Research Article

Impacts of education and immigration on the size and skills of the future workforce

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Impacts of education and immigration on the size and skills of the future workforce

Samuel Vézina¹
Alain Bélanger²

Abstract

BACKGROUND
In most developed countries the demographic regime is characterised by low fertility, high immigration, and population ageing. This regime impacts on the size of the labour force population, the ethno-cultural composition, and potential productivity.

OBJECTIVE
We examine the prospective impacts of two sociodemographic changes, namely increasing education and increasing immigration levels, on the size and skills of the workforce aged 25 to 64 years old for the time period 2011–2061.

METHODS
We use microsimulation models to project the population of Austria and Canada by age, sex, and several other socioeconomic and ethno-cultural dimensions. Using ‘what-if’ scenarios, we measure the impact of different immigration rates and immigrant selection patterns on the size and average literacy skills of the projected workforce. We also examine the potential effects of different assumptions about future educational attainment on the projection outcomes.

RESULTS
The results show that the volume of immigration heavily influences the expansion of the workforce. In terms of education among the general population, the rise in educational attainment is positively correlated with higher skill levels. However, increased immigration can override the effect that increasing education levels have on advancing workforce skills.

CONTRIBUTION
This paper shows the extent to which Austria and Canada have adopted diametrically opposed strategies regarding the future development of their workforces. Assuming that

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current policy goals hold true in the coming decades, Austria is likely to rely on a higher-skilled but declining (in size) workforce, whereas Canada will rely on a fast-growing but less-skilled and more culturally diversified workforce.

1. Introduction

In most developed countries, current immigration has reached historical levels: population growth is now mainly fuelled by immigration (Wilson et al. 2013). In the context of population ageing, where a growing number of people are about to withdraw from the labour market, pressures to keep immigration rates at high levels are likely to persist.

However, developed countries have different approaches to migration. On one hand, there are countries with a long history of immigration such as the United States, Canada, Australia, and New Zealand. Despite important variation over time, they have admitted record numbers of immigrants for centuries. The United States is still the OECD’s leading destination country, receiving about 20% of the global immigrant inflow, whereas Australia, New Zealand, and Canada have some of the highest proportions of foreign-born populations (OECD 2017a). These latter countries have a specific government-led selection system, with policies targeted to attract a larger share of higher-educated immigrants (Papademetriou and Sumption 2011).

In Western European countries migration issues have also been of great importance, but the immigrant-selection mechanisms and citizenship-granting philosophies are different. European policymakers have been challenged by issues resulting from geopolitical events such as the dissolution of the Soviet Union, European Union (EU) enlargement (and Schengen area implementation), and, most recently, the so-called refugee crisis, with millions of asylum seekers coming from the Middle East, Africa, and Asia. Since 2005 the European Commission has developed and implemented a Global Approach to Migration and Mobility to better manage migration through the development of ‘mobility partnerships’ with sending countries. The approach also aims to maximise the development impact of EU external migration and better address the issue of integrating the foreign-born into the labour force (Geddes and Scholten 2016).

The world’s most developed countries have recently seen not only an increased share of foreign-born populations but also more culturally diverse immigration (Coleman 2006, 2009; Lee et al. 2017; Vertovec 2007). At the same time these countries also face a unique opportunity, as older cohorts of workers are replaced by more-educated and more-productive young generations entering the labour market
(Barakat and Durham 2014; Meyer et al. 1977; Wils and Goujon 1998). Using projection scenarios, experts have shown how increased human capital (in terms of educational attainment) can influence the consequences of population ageing in terms of labour force population and public pensions and other health and social expenditures (Lutz, Butz, and KC 2014).

As both immigration and educational attainment are increasing, concerns about immigrant educational attainment and labour market integration have entered policy discussions. A large body of literature documents various aspects of the deteriorating economic outcomes among recent immigrants (Aydemir and Skuterud 2005; Borjas and Friedberg 2009; Clark and Drinkwater 2008; Rosholm, Scott, and Husted 2006). However, these studies mostly use standard educational attainment variables to assess the human capital of immigrants. The Organization for Economic Cooperation and Development’s (OECD) and its Programme for the International Assessment of Adult Competencies (PIAAC) has developed a more direct measure of skills, which assesses proficiency in key information-processing skills. Even though such an assessment does not reflect all human capital and skills (St. Clair 2012), the measure provides interesting insight into the human capital of the adult population, since it seems to better capture labour-market-relevant skills than standard measures of educational attainment (Bonikowska, Green, and Riddell 2008; Ferrer, Green, and Riddell 2006). In most OECD countries, immigrants have a lower literacy level but a higher overall education level than the native-born (Green and Riddell 2007; OECD 2016). This implies that the actual skills brought into labour markets through immigration are significantly lower than suggested by assessments based on immigrant level of education (Green and Worswick 2017).

Literacy proficiency varies across countries and across groups of individuals, and the determinants are numerous. Studies show that literacy declines with age (Desjardins and Warnke 2012; Green and Riddell 2013; Paccagnella 2016) but increases with education (Green and Riddell 2007; OECD 2016). Specific immigrant characteristics such as age at immigration (generation status), country of birth, mother’s education level, and the language spoken at home are all significant predictors of literacy proficiency.

The PIAAC survey also measures two other information-processing skills, namely numeracy and problem solving in technology-rich environments. This paper focuses on literacy skills, mainly because nowadays literacy is important even in jobs that are not primarily oriented to literacy skills. Moreover, in the PIAAC the three dimensions are highly correlated, mainly because the skills assessment methodology requires the use of literacy skills. For example, numeracy skills do not solely correspond to someone’s ability to solve a set of equations in isolation but rather correspond to how people deal with mathematical concept in the everyday life. Indeed, in the PIAAC numeracy is
defined as “the ability to access, use, interpret and communicate mathematical information and ideas, in order to engage in and manage the mathematical demands of a range of situations in adult life” (OECD 2012: 33). Thus, using numeracy instead of literacy skills would lead to very similar results.

