immigrants born in developed countries and their descendants present similar labour market outcomes to those of Belgian natives (Fays et al., 2021; Jacobs et al., 2021; Piton and Rycx, 2021; Pineda-Hernandez et al., 2022). Therefore, as they do not represent an intergenerational integration issue for Belgium, we merge F-G immigrants born in developed countries and their descendants into a unique group.

<sup>&</sup>lt;sup>27</sup> The category 'others' was created because workers born in developing countries with both parents born in developed countries earn more than any other group of origin (see Pineda-Hernández et al. (2022a)) and have better worker and employment characteristics than immigrants from developing countries (see Appendix 3). One might expect these workers to be the children of expatriates, who are generally highly educated and have a high socio-economic level. Consequently, considering these workers as F-G immigrants born in developing countries could lead to misleading conclusions.

In a subsequent specification, we split immigrants from developing countries into six geographical groups: i) Sub-Saharan African countries, ii) the Maghreb countries, iii) the Near and Middle Eastern countries, iv) non-EU Eastern European countries, v) emerging and developing Asian countries, and vi) Latin American and Caribbean countries (see Appendix 2 for a list of countries by geographical region). We aim to assess the effects of more fine-grained characteristics associated with workers' region of birth or that of their parents (e.g. patronymic, physical appearance, type of religion, political stability of the region and quality of higher education systems in the destination countries) on the likelihood of being overeducated for immigrants from developing countries.

Moreover, we investigate the moderating role of gender and part-time work in the intergenerational relationship between overeducation and origin in order to reveal potential immigrant-native differences that our benchmark specification may mask. To achieve this goal, we re-estimate equation (1) with an explanatory variable that varies according to origin and each moderator. Unlike gender-segregated regressions, this approach allows us to estimate potential double penalties resulting from being a woman with a foreign background or an immigrant in a part-time job. Finally, we simultaneously test the role of gender and part-time in the likelihood of being overeducated for immigrants from developing countries by reestimating equation (1) with an explanatory variable that varies according to origin, gender and part-time work (i.e. a three-way interaction).

#### 4. Data and descriptive statistics

Our empirical strategy uses a matched employer-employee database for the Belgian labour market between 1999 and 2016, provided by Statistics Belgium (STATBEL). This database was obtained by merging two datasets: the Structure of Earnings Survey (SES) and administrative data from the Belgian National Register (BNR). The SES covers all firms operating in Belgium with economic activities defined by the NACE-BEL 2008 Rev. 2 nomenclature.<sup>28</sup> Based on a sophisticated stratified sampling design, the SES provides a nationally representative sample of workers in the Belgian labour market.

<sup>&</sup>lt;sup>28</sup> The NACE-BEL 2008 Rev. 2 is the statistical classification of economic activities in the European Union (Belgian Version). More precisely, our data cover the following sectors: (B) mining and quarrying, (C) manufacturing, (D) electricity, gas, steam and air conditioning supply, (E) water supply, sewerage, waste management, and remediation, (F) construction, (G) wholesale and retail trade, repair of motor vehicles and motorcycles, (H) transportation and storage, (I) accommodation and food service activities, (J) information and

