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#### **ABSTRACT**

## Educational Mismatch: Are High-Skilled Immigrants Really Working at High-Skilled Jobs and the Price They Pay if They Aren't?\*

This paper examines the incidence of the mismatch of the educational attainment and the occupation of employment, and the impact of this mismatch on the earnings, of high-skilled adult male immigrants in the US labor market. Analyses for high-skilled adult male native-born workers are also presented for comparison purposes. The results show that overeducation is widespread in the high-skilled US labor market, both for immigrants and the native born. The extent of over-education declines with duration in the US as high-skilled immigrants obtain jobs commensurate with their educational level. Years of schooling that are above that which is usual for a worker's occupation are associated with very low increases in earnings. Indeed, in the first 10 to 20 years in the US years of over-education among high-skilled workers have a negative effect on earnings. This ineffective use of surplus education appears across all occupations and high-skilled education levels. Although schooling serves as a pathway to occupational attainment, earnings appear to be more closely linked to a worker's occupation than to the individual's level of schooling.

JEL Classification: I21, J24, J31, J61, F22

Keywords: immigrants, skill, schooling, occupations, earnings, rates of return

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### EDUCATIONAL MISMATCH: ARE HIGH-SKILLED IMMIGRANTS REALLY WORKING AT HIGH-SKILLED JOBS AND THE PRICE THEY PAY IF THEY AREN'T?

#### I. INTRODUCTION

The United States is a home to millions of immigrants. Her "Golden Door" has been open to many flows of immigrants that were "the wretched refuse of your teeming shore". At the same time, however, from Colonial times to the present, the US has attracted many skilled immigrants.<sup>1</sup> The high-skilled immigrants currently in the US are the subject of this study.

Figure 1 displays the legal permanent resident flow into the US between 1986 and 2007. These numbers reflect both new arrivals and adjustments of status among those who already lived in the US. Permanent residence status is primarily gained on the basis of family relationship with a US citizen or legal permanent resident (Immediate Relatives and Family-sponsored preferences), with skills serving as a much smaller, but the second largest, category (Employment preferences) (see Table 1 for 2007 admissions). Figure 1 also provides information on the number of legal permanent residents in the employment preference categories.<sup>2</sup> The number of immigrants entering the US in the employment preference categories has increased considerably over the past two decades. In 1986 they numbered 56,617, or 9.4 percent of the total immigration, while in response to 1990 legislation to increase their numbers, in 2007 they numbered 162,176, or 15.4 percent of

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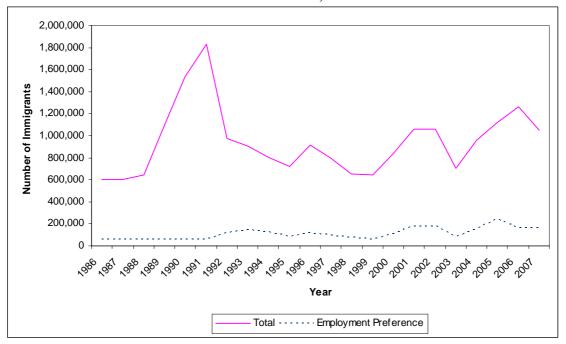
<sup>&</sup>lt;sup>1</sup> For a study of high-skilled immigrants to the US in the 19<sup>th</sup> and early 20<sup>th</sup> centuries, see Ferrie (2009).

<sup>&</sup>lt;sup>2</sup> The employment preference categories cover: (i) priority workers; (ii) professionals with advanced degrees or aliens with exceptional ability; (iii) skilled workers, professionals without advanced degrees and needed unskilled workers; (iv) special immigrants, such as religious workers; and (v) employment creation immigrants (*i.e.*, investors). The data include the immediate family members (spouse and minor children) of the principal applicant recipients of employment visas. They typically constituted about one-half of the category.

the total immigration, although about half of these in both years were the spouses and minor children of principal applicants.

Figure 1

Legal Permanent Resident and Employment Preference Visas,
Fiscal Years 1987 to 2007, United States



<u>Note</u>: The spike in permanent resident visas from 1989 to 1992 is related to the granting of amnesty to nearly 3 million illegal migrants under the 1986 Immigration Reform and Control Act. <u>Source</u>: 2004 and 2007 Yearbook of Immigration Statistics.

Table 1

Immigration by Type of Visa, United States, 2007

Catagory	Immigrants (in thousands)	
Category	mousanus)	
Immediate Relatives of US Citizens	494	}688
Family Sponsored Preferences	194	1000
Employment Based (and their families)	162	
Diversity	42	
Refugees, Asylees, Parolees	138	
Other	20	
Total	1,052	

Note: Detail may not add to total due to rounding.

<u>Source</u>: Immigration Statistics of the United States 2007, Department of Homeland Security, 2008.

Understanding how employment preference immigrants perform in the US labor market is important from the perspective of guiding the mix of immigrants: whether there are relatively more of "the wretched refuse of your teeming shore" or more high-skilled employment preference immigrants. Unfortunately, visa category information is not available in the data sets, such as from the Decennial Census, which are otherwise most useful for labor market analyses of immigrants in the US. Instead, therefore, this paper looks at all skilled foreign-born workers, regardless of their visa status, including those on temporary work visas (*e.g.*, H1-B visa recipients).

The study adopts perspectives from the over-education/under-education literature. This literature proposes that there is a "usual" education level for each occupation. Some workers will have this level of education, and will therefore be regarded as being matched to the typical educational requirements of their job. Other workers will have a higher level of education than that which is usual in their job. These workers with "surplus" years of schooling are viewed as being over-educated. Still other workers will have a lower level of education than that which is usual in their job. Such workers are viewed as being under-educated. Chiswick and Miller (2008)(2009a) show that, for analyses of the US and Australia, this framework yields important insights into the international transferability of human capital for immigrant workers across all skill levels. The focus here, however, is on high-skilled immigrant workers.

Section II presents a discussion of the determinants of the "mismatch" of education and occupation in the labor market. While the factors that bring about this

<sup>&</sup>lt;sup>3</sup> In the immigration literature this is frequently referred to as the non-recognition of foreign educational credentials.

mismatch for the native born also apply for the immigrants, two additional factors (skill transferability and selectivity in migration) also apply for immigrants.

Section III provides an overview of data on the education levels of the native born and foreign born. A selection of previous studies in the over-education/under-education literature is briefly reviewed in Section IV.<sup>4</sup> The broad aim of this review is to highlight methodological issues pertinent to a study of high-skilled immigrants. Section V outlines the empirical framework adopted in this study, and provides information on the data sources. The statistical analyses of the extent of the educational mismatch and the earnings consequences of these mismatches are presented in Section VI. Section VII concludes, with a summary and policy implications of the findings.

#### II. WHY WOULD THERE BE EDUCATIONAL MISMATCHES?

Consider the typical or usual level of education in an occupation. Why would there be educational mismatches, that is, individuals whose educational attainment differs from the "norm" in their occupation?