With a specific focus on two developed countries, Austria and Canada, this research uses microsimulation models to assess how education and immigration levels impact the size of the future workforce and its average literacy skill level. Austria and Canada are two developed OECD member countries, yet they generally have different immigration traditions, policies, approaches. Comparing these two countries provides an understanding of how the determinants of literacy skills can vary from one country to another and how different contexts (e.g., migration) can influence the average literacy skill level of the future workforce. The study presents two microsimulation models (PÖB and LSD–C), which project the future population by many more dimensions than age and sex. The models incorporate immigration status, ethno-cultural variables, and many other dimensions that are either important determinants of demographic behaviours such as education, or are closely linked to sociodemographic characteristics such as labour force participation and literacy skill proficiency. Using ‘what-if’ scenarios, the results disentangle potential impacts of likely sociodemographic changes – such as increasing education and immigration levels – on the overall literacy proficiency of the active population between 2011 and 2061. This research generates new knowledge relevant to policymaking on migration, social cohesion, and labour market needs and changes, as well as the needs involved in developing educational skills.

2. Theoretical framework

Our research draws on multiple theoretical perspectives. The first stems from the idea that population forecasts can also predict socioeconomic changes, in accordance with the Demographic Metabolism theory (Lutz 2013). This theory emphasises the importance of cohort succession in explaining social changes. Since many characteristics of individuals (education, values, language, etc.) tend to remain stable over their life course, social changes may occur through the continuous succession of cohorts, whose composition and social structure vary over time as cohorts join the population through births and immigration and others leave through death and emigration. Lutz (2013) has shown that Demographic Metabolism may also be used as a forecasting theory of socioeconomic changes with predictive power, the goal of which is not to explain or forecast specific demographic variables, but “to predict broad
socioeconomic change (ranging from values and religions to skills and productivity of the labour force) using a demographic paradigm” (Lutz 2013: 283–284).

In conjunction with the aggregate-level changes predicted by Demographic Metabolism, individual behavioural shifts may also act as a source of macro-level social change. Therefore, the theoretical basis of our research programme links to multiple micro-level theories (Human Capital and New Home Economics, Social Capital, Life Course, etc.), on which the different simulated transitions are modelled. The specific theoretical framework underpinning the literacy skills analysis derives from approaches developed by Desjardins (2003a) and Barone and van de Werfhorst (2011). These researchers assert causal relationships between schooling, skills, and economic and social outcomes, which have been studied extensively in the scientific literature (Chesters, Ryan, and Sinning 2013; Green and Riddell 2003; Hanushek et al. 2015). Education is considered to be the main driver of literacy skill formation (Green and Riddell 2007; OECD 2016), triggering the development of skills and maintaining them over the life course (Desjardins 2003b). Our framework recognises the functional definition of Human Capital theory (Becker 1964) regarding the connection between schooling pathway and productivity, while also asserting the influence of social structure and learning context during childhood (cultural and social capital) on the schooling pathway and economic outcomes (Bourdieu 1979; Bynner and Parsons 2009; Gimenez-Nadal and Molina 2012; Hart and Risley 1995; Hassrick, Raudenbush, and Rosen 2017; Kerckhoff 1986). Aside from the schooling pathway, we recognise the importance of the learning context during childhood for cognitive skill acquisition and maintenance during adulthood (Dickinson and Neuman 2006). Finally, the Practice-Engagement theory (Reder 1994) asserts that observed skill level is a function of the use of skills (incidence, frequency, complexity) on a daily basis in everyday life. The analysis of the determinants of literacy skill proficiency is based on the concept that the practice of reading skills at home and at work (known as life-wide factors\(^3\)) is a pertinent dimension worth consideration.

Our analysis is further rooted in the Segmented Assimilation thesis (Portes and Zhou 1993). From the original assimilation perspective, acculturation occurs through a linear process by which diverse ethnic groups gradually change their behaviours to adopt those of the majority and eventually share its culture. Studies have shown that most of the demographic and socioeconomic behaviours of immigrants tend to converge more or less rapidly with the level of the native-born population as the duration of residence in the host country increases (Adserà et al. 2012; Bloom, Grenier, Grenier, ...}

\(^3\) Desjardins (2003b) differentiates lifelong learning from life-wide learning. Lifelong learning refers to learning occurring outside of formal education contexts, while life-wide learning adds an element of space to the framework. Life-wide learning refers to learning occurring in different contexts “such as home, school, work, community and others” (Desjardins 2003b: 211).
According to classical assimilation theory, immigrants of the 1.5 generation and the second generation alike will integrate more rapidly and adopt mainstream values more easily because they are enrolled in the host country school system. Although the assimilation perspective has been challenged (Le Grand and Szulkin 2002; Licht and Steiner 1994), it nevertheless underlines the importance of taking the year of immigration into account when studying and projecting the behaviours of newcomers (Duleep and Dowhan 2002; Hall and Farkas 2008; Hudson 2007; Powers and Seltzer 1998). Moreover, time does not seem to operate uniformly on all immigrants. Divergent outcomes have been observed across both cohorts of immigrants and ethnic minority groups (Causa and Jean 2007; Dustmann and Fabbri 2005; Lemos 2017; Poot and Stillman 2016). The New Labour Market Segmentation theory provides an interesting framework to view these divergent outcomes, as it accounts for the possibility of different minority groups being either permanently confined to the underclass or advancing rapidly economically.

In its empirical application, this theoretical framework requires a projection model that goes beyond age and sex and allows for the simultaneous projection of several variables. As shown in the above paragraphs, age at immigration, along with length of stay in the host country and generation status, are important variables to consider when studying ethno-cultural differences in demographic or socioeconomic behaviours. Accordingly, this paper uses microsimulation models (see Section 3 on data and method), which capture the effect of changes in population composition better than traditional cohort-component projection models.

### 3. Data and research methods

#### 3.1 The microsimulation projection models

PÖB (Projektionsmodell der österreichischen Bevölkerung) and LSD–C (Laboratoire de simulations démographiques – Canada) are the two microsimulation models used in this research, which project the populations of Austria and of Canada respectively. These models project the total population by age, sex, immigrant status, country of birth, education status, labour force participation, literacy skills, and other variables related to immigrants such as age at immigration and length of stay in host country.