$\begin{tabular}{ c c c c c c } \hline \hline Belgium & Developing countries' \\ \hline First generation & Second generation' \\ \hline First generation & Second generation & Second generation' \\ \hline First generation & Second generation & Second generation & Second generation' \\ \hline First generation & Seco$		Sample of tertiary-educated workers born in or from:			
Belgium         First generation         Second generation*           Share of the sample (%)*         78.9         3.7         3.0           Observations         311,222         14,459         11,748           Region of birth (%)         100.0 $2220$ 14,459         11,748           Belgium         100.0 $25.1$ 29.9 $25.1$ 29.9           Near and Middle Eastern countries         25.1         29.9 $25.1$ 29.9           Near and Middle Eastern countries         11.0         9.7         Emerging and developing Asian countries $28.2$ $27.1$ $46.0$ Worker characteristics         9.8         2.7 $98.2$ $27.1$ $98.2$ $27.1$ $98.2$ $27.1$ Worker characteristics         9.8         2.7 $98.2$ $27.1$ $98.2$ $27.1$ $98.2$ $27.1$ $98.2$ $27.1$ $98.2$ $27.1$ $39.3$ $36.8$ $40.6$ $42.9$ $40.4$ $29.2$ $18.0$ $15.5$ $31.8$ $39.3$ $36.3$ $32.7$ $38.1$ $39.9$ $36.8$ $40.6$ <t< th=""><th></th><th colspan="2">Developing countries<sup>a</sup></th><th>ng countriesª</th></t<>		Developing countries <sup>a</sup>		ng countriesª	
Share of the sample (%)*         78.9         3.7         3.0           Observations         311,222         14,459         11,748           Region of birth (%)         100.0 $27.1$ 46.0           Maghreb countries         27.1         46.0           Maghreb countries         25.1         29.9           Near and Middle Eastern countries         11.0         9.7           Emerging and developing Asian countries         12.0         6.3           Latin American and Caribbean countries         9.8         2.7           Worker characteristics         9.8         2.7           Worker (%):         37.4         34.5         43.4           Age categories (%):         20-24         6.2         2.7         10.3           25-29         18.0         15.5         31.8         30-39         36.8         40.6         42.9           30-39         36.8         40.6         42.9         40-49         2.6         7         12.1         12.1           Education (%):         Bachelor         57.6         57.5         57.3         Master         39.9         Advanced Master or PhD         2.7         4.4         2.8           Other households*         1.6		Belgium	First generation	Second generation <sup>b</sup>	
Observations         311,222         14,459         11,748           Region of birth (%) Belgium         100.0         1748           Belgium         100.0         1748           Developing countries <sup>4</sup> 27.1         46.0           Maghreb countries         25.1         29.9           Near and Middle Eastern countries         15.1         5.4           Other Eastern European countries         12.0         6.3           Latin American and Caribbean countries         9.8         2.7           Worker characteristics         20.24         6.2         2.7         10.3           20-24         6.2         2.7         10.3         20.24           40.49         26.67         27.1         12.1         12.1           50+         12.4         14.0         2.9         12.1           50+         12.4         14.0         2.9         12.1           50+         12.4         14.0         2.9         12.1           50+         12.4         14.0         2.9         12.1           50+         12.4         14.0         2.9         12.1           50+         12.4         14.0         2.9         12.1	Share of the sample (%) <sup>c</sup>	78.9	3.7	3.0	
Image: 100.0           Belgium         100.0           Developing countries <sup>d</sup> 27.1         46.0           Maghreb countries         25.1         29.9           Near and Middle Eastern countries         11.0         9.7           Emerging and developing Asian countries         15.1         5.4           Other Eastern European countries         12.0         6.3           Latin American and Caribbean countries         9.8         2.7           Worker characteristics         9.8         2.7           Worker characteristics         20.24         6.2         2.7         10.3           20-24         6.2         2.7         10.3         35.3         30.3           20-39         36.8         40.6         42.9         20.4         40.49         26.7         27.1         12.1           20-4         2.7         12.4         14.0         2.9         140.49         2.9         140.49         2.6         12.4         14.0         2.9         140.49         2.6         12.4         14.0         2.9         140.49         2.6         13.8         30.3         33.3         32.7         Matter         30.3         33.3         32.7         With chithere Invin	Observations	311,222	14,459	11,748	
Belgium         100.0           Developing countries <sup>4</sup> 27.1         46.0           Maghreb countries         27.1         46.0           Maghreb countries         25.1         29.9           Near and Middle Eastern countries         15.1         5.4           Other Eastern European countries         9.8         2.7           Worker characteristics         9.8         2.7           Worker characteristics         9.8         2.7           20-24         6.2         2.7         10.3           25-29         18.0         15.5         31.8           30-39         36.8         40.6         42.9           40-49         26.7         27.1         12.1           50+         12.4         14.0         2.9           renure in years         7.8         5.2         4.3           Education (%):         39.7         38.1         39.9           Advanced Master or PhD         2.7         4.4         2.8           Household (%):         1.6         7.3         3.1           Without chidren living at home         30.3         33.3         32.7           Without chidren living at home         3.6         8.5         7.3	Region of birth (%)				
Decempting countries         27.1         46.0           Sub-Saharan African countries         25.1         29.9           Near and Middle Eastern countries         15.1         5.4           Other Eastern European countries         12.0         6.3           Lain American and Caribbean countries         9.8         2.7           Worker characteristics         9.8         2.7           Worker characteristics         2.2         2.7         10.3           20:24         6.2         2.7         10.3           20:24         6.2         2.7         10.3           25:29         18.0         15.5         31.8           30:39         36.8         40.6         42.9           40:49         26.7         2.1         12.1           50+         12.4         14.0         2.9           Tenure in years         7.8         5.2         4.3           Education (%):         3.7         33.3         32.7           Without children living at home         68.1         59.4         64.3           Household (%):         1.6         7.3         3.1           Type of contract (%):         9.9         90.6         91.9           Fixe	Belgium	100.0			
Sub-Saharan African countries         27,1         46.0           Maghreb countries         25.1         29.9           Near and Middle Eastern countries         11.0         9.7           Emerging and developing Asian countries         12.0         6.3           Latin American and Caribbean countries         9.8         2.7           Worner (%):         37.4         34.5         43.4           Age categories (%):         -         -         -           20-24         6.2         2.7         10.3           25-29         18.0         15.5         31.8           30.39         36.8         40.6         42.9           40.49         26.7         12.1         12.1           50+         12.4         14.0         2.9           Tenure in years         7.8         5.2         4.3           Education (%):         -         -         -           Bachelor         57.6         57.5         57.3           Master         30.3         33.3         32.7           With children living at home         68.1         59.4         64.3           Other households <sup>6</sup> 1.6         7.3         3.1           Permanent t	Developing countries <sup>d</sup>				
Maghreb countries       25.1       29.9         Near and Middle Eastern countries       11.0       9.7         Emerging and developing Asian countries       15.1       5.4         Other Eastern European countries       9.8       2.7         Worker characteristics       9.8       2.7         Worker characteristics       9.8       2.7         Worker characteristics       0.43.5       43.4         Age categories (%):       37.4       34.5       43.4         20-24       6.2       2.7       10.3         25.29       18.0       15.5       31.8         30-39       36.8       40.6       42.9         40-49       26.7       27.1       12.1         50+       12.4       14.0       2.9         Penure in years       7.8       5.2       4.3         Education (%):       15.1       57.5       57.3         Master       39.7       38.1       39.9         Advanced Master or PhD       2.7       4.4       2.8         Household (%):       1.6       7.3       3.1         Other bousehold (%):       1.6       7.3       3.1         Perof contract (%):       1.9	Sub-Saharan African countries		27.1	46.0	
Near and Middle Eastern countries         11.0         9.7           Emerging and developing Asian countries         12.0         6.3           Latin American and Caribbean countries         9.8         2.7           Worker characteristics         9.8         2.7           Wormen (%):         37.4         34.5         43.4           Age categories (%):         -         -         - $20-24$ 6.2         2.7         10.3           25-29         18.0         15.5         31.8           30-39         36.8         40.6         42.9 $40.49$ 26.7         27.1         12.1 $50+$ 12.4         14.0         2.9           Tenure in years         7.8         5.2         4.3           Education (%):         -         -         -           Bachelor         57.6         57.5         57.3           Master         39.7         38.1         39.9           Advanced Master or PhD         2.7         4.4         2.8           Household (%):         -         -         -           With children living at home         68.1         59.4         64.3	Maghreb countries		25.1	29.9	
Emerging and developing Asian countries       15.1       5.4         Other Eastern European countries       12.0       6.3         Latin American and Caribbean countries       9.8       2.7         Worker characteristics       9.8       2.7         Worker characteristics       20-24       6.2       2.7       10.3         25-29       18.0       15.5       31.8         30-39       36.8       40.6       42.9         40-49       26.7       27.1       12.1         50+       12.4       14.0       2.9         Tenure in years       7.8       5.2       4.3         Education (%):       39.7       38.1       39.9         Advanced Master or PhD       2.7       4.4       2.8         Household (%):            Without children living at home       30.3       33.3       32.7         With children living at home       36.6       8.5       7.3         Differenceship       0.5       1.0       0.8         Permanent       95.9       90.6       91.9         Fixed-term       3.6       8.5       7.3         Internetion wast (%) <sup>5</sup> 1.9       2.7	Near and Middle Eastern countries		11.0	9.7	
Other Eastern European countries       12.0       6.3         Latin American and Caribbean countries       9.8       2.7         Worker characteristics $37.4$ 34.5       43.4         Wormen (%):       37.4       34.5       43.4         Age categories (%): $20-24$ 6.2       2.7       10.3 $20-24$ 6.2       2.7       10.3 $25-29$ 18.0       15.5       31.8 $30.39$ 36.8       40.6       42.9 $40.49$ 26.7       27.1       12.1 $50+$ 12.4       14.0       2.9         Tenure in years       7.8       5.2       4.3         Education (%): $Wither S7.6$ 57.5       57.3         Master       39.7       38.1       39.9         Advanced Master or PhD       2.7       4.4       2.8         Household (%): $Wither Iniving at home       68.1       59.4       64.3         Other households*       1.6       7.3       3.1         Employment characteristics       7.7       4.4       2.8         Type of contract (%):       7.7 7.7 4.4 2.8 $	Emerging and developing Asian countries		15.1	5.4	
Latin American and Caribbean countries       9.8       2.7         Worker characteristics $37.4$ 34.5       43.4         Wormen (%): $37.4$ 34.5       43.4         Age categories (%): $20.24$ $6.2$ $2.7$ $10.3$ $20.24$ $6.2$ $2.7$ $10.3$ $35.2$ $31.8$ $30.39$ $36.8$ $40.6$ $42.9$ $40.49$ $26.7$ $27.1$ $12.1$ $50+$ $12.4$ $14.0$ $2.9$ $78$ $5.2$ $4.3$ Education (%): $78$ $5.2$ $4.3$ $39.9$ Advanced Master or PhD $2.7$ $4.4$ $2.8$ Household (%): $30.3$ $33.3$ $32.7$ With children living at home $68.1$ $59.4$ $64.3$ Other households <sup>4</sup> $1.6$ $7.3$ $3.1$ Employment characteristics $7.9$ $90.6$ $91.9$ $97.7$ $24.4$ $28.5$ Other households <sup>4</sup> $1.6$ $7.3$ $3.1$ $27.7$ $44.4$ $28.5$ $7.3$ $31.6$ $8.5$ <td>Other Eastern European countries</td> <td></td> <td>12.0</td> <td>6.3</td>	Other Eastern European countries		12.0	6.3	
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Worner (%): $37.4$ $34.5$ $43.4$ Age categories (%): $20.24$ $6.2$ $2.7$ $10.3$ $25.29$ $18.0$ $15.5$ $31.8$ $30.39$ $36.8$ $40.6$ $42.9$ $40.49$ $26.7$ $27.1$ $12.1$ $50+$ $12.4$ $14.0$ $2.9$ Tenure in years $7.8$ $5.2$ $4.3$ Education (%): $85.2$ $4.3$ $26.7$ $27.1$ Bachelor $57.6$ $57.5$ $57.3$ Master $39.7$ $38.1$ $39.9$ Advanced Master or PhD $2.7$ $4.4$ $2.8$ Household (%): $1.6$ $7.3$ $3.1$ <b>Employment characteristics</b> $1.6$ $7.3$ $3.1$ <b>Employment characteristics</b> $1.6$ $7.3$ $3.1$ Type of contract (%): $95.9$ $90.6$ $91.9$ Permanent $95.9$ $90.6$ $91.9$ Fixed-term $36.6$ $8.5$ $7.3$ Internship or apprenticeship $0.5$ $1.0$ $0.8$ Part-time work (%) <sup>4</sup> $4.7$ $7.7$ $4.4$ Overtime work (%) <sup>4</sup> $1.9$ $2.7$ $2.0$ Occupational categories - ISCO1 (%): $7.9$ $14.0$ $19.1$ Clerical support $23.1$ $19.4$ $23.4$ Service and sales workers $4.0$ $6.2$ $5.0$ Craft and related trades workers $2.0$ $4.2$ $1.9$ Plant and machine operators and assemblers $1.8$ $4.6$ $1.8$ Elementer Co	Worker characteristics				
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20-246.22.710.320-246.22.710.325-2918.015.531.830-3936.840.642.9 $40-49$ 26.727.112.150+12.414.02.9Tenure in years7.85.24.3Education (%):899.738.139.9Advanced Master or PhD2.74.42.8Household (%):2.74.42.8Without children living at home30.333.332.7With children living at home68.159.464.3Other households*1.67.33.1Employment characteristicsType of contract (%):95.990.691.9Pixed-term3.68.57.3Internship or apprenticeship0.51.00.8Part-time work (%) <sup>6</sup> 1.92.72.0Occupational categories - ISCO1 (%):19.92.72.0Managers12.59.08.9Professionals38.135.239.0Technicians and associate professionals17.914.019.1Clerical support23.119.423.4Service and sales workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementer of Conversion0.67.40.0	Age categories (%):	57.1	5 115	15.1	
25-29       18.0       15.5       31.8         30-39       36.8       40.6       42.9 $40-49$ 26.7       27.1       12.1         50+       12.4       14.0       2.9         Tenure in years       7.8       5.2       4.3         Education (%):	20-24	62	2.7	10.3	
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50+       12.4       14.0       2.9         Tenure in years       7.8       5.2       4.3         Education (%):       3       3       3         Bachelor       57.6       57.5       57.3         Master       39.7       38.1       39.9         Advanced Master or PhD       2.7       4.4       2.8         Household (%):	40-49	26.7	27.1	12.1	
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With children living at home Other households <sup>e</sup> 68.1       59.4       64.3         Other households <sup>e</sup> 1.6       7.3       3.1         Employment characteristics       7.3       3.1         Type of contract (%): Permanent       95.9       90.6       91.9         Fixed-term       3.6       8.5       7.3         Internship or apprenticeship       0.5       1.0       0.8         Part-time work (%) <sup>f</sup> 4.7       7.7       4.4         Overtime work (%) <sup>g</sup> 1.9       2.7       2.0         Occupational categories - ISCO1 (%): Managers       12.5       9.0       8.9         Professionals       38.1       35.2       39.0         Technicians and associate professionals       17.9       14.0       19.1         Clerical support       23.1       19.4       23.4         Service and sales workers       4.0       6.2       5.0         Craft and related trades workers       2.0       4.2       1.9         Plant and machine operators and assemblers       1.8       4.6       1.8	Without children living at home	30.3	33.3	32.7	
Other households <sup>e</sup> 1.6       7.3       3.1         Employment characteristics       Type of contract (%):       90.6       91.9         Permanent       95.9       90.6       91.9         Fixed-term       3.6       8.5       7.3         Internship or apprenticeship       0.5       1.0       0.8         Part-time work (%) <sup>f</sup> 4.7       7.7       4.4         Overtime work (%) <sup>g</sup> 1.9       2.7       2.0         Occupational categories - ISCO1 (%):       1       9.0       8.9         Professionals       38.1       35.2       39.0         Technicians and associate professionals       17.9       14.0       19.1         Clerical support       23.1       19.4       23.4         Service and sales workers       4.0       6.2       5.0         Craft and related trades workers       2.0       4.2       1.9         Plant and machine operators and assemblers       1.8       4.6       1.8         Flowmenterre       0.6       7.4       0.0	With children living at home	68.1	59.4	64.3	
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Internship or apprendices hip $0.5$ $1.0$ $0.8$ Part-time work (%) <sup>f</sup> $4.7$ $7.7$ $4.4$ Overtime work (%) <sup>g</sup> $1.9$ $2.7$ $2.0$ Occupational categories - ISCO1 (%): $12.5$ $9.0$ $8.9$ Professionals $38.1$ $35.2$ $39.0$ Technicians and associate professionals $17.9$ $14.0$ $19.1$ Clerical support $23.1$ $19.4$ $23.4$ Service and sales workers $4.0$ $6.2$ $5.0$ Craft and related trades workers $2.0$ $4.2$ $1.9$ Plant and machine operators and assemblers $1.8$ $4.6$ $1.8$	Fixed-term	5.0	8.5	1.5	
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Managers12.59.08.9Professionals38.135.239.0Technicians and associate professionals17.914.019.1Clerical support23.119.423.4Service and sales workers4.06.25.0Craft and related trades workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementory Operators0.67.40.0	Overtime work (%) <sup>e</sup>	1.9	2.7	2.0	
Managers12.59.08.9Professionals38.135.239.0Technicians and associate professionals17.914.019.1Clerical support23.119.423.4Service and sales workers4.06.25.0Craft and related trades workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementary Operators0.67.40.0	Occupational categories - ISCOI (%):	10.5	0.0	8.0	
Folessionals58.153.259.0Technicians and associate professionals17.914.019.1Clerical support23.119.423.4Service and sales workers4.06.25.0Craft and related trades workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementary Operators0.67.40.0	Managers	12.5	9.0	8.9	
Technicians and associate professionals17.914.019.1Clerical support23.119.423.4Service and sales workers4.06.25.0Craft and related trades workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementary Operators0.67.40.0	Technicians and associate professionals	30.1 17.0	55.2	59.0 10.1	
Clerical support25.119.425.4Service and sales workers4.06.25.0Craft and related trades workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementary Operators0.67.40.0	Clarical support	17.9	14.0	19.1	
Service and sales workers4.00.25.0Craft and related trades workers2.04.21.9Plant and machine operators and assemblers1.84.61.8Elementary Operators0.67.40.0	Service and sales workers	23.1	19.4	∠3.4 5.0	
Plant and machine operators and assemblers2.04.21.9Elementary Competitions1.84.61.8	Service and sales workers Craft and related trades workers	4.0	0.2	5.0	
Figure 1.8 1.8 1.8 1.8 1.8 1.8 1.8 1.8 1.8 1.8	Diant and machine operators and assemblars	2.U 1 9	4.Z	1.9	
	Flementary Occupations	1.0	4.0 7 /	1.0	