The usual level or norm is merely a measure of central tendency. Depending on the particular technology that they employ, or the educational attainment of the labor market from which they draw their labor supply, firms may have a different optimal level of education for their workers in a particular occupation compared to the occupation as a whole nationwide. Workers also differ by age and hence there are cohort differences in when they received their formal schooling, when they joined the labor force, and the extent of their labor market experience. Mismatches related to cohorts may arise if there has been an upgrading of educational requirements for new hires, but longer term employees are retained

<sup>&</sup>lt;sup>4</sup> For a fuller review, see Chiswick and Miller (2008).

because of their seniority or for whom the greater on-the-job training (labor market experience) compensates for their falling behind the educational norms for new hires. The mismatches here would be over-educated new hires and under-educated established workers compared to the average worker currently in place.

Workers clearly differ in characteristics that may be difficult, if not impossible, to measure in survey or census data, but which may be revealed in the labor market. These unmeasured characteristics include dimensions of worker and allocative (decision making ability) efficiency, ambition, aggressiveness, energy, job dedication, favorable and unfavorable personality traits, etc. Those with higher levels of desirable unmeasured abilities can attain a higher level occupation for the same level of schooling, and thereby appear to be under-educated. On the other hand, those who the market evaluates as being deficient in beneficial unmeasured traits are more likely to be relegated to occupations that are at a lower level compared to their schooling, and hence appear to be over-educated given their occupation.

The reasons just discussed for educational mismatches would apply equally well to native-born and foreign-born workers. There are, however, immigrant-specific factors that may contribute to a greater mismatch of education and occupation among the foreign born in the labor market – the limited international transferability of skills and selectivity in migration.

For most immigrants to a destination, skills acquired in the country of origin are not perfectly transferable. These skills include information about how labor markets operate, as well as destination language skills. There may be occupation-specific skills that are not readily transferable because of differences in type of technology (*e.g.*, English measures vs.

metric system, legal systems based on English common law vs. Napoleonic code). There may be differences in level of technology because of differences in capital/labor ratios or relative factor prices (*e.g.*, consider high-technology medicine in the US vs. low-technology medicine in the former Soviet Union and LDCs). Moreover, there may be barriers to entry into the destination occupations that immigrants trained for and practiced in their origin (*e.g.*, occupational licensing, union regulations, and governmental requirements, such as citizenship). In addition, there may be cultural differences that make it difficult for immigrants in certain occupations to "transfer" their skills to the destination labor market.<sup>5</sup>

A frequent concern expressed by immigrants, and those who assist their integration into the destination labor market, is the non-recognition of the immigrants' pre-migration skills, whether acquired in school or on the job. In some instances this is due to occupational licensing, but in other instances it may arise from understandably risk averse employers and consumers not knowing how to evaluate foreign credentials compared to the credentials of workers trained in the destination.<sup>6</sup>

Finally, one cannot rule out discrimination against immigrants reducing their ability to transfer their skills in whole or in part to the destination.

The lesser the degree of transferability of skills from the origin to the destination the greater would be the occupational downgrading of the immigrant, and hence the greater

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<sup>&</sup>lt;sup>5</sup> For, example, Remennick (2008) found that primary and secondary school teachers from the former Soviet Union who immigrated to Israel generally could not make the adjustment from the rigid, highly disciplined, highly structured Soviet classroom to the informal, flexible, Israel classroom with little structure. It was not the teaching of the subject matter or the language issues that were so difficult to overcome, but the school and classroom cultural gap was too great for the teaching skills to be transferable.

<sup>&</sup>lt;sup>6</sup> The issue of the non-recognition of the skills of immigrant physicians in the US and Canada is the theme of McDonald, *et al.* (2009). For a study of the adjustment of high-skilled immigrants in Israel, see Cohen-Goldner and Weiss (2009).

would be the appearance of over-education of immigrants in their occupations. With the passage of time in the destination, however, investments are made in destination human capital, either to modify (increase the transferability of) pre-migration skills, or to acquire new skills, occupational upgrading occurs and the extent of over-education would diminish.

A second immigrant-specific consideration is selectivity in immigration. For several reasons, there is a tendency for economic migrants to be favorably selected for labor market success in the destination (Chiswick 1999, 2008). Indeed, economic migrants by definition have success in the destination as their primary goal (supply of immigrants). Moreover, some immigrants are specifically granted visas (demand for immigrants) on the basis of their high levels of skill, although the relative importance of employment-based visas varies across destinations. Combining the self-selection (supply) and employment visas (demand for high-skilled immigrants) considerations suggests that there is, in general, favorable selectivity among immigrants.

Other measured variables the same, including educational attainment, this suggests more favorable unmeasured dimensions of ability among immigrants compared to the others in the origin who do not migrate. If these unmeasured dimensions of ability have a similar distribution among the native-born population in the origin and the destination, by implication the migrants have, on average, a higher level of unmeasured dimensions of ability than do the native born in the destination. Then, if the usual educational attainment in an occupation is based on the native-born population, the higher level of unmeasured ability would enable the immigrants to attain a higher occupational level than the destination native born with the same level of schooling, or alternatively gain employment in the same occupation as more highly educated natives. Hence they would appear to be under-

educated.

In summary, in the labor market one would expect to observe workers who appear to be over-educated and under-educated relative to the usual educational attainment in their occupation. In addition to the factors relevant for the native born, immigrants have two additional reasons for the education-occupation mismatch. The less than perfect international transferability of skill will tend to result in the over-education of immigrants, that is, a tendency for them to be in occupations in which the usual schooling level is less than theirs. On the other hand, the favorable selectivity of immigrants will tend to result in their being under-educated, that is, working in occupations in which the usual education level is higher than theirs. The issue of skill transferability is more intense the higher the level of skill, while the issue of selectivity is more intense the higher is the ratio of out-of pocket or direct costs of immigration to the opportunity cost of time, that is, it is more intense for lower-skilled workers (Chiswick 1999, 2008; Chiswick and Miller 2008). As a result, in a study of high-skilled immigrants it is to be expected that the dominant educational mismatch will be over-educated immigrants.

#### III. EDUCATION LEVELS OF THE NATIVE BORN AND FOREIGN BORN

Figure 2 presents information on the distribution of education levels of native-born and foreign-born males, aged 25 years and over in 2000.<sup>7</sup> This figure shows that only around 14 percent of native-born males left school before completing high school, while 33 percent are classified as high school graduates, 18 percent attended college but did not receive a degree, seven percent attained an Associate degree, 18 percent a

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<sup>&</sup>lt;sup>7</sup> See Appendix A for the definition of the various educational categories. Sensitivity tests were performed for alternate measures of years of schooling for those with Master's, Professional, and Doctorate degrees as their highest level of schooling. The findings are essentially invariant with respect to these alternative values.

Bachelor's degree, six percent a Master's degree and four percent either Professional or Doctorate degrees.