PÖB and LSD–C are built on a similar framework, with the shared broad objective of projecting ethno-cultural diversity and the future composition of the population and

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4 Portes and Zhou (1993) define the 1.5 generation of immigrants as those who arrived as children.
labour force. They were developed using Modgen 12, which is a C++ based microsimulation language developed and maintained by Statistics Canada (Statistics Canada 2017). Both models are case-based in the sense that every individual is simulated separately from other individuals and no interactions between individuals are allowed (except for interactions between mother and children). PÖB and LSD–C are dynamic: they allow for changes in individual characteristics over the life course as well as for inter-generational transfers of some characteristics from mother to child.

The models are in continuous time and characteristics of individuals are modified continuously in ‘real time’. The starting year for both models is 2011; the Austrian population is based on the Austrian Labour Force Survey (LFS) and the Canadian population on the 2011 National Household Survey public-use microdata file (NHS–PUMF). Individuals from the base population are simulated one by one and their characteristics are modified through events determined by the values of their specific input parameters at any given time during the projection period.

PÖB and LSD–C are open to international migration, which has been a crucial component of population change in contemporary Western societies and is a driving factor in changing the stock of human capital. The immigration module includes all classifications, state variables, and parameters relevant to immigrants and immigration: immigration level and composition, immigrant status, age at immigration, duration since immigration, generation status, and place of birth. The module works as follows: in every projection year, a new immigrant cohort enters the simulation. All simulated immigrant cohorts from 2012 to 2061 have the characteristics of the recent immigrants as observed in 2011. These recent immigrants are defined as those having lived in the host country for less than 5 years. This method for generating immigrants implies that all immigrant cohorts have the same characteristics as recently observed immigrants. To allow for some variation, the individual weights are adjusted to match desired distributions. Thus, through re-weighting, it is possible to change the immigrant distribution according to age, education, and other variables. Finally, the volume of

For a thorough description of this framework, readers are invited to consult the publication Bélanger et al. (2018). The Canadian model (LSD–C) projects the population by more variables than the Austrian model (PÖB). For example, LSD–C simulates internal migrations, and therefore projection results can be disaggregated by province of residence. LSD–C also projects the population by knowledge of official languages, visible minority group, country of highest diploma, and religion. In both microsimulation models the list of variables is kept relatively short, leading to aggregate projections that are generally similar to those obtained by traditional population projections, while adding valuable details.


For a more detailed introduction to microsimulation in the social sciences and population projections, see van Imhoff and Post (1998).

At its point of departure in 2011, the age structure in Austria is slightly older than that in Canada. For example, in Austria the old age dependency ratio in 2011 is 0.26 (1 elderly for 4 working-age persons), while in Canada this ratio is 0.20 (1 elderly for 5 working-age persons).

http://www.demographic-research.org
immigration, i.e., the number of immigrants entering the simulation each year, can be modified as well.

Demographic assumptions of future mortality and fertility are identical to the assumptions of the Medium (SSP2) scenario for Austria (Wittgenstein Centre for Demography and Global Human Capital 2015) and to the medium-growth (M1) scenario of most recent official Canadian projections (Statistics Canada 2015). In Canada, projections assume that the Total Fertility Rate (TFR) will remain at around 1.67 children per woman; in Austria, the TFR starts at 1.45 in 2011 and reaches 1.61 at the end of the projection in 2061. Regarding mortality, life expectancy at birth increases from 79 to 88 years for Canadian men and from 84 to 89 years for Canadian women. Assumptions are similar in Austria, although a little more optimistic: life expectancy is projected to increase from 79 to 89 years for men, and from 84 to 93 years for women.

All other parameters are derived from a variety of sources and methods: describing all of them is beyond the scope of this paper. In general, model parameters are estimated using logistic regressions from several data sources: censuses, vital statistics, social surveys, labour force surveys, and population estimates. For a more detailed description of the models, including methods and assumptions, see Bélanger et al. (2018). More details on the projections (parameters and methodology) of literacy skill proficiency are provided in the following section.

3.2 Variable categories and conventions

Despite the traditional definition of the workforce as encompassing the economically active population 15 years of age and over, the age range considered throughout this study is limited to persons aged 25 to 64 years old. The 15+ population is too heterogeneous in terms of age-specific participation rates, education level, and literacy skill proficiency to adequately tackle the research objectives of this paper. Instead the 25–64 years cut-off was selected in order to focus on the age range where individuals have the highest participation rates, with the assumption that by the age of 25 most people have completed their education pathway. In both Austria and Canada the retirement age is 65 years, justifying the other end of the range used. Finally, this cut-off point is necessary, as data on adult literacy skills (PIAAC Survey) is not collected for persons older than 65 years.

Both microsimulation models have a literacy skills projection module. During the simulation, only the population aged 25 to 64 is assigned a literacy score. Literacy scores are updated whenever the characteristics of relevant actors undergo changes.

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9 In Austria the normal pension age is 65 for men. The retirement age for women is currently 60 years, but will be gradually increased to 65 between 2024 and 2033 (OECD 2013b).
Parameters were assessed using the most recent cycle (2012) of PIAAC survey data. In this survey the respondents’ literacy proficiency is measured along a continuous scale ranging from 0 to 500, where a higher score indicates greater proficiency. The series of tasks used for assessing literacy range from reading a product label or locating a single piece of information, to reading newspaper article and distilling information. To help organize and interpret the scores, the scale is divided into proficiency levels. For example, Level 3 ranges between 276 and 325 points on the 0–500 literacy scale. At any given level, individuals can successfully complete certain types of task. For example, at Levels 1 and 2 respondents can read and understand relatively short texts, whereas Levels 4 and 5 require multistep operations to integrate, synthesise, and interpret information (OECD 2016). The international ranking shows that Canada (with an average literacy score of 273.5) remains close to the OECD average (272.8), behind countries such as Japan (296.4) and Finland (287.6), but ahead of countries such as France (262.1) and Italy (250.5). With an average literacy score of 269.5, Austria ranks just below Germany and the United States, which both have 269.8 points on average (OECD 2013a).

Regression analyses of the log\textsubscript{10} scores were conducted and stratified by immigration status in order to consider specific characteristics of the foreign-born population. Literacy proficiency depends on sociodemographic variables (age, sex), human capital variables (education level), socioeconomic variables (mother’s education level, labour force status), and life-wide factors, and on the variables linked to characteristics specific to immigrants, such as age at immigration, years in host country since migration, and country of birth.\textsuperscript{11} In the LSD–C, parameters for the literacy skills module also vary by province of residence, mother tongue, language spoken at home, and country of highest diploma. In the Appendix we provide more details on our analyses, the factors linked to literacy proficiency, and our estimations of the projection parameters.