#### Table 1. Descriptive statistics by origin and generation – means and percentages

Table 1. (Continued)

communication, (K) financial and insurance activities, (L) real estate activities, (M) professional, scientific and technical activities, (N) administrative and support service activities, (P) Education, (Q) Human health and social work activities, (R) Arts, entertainment and recreation and (U) Other service activities. It should be noted that the SES provides information on public sectors (P-Q-R) every four years.

#### **Table 1. Continued**

	Sample of tertiary-educated workers born in or from:		
_		Developir	ıg countries <sup>a</sup>
	Belgium	First generation	Second generation <sup>b</sup>
Firm characteristics			
Sector of activity - NACE1 (%):			
B - Mining and Quarrying	0.1	0.1	0.1
C - Manufacturing	29.8	24.5	20.0
D - Electricity, gas, steam, and air conditioning supply	2.1	0.8	1.9
E - Water supply, sewerage, waste management and remediation activities	0.8	0.6	0.5
F - Construction	4.2	2.5	3.5
G - Wholesale and retail trade, repair of motor vehicles and motorcycles	18.5	15.7	16.7
H - Transportation and storage	5.2	6.2	5.0
I - Accommodation and food service activities	0.9	3.4	1.6
J - Information and communication	11.7	12.1	15.9
K - Financial and insurance activities	2.6	4.3	4.3
L - Real Estate activities	0.3	0.4	0.4
M - Professional, scientific, and technical activities	12.4	14.5	14.2
N - Administrative and support service activities	8.3	11.6	13.3
P - Education	0.3	0.3	0.3
Q - Human Health and social work activities	2.4	2.0	1.8
R - Arts, entertainment, and recreation	0.1	0.3	0.1
U - Other service activities	0.3	0.6	0.5
Size of the firm (FTE number of employees)	459.0	505.3	481.2
Firm-level collective agreement (Yes) (%)	28.2	27.0	26.0
More than 50% privately owned (%)	95.1	96.1	94.9
Region where the firm is located (%):			
Brussels	19.2	34.6	37.3
Flanders	64.8	45.3	38.8
Wallonia	16.1	20.1	24.0

*Notes:* Worker and firm weights are used in the calculation of means and percentages. <sup>a</sup> By 'developing countries', we mean either transition and developing countries listed in the United Nations' (2020) classification and/or emerging market and developing economies listed in the IMF's (2020) classification. <sup>b</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>c</sup> The groups 'immigrants from developed countries' and 'others' are also considered in our empirical strategy. They represent 13.3% and 1.1% of our sample, respectively. Descriptive statistics for these groups are shown in Appendix 3. <sup>d</sup> Appendix 2 shows the list of developing countries by region of birth. For the sake of accuracy in correctly classifying immigrants by geographical origin and economic development level, we construct our geographical classification of countries based on both the United Nations' (2020) classification and the IMF's (2020) classification. <sup>e</sup> 'Other households' refer to brothers/sisters living together, friends living together, students or workers' homes, etc. <sup>f</sup> A worker is recognized as a part-time employee if he/she works less than 30 hours per week. <sup>g</sup> Overtime is when an employee works more than his/her contractual working hours. Source: STATBEL, 1999-2016.

The SES also contains granular information on the structural characteristics of the firms (e.g. sector of activity, number of employees and type of collective agreement) and the demographic and employment characteristics of workers (e.g. age, gender, educational attainment, tenure, occupation and type of contract). Regarding the administrative data from the BNR, they provide reliable information on workers' country of birth, that of their parents and the type of household they live. The original sample contains information on 1,604,835 workers employed in 20,375 firms.

After calculating educational mismatches by occupation-age-sector cell (see Section 3.1 for more details), two filters were applied to the original sample. First, we restricted our sample to

tertiary-educated workers to focus exclusively on the population for which overeducation is more likely to be an issue (1,147,473 observations deleted).<sup>29</sup> Second, to avoid misclassification issues in the design of groups by origin and generation, we dropped workers born in Belgium but with missing information on the country of birth of at least one of their parents (33,826 observations deleted).<sup>30</sup> Therefore, the final sample consists of 396,462 workers employed in 15,628 firms.

Moreover, STATBEL has also given us access to the 2021 Labour Force Survey for Belgium and its ad-hoc module on '*Migration and labour market*'. This dataset contains around 22,000 observations and is representative of the working-age population in Belgium. We use descriptive statistics from this dataset as a complementary information source to better understand our results.

Table 1 shows the statistical profiles of tertiary-educated workers by origin and generation.<sup>31</sup> 78.9% of workers in our sample are Belgian natives, while F-G immigrants born in developing countries and their descendants represent 3.7% and 3%, respectively. Moreover, the groups 'immigrants from developed countries' and 'others' constitute 13.3% and 1.1% of our sample, respectively (see Appendix 3). Within the cohort of immigrants from developing countries, we observe that around 3 in 4 workers are geographically originating from Sub-Saharan Africa and the Maghreb, which accords with the total distribution of the working-age population in Belgium by geographical origin between 2008 and 2016 (FPS Employment and Unia, 2019).

Regarding worker characteristics, S-G immigrants from developing countries are younger than the other groups of workers. For instance, the proportion of S-G immigrants from developing countries aged 20-39 amounts to 42.1%, while those of natives and their F-G peers are 24.2% and 18.2%, respectively. In line with their age, S-G immigrants also have, on average, less tenure than natives and their F-G peers (4.3% vs. 5.2% and 7.8%, respectively). The shares of

<sup>&</sup>lt;sup>29</sup> Using our database, we find that within the cohort of workers with at most upper secondary education, F-G immigrants born in developing countries and their S-G peers are just as likely to be overeducated as natives. Those estimates can be obtained on request.

<sup>&</sup>lt;sup>30</sup> Around 8% of tertiary-educated workers born in Belgium were dropped because of missing information on their parental country of birth. Our estimates show that these workers face a 2% points higher likelihood of being overeducated than natives. Therefore, if we assume that a substantial part of these workers born in Belgium have at least a foreign-born parent, our findings on the overeducation gap between natives and S-G immigrants must be interpreted as a lower bound.

<sup>&</sup>lt;sup>31</sup> To shorten the term that describes tertiary-educated workers, tertiary-educated natives and tertiary-educated immigrants in the remainder of this paper, we refer to them simply as workers, natives and immigrants, respectively.

workers by tertiary education level are similar between natives and immigrants from developing countries across two generations. It should also be noted that around two-thirds of workers belong to a couple with children living at home, irrespective of origin and generation.