The data for foreign-born males show a much lower mean and a greater inequality in the distribution of schooling. A major difference occurs among the very early school leavers. Only 14 percent of native-born adults did not complete high school, whereas 34 percent of the foreign born are in this category. It is, therefore, this relatively high representation in the early school leaver category that is responsible for the mean level of education for the foreign born (11.76 years in 1999) being around 1.5 years less than the mean level of education for the native born (13.13 years).

The foreign born and native born have similar proportions with higher education. Among the foreign born, 15 percent have only a Bachelor's degree, and for the native born it is 18 percent. Seven percent of the foreign born have a Master's degree, compared to six percent among the native born. Finally, whereas four percent of the native born have Professional degrees or Doctorates, five percent of the foreign born fall into this category. Thus, the foreign born are more heavily represented at the lowest and, to a smaller extent, the very highest, educational levels.

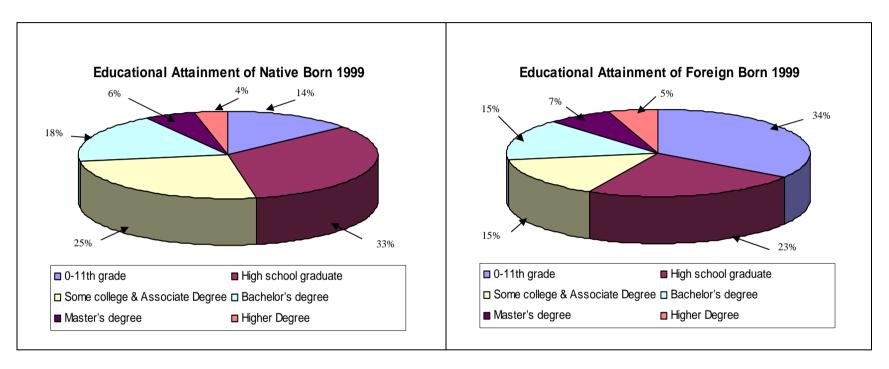
The skilled immigrant group that is the focus of this study can be defined in various ways. There could be a focus on the approximately 28 percent of the population of each birthplace group with Bachelor's or higher degrees. Or a more restrictive definition covering those with Master's or higher degrees could be considered, 12 percent of the immigrant population and 10 percent of the native-born population. Both definitions are considered in the analyses that follow.<sup>8</sup>

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<sup>&</sup>lt;sup>8</sup> See Ferrie (2009) for discussion of why the definition of skilled immigration is time and place specific.

Figure 2

Distribution of Education Levels of the Males Aged 25 Years and Over, by Nativity, 1999



Note: Higher degree includes those with degrees above the Master's level, including Professional (*e.g.*, MD, LLB) and Doctorate (PhD) degrees. Source: Current Population Survey, 1999.

#### IV. LITERATURE REVIEW ON THE ORU TECHNIQUE

The over-education/under-education literature has been used to examine the allocation of workers across the over-educated, under-educated and correctly matched job categories in the US. This literature has also examined the impacts on earnings of educational mismatches. The latter research has been based on a variant of the human capital earnings function that has been termed the ORU (Over-education/Required education/Under-education) specification. In this model, the dependent variable is the natural logarithm of earnings  $(\ln Y_i)$  and the variable for actual years of education is decomposed into three terms. That is,

(1) 
$$\ln Y_i = \alpha_0 + \alpha_1 \mathbf{O} \text{ver\_Educ}_i + \alpha_2 \mathbf{Req\_Educ}_i + \alpha_3 \mathbf{U} \text{nder\_Educ}_i + \dots + u_i$$

where Over\_Educ = years of surplus or over education,

**R**eq\_Educ = the usual or reference years of education,

Under\_Educ = years of deficit or under education,

and the actual years of education equals **O**ver\_Educ + **R**eq\_Educ - **U**nder\_Educ. Note that for each individual, "**O**ver\_Educ" and "**U**nder\_Educ" cannot both be positive. Either one or both must be zero. Equation (1) will also contain other variables generally included in earnings functions, such as years of labor market experience, marital status, location, veteran of the US Armed Forces, race/ethnicity, and variables specific to the foreign born, such as duration of residence in the US and citizenship status.

All studies report that there is a high incidence of educational mismatches in the US labor market. In most studies equation (1) is estimated on samples of all workers,

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<sup>&</sup>lt;sup>9</sup> The standard equation,  $\ln Y_i = \beta_0 + \beta_1 \text{Actual Educ}_i + ... + \upsilon_i$ , forces  $\alpha_1 = \alpha_2 = |\alpha_3|$ . As this condition does not hold, the ORU specification results in a higher R-squared and  $\alpha_2 > \beta_1$ .

though separate analyses are often undertaken for particular groups of interest. For example, Rumberger (1987) reported findings from estimations undertaken on separate samples of men and women. Duncan and Hoffman (1981) present results for four gender-race groups (White men, Black men, White women, Black women). Chiswick and Miller (2008) conduct separate analyses for foreign-born and native-born male workers, and among the foreign born by country of origin.

Some analyses extend the disaggregation of the sample beyond that based on nativity, gender or race to consider occupations (Rumberger (1987) and Verdugo and Verdugo (1989)). Rumberger (1987, p.31), for example, argued that "we would expect the estimated return to required and surplus schooling to vary across occupations just as the estimated return to actual schooling varies across occupations". Rubb (2003, p.54) explains that "The theory behind the occupational analysis is that some occupational groups may be better suited than others in using the surplus human capital of the overeducated workers". Rumberger's (1987) study was based on only five broad categories of occupations: (i) Professional/Managerial; (ii) Support; (iii) Craft; (iv) Operative; and (v) Service. Verdugo and Verdugo (1989) expanded the occupation-specific analyses to nine occupations. Other studies have focused only on particular skill segments of the labor force. Rubb (2003) and Duncan and Hoffman (1981), for example, studied the links between over-education and earnings among workers with post-college schooling.

In analyses of earnings, the return to years of education that are usual in an occupation ( $\alpha_2$ ) is typically much higher than the return to actual years of education ( $\beta_1$ ) (see Hartog, 2000). Years of education above those that are usual in a person's job are associated with a payoff that is much lower than the payoff to the education levels that

are usual for an occupation  $(\alpha_2 > \alpha_1)$ , whereas years of under-education are associated with an earnings penalty compared to those correctly matched  $(\alpha_2 > |\alpha_3|)$ . These earnings effects, however, have been shown to vary by nativity, occupation and skill level.

Chiswick and Miller (2008) report that the payoff to an actual year of education in the US 2000 Census was 10.6 percent for native-born males, and only 5.2 percent for foreign-born males. The payoff to a year of education that is usual in a person's job did not differ by nativity: it was 15.4 percent for the native born and 15.3 percent for the foreign born. A year of surplus schooling was associated with a payoff of 5.6 percent for the native born and of 4.4 percent for the foreign born. In comparison, the earnings penalty associated with a year of under-education was -6.7 percent for the native born and only -2.1 percent for the foreign born.