4. Hypotheses and scenarios

Projection scenarios are formulated according to specific sets of assumptions about both general-level phenomena and characteristic-specific differentials between

\textsuperscript{10} A simple logarithmic transformation is made on the dependent variable (literacy score) to ensure that the linear regression model will not lead to an illogical predicted score (below 0) during the simulation.

\textsuperscript{11} In our analyses as well as in both microsimulation models, the country of birth variable has three categories: individuals born in host country (whether Austria or Canada), individuals born in another rich country, and individuals born elsewhere. The richest countries of the world are grouped together and correspond to Western European and North American countries together with Australia, Japan, New Zealand, Singapore, and South Korea.
individuals with the likelihood of experiencing an event. For example, models allow for changes in total fertility level over time and fertility differentials linked to certain characteristics such as education level or immigration status. Table 1 sums up the underlying hypotheses of the different scenarios, designed according to immigration, education, and labour market participation rates.

### Table 1: Summary of projection scenario assumptions

<table>
<thead>
<tr>
<th>Scenario</th>
<th>Immigration volume</th>
<th>Immigration composition</th>
<th>Education</th>
<th>Labour market participation rates</th>
</tr>
</thead>
<tbody>
<tr>
<td>IM_USRATE</td>
<td>Immigration rate set at 0.35%</td>
<td>Characteristics of immigrants arrived between 2006–2010</td>
<td>Educational attainment set at observed rate in 2011</td>
<td>Recent trends</td>
</tr>
<tr>
<td>IM_OFFICIAL</td>
<td>Official immigration volume projected by National Statistical agencies</td>
<td>Characteristics of immigrants arrived between 2006–2010</td>
<td>Educational attainment set at observed rate in 2011</td>
<td>Recent trends</td>
</tr>
<tr>
<td>IM_ZERO</td>
<td>No immigration</td>
<td>Characteristics of immigrants arrived between 2006–2010</td>
<td>Educational attainment set at observed rate in 2011</td>
<td>Recent trends</td>
</tr>
<tr>
<td>EDUCATION</td>
<td>Immigration rate set at 0.35%</td>
<td>Characteristics of immigrants arrived between 2006–2010</td>
<td>Recent trends reflecting the observed rise of educational attainment of cohorts</td>
<td>Recent trends</td>
</tr>
</tbody>
</table>
Canada: Immigrants come in with more 'literacy-oriented' characteristics in terms of age, education, language skills, and country of highest diploma | Educational attainment set at observed rate in 2011       | Recent trends                    |
Canada: Characteristics of immigrants arrived between 2006–2010 | Recent trends reflecting the observed rise of educational attainment of cohorts | Recent trends                    |

Three scenarios were designed as variations in volume of immigration:

- **IM_USRATE**: The immigration rate is set at 0.35% in both models for the years 2011–2061. This immigration rate corresponds to that of the United States in recent years. Applying such a rate in Austria and Canada translates into 30,000 and 120,000 immigrants per year at the beginning of the simulation, respectively. This scenario provides a basis for comparing the demographic dynamic prevailing in both countries. In reality, these two countries have very different immigration targets and policies, which affect the projected outcomes (workforce size and skills). Therefore, this scenario shows the differences attributable to dimensions other than immigration, since it applies the same immigration rate to both countries.
IM_OFFICIAL: The absolute number of immigrants is set to the officially observed (2011–2015) and projected (2015–2061) numbers published by Statistik Austria (2016) and Statistics Canada (2015). In Canada the official targets correspond to an immigration rate of 0.75%. In Austria the immigration rate surpassed 1.3% in 2015, but is projected to drastically decrease to below 0.35% by 2026 and to remain constant at around 0.25% from then onward. These numbers reflect current official national immigration targets.

IM_ZERO: The immigration rate is set to zero. This scenario is highly unrealistic but serves the purpose of measuring the range of influence that migration has over the projected outcomes.

In these scenarios the educational attainment of young adults is set to the level observed in 2011. However, historical data shows a steady increase in the educational attainment of cohorts over the past decades. Therefore, the following scenario was designed:

EDUCATION: Designed to measure the impact of these trends on the projected outcomes. More specifically, the EDUCATION scenario assumes that the educational attainment of cohorts will continue to rise between 2011 and 2026 and will remain constant after 2026. The rise observed in recent decades was too substantial to be maintained for three more decades, i.e., between 2026 and 2061. Under this scenario the immigration rate is set at 0.35% between 2011 and 2061 for both models, as in the IM_USRATE scenario.

In the four scenarios described above, all simulated immigrant cohorts from 2012 to 2061 have the same characteristics as recent immigrants from 2011, i.e., immigrants arrived in host countries less than 5 years prior to 2011. However, the volume and composition of immigrants are subject to change. The recent migration flow into Austria is a clear demonstration of this. In 2011, recent immigrants living in Austria were highly educated (more than the native-born Austrians), fluent in German language, and mainly from developed countries (Statistik Austria 2017). However, the 2015 refugee flow drastically modified the characteristics of the recent immigrants to Austria. The IM_OFFICIAL scenario, which assumes relatively high immigration rates between 2011 and 2020, assumes that these simulated immigrant cohorts have the characteristics of recent immigrants prior to the ‘refugee crisis’. Therefore, under the IM_CHARACT scenario the simulated distribution of immigrant cohorts is modified to reflect the characteristics of the total immigrant population admitted to Austria in
2015/2016. In the Canadian model, under the IM_CHARACT scenario, the simulated distribution of immigrant cohorts is slightly modified to reflect a plausible change in Canada’s immigrant selection policies that would foster the admittance of immigrants with even more ‘literacy-oriented’ characteristics (younger, more-educated, higher language skills, etc.). More precisely, the distribution of simulated immigrants is modified as follows: it has 5% more persons aged 0 to 30 years old, 5% more university graduates, 7% more individuals whose highest diploma was obtained in Canada, and 11% more people whose mother tongue is either French or English.