Regarding employment and firm characteristics, F-G immigrants born in developing countries and their S-G peers are more clustered in fixed-term contracts than natives. The share of S-G immigrants from developing countries working part-time is close to that of natives (4.4% vs. 4.7%) and two times smaller than that of their F-G peers (7.7%). We also observe that the distribution of immigrants from developing countries by occupation converges towards that of natives across two generations, except for managerial positions. Focusing on the most representative sectors of activity in Belgium, the share of immigrants from developing countries in sector C (Manufacturing) decreases across two generations, while those in sector G (wholesale and retail trade, including repair of motor vehicles and motorcycles), sector J (information and communication) and sector N (administrative and support service activities) increase. In addition, the share of immigrants from developing countries in sector M (professional, scientific and technical activities) remains stable across two generations (between 14.2% and 14.5%) and relatively close to that of natives (12.4%).

Table 2 documents the incidence of overeducation by origin and generation. Column (1) shows that the share of overeducated workers is greater within the cohort of F-G immigrants born in developing countries than within that of natives (53.7% vs. 43.5%). By contrast, S-G immigrants from developing countries reverse this gap, as their incidence of overeducation is 42.1%. Moreover, breaking down the groups by gender, there is a gender overeducation gap among natives (7.7%) and among S-G immigrants from developing countries (4.1%) (see columns (2)-(3)). In contrast, the incidence of overeducation for F-G female immigrants born in developing countries is somewhat lower than that of their F-G male peers (51.0% vs. 52.7%). Focusing on the type of employment, the share of overeducated workers is much more prominent among natives in part-time jobs than natives in full-time jobs (59.9% vs. 42.7%) (see columns (4)-(5)). Moreover, when it comes to immigrants in part-time jobs, their incidence of overeducation is much larger than that of natives in part-time jobs, irrespective of generation.

		Ger	Gender Employment		Gender and employment				
	Total			Full_time	Part_time	Full-tir	ne work	Part-tii	ne work
	Total	Men Wo	Women	nen work	work	Men	Women	Men	Women
Tertiary-educated workers born in/from:	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Belgium	43.5	40.5	48.2	42.7	59.9	40.3	46.9	55.4	61.1
	(135,295 obs.)	(77,760 obs.)	(57,535 obs.)	(126,884 obs.)	(8,411 obs.)	(76,067 obs.)	(50,817 obs.)	(1,693 obs.)	(6,718 obs.)
Developing countries <sup>a</sup>									
First generation	53.7	52.7	51.0	50.0	80.1	51.4	47.3	83.5	78.0
	(7,534 obs.)	(4,944 obs.)	(2,590 obs.)	(6,736 obs.)	(798 obs.)	(4,621 obs.)	(2,115 obs.)	(323 obs.)	(475 obs.)
Second generation	42.1	40.7	44.8	41.7	71.7	40.7	43.1	79.7	68.3
	(5,052 obs.)	(2,718 obs.)	(2,334 obs.)	(4,693 obs.)	(359 obs.)	(2,600 obs.)	(2,093 obs.)	(118 obs.)	(241 obs.)

## Table 2. Incidence of overeducation (%) by origin and generation across moderating variables

*Notes:* Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). <sup>a</sup> S-G immigrants' origin is defined based on the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. Source: STATBEL, 1999-2016.

More precisely, 80.1% of F-G immigrants born in developing countries working part-time are overeducated, and 71.7% of their S-G counterparts are overeducated.

Finally, we observe that the shares of overeducated workers in full-time jobs by gender (see columns (6)-(7)) are similar to those of the total sample by gender (see columns (2)-(3)). However, there are important gender differences in the incidence of overeducation among part-time workers. Specifically, the share of overeducated female natives in part-time jobs is larger than that of their male peers in part-time jobs (61.1% vs. 55.4%) (see columns (8)-(9)). In contrast, the share of overeducated F-G female immigrants born in developing countries working part-time is smaller than that of their male peers working part-time (78.0% vs. 83.5%). Similarly, the incidence of overeducation for S-G female immigrants from developing countries working part-time is smaller than that of their male peers working part-time (68.3% vs. 79.7%). However, it should be highlighted that in absolute terms, female workers are more representative in part-time jobs than male workers, irrespective of origin and generation (see the number of observations between parentheses in Table 2).

#### 5. Results

#### 5.1 Benchmark scenario

To investigate the intergenerational relationship between origin and overeducation, Table 3 shows the marginal effects of our benchmark GOLOGIT regression (see equation (1)). We find that the likelihood of being overeducated for a worker increases by 10.5% points if the worker was born in a developing country. In other words, F-G immigrants born in developing countries are much more likely to be overeducated than natives, whose overeducation incidence is 43.5% (see column 1 in Table 2).<sup>32</sup> We cannot ultimately assert whether this labour market disadvantage for F-G immigrants born in developing countries is completely related to their origin or unobserved individual heterogeneity (e.g. motivation, language proficiency and organization skills). However, given the granular realized matches approach used in the identification of overeducation and the large number of covariates included in our GOLOGIT regression, we feel confident to attribute this finding at least partially to a penalty related to immigrants' origin.

<sup>&</sup>lt;sup>32</sup> Our results in Table 3 also show that immigrants from developed countries are 1.3% less likely to be overeducated than natives, and the group 'others' is just as likely to be overeducated as natives.

Moreover, it should be noted that the estimates of standard econometric models go in the same direction as those of our GOLOGIT regressions, although their magnitude differs (Appendix 4). For instance, OLS estimates suggest that F-G immigrants born in developing countries face a 12.1% points higher likelihood of being overeducated than natives, whereas OLOGIT estimates quantify this penalty in 9.4%. However, these standard econometric models face significant constraints (e.g. linearity and parallel regression assumptions), which may lead to biased estimates. Therefore, our GOLOGIT regressions provide more precise estimates by dealing with these constraints.

Turning to S-G immigrants from developing countries, we find no significant relationship between their origin and their likelihood of being overeducated. Put differently, S-G immigrants from developing countries are just as likely to be overeducated as natives.<sup>33</sup> Therefore, our GOLOGIT estimates align, on average, with the classical assimilation theory, which suggests that S-G immigrants from developing countries are treated as natives in the recruitment or promotion process.

The following assumptions can explain the disappearance of immigrant overeducation across two generations: i) S-G immigrants are likely to have good proficiency in at least one of the languages used in the Belgian labour market (i.e. French, Dutch or English), as they completed their whole education in Belgium; ii) employers may consider S-G immigrants as natives since they have accumulated human capital specific to Belgium (e.g. tertiary diplomas from Belgian universities or internships in firms located in Belgium); and iii) S-G immigrants have probably built better social networks than their parents due to a process of socialization from childhood to adulthood, which can be helpful in their job search and progress along the job ladder.

<sup>&</sup>lt;sup>33</sup> In addition, standard econometric models (i.e. OLS and OLOGIT regressions) also suggest that S-G immigrants face the same likelihood of being overeducated as natives.

	Probability of being overeducated
Tertiary-educated workers born in/from:	(1)
<b>Belgium</b> (n = 311,222)	Reference
Developing countries <sup>a</sup>	
First generation $(n = 14,459)$	0.105*** (0.010)
Second generation $(n = 11,748)$	0.008 (0.007)
<b>Developed countries</b> <sup>b</sup> ( $n = 54,524$ )	-0.013*** (0.005)
<b>Others</b> <sup>c</sup> (n = 4,509)	0.008 (0.010)
Control variables	
Women	0.050*** (0.005)
Tenure	-0.002*** (0.001)
Squared tenure	0.000*** (0.000)
Level of education (ref. Bachelor)	
Master Advanced Master or PhD	-0.060*** (0.010) 0.543***
	(0.009)
Type of household (ref. without children living at home)	
With children living at home	0.009*** (0.003)
Other households	-0.014*
Type of contract (ref. permanent)	(0.008)
Fixed term	$0.104^{***}$ (0.029)
Internship	0.067** (0.028)
Part-time work	0.156*** (0.011)
Overtime work	0.145*** (0.015)
Size of the firm (FTE number of employees in log)	-0.021*** (0.003)
Firm-level collective agreement (Yes)	0.018** (0.008)
More than 50% privately owned	0.008 (0.024)
Region (ref. Brussels)	
Flanders	0.076*** (0.010) 0.084***
	(0.013)
Year fixed effects <sup>u</sup>	Yes
Occupation-age-sector cells <sup>e</sup> Observations	13,628 396 462

#### Table 3. Benchmark: average marginal effects - GOLOGIT regression

*Notes*: \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Worker and firm weights are used in the GOLOGIT regression. Robust standard errors are in parentheses, which are clustered at the firm level. <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> The group 'immigrants from developed countries' includes F-G immigrants born in developed countries and their S-G peers. <sup>c</sup> The group 'others' refers to workers born in developing countries with both parents born in Belgium (e.g. children of expatriates). <sup>d</sup> 17 year dummies. <sup>e</sup> Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). Source: STATBEL, 1999-2016.

		Probability of being overeducated
Tertiary-educated workers born in/from:		(1)
<b>Belgium</b> (n = 311,222)		Reference
Developing countries <sup>a</sup>		
Sub-Saharan African countries	First generation (n = 3,913)	0.121*** (0.015)
	Second generation ( $n = 5,403$ )	-0.020** (0.009)
Maghreb countries	First generation $(n = 3,634)$	0.180*** (0.016)
	Second generation $(n = 3,517)$	0.032** (0.015)
Near and Middle Eastern countries	First generation (n = 1,589)	0.093*** (0.023)
	Second generation (n = 1,138)	0.032 (0.021)
Emerging and developing Asian countries	First generation $(n = 2, 175)$	0.020 (0.016)
	Second generation $(n = 634)$	-0.023 (0.029)
Non-EU Eastern European countries	First generation $(n = 1,733)$	0.092*** (0.019)
	Second generation (n = 742)	0.026 (0.024)
Latin American and Caribbean countries	First generation $(n = 1,415)$	0.018 (0.021)
	Second generation (n = 314)	0.019 (0.035)
Control variables Year fixed effect <sup>b</sup>		Yes
Worker characteristics <sup>c</sup>		Yes
Employment characteristics <sup>d</sup>		Yes
Firm characteristics		res
Occupation-age-sector cells <sup>1</sup>		13,028

#### Table 4. Geographical origin: average marginal effects - GOLOGIT regression

*Notes*: \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Worker and firm weights are used in the GOLOGIT regression. Robust standard errors are in parentheses, which are clustered at the firm level. The groups 'immigrants from developed countries' and 'others' are also included in the regression, but their estimates are not portrayed in this table. <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> 17 year dummies. <sup>c</sup> Gender, level of tertiary education, tenure, squared tenure and type of household. <sup>d</sup> Type of contract and dummies for part-time and overtime work. <sup>e</sup> Size of the firm (FTE number of workers in log), dummy for more than 50% privately owned, dummy for a firm-level collective agreement and region where the firm is located (Brussels, Flanders or Wallonia). <sup>f</sup> Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). Source: STATBEL, 1999-2016.