Vahey (2000) examined the incidence and returns to educational mismatch in Canada with a modification to the ORU model. Thus, the estimating equation in Vahey (2000) was:

(2) 
$$\ln Y_i = \gamma_0 + \gamma_1 \mathbf{O} \text{ver} \underline{-} \text{Educ}_i^A + \gamma_2 \mathbf{Req} \underline{-} \text{Educ}_i^A + \gamma_3 \mathbf{U} \text{nder} \underline{-} \text{Educ}_i^A + \dots + u_i$$

where the superscript A on the ORU variables simply indicates an alternative definition. In particular, Vahey (2000) defined  $\mathbf{Req\_Educ}_i^A$  as a vector of dichotomous variables for each usual level of education. Because the usual level of education was rarely more than one level from the attained level of education, in Vahey's (2000) empirical analysis a restricted specification was employed, where  $\mathbf{Over\_Educ}_i^A$  and  $\mathbf{Under\_Educ}_i^A$  comprised, for each usual level of schooling, single dichotomous variables for over-education and under-education regardless of the number of years.

Thus, the analyses of over-education and under-education have shown that knowledge of educational mismatch can enhance understanding of labor market outcomes. The efforts to extend the analyses to consider variation across education levels and across occupations revealed that this extension can be useful, although the limitations of these earlier studies prevent strong conclusions from being drawn. The analyses presented below, based on the large Public Use Microdata Sample from the 2000 Census, overcome these limitations, and demonstrate the considerable potential of study disaggregated by occupation and using more detailed information for the required level of school and for schooling mismatches.

#### V. MEASUREMENT OF MISMATCHES AND DATA

#### A. <u>Measurement</u>

A method is needed to identify the "required" or "usual" level of education in an occupation. For the purposes of this study, the Realized Matches (RM) technique is used. This is based on the actual educational attainments of workers in each occupation, and therefore reflects the outcome of the labor market matching process. Either the mean of educational attainments within each occupation (*e.g.*, Verdugo and Verdugo, 1989) or the modal educational attainment (*e.g.*, Cohn and Khan, 1995) may be used.

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<sup>&</sup>lt;sup>10</sup> Two other techniques are the Worker Self-Assessment (WSA) and the Job Analyst (JA) techniques, where the latter is based on "objective" evaluations of experts. For a comparative analysis of the WSA and RM techniques, see the methodological note in Chiswick and Miller (2009b). This shows there is a high degree of correlation between the WSA and RM data series, with the simple correlation coefficient between these measures being around 0.8 for all skill-nativity groups considered in this study. Under each of the three assessment methods, the "typical" or "required" level of education is related to the technology employed, relative factor prices, and the educational distribution of the population under study. There is no fixed or unique required level of education in an occupation, across either time or space.

#### B. The Data

The analyses reported below are based on the 2000 US Census five percent Public Use Microdata Sample, using the approximately 500 occupations that are separately identified. This data set contains information on labor market outcomes (earnings, occupation) and demographic characteristics (educational attainment, age, marital status, veteran of US Armed Forces, English proficiency, location, and among the foreign born, citizenship and duration of residence in the US). While this data source covers the entire population, the analyses are based on men aged 25 to 64 years who worked in paid employment in 1999. The analyses are restricted to those in non-military occupations, as these are the most likely to respond to market forces. Separate analyses are conducted for native-born workers and for foreign-born workers. Both wage and salary earners and the self-employed are covered by the study. All foreign-born men, and a 0.15 random sample of native-born men, meeting the sample restrictions are included in the analysis.

The modal level of education of native-born workers in the 2000 Census data is used to determine the usual level of education in each of the approximately 500 occupations. The focus on native-born male workers is appropriate where the economic majority group sets the norm for all workers in the occupation. This RM measure ranges from 12 years of schooling to the Professional and Doctorate degree categories (seven categories in total).

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<sup>&</sup>lt;sup>11</sup> Conventionally, a 64-year upper threshold has been used to minimise any selection bias associated with retirement from the paid labor force. Using a lower threshold of 54 years has no material effect on the regression estimates presented in Tables 4 and 5.

<sup>&</sup>lt;sup>12</sup> Chiswick and Miller (2008) report that tests of robustness with respect to alternative definitions of the population for defining the modal education showed virtually no substantive differences.

#### VI. STATISTICAL ANALYSIS

The statistical analyses that follow have several main sections. Section VI.A contains a brief overview of the incidence of educational mismatch in the US labor market. Section VI.B presents the analyses of the determinants of earnings for high-skilled workers: workers with a Bachelor's or higher degree, and workers with a Master's or higher degree. The analyses of earnings for the skilled workers are conducted separately by major occupation in Section VI.C. This will permit assessment of whether some occupations are able to utilize more effectively any surplus educational attainments. In VI.D the analysis of earnings is undertaken using the more flexible specification of the ORU model introduced by Vahey (2000). This approach offers advantages in terms of understanding whether the apparent inability of the labor market to effectively utilize surplus schooling depends on the level of schooling. Finally, Section VI.E reports findings from an analysis of the effects of education—actual years, usual years and surplus years—on earnings by duration of residence in the US.

#### A. The Incidence of Skill Mismatch

Table 2 lists the incidence of correctly matched education and mismatched education in the US labor market, based on the modal education in their occupation, by nativity, skill level and occupation, using data on adult males from the 2000 Census. The data for the native born are in standard font (first row) and the data for the foreign born are in italics (second row) for each occupation. The first three columns of the table cover all educational attainments, while the final two columns are for the two definitions of high-skilled workers employed in this study. When all workers are considered information is presented on under-education, correctly matched education and over-

education. When only high-skilled workers are considered, however, under-education is not a material issue as very few workers are in this category, and so only the incidence of over-education is presented, with the balance of the workforce being considered correctly matched.

Across all occupations (see the first row of data in Table 2) the rate of correctly matched education among the native born is around 40 percent, while the rates of undereducation and over-education are 26 percent and 33 percent, respectively. The rate of being over-educated among the foreign born is similar to that of the native born (29 percent). The foreign born, however, are far more likely than the native born to be undereducated (45 percent compared to 26 percent) and are far less likely than the native born to be in the correctly matched group (26 percent compared to 40 percent).

The patterns in the incidence of educational match/mismatch across occupations are affected by two sets of factors. First, the usual level of education varies by occupation, from 12 years in some occupations (*e.g.*, Sales and related) to a Doctorate in other occupations (*e.g.*, Life, Physical, and Social Science). Second, the proportion of highly educated workers varies across occupations. Hence the mean actual years of education by occupational group in Table 2 ranges from 12.19 years to 18.05 years among the native born, and from 9.24 years to 17.77 among the foreign born. <sup>13</sup>

In the fourth column of Table 2 the analysis is restricted to workers with at least a Bachelor's degree. Thus these workers will have, by definition, a higher mean level of actual years of education than the sample of all workers. This will tend to increase the incidence of over-education.