Finally, the most likely assumptions with respect to immigration and education in Austria and Canada were merged into one last scenario (BASELINE). More precisely, immigration volume is set according to the reference scenario of the most recent projections from the national statistical agencies (Statistics Canada 2015; Statistik Austria 2016), which reflect official national immigration targets. Regarding the composition of immigration, the simulated distribution of immigrant cohorts is modified to reflect the characteristics of the total immigrant population admitted to Austria during the 2011–2016 period. This distribution is therefore a blend of the pre-2015 flows and the 2015 refugee flow. In the Canadian model the hypotheses used in the BASELINE scenario expect that the characteristics of future immigrant cohorts will be similar to the characteristics of recent immigrants. Finally, the BASELINE scenario assumes that the educational attainment of future cohorts will continue to rise between 2011 and 2026, remaining constant after 2026. Indeed, experts have shown that the increases observed in the last decades are likely to hold in the future (Barakat and Durham 2014; KC et al. 2010; Lutz, Butz, and KC 2014).

No specific scenario was designed with respect to labour force participation rates, as it would fall beyond the scope of this paper. The reference hypotheses assume lower participation rates for foreign-born workers, increased female labour participation, and increased participation of the 55+ population. These hypotheses extrapolate from the most recent trends observed in both Austria and Canada (Loichinger 2015; Martel et al. 2011).

5. Results

5.1 Projection results under the different immigration and education scenarios

In this section we study the effect of alternative assumptions about immigration and education trends on the size and average literacy proficiency of the workforce in Austria and Canada between 2011 and 2061. At this point it is important to point out that all the projection scenarios and results presented in this study illustrate what would
happen if the assumptions were proven correct. They should be considered as prospective exercises, designed to support decision-making and policy planning rather than to predict the future.

Figure 1 presents the projection results of all six scenarios for both countries. The projected size of the future workforce is illustrated on the left-hand side and the projected literacy proficiency on the right-hand side. The projected active population aged between 25 and 64 years old is expressed in terms relative to its 2011 size (base_{100} = 2011),\textsuperscript{12} and projected literacy proficiency is the mean score of the workforce aged 25 to 64 years old.

\textbf{Figure 1: Projected size of the workforce (left) and its average literacy proficiency score (right), according to different scenarios, 25–64 years old, 2011–2061, Austria and Canada}

\begin{table}
<table>
<thead>
<tr>
<th>Scenario</th>
<th>Size of the workforce (base 100 in 2011)</th>
<th>Average literacy skill score of the workforce</th>
</tr>
</thead>
<tbody>
<tr>
<td>IM_USRATE</td>
<td></td>
<td></td>
</tr>
<tr>
<td>IM_OFFICIAL</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

\textsuperscript{12} Index numbers are used to express the difference between two measurements by designating one number as the ‘base’, giving it the value 100, and then expressing the projected numbers as a percentage of the first. This is useful for comparing Austria and Canada since their absolute workforce sizes are different: In 2011 the active population (aged 25 to 64 years old) was 3.6 million in Austria and 15.2 million in Canada. This 2011 level was designated the base period and given the value 100. The index numbers for the measurement at all other points in time indicate the percentage change from the base period.
In the IM_USRATE scenario the projected size of the workforce is similar in both countries. Compared to its 2011 level, the active population slightly increases up to 2021 and then decreases. The total labour force population is smaller in 2061 than in
2011 by 4% in Austria and by 10% in Canada. In other words, when keeping the immigration rate at 0.35% and assuming that educational attainment of future cohorts will not increase and will remain constant at the 2011 level, the projected workforce of both countries is likely to decrease in the coming decades. On the other hand, under this scenario the average literacy proficiency is likely to increase in both countries: we observe a 5-point increase in Austria from 274 to 279 and a 3-point increase in Canada from 273 to 276. This scenario reveals that under equivalent assumptions of immigration and education the general demographic dynamics are not favourable to growth of the active population, but it fosters increases in literacy skills proficiency.

In the IM_OFFICIAL scenario all hypotheses of the IM_USRATE scenario are kept the same, apart from the absolute number of immigrants, which is set to the drastically contrasting official national immigration targets of the two countries: between 2011 and 2061 Canada’s immigration intake objectives are twice as high as Austria’s. This difference has a clear impact of the projected workforce and their skills. The total active population projected in Austria by 2061 is close to the 2011 level, and there is a clear increase in average literacy skills proficiency. On the other hand, Canada’s high immigration rate – which is the highest by far among the G7 countries (OECD 2017b) – leads to a 27% increase in its active population between 2011 and 2061, growing from 15.2 to 19.2 million persons. Furthermore, the projected literacy score is likely to decrease in the next 50 years by almost 3 points on average.

The results obtained under the no-immigration scenario (IM_ZERO) amplify those of the IM_USRATE scenario. Over the 2011–2061 period the total workforce steadily decreases in both countries, by 30% in Austria and by 34% in Canada, while the average skill level increases by 11 points in Austria and 8 points in Canada. This scenario shows the extent to which the growth of the population is dependent on immigration intakes in both countries. On the skills side, this projection shows that despite the fact that the IM_ZERO scenario does not assume that the trend of increasing education will continue, there is already significant inertia in the demographic dynamic which fosters rising skills among the workforce. In both countries, retiring older workers are likely to be replaced by more-educated younger workers, regardless of immigration intakes, which increases the projected average skill level.

In contrast to the projection results obtained under the IM_USRATE scenario, the EDUCATION scenario illustrates the impact of extrapolating the rising trends in the educational attainment of cohorts. In EDUCATION the total projected workforce is very close in size to the IM_USRATE scenario. In other words, the rising education levels are not likely to significantly affect the size of the future workforce. In the EDUCATION scenario the projected average literacy proficiency is likely to be higher in both Austria and Canada, by 2 points and 1 point respectively, in comparison to the IM_USRATE scenario projection results.
In the IM_CHARACT scenario all hypotheses of the IM_OFFICIAL scenario are kept the same, apart from the composition of the simulated newcomers. When playing with assumptions about the characteristics of future immigrants (and not the immigration level per se), we observe no significant impact on the projected size of the workforce. However, notable differences are observed in the skills of the projected workforce. In Austria the IM_CHARACT scenario assumes that immigrants have the characteristics of those who arrived during the 2015 ‘refugee crisis’ (not only the refugees, but all immigrants admitted in 2015 and 2016). Under this scenario the projected average skills proficiency declines between 2011 and 2021 but recovers in 2061, reaching 277 points. In Canada the IM_CHARACT scenario assumes that future admitted immigrants will have characteristics that foster higher literacy skills. The results suggest that changes to immigrant selection policies could prevent the projected skills decline of the Canadian workforce observed in the IM_OFFICIAL scenario while keeping its immigration targets at a very high rate. Indeed, under the IM_CHARACT scenario the projected average literacy score in 2061 is 273, which is equivalent to the 2011 level.