As workers' origin does not exclusively explain their likelihood of being overeducated, we also find that the coefficients associated with our covariates are statistically significant and have the expected signs.<sup>34</sup> These complementary results further support our choice of moderating variables. Specifically, being a woman increases the likelihood of being overeducated for a worker by 5% points. Similarly, part-time workers are 15.6% points more likely to be overeducated than full-time workers.

Using a more fine-grained geographical classification in equation (1), Table 4 reports the marginal effects of our GOLOGIT estimates regarding the intergenerational relationship between geographical origin and overeducation. We find that F-G immigrants born in the Maghreb are the geographical group that experiences the highest likelihood of being overeducated (18.0% points higher than natives). To a lesser extent, F-G immigrants born in Sub-Saharan Africa are also more likely to be overeducated than natives by 12.1% points. Regarding F-G immigrants born in the Near and Middle East and non-EU Eastern Europe, their likelihood of being overeducated is around 9.0% points higher than that of natives. In contrast, F-G immigrants born in emerging and developing Asia, Latin America and the Caribbean are just as likely to be overeducated as natives.

Turning to S-G immigrants from developing countries, we find three intergenerational patterns. First, immigrant overeducation vanishes across two generations for workers from the Near and Middle East and non-EU Eastern Europe. In addition, in line with their F-G peers, S-G immigrants from emerging and developing Asia, Latin America and the Caribbean perform on a par with natives (i.e. no immigrant overeducation across two generations for these geographical groups). Second, although S-G immigrants from the Maghreb fare better than their F-G peers, their likelihood of being overeducated remains 3.2% points higher than that of natives. Third, S-G immigrants from Sub-Saharan Africa experience a 2.0% points lower probability of being overeducated than natives, thus reversing the initial penalty faced by their F-G peers. Statistically, the good performance of S-G immigrants from Sub-Saharan Africa may be explained by their important share of workers holding a master's degree (47.2%), which

<sup>&</sup>lt;sup>34</sup> More precisely, having children at home or having a non-permanent contract is positively linked to overeducation. Also, working overtime and in a firm located outside Belgium's capital (i.e. Brussels) increases the likelihood of being overeducated. By contrast, the size of the firm is negatively associated with overeducation. Similarly, tenure reduces the likelihood of being overeducated, albeit up to a certain level, because their relationship is quadratic. Regarding the type of tertiary education, a worker holding a master's degree is less likely to be overeducated than a worker holding a bachelor's degree. However, holding advanced master's or PhD's degrees substantially increases the likelihood of being overeducated.

is even higher than that of natives (39.7%) (see Appendix 5). Indeed, the probability of being overeducated is lower for workers holding a master's degree than a bachelor's degree (see the estimates associated with education in Table 3).

#### 5.2 Gender and immigrant overeducation

In this Subsection, we investigate the moderating effect of gender on the intergenerational nexus between origin and overeducation using equation (2). Results are displayed in Table 5. Our gender-interacted estimates suggest that while F-G male immigrants born in developing countries experience a 14.5% points higher likelihood of being overeducated than male natives, S-G male immigrants from developing countries are just as likely to be overeducated as male natives. These estimates go in the same direction as those of our benchmark GOLOGIT regression in Table 3, where immigrant overeducation also disappears across two generations.

However, when it comes to female workers, the findings are more nuanced. More precisely, our gender-interacted estimates suggest that female natives experience a 5.6% points higher likelihood of being overeducated than male natives (i.e. evidence of a gender overeducation gap). We also find that F-G female immigrants born in developing countries are 5.5% points more likely to be overeducated than female natives (i.e. F-G female immigrants born in developing countries experience an 11.1% points higher likelihood of being overeducated than male natives). In addition, the coefficient associated with F-G female immigrants born in developing countries is statistically different from that associated with female natives (see the test for equality of coefficients at the bottom of Table 5). Thus, there are solid grounds for suggesting that F-G female immigrants born in developing countries undergo a double penalty in their likelihood of being overeducated due to their gender and foreign background.

Regarding S-G female immigrants from developing countries, our gender-interacted estimates suggest that they no longer face a penalty in their likelihood of being overeducated due to their origin. More precisely, S-G female immigrants from developing countries are 1.8% points less likely to be overeducated than female natives (i.e. S-G female immigrants from developing countries undergo a 3.8% points higher likelihood of being overeducated than male natives). However, although S-G female immigrants from developing countries seem to perform better than female natives, it is worth noting that their coefficient associated with their origin is not statistically different from that for female natives (see the test for equality of coefficients at the

bottom of Table 5). Therefore, S-G female immigrants from developing countries are just as likely to be overeducated as female natives (i.e. only evidence of a gender penalty in their likelihood of being overeducated).

		Probability of being overeducated
Tertiary-educate	ed workers born in/from:	(1)
Belgium		
Men (n = 191	,913)	Reference
Women (n =	119,309) [1]	0.056*** (0.005)
Developing cou	intries <sup>a</sup>	
Men	First generation $(n = 9,382)$	0.145*** (0.012)
	Second generation $(n = 6,541)$	0.015 (0.010)
Women	First generation $(n = 5,077)$ [2]	0.111*** (0.012)
	Second generation (n = 5,207) [3]	0.038*** (0.013)
Control variable	20	
Year fixed ef	fects <sup>b</sup>	Yes
Worker chara	acteristics <sup>c</sup>	Yes
Employment	characteristics <sup>d</sup>	Yes
Firm characte	eristics <sup>e</sup>	Yes
Test for equality	y of coefficients (p-value) <sup>f</sup>	
[1] = [2]	, or econnecting (p ( and c)	0.00
[1] = [3]		0.12
[2] = [3]		0.00
Occupation-sec	tor-age cells <sup>g</sup>	13,628
Observations		396,462

#### Table 5. Gender and origin: average marginal effects - GOLOGIT regression

*Notes*: \*\*\* p < 0.01, \*\* p < 0.05, \* p < 0.1. Worker and firm weights are used in the GOLOGIT regression. Robust standard errors are in parentheses, which are clustered at the firm level. The groups 'immigrants from developed countries' and 'others' are also included in the regression, but their gender-interacted estimates are not portrayed in this table (available on request). <sup>a</sup> Secondgeneration immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> 17 year dummies. <sup>c</sup> Level of tertiary education, tenure, squared tenure and type of household. <sup>d</sup> Type of contract and dummies for part-time and overtime work. <sup>c</sup> Size of the firm (FTE number of workers in log), dummy for more than 50% privately owned, dummy for firm-level collective agreement and region where the firm is located (Brussels, Flanders or Wallonia). <sup>f</sup> The null hypothesis of the test specifies that the estimates are not statistically different from each other if the p-value is higher than 0.10. <sup>g</sup> Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). Source: STATBEL, 1999-2016.

#### 5.3 Part-time work and immigrant overeducation

Column (5) in Table 2 shows that in our sample, about 60% of natives in part-time jobs are overeducated. In addition, this incidence is higher if the worker has a foreign background (80.1% for F-G immigrants born in developing countries and 71.7% for their S-G peers). Thus,

we leverage equation (3) to study the moderating role of part-time work in the intergenerational nexus between origin and overeducation. Results are shown in Table 6. We find that within the cohort of full-time workers, F-G immigrants born in developing countries are 9.4% points more likely to be overeducated than natives. In contrast, S-G immigrants from developing countries working full-time perform similarly to natives working full-time in terms of overeducation. These findings largely mirror those of our benchmark GOLOGIT regression in Table 3.

Moreover, our estimates suggest that natives working part-time face a 14.1% points higher likelihood of being overeducated than natives working full-time. However, it should be noted that working part-time affects more immigrants than natives in terms of overeducation. More precisely, F-G immigrants born in developing countries working part-time are 25.7% points more likely to be overeducated than natives working part-time (i.e. F-G immigrants born in developing countries working part-time face a 39.8% higher likelihood of being overeducated than natives working full-time). Although this overeducation gap between natives and immigrants reduces across two generations, it remains sizeable. In other words, S-G immigrants from developing countries working part-time are 12.4% points more likely to be overeducated than natives working part-time (i.e. S-G immigrants from developing countries working full-time).

However, caution must be exercised when interpreting these results. Indeed, the relationship between part-time work and overeducation might suffer from endogeneity (e.g. reverse causality). Although 2SLS regressions could tackle this econometric issue, it remains challenging to find valid instruments, namely variables that are both relevant (i.e. goods predictors of part-time work) and exogenous (i.e. uncorrelated with the error of being overeducated). In addition, our potential endogenous variable is represented by nine categories in the regression (i.e. the interactions between origin and part-time work), which implies the search of at least nine instruments to guarantee identification in 2SLS regressions (i.e. having as many instruments as endogenous variables to satisfy one of 2SLS regressions' conditions). Accordingly, although our estimates provide solid evidence of a relationship between part-time work and immigrant overeducation, they should not be interpreted as causal.