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<sup>&</sup>lt;sup>13</sup> These are based on imputed years of schooling where a Bachelor's degree is assumed to require 16 years, a Master's degree 17.5 years, a Professional degree 18.5 years, and a Doctorate degree 20 years.

Table 2

Incidence of Over-education, Correctly Matched Education and Under-education by Nativity, Skill Level and by Occupation, 25-64 Year Old Males, 2000 US Census

		of occupation	<u> </u>	Bachelor's	Master's
		All Skill Levels		Degree +	Degree +
Occupation	Under- educated (i)	Correctly Matched (ii)	Over- educated (iii)	Over- educated (iv)	Over- educated (v)
All Occupations	0.263 0.450	0.402 0.260	0.334 0.291	0.503 0.625	0.697 0.790
Management, Business and Financial Operations	0.323 0.369	0.361 0.281	0.315 0.350	0.452 0.578	0.867 0.965
Business and Financial Operations	0.239 0.335	0.452 0.404	0.309 0.261	0.353 0.460	1.000 1.000
Professional and Related	0.223 0.460	0.402 0.386	0.375 0.154	0.378 0.555	0.975 0.992
Architecture and Engineering	0.182 <i>0.411</i>	0.429 0.384	0.389 0.206	0.338 <i>0.553</i>	1.000 1.000
Life, Physical, and Social Science	0.393 <i>0.414</i>	0.438 0.406	0.169 0.180	0.424 <i>0.434</i>	0.694 <i>0.507</i>
Community and Social Services	0.176 0.209	0.387 0.332	0.437 0.459	0.240 0.296	0.428 <i>0.510</i>
Legal	0.116 <i>0.237</i>	0.790 0.572	0.094 <i>0.191</i>	0.122 0.266	0.073 0.181
Education, Training, and Library	0.460 0.520	0.411 0.329	0.129 0.151	0.488 0.552	0.804 <i>0.716</i>
Arts, Design, Entertain., Sports, and Media	0.160 0.222	0.383 0.302	0.458 0.475	0.301 <i>0.433</i>	1.000 1.000
Healthcare Practitioner and Technical	0.145 <i>0.190</i>	0.625 0.656	0.230 0.154	0.196 0.228	0.204 0.235
Healthcare Support	0.511 <i>0.517</i>	0.312 0.227	0.177 0.256	1.000 1.000	1.000 1.000
Protective Service	0.343 <i>0.413</i>	0.260 0.252	0.397 0.335	0.844 <i>0.916</i>	1.000 1.000
Food Preparation	0.395 0.205	0.343 0.208	0.262 0.587	1.000 1.000	1.000 1.000

Building and Grounds	0.314 <i>0.157</i>	0.442 0.196	0.245 0.647	1.000 1.000	1.000 1.000
Cleaning and Maintenance					
Personal Care and Service	0.442 <i>0.353</i>	0.323 0.254	0.235 <i>0.393</i>	0.784 <i>0.914</i>	1.000 1.000
Sales and Related	0.435 <i>0.443</i>	0.295 0.217	0.271 0.340	0.638 <i>0.812</i>	1.000 1.000
Office and Administrative Support	0.470 0.453	0.248 0.184	0.282 0.363	0.998 <i>0.99</i> 8	1.000 1.000
Farming, Fishing, and Forestry	0.263 0.051	0.407 0.092	0.330 0.857	1.000 1.000	1.000 1.000
Construction and Extraction	0.332 <i>0.171</i>	0.441 <i>0.219</i>	0.226 0.610	1.000 1.000	1.000 1.000
Installation, Maintenance, and Repair	0.336 0.302	0.406 0.246	0.258 0.453	1.000 1.000	1.000 1.000
Production, Transport, and Material Moving	0.376 0.259	0.460 0.228	0.164 0.513	1.000 1.000	1.000 1.000
Transportation and Material Moving	0.301 0.248	0.477 0.253	0.222 0.499	0.784 <i>0.953</i>	1.000 1.000

<u>Note</u>: For each occupation the data in the first row are for the native born and the data in the second row (italics) are for the foreign born. Based on realized matches (RM) procedure (mode). <u>Source</u>: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

These Table 2 column (iv) results show that in about one-third of the occupational groups all of the workers with at least a Bachelor's degree are over-educated, regardless of nativity group. The foreign born have a greater rate of over-education than the native born in the remaining occupations. Furthermore, when the analysis focuses on the group with a Master's degree or higher (see Table 2 column (vi)), all the workers in each nativity group are over-educated in over half of the occupational groups. The incidences of over-education are similar for both the native born and the foreign born in the remaining occupations, with the exception of the Legal occupation, where the rate of

over-education for the foreign born is only 18 percent and that for the native born is even lower, at 7 percent.

Table 3

Incidence of Over-education, Correctly Matched Education and Under-education for 25-64 Year Old Foreign-Born Males by Duration of Residence and Skill Level, 2000 US Census

		All Skill Levels		Bachelor's Degree +	Master's Degree +
Duration	Under- educated (i)	Correctly Matched (ii)	Over- educated (iii)	Over- educated (iv)	Over- educated (v)
All Durations	0.450	0.260	0.291	0.625	0.790
0-9	0.426	0.272	0.302	0.627	0.800
10-19	0.485	0.240	0.275	0.671	0.837
20-29	0.465	0.253	0.281	0.594	0.755
30+	0.383	0.295	0.322	0.578	0.734

Note: Based on realized matches (RM) procedure (mode).

Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

The incidence of educational mismatches can also be considered by duration in the United States, as is done in Table 3. Among high-skilled workers in the US for 10 or more years in 2000, the extent of over-education declines with duration of residence. This suggests that with duration in the US labor market immigrants are more likely to acquire the US-specific skills, credentials, and reputation that permit more workers to get jobs in occupations commensurate with their educational attainment. Note, however, that the degree of over-education is lower for those in the US fewer than 10 years in 2000 compared to those with a 10 to 19 years duration. The better occupational matching of the foreign born who came to the US in the 1990's may reflect cohort differences arising from the 1990 Immigration Act. This legislation had two major effects on this issue. One is that it increased the number of labor certification/employer sponsored visas, and

workers entering under these visas are more likely to be better matched than those entering under other visas, such as the family based, diversity, or refugee visas. The second is that the act created the H1-B (temporary worker) visas for employer sponsored high-skilled workers, where again, there would be a better matching (fewer overeducated workers).

Thus, educational mismatch, especially for over-educated workers, is a major feature of the US labor market. Its importance increases when the focus is on the most highly skilled workers. Indeed, in many occupations, all of the most highly educated workers are categorized as over-educated. This would be expected to have major implications for the earnings of these workers. These implications are explored in the following sub-sections.

#### B. <u>Analyses for High-Skilled Workers</u>

Table 4 presents results from the estimation of the standard and ORU models of earnings determination on a sample restricted to workers with at least a Bachelor's degree.