Finally, under the BASELINE scenario the total Austrian workforce is likely to remain more or less constant between 2011 and 2061. By 2061 the projected active population will increase by 0.1% of its 2011 level. On the other hand, high immigration targets in Canada lead to a 27% increase in its active population between 2011 and 2061, growing from 15.2 to 19.4 million persons. While the Canadian workforce is likely to increase due to its high immigration rate, the skills of its workforce are likely to decrease. The opposite effect is likely to be observed in Austria: the future Austrian workforce is likely to have higher literacy skills as the average rises from 274 in 2011 to 279 in 2061. With such a score, Austria would be pushed up to the level of the Scandinavian countries\textsuperscript{13} in international ranking, assuming that the average literacy level in the other OECD countries remains at the 2012 level.

\section*{5.2 Composition and skills of the future workforce}

In this section we study the impact of likely sociodemographic changes on the composition and skills of the future workforce. Figure 2 shows the projection results obtained under the BASELINE scenario, disaggregated by immigration status (and country of birth) and literacy skill level. As explained in the appendix, immigrants born in the most developed (richest) countries of the world are grouped together and correspond to immigrants born in North American and Western European countries as well as Australia, Japan, New Zealand, Singapore, and South Korea. Immigrants born

\textsuperscript{13} Norway and Sweden have average literacy scores of 278 and 279 respectively (OECD 2013a).
in other countries fall into the category ‘Foreign-born (less developed countries)’. Regarding literacy level, individuals scoring level 3 or higher, i.e., 276 points or over on the literacy scale, are grouped together and are compared with those scoring below 276 points.

**Figure 2:** Projected workforce population aged between 25 and 64 years old, by immigration status (country of birth) and literacy level, BASELINE Scenario, 2011–2061, Austria and Canada

As expected, under the BASELINE scenario the Austrian workforce slightly increases up to 2021, decreases afterwards, and resumes its 2011 size by 2061 at around 3.65 million persons. By contrast, the size of the Canadian workforce grows steadily from 15.2 to 19.4 million persons between 2011 and 2061. When these projection results are broken down by immigration status and country of birth, the characteristics of the Austrian workforce in 2061 are relatively close to its 2011 composition. The
proportion of native-born Austrians among the 25–64 workforce is projected to decrease from 80% in 2011 to 69% in 2031, but rises up to 78% in 2061. If the assumptions about future Austrian immigration composition hold true, the share of foreign-born in Austria (from most-developed countries) is likely to decline by almost half (from 7% in 2011 to 4% in 2061). As for Canada, the proportion of foreign-born workers (regardless their country of birth) will increase from 25% to 41%. This proportion is almost twice as high as the proportion projected for Austria in 2061 (22%). The increase in Canada is fuelled entirely by immigrants coming from less-developed countries. As a result, the proportion of native-born Canadians in the 25–64 workforce decreases from 75% in 2011 to 59% in 2061. Interestingly, in both countries the absolute number of native-born workers is likely to remain quite stable over time. The projected number of immigrants (immigration volume) only impacts the relative prevalence of native-born workers among the total workforce.

When looking at these projections by literacy level, in Canada the distribution of the workforce remains fairly stable. The proportion of workers scoring level 3 or higher slightly decreases, from 53% in 2011 to 49% in 2061. In Austria the rise of the overall skill level of the workforce is clear. The proportion of workers scoring level 3 or higher increases from 50% in 2011 to 61% in 2061.

### 6. Discussion

With a specific focus on two developed countries, Austria and Canada, this research uses microsimulation models to assess how education and immigration levels impact workforce size and average literacy skill level. Comparing countries is not always an easy task. The use of different ‘what-if’ scenarios provides a basis for making such comparisons.

The IM_USRATE and the IM ZERO scenarios reveal strong similarities between Austria and Canada regarding the demographic dynamic (population ageing) that drives the renewal of their workforces. Indeed, when immigration rates are kept constant at 0% or 0.35%, the projected workforce (expressed in relative terms) evolves in a similar manner for both countries. Under constant and comparable immigration assumptions, the results show that growth of workforce size relies heavily on immigration intake. Regarding education, whether or not increasing trends hold true in the future has very little impact on the projected expansion of the future workforce. On the other hand, rising educational attainment trends have a significant influence on the average literacy skills of the workforce and the overall adult population aged 25 to 64 years old.

Applying current official immigration assumptions and extrapolating the rising education trends in a projection scenario (BASELINE) shows that Austria and Canada
have adopted different strategies regarding the future development of their workforces. Under actual Austrian immigration targets, the total workforce (25–64 years old) is likely to be the same size in 2061 as it was in 2011 (stable workforce population). This strategy appears to foster improvement in the skills of the active Austrian population. Again, these projections do not predict the future; official national immigration targets are likely to evolve and might in fact change rather quickly to better adapt to future socioeconomic contexts.

In Canada the growth of the workforce seems to be the most important objective, despite probable negative impacts on the average skills proficiency of the workers, and possibly on labour in general. While the contributions of immigration are still the subject of debate in public discourse, the current Canadian policy does not seem to be supported by evidence-based information but rather appears to be determined by faith in the long-term benefits of high immigration levels (Green and Green 2004; Romaniuk 2017). As Beaujot (2017) points out, admitting more people fuels Gross Domestic Product (GDP) growth in absolute terms but does not have a significant positive impact on GDP per capita. Moreover, immigration has no effect on employment, wages, innovation, or patenting (Green and Worswick 2017; Picot 2013). In our view, the persistent argument put forward since the late 1980s for ever-increasing immigration levels in Canada mostly relies on subjective arguments (related to nation-building and the place of Canada in the world) rather than scientific evidence on economic benefits for Canadians (GDP per capita growth).