		Probability of being overeducated
Tertiary-educated workers born in/from:		(1)
Belgium		
Full-time work ( $n = 297, 169$ )		Reference
Part-time work $(n = 14,053)$		0.141***
		(0.011)
Developing countries <sup>a</sup>		
Full-time work	First generation $(n = 13,463)$	0.094*** (0.010)
	Second generation ( $n = 11,247$ )	-0.001 (0.008)
Part-time work	First generation $(n = 996)$	0.398*** (0.022)
	Second generation (n = 501)	0.265*** (0.033)
Control variables		
Year fixed effects <sup>b</sup> Worker characteristics <sup>c</sup> Employment characteristics <sup>d</sup> Firm characteristics <sup>6</sup>		Yes Yes Yes
Occupation-sector-age cells <sup>f</sup> Observations		13,628 396,462

#### Table 6. Part-time work and origin: average marginal effects - GOLOGIT regression

*Notes:* \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Worker and firm weights are used in the GOLOGIT regression. Robust standard errors are in parentheses, which are clustered at the firm level. The groups 'immigrants from developed countries' and 'others' are also included in the regression, but their estimates associated with full- or part-time work are not portrayed in this table (available on request). <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> 17 year dummies. <sup>c</sup> Gender, level of tertiary education, tenure, squared tenure and type of household. <sup>d</sup> Type of contract and a dummy for overtime work. <sup>e</sup> Size of the firm (FTE number of workers in log), dummy for more than 50% privately owned, dummy for a firm-level collective agreement and region where the firm is located (Brussels, Flanders or Wallonia). <sup>f</sup> Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). Source: STATBEL, 1999-2016.

#### 5.4 Three-way interaction: origin, gender and part-time work

The literature suggests that part-time work is mainly an issue for women because they are more likely to work part-time due to motherhood and home production (e.g. Kifle et al., 2014; Piton, 2022). Our LOGIT estimates accord with this statement, as they suggest that the likelihood of working part-time for female workers significantly increases when they have children at home: by 7.1% points for female natives, by 7.2% points for F-G female immigrants born in developing countries and by 3.9% points for S-G female immigrants from developing countries (see Appendix 6). In contrast, we find no significant effect of parenthood on the likelihood of working part-time for male workers, irrespective of origin and generation. These results remain largely stable after controlling for a large set of covariates. Accordingly, we use equation (4) to

implement a three-way interaction (origin, part-time work and gender) in our GOLOGIT regression. Results are presented in Table 7.

Within the cohort of full-time female and male workers, our findings largely align with the gender-interacted estimates in Table 5. However, focusing on part-time workers, our three-way interaction reveals interesting new outcomes. First, we find that female natives working part-time are 3.3% points more likely to be overeducated than their male peers working part-time, who already face a 15.7% points higher likelihood of being overeducated than male natives working full-time.

Second, being born in a developing country skyrockets the likelihood of being overeducated for a part-time worker, regardless of gender. More precisely, F-G male immigrants born in developing countries working part-time are 29.2% points more likely to be overeducated than male natives working part-time (i.e. F-G male immigrants born in developing countries working full-time). Similarly, F-G female immigrants born in developing countries working part-time are 22.3% points more likely to be overeducated than female natives working part-time (i.e. F-G female immigrants born in developing countries working part-time are 22.3% points more likely to be overeducated than female natives working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working part-time (i.e. F-G female immigrants born in developing countries working full-time). In this regard, it should be noted that there is no significant difference between the likelihood of being overeducated for F-G male immigrants born in developing countries and that of their F-G female peers (see the test for equality of coefficients at the bottom of Table 7).

Third, immigrant overeducation is a persistent intergenerational phenomenon within the cohort of part-time female and male workers. More precisely, S-G male immigrants from developing countries working part-time are 21.6% points more likely to be overeducated than male natives working part-time (i.e. S-G male immigrants from developing countries working part-time undergo a 37.3% higher likelihood of being overeducated than male natives working full-time). Similarly, S-G female immigrants from developing countries working part-time are 8% points more likely to be overeducated than female natives working part-time (i.e. S-G female immigrants from developing countries working part-time are 8% points more likely to be overeducated than female natives working part-time (i.e. S-G female immigrants from developing countries working part-time (i.e. S-G female immigrants from developing countries working part-time (i.e. S-G female immigrants from developing countries working part-time (i.e. S-G female immigrants from developing countries working part-time (i.e. S-G female immigrants from developing countries working part-time (i.e. S-G female immigrants from developing countries working part-time face a 27% higher likelihood of being overeducated than female natives working full-time). In addition, the test for equality of coefficients at the bottom of Table 7 shows that S-G male immigrants from developing

countries working part-time fare worse than their S-G female peers working part-time in terms of overeducation.

The findings associated with our three-way interaction are in line with the literature, which suggests that family duties (e.g. informal childcare and home production) and ethnic norms (e.g. fertility rate and family hierarchy) affect the labour market expectations of female immigrants more than those of female natives (Blau et al., 2013; Fernández and Fogli, 2009; Jacobs et al. 2022c). However, it seems less clear that male immigrants revisit the labour market trajectories due to parenthood or ethnic identities (Nadim and Midtbøen, 2023). Therefore, other underlying mechanisms linking part-time work, origin and overeducation among male workers need to be considered. For instance, an excess supply of labour from male immigrants, which could foster employers' incentive to offer them jobs below their level of education or generate additional barriers preventing them from finding full-time jobs (i.e. jobs where the likelihood of being overeducated is less substantial).

Incidentally, it should be noted that descriptive statistics of the 2021 ad-hoc module of the Labour Force Survey (LFS) concerning migration in Belgium seem to be supportive of this interpretation. Figure 1 shows indeed that in 2021, the labour market participation rate of F-G male immigrants born in developing countries was more than 20% points higher than that of their F-G female peers. Moreover, this gender gap appears to be intergenerational persistent among immigrants (i.e. the gender gap in participation rates of S-G immigrants from developing countries is still as high as 14% points).

Next, Figure 2 reveals that in 2021, the involuntary part-time employment rate of F-G male immigrants born in developing countries was substantially higher than that of natives (57.4% vs. 15.4%). This employment issue extends to S-G male immigrants from developing countries, whose involuntary part-time employment rate was 45.6%. However, when it comes to female immigrants, involuntary part-time work seems less problematic. Indeed, 20% of F-G female immigrants born in developing countries and 21.7% of their S-G female peers declared themselves in an involuntary part-time job. Thus, the additional difficulties immigrants face to find full-time jobs are likely to explain their extremely high likelihood of being overeducated when working part-time.

born in/from:			
			(1)
Belgium			
	Full-time jobs (n =	= 188,859)	Reference
Men	Part-time jobs (n =	= 3,054)	0.157*** (0.011)
	Full-time jobs (n =	= 108,310)	0.054***
Women	Part-time jobs (n =	= 10,999)	0.190*** (0.006)
Developing countries <sup>a</sup>			
		First generation $(n = 8,995)$	0.128*** (0.007)
	Full-time jobs	Second generation (n = 6,393)	0.018** (0.007)
Men		First generation (n = 387) [1]	0.449*** (0.024)
	Part-time jobs	Second generation (n = 148) [2]	0.373*** (0.048)
		First generation $(n = 4,468)$	0.080*** (0.010)
	Full-time jobs	Second generation (n = 4,854)	0.028*** (0.010)
Women		First generation $(n = 609)$ [3]	0.413*** (0.018)
	Part-time jobs	Second generation (n = 353) [4]	0.270*** (0.035)
Year fixed effects <sup>b</sup> Worker characteristic Employment character	eristics <sup>d</sup>		Yes Yes Yes
Firm characteristicse			Yes
Test for equality of coer [1] = [3] [2] = [4]	fficients (p-value) <sup>f</sup>		0.23
Occupation-sector-age Observations	cells <sup>g</sup>		13,628 396,462

#### Table 7. Gender, part-time work and origin: average marginal effects - GOLOGIT regression

Probability of being overeducated

Tertiary-educated workers

*Notes:* \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Worker and firm weights are used in the GOLOGIT regression. Robust standard errors are in parentheses, which are clustered at the firm level. The groups 'immigrants from developed countries' and 'others' are also included in the regression, but their gender-interacted estimates associated with full- or part-time work are not portrayed in this table (available on request). <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> 17 year dummies. <sup>c</sup> level of tertiary education, tenure, squared tenure and type of household. <sup>d</sup> type of contract and a dummy for overtime work. <sup>e</sup> Size of the firm (FTE number of workers in log), dummy for more than 50% privately owned, dummy for a firm-level collective agreement and region where the firm is located (Brussels, Flanders or Wallonia). <sup>f</sup> The null hypothesis of the test specifies that the estimates are not statistically different from each other if the p-value is higher than 0.10. <sup>g</sup> Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). Source: STATBEL, 1999-2016.



#### Figure 1. Labour market participation rate in 2021 among people aged 20-64 in Belgium.

*Notes:* The labour market participation rate is calculated as the share of employed and unemployed persons aged 20-64 in the total population (active and inactive persons) of that same age. <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. Source: 2021 Labour Force Survey, STATBEL.

# Figure 2. Involuntary part-time employment rate in 2021 among part-time workers aged 20-64 in Belgium.



*Notes:* The involuntary part-time employment rate is defined as the percentage of part-time workers (i.e. workers who work less than 30h per week) who want to work more hours and are available in the next two weeks at the time of the survey. <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. Source: 2021 Labour Force Survey, STATBEL.