Table 4
Estimates of Standard and ORU Models of Earnings by Nativity, Skilled (Bachelor's or Higher Degree) 25-64 Year Old Males, 2000 US Census

Variable         Standard         ORU         Standard         ORU           Constant         4.073         4.131         4.669*         4.297           (52.08)         (54.16)         (71.98)         (67.29)           Educational         0.111         (a)         0.106         (a)           Attainment         (42.49)         (49.52)         (49.52)           Usual Level of Education         (47.85)         (64.45)           Years of Overeducation         (7.15)         (8.34)           Experience         0.057         0.059         0.031*         0.039*           (48.12)         (50.65)         (25.67)         (32.55)           Experience         -0.122         -0.124         -0.074*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061		Native Born		Foreign	Born
Educational 0.111 (a) 0.106 (a) (47.29)  Educational (42.49) (49.52)  Usual Level of (a) 0.122 (a) 0.140*  Education (47.85) (64.45)  Years of Over- (a) 0.020 (a) 0.019  education (7.15) (8.34)  Experience 0.057 0.059 0.031* 0.039*  (48.12) (50.65) (25.67) (32.55)  Experience -0.122 -0.124 -0.074* -0.085*  Squared/100 (39.64) (41.38) (24.61) (29.17)  Log Weeks Worked 0.999 0.979 0.972 0.945  (59.77) (59.75) (73.07) (72.22)  Married 0.302 0.271 0.232* 0.215*  (48.67) (44.59) (36.23) (34.64)  South -0.031 -0.034 -0.061* -0.054*  (5.43) (6.01) (9.86) (9.12)  Metropolitan 0.333 0.308 0.147* 0.154*  (36.82) (34.82) (8.28) (8.96)  Veteran of US -0.056 -0.043 -0.128* -0.106*  Armed Forces (7.22) (5.68) (8.97) (7.70)  Black -0.188 -0.162 -0.296* -0.262*  (17.17) (14.98) (30.40) (27.70)  English Very Well -0.072 -0.064 -0.141* -0.110*  (5.37) (4.79) (18.76) (32.84)  English Not -0.109 -0.099 -0.690* -0.492*  Well/Not at All (2.57) (2.35) (49.40) (35.70)  Years Since (a) (a) 0.035 0.024  Migration (YSM) (2.41) (6.07)  Citizen (a) (a) 0.259 0.278 0.322	Variable	Standard	ORU	Standard	ORU
Educational Attainment         0.111 (42.49)         (49.52)           Usual Level of Education         (a)         0.122 (49.52)           Education         (47.85)         (64.45)           Years of Overeducation         (a)         0.020 (a)         0.019           Experience         0.057 (48.12)         (50.65)         (25.67)         (32.55)           Experience         -0.122 (48.12)         -0.065         (25.67)         (32.55)           Experience         -0.122 (48.12)         -0.074* (40.074* (29.17)         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999 (0.979)         0.972 (0.945)         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302 (0.271)         0.232* (0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031 (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333 (0.308)         0.147* (0.154*)         0.154*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333 (0.308)         0.147* (0.154*)	Constant	4.073	4.131	4.669*	4.297
Attainment         (42.49)         (49.52)           Usual Level of Education         (a)         0.122         (a)         0.140*           Years of Over- education         (a)         0.020         (a)         0.019           Experience         0.057         0.059         0.031*         0.039*           (48.12)         (50.65)         (25.67)         (32.55)           Experience         -0.122         -0.124         -0.074*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061*         -0.054*           (543)         (60.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           (543)         (60.01)         (9.86)         (9.12)           Metropolitan         0.333		(52.08)	(54.16)	(71.98)	(67.29)
Usual Level of Education         (a)         0.122 (47.85)         (a)         0.140* (64.45)           Years of Overeducation         (a)         0.020 (7.15)         (a)         0.019 (8.34)           Experience         0.057 (50.65)         0.059 (25.67)         0.32.55)           Experience         -0.122 (50.65)         (25.67)         (32.55)           Experience         -0.122 (0.124)         -0.074* (0.085*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999 (0.979)         0.972 (0.945)         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302 (0.271)         0.232* (0.215*)           South         -0.031 (0.034)         -0.061* (0.064*)         -0.054*           (54.3)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333 (0.308)         0.147* (0.154*)         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128* (0.06*)         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)	Educational	0.111	<b>(a)</b>	0.106	(a)
Education         (47.85)         (64.45)           Years of Overeducation         (a)         0.020         (a)         0.019           education         (7.15)         (8.34)           Experience         0.057         0.059         0.031*         0.039*           (48.12)         (50.65)         (25.67)         (32.55)           Experience         -0.122         -0.124         -0.074*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162 <t< th=""><th>Attainment</th><th>(42.49)</th><th></th><th>(49.52)</th><th></th></t<>	Attainment	(42.49)		(49.52)	
Years of Overeducation         (a)         0.020         (a)         0.019           education         (7.15)         (8.34)           Experience         0.057         0.059         0.031*         0.039*           (48.12)         (50.65)         (25.67)         (32.55)           Experience         -0.122         -0.124         -0.074*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black	<b>Usual Level of</b>	(a)	0.122	(a)	0.140*
education         (7.15)         (8.34)           Experience         0.057         0.059         0.031*         0.039*           (48.12)         (50.65)         (25.67)         (32.55)           Experience         -0.122         -0.124         -0.074*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (	Education		<b>(47.85)</b>		(64.45)
Experience	Years of Over-	(a)	0.020	(a)	0.019
Care	education		<b>(7.15)</b>		(8.34)
Experience         -0.122         -0.124         -0.074*         -0.085*           Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (14.98)         (30.40)         (27.70)           English Very Well         -0.072         -0.064         -0.141*         -0.110*	Experience	0.057	0.059	0.031*	0.039*
Squared/100         (39.64)         (41.38)         (24.61)         (29.17)           Log Weeks Worked         0.999         0.979         0.972         0.945           (59.77)         (59.75)         (73.07)         (72.22)           Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (14.98)         (30.40)         (27.70)           English Very Well         -0.072         -0.064         -0.141*         -0.110*           (5.37)         (4.79)         (18.76)         (14.98)           English Not		(48.12)	(50.65)	(25.67)	(32.55)
Log Weeks Worked	Experience	-0.122	-0.124	-0.074*	-0.085*
Married	Squared/100	(39.64)	(41.38)	(24.61)	(29.17)
Married         0.302         0.271         0.232*         0.215*           (48.67)         (44.59)         (36.23)         (34.64)           South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (14.98)         (30.40)         (27.70)           English Very Well         -0.072         -0.064         -0.141*         -0.110*           (5.37)         (4.79)         (18.76)         (14.98)           English Well         -0.068         -0.055         -0.403*         -0.304*           English Not         -0.109         -0.099         -0.690*         -0.492*           Well/Not at All         (2.57)         (2.35)         (49.40)         (35.70)	Log Weeks Worked	0.999	0.979	0.972	0.945
South		(59.77)	(59.75)	(73.07)	(72.22)