In terms of immigration rate, the Austrian targets are close to most G7 and other developed countries (OECD 2017b). This might come with a certain macroeconomic price, as it will not foster sustained growth of the workforce. A stable workforce as opposed to a growing workforce might hinder long-term GDP growth in absolute terms, and thus general expansion of the national economy. However, a more productive workforce might mitigate this potential missed growth due to the increasing education level and average literacy. This strategy could in fact be beneficial in terms of GDP per capita, as a higher skilled population could result in increased productivity, particularly in the context of the knowledge economy and knowledge-based societies.

Inversely, the Canadian workforce is likely to increase in size due to Canada’s very high immigration targets. The challenge of this strategy resides in its negative impact on the average literacy skills of the population, despite a projected increase in overall educational attainment. The larger workforce will be not only less skilled but also more diverse: the proportion of foreign-born Canadians among the workforce will increase from 25% to 41%. This means a more culturally diverse workforce in terms of religious affiliation, visible minority groups, mother tongue, etc. Although diversity can bring opportunity to national economies, studies on ethno-cultural diversification show
that it can also have negative economic and social effects (Collier 2013; Levy 2017; Patsiurko, Campbell, and Hall 2012; Portes and Vickstrom 2011).

Microsimulation models constitute innovative tools to better inform “the public, but also to provide the means for policy makers and advisors to assess the impacts of their policies and programmes and to increase their effectiveness and cost-efficiency” (Wolfson 2011: 3). The models and their projection results presented in this paper provide clear examples. They project the impacts of different plausible immigration policies regarding immigration volume and characteristics in conjunction with future demographic trends. Indeed, the models show the apparent relationship between immigration volume and population growth. The recent ‘refugee crisis’ has modified Austria’s immigration rate – momentarily – with a significant impact on its population structure. Results also show the extent to which the characteristics of newcomers can impact the literacy skills of the total workforce. In a simulation where the Canadian points-based system selects immigrants with higher language skills (the proportion of immigrants whose mother tongue is either French or English is increased by 11%), the projected skills decline of the Canadian workforce is completely mitigated, while immigration targets are kept relatively high (official targets).

On labour policies, the projections presented in this paper assess skills on the supply side. More detailed studies could analyse future potential shortages (or oversupply) in terms of number of workers, as well as in terms of skills and education level. Further policies and programmes targeting adult training or linguistic, economic, and social integration of immigrants can also be tested with our models.

However, the models have some limitations. First, the labour force participation module of both models derives participation rates to provide adequate aggregate cross-sectional descriptions. However, the module yields incoherent individual life courses, as there are no specific transition probabilities between the active and the inactive states. Indeed, individual actors are susceptible to changing their participation status from one year to another without any consideration in the calculation of the previous activity status. Further development of the models should introduce transition probabilities between states according to characteristics of the actor as well as according to duration in a given state.

Likewise, it would be interesting to have a dynamic projection module for literacy skills proficiency that could simulate the individual skill trajectories. The parameter estimations of the actual projection module provide some hint of the forces that underlie skill gain and loss for people aged between 25 and 64 years old. However, the model generates statically derived projections as it imputes a literacy score to individuals based on many characteristics. Longitudinal data on adult literacy skills\(^{14}\) would be needed to generate dynamic literacy projection results.

\(^{14}\) To our knowledge such data only exists for Germany (see Rammstedt et al. 2017).
In the actual version of the models the projected variation of the literacy score is constrained by the regression models’ coefficients shown in the Appendix. Consequently, the models do not correctly account for variance observed in the PIAAC survey data. A component of uncertainty could be added in further development of the models, although the projection outcomes would not be substantially different. To test the extent to which different sociodemographic characteristics influence the literacy skills of the future workforce, the version of the models used in this study is convenient and generates pertinent results.

7. Conclusion

This paper shows that under a constant immigration rate of 0.35%, microsimulation models project a declining labour force population for both Austria and Canada between 2011 and 2061. They project that the active population aged 25 to 64 years old will decrease by 4% in Austria and by 10% in Canada during the period. Due to demographic ageing the overall participation rates (of the 15+ population) are also likely to decline. The rising trends in educational attainment are positively affecting the overall skill level of the workforce, but changes in immigration level can offset this effect. The projected labour force populations of Austria and Canada are different, mainly due to their divergent immigration contexts and targets. The two countries appear to have adopted diametrically opposed strategies with regards to the future development of their workforce: in the coming decades, Austria will rely on a higher-skilled workforce, whereas Canada will count on a fast-growing workforce. This research provides an analytical tool for understanding the evolution of skills in the Austrian and Canadian contexts based on foreseeable changes in population characteristics.

Microsimulation models, though heavily dependent on data, are powerful tools that can inform research in an innovative way, for a broad range of issues. The models presented in this paper are built according to a research framework (Bélanger et al. 2019) that aims to assess the impact of population dynamics in Western countries and the changes that are likely to happen in terms of ethno-cultural and sociodemographic characteristics. Capable of projecting the population according to numerous variables, these models generate workforce-relevant results to guide policymakers in their decisions regarding immigration policy and the conditions for education skills formation. For example, ‘what-if’ scenarios can be used to illustrate the effects of different immigrant selection patterns on the future size of the workforce, and also on the literacy proficiency of the projected labour force.
The methodology presented in this paper could easily be applied to other countries (or regions) that participated in the PIAAC survey. Information-processing skills such as those measured in the PIAAC survey are important assets in contemporary society. There are numerous positive relationships between these skills and many indicators of wellbeing and labour market success (OECD 2016). Understanding better how and why there is a significant gap in literacy proficiency between the foreign- and native-born in the OECD countries is crucial. Immigrants’ lower literacy proficiency has important implications for the overall skills of the future workforce, since this gap is comparable in magnitude to the difference between respondents having upper secondary education or tertiary education (Bonikowska, Green, and Riddell 2008; OECD and European Union 2014).