Finally, we cannot exclude an employment selection according to female and male workers' origin (i.e. taste-based, statistical or monopsonistic discrimination) as a potential channel to explain the intergenerational connection between part-time work and immigrant overeducation. For instance, it might be possible that when hiring tertiary-educated workers, some employers have preferences for natives or negative stereotypes against immigrants.

#### 6. Conclusion

In the OECD area, the population with tertiary education increased from 26.1% in 2000 to 47.1% in 2020 (OECD, 2022). However, although educational expansion accords with the growing demand for tertiary-educated people in developed countries, it could also lead to many overeducation cases (e.g. an economist employed as a cashier in a supermarket) (Green and Henseke, 2016). Indeed, overeducation has become a persistent social and economic phenomenon in the developed world (McGuinness et al., 2018; Nugent, 2022). Moreover, the literature shows that overeducation rates are higher in developed countries with more immigrant labour because F-G immigrants are more likely to be overeducated than natives (e.g. Davia et al., 2017; Jacobs et al., 2020; Wen and Maani, 2018). However, the evolution of immigrant overeducation across generations remains largely unexplored. Put differently, very little is known about the overeducation of S-G immigrants (e.g. Falcke et al., 2020; Fernández-Reino et al., 2018). Therefore, we leverage rich employer-employee data, covering almost two decades (1999-2016), a granular realized matches approach and GOLOGIT regressions to investigate the intergenerational interplay between origin and overeducation among tertiary-educated workers in Belgium.

After accounting for differences in worker, employment and firm characteristics (e.g. gender, tenure, level of tertiary education, type of contract, overtime, firm size and firm-level collective agreements), our GOLOGIT estimates suggest that F-G immigrants born in developing countries are substantially more likely to be overeducated than natives, who face an important likelihood of being overeducated (43.5%). However, when it comes to S-G immigrants from developing countries, their likelihood of being overeducated is not statistically different from that of natives. In other words, immigrant overeducation disappears across two generations, thus illustrating the positive effect of being born, studying and socialising from childhood to adulthood in the host country (i.e. the classical assimilation theory). Our findings also hold for all geographical groups of immigrants, with the exception of workers from the Maghreb. More

precisely, although S-G immigrants from the Maghreb perform much better than their F-G peers, their likelihood of being overeducated remains somewhat higher than that of natives.

Moreover, our gender-interacted estimates suggest that F-G female immigrants born in developing countries experience a double penalty in their likelihood of being overeducated due to their gender and migration background. In contrast, S-G female immigrants from developing countries perform similarly to female natives, who experience a higher likelihood of being overeducated than male natives (i.e. a gender overeducation gap). Put differently, our gender-interacted estimates show that S-G female immigrants from developing countries only face a gender penalty in their likelihood of being overeducated.

However, it should be noted that all these findings only apply to the cohort of full-time female and male workers. Indeed, our GOLOGIT estimates show that part-time work makes immigrant overeducation intergenerationally persistent, in line with the segmented assimilation theory. More precisely, although S-G female and male immigrants from developing countries working part-time are less likely to be overeducated than their F-G same-gender peers, their likelihood of being overeducated is much higher than that of same-gender natives working part-time. The following underlying channels are likely to explain the nexus between part-time work and immigrant overeducation: i) the intergenerational transmission of ethnic and gender norms among female immigrants (e.g. fertility choices, informal childcare and family hierarchy); ii) the high incidence of involuntary part-time work among male immigrants from developing countries; and iii) an employment selection based on workers' origin (i.e. taste-based, statistical or monopsonistic discrimination).

Overall, our paper provides solid evidence on the disappearance of immigrant overeducation across two generations in Belgium. However, the main finding of our paper highly depends on the type of employment (i.e. working full-time or part-time), which emphasizes the importance of accounting for employment conditions when designing integration policies aimed at promoting the labour market integration of immigrants. Last but not least, our paper also opens promising avenues for further intergenerational research on other barriers that workers of foreign origin may face at the workplace (e.g. mismatches between the job and the field of education and overeducation persistence).

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# **Appendices – Chapter 3: Immigrant overeducation across generations: The role of gender and part-time work**

**Appendix 1: Chart of developed and developing countries** 



Notes: Overseas territories are classified depending on their neighbouring countries. No data stipulates that no observation for workers born in or from these countries (Greenland (Denmark), Oman, Papua New Guinea, Tajikistan and Turkmenistan) is presented in our database.

## Appendix 2: List of countries by geographical region in our database

#### **Developed countries**

#### Belgium

**EU-14 countries**<sup>a</sup>: Austria, Denmark, Finland, France, Germany, Greece, Ireland, Italy, Luxembourg, Netherlands, Portugal, Spain, Sweden and United Kingdom.

**Other EU countries:** Bulgaria, Croatia, Cyprus, Czech Republic, Estonia, Hungary, Latvia, Lithuania, Malta, Poland, Romania, Slovak Republic and Slovenia.

**Other developed countries:** Andorra, Australia, Canada, Iceland, Japan, Liechtenstein, Monaco, New Zealand, Norway, Saint-Marin, Singapore, South Korea, Switzerland, Taiwan and United States.

#### **Developing countries**

The Maghreb countries: Algeria, Libya, Mauritania, Morocco, and Tunisia.

**Sub-Saharan African countries:** Angola, Benin, Botswana. Burkina Faso, Burundi, Cabo Verde, Cameroon, Central Africa Republic, Chad, Comoros, Congo, Congo DRC, Côte d'Ivoire, Djibouti, Equatorial Guinea, Eritrea, Ethiopia, Gabon, Gambia, Ghana, Guinea, Guinea-Bissau, Kenya, Liberia, Madagascar, Malawi, Mali, Mauritius, Mozambique, Namibia, Niger, Nigeria, Reunion (French Department), Rwanda, Sao Tome and Principe, Sierra Leone, Somalia, South Africa, Sudan, Tanzania, Togo, Uganda, Zambia and Zimbabwe.

**Near and Middle Eastern countries:** Afghanistan, Bahrein, Egypt, United Arab Emirates, Iran, Iraq, Israel, Jordan, Kuwait, Pakistan, Palestine, Qatar, Saudi Arabia, Syria, Turkey and Yemen.

**Non-EU Eastern European countries:** Albania, Armenia, Azerbaijan, Belarus, Bosnia and Herzegovina, Georgia, Kosovo, North Macedonia, Moldova, Montenegro, Russia, Serbia and Ukraine.

**Emerging and developing Asian countries:** Bangladesh, Bhutan, Brunei Darussalam, Cambodia, China, Fiji, French Polynesia (French Department), India, Indonesia, Kazakhstan, Kyrgyz Republic, Laos, Malaysia, Mongolia, Myanmar, Nauru, Nepal, North Korea, Nauru, New Caledonia, Philippines, Sri Lanka, Thailand, Uzbekistan, Vanuatu, Vietnam and Wallis and Futana (French Department).

Latin American and Caribbean countries: Argentina, Bahamas, Barbados, Belize, Bermuda, Bolivia, Brazil, Chile, Colombia, Costa Rica, Cuba, Dominican Republic, Dutch Antilles, Ecuador, Grenada, Guadeloupe (French Department), Guatemala, Guyana, French Guyana (French Department), Haiti, Honduras, Jamaica, Martinique (French Department), Mexico, Nicaragua, Panama, Paraguay, Peru, Saint Lucia, Suriname, Trinidad and Tobago, Uruguay and Venezuela.

<sup>&</sup>lt;sup>a</sup> EU countries are defined as during the time span of the database (1999-2016). Therefore, the United Kingdom is still considered an EU country.

	Sample of tertiary-educated workers born in or from:		
	Developed countries <sup>a</sup>	Others <sup>b</sup>	
Share of the sample (%)	13.8	1.1	
Observations	54,524	4,509	
Region of birth (%)	- 7-	,	
Belgium	53.5		
Developed countries <sup>c</sup>			
EU-14 countries	38.8	40.9	
Other EU countries	4.0	0.0	
Other developed countries	3.7	7.0	
Developing countries			
Sub-Saharan African countries		46.8	
Maghreb countries		1.7	
Near and Middle Eastern countries		0.5	
Emerging and developing Asian countries		0.7	
Non-EU Eastern European countries		0.0	
Latin American and Caribbean countries		2.4	
Worker characteristics	20.0	21.0	
Women (%):	38.8	34.0	
Age categories (%):	2.5		
20-24	3./	0.0	
25-29	16.1	2.4	
30-39	38.5 26 0	10.5	
40-49	26.9	28.3	
50+ Tanura in Maara	15.0	34.0 24.5	
Education (%):	7.1	24.3	
Bachelor	55 /	/03	
Master	40.7	47.3	
Advanced Master or PhD	3.9	34	
Household (%):	5.7	5.4	
Without children living at home	35.7	29.1	
With children living at home	58.0	69.2	
Other households <sup>d</sup>	63	18	
Employment characteristics	0.0	1.0	
Type of contract (%):			
Permanent	94.7	96.2	
Fixed-term	24.7 4 7	33	
Internship or apprenticeship	0.6	0.6	
Part-time work (%) <sup>e</sup>	4 4	49	
Overtime work $(\%)^{f}$	16	14	
Occupational categories - ISCO1 (%):	1.0		
Managers	15.5	15.9	
Professionals	39.9	41.1	
Technicians and associate professionals	15.6	16.4	
Clerical support	20.8	19	
Service and sales workers	3.9	3.9	
Craft and related trades workers	1.8	1.6	
Plant and machine operators and assemblers	1.6	1.6	
Elementary Occupations	0.9	0.6	

## Appendix 3. Descriptive statistics by origin – means and percentages

Appendix 3. (Continued)