South         -0.031         -0.034         -0.061*         -0.054*           (5.43)         (6.01)         (9.86)         (9.12)           Metropolitan         0.333         0.308         0.147*         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (14.98)         (30.40)         (27.70)           English Very Well         -0.072         -0.064         -0.141*         -0.110*           (5.37)         (4.79)         (18.76)         (14.98)           English Well         -0.068         -0.055         -0.403*         -0.304*           (1.97)         (1.64)         (42.61)         (32.84)           English Not         -0.109         -0.099         -0.690*         -0.492*           Well/Not at All         (2.57)         (2.35)         (49.40)         (35.70)           Years Since         (a)         (a)         0.009         0.011	Married	0.302	0.271	0.232*	0.215*
Metropolitan       (5.43)       (6.01)       (9.86)       (9.12)         Metropolitan       0.333       0.308       0.147*       0.154*         (36.82)       (34.82)       (8.28)       (8.96)         Veteran of US       -0.056       -0.043       -0.128*       -0.106*         Armed Forces       (7.22)       (5.68)       (8.97)       (7.70)         Black       -0.188       -0.162       -0.296*       -0.262*         (17.17)       (14.98)       (30.40)       (27.70)         English Very Well       -0.072       -0.064       -0.141*       -0.110*         (5.37)       (4.79)       (18.76)       (14.98)         English Well       -0.068       -0.055       -0.403*       -0.304*         (1.97)       (1.64)       (42.61)       (32.84)         English Not       -0.109       -0.099       -0.690*       -0.492*         Well/Not at All       (2.57)       (2.35)       (49.40)       (35.70)         Years Since       (a)       (a)       (a)       0.009       0.011         Migration (YSM)       (9.81)       (12.23)         YSM Squared/100       (a)       (a)       -0.005       -0.011 <td></td> <td>(48.67)</td> <td>(44.59)</td> <td>(36.23)</td> <td>(34.64)</td>		(48.67)	(44.59)	(36.23)	(34.64)
Metropolitan         0.333         0.308         0.147*         0.154*           (36.82)         (34.82)         (8.28)         (8.96)           Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (14.98)         (30.40)         (27.70)           English Very Well         -0.072         -0.064         -0.141*         -0.110*           (5.37)         (4.79)         (18.76)         (14.98)           English Well         -0.068         -0.055         -0.403*         -0.304*           English Not         -0.109         -0.099         -0.690*         -0.492*           Well/Not at All         (2.57)         (2.35)         (49.40)         (35.70)           Years Since         (a)         (a)         0.009         0.011           Migration (YSM)         (9.81)         (12.23)           YSM Squared/100         (a)         (a)         -0.005         -0.011           (2.41)         (6.07)         (6.07)           Citizen         (a)<	South	-0.031	-0.034	-0.061*	-0.054*
Veteran of US		(5.43)	(6.01)	(9.86)	(9.12)
Veteran of US         -0.056         -0.043         -0.128*         -0.106*           Armed Forces         (7.22)         (5.68)         (8.97)         (7.70)           Black         -0.188         -0.162         -0.296*         -0.262*           (17.17)         (14.98)         (30.40)         (27.70)           English Very Well         -0.072         -0.064         -0.141*         -0.110*           (5.37)         (4.79)         (18.76)         (14.98)           English Well         -0.068         -0.055         -0.403*         -0.304*           (1.97)         (1.64)         (42.61)         (32.84)           English Not         -0.109         -0.099         -0.690*         -0.492*           Well/Not at All         (2.57)         (2.35)         (49.40)         (35.70)           Years Since         (a)         (a)         0.009         0.011           Migration (YSM)         (9.81)         (12.23)           YSM Squared/100         (a)         (a)         -0.005         -0.011           (2.41)         (6.07)         (6.07)         (2.41)         (6.07)           Citizen         (a)         0.259         0.278         0.322	Metropolitan	0.333	0.308	0.147*	0.154*
$\begin{array}{c ccccccccccccccccccccccccccccccccccc$		(36.82)	(34.82)	(8.28)	(8.96)
Black	Veteran of US	-0.056	-0.043	-0.128*	-0.106*
English Very Well $(17.17)$ $(14.98)$ $(30.40)$ $(27.70)$ $(27.70)$ English Very Well $(5.37)$ $(4.79)$ $(18.76)$ $(14.98)$ $(14.98)$ English Well $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ English Not $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ English Not $(2.57)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since $(a)$	Armed Forces	(7.22)	(5.68)	(8.97)	(7.70)
English Very Well $-0.072$ $-0.064$ $-0.141*$ $-0.110*$ $(5.37)$ $(4.79)$ $(18.76)$ $(14.98)$ English Well $-0.068$ $-0.055$ $-0.403*$ $-0.304*$ $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ English Not $-0.109$ $-0.099$ $-0.690*$ $-0.492*$ Well/Not at All $(2.57)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since (a) (a) $0.009$ $0.011$ Migration (YSM) $(9.81)$ $(12.23)$ YSM Squared/100 (a) (a) $-0.005$ $-0.011$ $(2.41)$ $(6.07)$ Citizen (a) $(3)$	Black	-0.188	-0.162	-0.296*	-0.262*
English Well $(5.37)$ $(4.79)$ $(18.76)$ $(14.98)$ $(14.98)$ $(1.97)$ $(1.64)$ $(4.61)$ $(32.84)$ $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ $(32.84)$ English Not $(2.57)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since $(3)$ $(3)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since $(3)$		(17.17)	(14.98)	(30.40)	(27.70)
English Well $-0.068$ $-0.055$ $-0.403*$ $-0.304*$ $(1.97)$ $(1.64)$ $(42.61)$ $(32.84)$ English Not $-0.109$ $-0.099$ $-0.690*$ $-0.492*$ Well/Not at All $(2.57)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since (a) (a) $0.009$ $0.011$ Migration (YSM) $(9.81)$ $(12.23)$ YSM Squared/100 (a) (a) $-0.005$ $-0.011$ $(2.41)$ $(6.07)$ Citizen (a) (a) $0.035$ $0.024$ $(4.95)$ $(3.42)$ Adjusted $R^2$ $0.230$ $0.259$ $0.278$ $0.322$	English Very Well	-0.072	-0.064	-0.141*	-0.110*
English Not $-0.109$ $-0.099$ $-0.690*$ $-0.492*$ Well/Not at All $(2.57)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since (a) (a) $0.009$ $0.011$ Migration (YSM) $(9.81)$ $(12.23)$ YSM Squared/100 (a) (a) $-0.005$ $-0.011$ $(2.41)$ $(6.07)$ Citizen (a) (a) $0.035$ $0.024$ $(4.95)$ $(3.42)$ Adjusted $R^2$ $0.230$ $0.259$ $0.278$ $0.322$		(5.37)	(4.79)	(18.76)	(14.98)
English Not $-0.109$ $-0.099$ $-0.690*$ $-0.492*$ Well/Not at All $(2.57)$ $(2.35)$ $(49.40)$ $(35.70)$ Years Since (a) (a) $0.009$ $0.011$ Migration (YSM) $(9.81)$ $(12.23)$ YSM Squared/100 (a) (a) $-0.005$ $-0.011$ $(2.41)$ $(6.07)$ Citizen (a) (a) $0.035$ $0.024$ $(4.95)$ $0.230$ $0.259$ $0.278$ $0.322$	English Well	-0.068	-0.055	-0.403*	-0.304*
$ \begin{array}{c ccccccccccccccccccccccccccccccccccc$		(1.97)	(1.64)	(42.61)	(32.84)
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	English Not	-0.109	-0.099	-0.690*	-0.492*
$ \begin{array}{cccccccccccccccccccccccccccccccccccc$		(2.57)	(2.35)	(49.40)	(35.70)
YSM Squared/100 (a) (a) $-0.005$ $-0.011$ (2.41) (6.07) Citizen (a) (a) $0.035$ $0.024$ (4.95) (3.42) Adjusted $R^2$ 0.230 0.259 0.278 0.322	Years Since	(a)	(a)	0.009	
YSM Squared/100 (a) (a) $-0.005$ $-0.011$ (2.41) (6.07) Citizen (a) (a) $0.035$ $0.024$ (4.95) (3.42) Adjusted $R^2$ 0.230 0.259 0.278 0.322	Migration (YSM)			(9.81)	(12.23)
Citizen (a) (a) (2.41) (6.07) (0.035 (0.024) (4.95) (3.42) Adjusted $R^2$ (0.230 (0.259 (0.278 (0.322))		(a)	(a)	-0.005	-0.011
				(2.41)	(6.07)
Adjusted $R^2$ 0.230 0.259 0.278 0.322	Citizen	(a)	(a)	0.035	0.024
Adjusted $R^2$ 0.230 0.259 0.278 0.322				(4.95)	(3.42)
	Adjusted $R^2$	0.230	0.259	0.278	
		100,885	100,885	100,968	100,968