Relationships between social and economic outcomes and policies are generally complex and multifactorial. Microsimulation models based on detailed microdata provide innovative tools to answer carefully constructed ‘what-if’ questions and to assess the specific impact of underlying processes on key outcomes. By comparing the outcomes of these scenarios with the baseline situation, it is possible to measure the influence of specific processes and dynamics. Microsimulation models that project the population (and its sociodemographic characteristics) can be developed and used to compare the impacts of different immigration histories, different foreign-born populations in terms of sociodemographic characteristics, different integration policies, and different educational systems. The projection methodology and models can always be further improved, elaborated, and above all adapted to country-specific contexts. Microsimulation models fed with PIAAC data can significantly help to address policy issues and tackle broad societal challenges in contemporary societies.

8. Acknowledgments

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15 The Survey of Adult Skills (PIAAC) was conducted in over 40 countries and regions, including most OECD countries.
programming and design. The authors are also grateful to Mr. Nick Gailey for language editing and his valuable comments. Finally, the authors would like to thank the anonymous reviewers and conference discussants for their helpful and constructive comments that greatly contributed to improving the final version of the paper.
References


Appendix: The calculation of literacy skill module parameters

This section presents the regression analyses conducted to identify the determinants of literacy proficiency. The analyses are stratified by immigration status in order to consider variables specific to the foreign-born population, such as age at immigration, country of birth, etc. Based on both economic and sociodemographic literature and on our theoretical framework, the most pertinent independent variables fall into four categories:

1. Sociodemographic variables (age, sex, province, area of residence)
2. Human capital variables (education level, mother tongue, language spoken at home)
3. Socioeconomic variables and life-wide factors (mother’s education level, literacy skills used at home, at work (labour force status), and in everyday life)
4. Variables linked with characteristics that are specific to immigrants, such as age at immigration, years in host country since migration, immigration category, country of birth, and country of highest diploma.

Table A-1 presents the regression coefficients for both Austria and Canada. The complete model contains all pertinent independent variables, while the simplified model coefficients constitute the parameters used in the microsimulation models to derive literacy skills proficiency. Even though this simplified model does not include all significant variables from the complete model, notably mother’s education level and life-wide factors, the predictive power of the model is not substantially affected. More precisely, compared to the complete model the adjusted R² of the simplified models is reduced by four to five points.

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16 Mother’s education level is also a measure of social and cultural capital, which directly and indirectly impacts on children’s skill development and socioeconomic status. See Augustine and Negraia (2018).
17 Reading books at home; writing letters, memos, or emails at work.
18 The categories of both country of birth and country of highest diploma variables are defined in the same way. The most developed (richest) countries of the world are grouped together and correspond to Western Europe, North America, Australia, Japan, New Zealand, Singapore, and South Korea.
Table A-1: Estimated coefficient from linear regressions with log of literacy score as the dependent variable, stratified by immigration status, 25–64 years old, Canada and Austria, PIAAC 2012

<table>
<thead>
<tr>
<th>Variables</th>
<th>Canada Complete model</th>
<th>Canada Simplified model</th>
<th>Austria Complete model</th>
<th>Austria Simplified model</th>
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<td>For.-born</td>
<td>Nat.-born For.-born</td>
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<tr>
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<td>β In(Literacy score)</td>
<td>β In(Literacy score)</td>
<td>β In(Literacy score)</td>
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<td>0.058*</td>
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http://www.demographic-research.org
Table A-1: (Continued)

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<td>Nat.-born For.-born</td>
<td>Nat.-born For.-born For.-born</td>
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<td>ln(Literacy score)</td>
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<td>Other immigration</td>
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<td>Country of highest diploma</td>
<td>Host country (Ref. cat.)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Most developed countries</td>
<td>–0.032</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Less developed countries</td>
<td>–0.063*</td>
<td>–0.062*</td>
<td>–0.010</td>
</tr>
<tr>
<td>Constant</td>
<td></td>
<td>5.698*</td>
<td>5.708*</td>
<td>5.668*</td>
</tr>
<tr>
<td>Sample size (n)</td>
<td>16,132</td>
<td>3,463</td>
<td>16,132</td>
<td>3,463</td>
</tr>
<tr>
<td>Adjusted $R^2$</td>
<td>0.339</td>
<td>0.385</td>
<td>0.291</td>
<td>0.345</td>
</tr>
</tbody>
</table>

Note: * Coefficient significant at 5% significance level; † Reference category is ‘Active.’

More importantly, the value of the coefficients and the significance level of the simplified models are quite close to those of the complete models. The sample size is bigger in Canada and the foreign-born were oversampled, increasing the predictive power of the regression analyses.

In a nutshell, education is the main driver of literacy skills, with a stronger effect for immigrants. The language spoken at home is also a significant predictor of literacy proficiency, but to a lesser extent. The results also show that literacy proficiency declines with age. Furthermore, the life-wide factors (use of literacy skills in everyday life) have a statistically significant difference from zero at the 5% level. This result suggests that literacy skills have the ‘use it or lose it’ dimension put forward in the Practice-Engagement theory. The regression coefficients show a significant impact of the mother’s education variable on literacy scores. This impact is slightly stronger for the foreign-born population than for the native-born.

Finally, of the immigration and integration variables, both age at immigration and length of stay in Canada remain statistically insignificant dimensions. In the simplified model, country of birth becomes significant, as it probably captures some of the effect
of immigration category (omitted variable). The positive impact on skills proficiency of having a Canadian diploma is significant and of similar strength to that observed in the complete model. In Austria, immigration variables are not significant in the complete model, but coefficients point in the expected direction. For example, immigrants who arrive before the age of 15 have higher skills than those who arrive when older and immigrants from rich countries have higher skills than those from poor countries. However, these variables are correlated with language skills and multicollinearity makes it hard to get significant coefficients. Furthermore, the foreign-born sample is small, which reduces the statistical power of the regression analyses. In the simplified regression model, language as well as other dimensions is not taken into account, and the immigrants’ characteristics (age at immigration and country of birth) therefore become significant predictors of literacy skills.

Additional analyses (results not shown here) investigated education-related effects. Regression results stratified by education level did not reveal significant variation across education groups. Since the sample size among the strata was low – especially among the least-educated subgroup of immigrants – these regression results were not robust enough to be used as parameters for the microsimulation model.