#### **Appendix 3. Continued**

	Sample of tertiary-educated workers born in or from		
	Developed countries <sup>a</sup>	Others <sup>b</sup>	
Firm characteristics			
Sector of activity - NACE1 (%):			
B - Mining and Quarrying	0.1	0.2	
C - Manufacturing	26.5	25.4	
D - Electricity, gas, steam and air conditioning supply	1.7	2.3	
E - Water supply, sewerage, waste management and remediation activities	0.5	0.8	
F - Construction	3	4.4	
G - Wholesale and retail trade, repair of motor vehicles and motorcycles	18.1	18.6	
H - Transportation and storage	4.8	5.6	
I - Accommodation and food service activities	1.7	1.3	
J - Information and communication	12.6	14.1	
K - Financial and insurance activities	3.3	3.4	
L - Real Estate activities	0.3	0.6	
M - Professional, scientific and technical activities	16.2	13	
N - Administrative and support service activities	8.8	7.5	
P - Education	0.2	0.6	
Q - Human Health and social work activities	1.7	1.9	
R - Arts, entertainment and recreation	0.1	0.1	
U - Other service activities	0.3	0.3	
Size of the firm (FTE number of employees)	482.4	444.0	
Firm-level collective agreement (Yes) (%)	27.5	27.5	
More than 50% privately owned (%)	95.4	94.8	
Region where the firm is located (%):			
Brussels	29.5	30.3	
Flanders	41.4	47.3	
Wallonia	29.2	22.4	

*Notes:* Worker and firm weights are used in the calculation of means and percentages. <sup>a</sup> By "immigrants from developed countries, we mean first- and second-generation immigrants from developed countries. <sup>b</sup> By 'others', we mean workers born abroad with both parents born in Belgium (e.g. children of expatriates). <sup>c</sup> Appendix 2 shows the list of developed and developing countries by region of birth. <sup>d</sup> 'Other households' refer to brothers/sisters living together, friends living together, students or workers' homes, etc. <sup>e</sup> A worker is recognized as a part-time employee if he/she works less than 30 hours per week. <sup>f</sup> Overtime is when an employee works more than his/her contractual working hours. Source: STATBEL, 1999-2016.

	Probability of be	ing overeducated
Tertiary-educated workers born in/from:	(1)	(2)
<b>Belgium</b> (n = 311,222)	Reference	Reference
Developing countries <sup>a</sup>		
First-generation $(n = 14,459)$	0.121*** (0.014)	0.094*** (0.011)
Second-generation $(n = 11,748)$	0.011 (0.010)	0.007 (0.007)
<b>Developed countries</b> <sup>b</sup> ( $n = 54,524$ )	-0.039*** (0.006)	-0.025*** (0.004)
<b>Others</b> <sup>c</sup> (n = 4,509)	0.008 (0.014)	0.007 (0.010)
Estimator	OLS	OLOGIT
<u>Control variables</u> Year fixed effects <sup>d</sup> Worker characteristics <sup>e</sup> Employment characteristics <sup>f</sup> Firm characteristics <sup>g</sup>	Yes Yes Yes Yes	Yes Yes Yes Yes
Occupation-age-sector cells <sup>h</sup> Observations	13,628 396,462	13,628 396,462

#### Appendix 4. Average marginal effects - OLS and OLOGIT regressions

*Notes*: \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Worker and firm weights are used. Robust standard errors are in parentheses, which are clustered at the firm level. <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> The group 'immigrants from developed countries' includes F-G immigrants born in developed countries and their S-G counterparts. <sup>c</sup> The group 'others' refers to workers born in developing countries with both parents born in Belgium. <sup>d</sup> 17 year dummies. <sup>e</sup> Gender, level of tertiary education, tenure, squared tenure and type of household. <sup>f</sup> Type of contract and dummies for part-time work and overtime work. <sup>g</sup> Size of the firm (FTE number of workers in log), dummy for more than 50% privately owned, dummy for a firm-level collective agreement and region where the firm is located (Brussels, Flanders or Wallonia). <sup>h</sup> Using a realized matches approach, overeducation is identified within each occupation-age-sector cell (see Section 3.1 for more details). Source: STATBEL, 1999-2016.

		Bachelor	Master	Advanced master or PhD
Tertiary-educated workers born in/from:	-	(1)	(2)	(3)
Belgium		57.6	39.7	2.7
Developing countries <sup>a</sup>				
Sub-Saharan African countries	First generation	58.8	37.2	4.0
	Second generation	49.2	47.2	3.6
Maghreb countries	First generation	64.0	32.4	3.6
	Second generation	66.8	31.8	1.5
Near and Middle Eastern countries	First generation	57.6	39.0	3.4
	Second generation	69.1	30.1	0.7
Emerging and developing Asian	First generation	51.2	43.3	5.5
countries	Second generation	46.8	47.6	5.6
Non-EU Eastern European	First generation	55.3	40.8	3.9
countries	Second generation	59.8	36.4	3.9
Latin American and Caribbean	First generation	48.8	43.2	8.0
countries	Second generation	59.5	38.4	2.2

## Appendix 5. Shares of tertiary-educated workers (%) by type of diploma, 1999-2016

*Notes:* Worker and firm weights are used in the calculation of percentages. <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. Source: STATBEL, 1999-2016.

		Probability of being in a part-time job	
Tertiary-educated workers born in/from:		(1)	(2)
Without child(-ren) (n	= 56,415)	Reference	Reference
With child(-ren) (n =	132,102)	-0.002*** (0.001)	-0.002* (0.001)
Without child(-ren) $(n = 38,762)$		0.031*** (0.002)	0.024*** (0.002)
Women With child(-ren) (n =	78,868)	0.102*** (0.002)	0.078*** (0.002)
ıntries <sup>a</sup>			
	First generation $(n = 2,969)$	0.034*** (0.005)	0.011*** (0.004)
without child(-ren)	Second generation $(n = 2,139)$	-0.003 (0.003)	0.001 (0.004)
With child(-ren)	First generation $(n = 5,676)$	0.028*** (0.004)	0.008*** (0.003)
	Second generation (n = 4,190)	0.008** (0.003)	0.012*** (0.004)
Without child(-ren)	First generation (n = 1,840)	0.072*** (0.009)	0.042*** (0.006)
	Second generation (n = 1,779)	0.029*** (0.007)	0.031*** (0.006)
	First generation $(n = 2,932)$	0.144*** (0.009)	0.064*** (0.005)
with child(-refi)	Second generation $(n = 3,294)$	0.068*** (0.007)	0.061*** (0.006)
<u>Control variables</u> Year fixed effects <sup>b</sup> Worker characteristics <sup>c</sup> Employment characteristics <sup>d</sup> Firm characteristics <sup>e</sup>		Yes No No No 286 555	Yes Yes Yes Yes
	ed workers born in/from         Without child(-ren) (n         With child(-ren) (n =         Without child(-ren) (n =         With child(-ren) (n =         mtries <sup>a</sup> Without child(-ren)         With child(-ren)         With child(-ren)         With child(-ren)         With child(-ren)         With child(-ren)         With child(-ren)         Setects <sup>b</sup> with child(-ren)	ed workers born in/from:Without child(-ren) (n = 56,415)With child(-ren) (n = 132,102)Without child(-ren) (n = 38,762)With child(-ren) (n = 78,868)mtries <sup>a</sup> Without child(-ren)First generation (n = 2,969)Second generation (n = 2,139)With child(-ren)First generation (n = 5,676)With child(-ren)First generation (n = 5,676)Second generation (n = 4,190)Without child(-ren)First generation (n = 1,840)Second generation (n = 1,779)First generation (n = 2,932)With child(-ren)Second generation (n = 3,294)Second generation (n = 3,294)	Probability of bein           ed workers born in/from:         (1)           Without child(-ren) (n = 56,415)         Reference           With child(-ren) (n = 132,102)         -0.002***           (0.001)         Without child(-ren) (n = 38,762)         0.031***           Without child(-ren) (n = 78,868)         0.102***           (0.002)         with child(-ren) (n = 78,868)         0.102***           Without child(-ren)         First generation (n = 2,969)         0.034***           Without child(-ren)         First generation (n = 2,139)         -0.003           Muth child(-ren)         First generation (n = 5,676)         0.028***           With child(-ren)         First generation (n = 4,190)         0.008**           Second generation (n = 1,840)         0.072***         (0.009)           Second generation (n = 1,779)         0.029***         (0.007)           With child(-ren)         First generation (n = 1,779)         0.029***           With child(-ren)         First generation (n = 3,294)         0.068****           With child(-ren)         Second generation (n = 3,294)         0.068****           With child(-ren)         First generation (n = 3,294)         0.068****           With child(-ren)         Second generation (n = 3,294)         0.068****

#### Appendix 6. Part-time work and parenthood: average marginal effects – LOGIT regression

*Notes*: \*\*\* p<0.01, \*\* p<0.05, \* p<0.1. Worker and firm weights are used in the LOGIT regression. Robust standard errors are in parentheses, which are clustered at the firm level. The groups 'immigrants from developed countries' and 'others' are also included in the regression but their gender-interacted estimates associated with or without having children are not portrayed in this table (available on request). Workers living in other types of households (e.g. brothers/sisters living together, friends living together, students or workers' homes) are not included in the regression <sup>a</sup> Second-generation immigrants' origin is defined according to the father's country of birth. However, if the father was born in Belgium and the mother was born in a developing country, the mother's country of birth is retained. <sup>b</sup> 17 year dummies. <sup>c</sup> age, level of tertiary education, tenure and squared tenure. <sup>d</sup> occupation, type of contract and dummy for overtime. <sup>e</sup> sector of activity, size of the firm (FTE number of workers in log), dummy for more than 50% privately owned, dummy for a firm-level collective agreement and region where the firm is located (Brussels, Flanders or Wallonia). Source: STATBEL, 1999-2016.