Notes: Heteroskedasticity-consistent 't' statistics in parentheses; RM = Realized Matches, \* = Estimated coefficient for the foreign born is significantly different from that for the native born. Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

Table 5
Estimates of Standard and ORU Models of Earnings by Nativity, Highly-Skilled (Master's or Higher Degree) 25-64 Year Old Males, 2000 US Census

	Native Born		Foreign	Born Born
Variable	Standard	ORU	Standard	ORU
Constant	3.775	3.695	5.663*	5.231*
	(26.57)	(26.72)	(49.87)	(46.52)
Educational	0.110	<b>(a)</b>	0.055*	(a)
Attainment	<b>(19.37)</b>		(13.43)	
Usual Level of	(a)	0.132	(a)	0.091*
Education		(23.67)		(22.00)
Years of Over-	(a)	0.027	(a)	-0.018*
education		(4.50)		(4.20)
Experience	0.069	0.069	0.034*	0.041*
	(31.39)	(32.16)	(18.51)	(22.75)
Experience	-0.154	-0.153	-0.076*	-0.087*
Squared/100	(27.66)	(28.14)	(16.62)	(19.51)
Log Weeks	1.056	1.024	0.936*	0.909*
Worked	(39.25)	(38.85)	(45.18)	(44.64)
Married	0.326	0.295	0.268*	0.245*
	(27.87)	(25.78)	(26.89)	(25.11)
South	-0.030	-0.033	-0.054	-0.049
	(2.97)	(3.27)	(5.93)	(5.53)
Metropolitan	0.336	0.331	0.097*	0.133*
	(20.66)	(21.06)	(3.71)	(5.28)
Veteran of US	-0.020	0.000	-0.153*	-0.132*
Armed Forces	(1.49)	(0.03)	(5.91)	(5.28)
Black	-0.179	-0.143	-0.368*	-0.334*
	(7.81)	(6.39)	(24.02)	(22.59)
English Very Well	-0.072	-0.060	-0.094	-0.078
	(2.96)	(2.51)	(8.30)	(6.97)
English Well	-0.026	-0.011	-0.424*	-0.336*
	(0.44)	(0.19)	(29.36)	(23.54)
English Not	-0.191	-0.021	-0.816*	-0.590*
Well/Not at All	(2.00)	(2.17)	(35.27)	(25.38)
Years Since	(a)	(a)	0.016	0.016
Migration (YSM)			(11.51)	(12.21)
YSM Squared/100	(a)	(a)	-0.017	-0.021
			(6.22)	(7.92)
Citizen	(a)	(a)	0.068	0.053
			(6.08)	(4.86)
Adjusted $R^2$	0.221	0.251	0.269	0.307
Sample Size	36,572	36,572	47,539	47,539

Notes: Heteroskedasticity-consistent 't' statistics in parentheses; RM = Realized Matches, \* = Estimated coefficient for the foreign born is significantly different from that for the native born. Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

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The payoff to actual years of education is 11.1 percent for the native born and 10.6 percent for the foreign born. These estimates are greater than those for the full sample of all male workers (of 10.3 and 5.3 percent, respectively), indicating a non-linearity in the returns to education, particularly among the foreign born. At first glance this might suggest that the limited international transferability of formal schooling is less of an issue for high-skilled immigrants than for less-skilled immigrants. Chiswick and Miller (2008), however, present a decomposition of the lower payoff to schooling for the foreign born than for the native born into components due to the international transferability of human capital skills and due to selection in migration. They suggest that the latter factor, which is likely to be more prevalent among the less-well educated, is of far greater importance than the former factor. The finding in Table 4, which excludes those with less than a Bachelor's degree, appears to reinforce the findings from the Chiswick and Miller (2008) analyses.

The payoff to labor market experience is higher in the analyses for the high-skilled group of workers than for all workers. It is 3.26 percent for native-born skilled workers per year of experience (evaluated at 10 years) compared to 2.20 percent for all native-born workers. The payoff to pre-immigration labor market experience is 1.62 percent for foreign-born skilled workers, compared to 0.86 percent for all foreign-born workers. Thus, there appear to be complementarities between formal education and labor market experience, particularly among the foreign born. This suggests that with additional years of formal schooling, immigrants receive greater earnings for skills acquired on the job prior to immigration.

The earnings payoff to an additional year of living in the US, holding constant total labor market experience, among the high-skilled immigrants is 0.80 percent, which is about the same as that (0.82 percent) received by all immigrants.

Finally, the earnings penalties associated with limited English skills are greater when the focus is on skilled immigrants than when all immigrants are considered. For example, among immigrants, skilled workers who self report that they speak English well have earnings 40 percent less than the earnings of skilled immigrants who speak only English at home. When all immigrants are used in the analysis, this earnings penalty was only 25 percent. To put this another way, among the immigrants there is evidence of a complementarity between English language skills and formal education, with there being a greater earnings return to English proficiency among skilled immigrants. Among the native born, almost all of whom speak only English at home, regardless of schooling level, the change in sample from all workers to skilled workers (BA and above) is associated with only minor changes to the estimated coefficients of the English language variables.

The coefficients on the ORU variables in Table 4 differ by up to four percentage points compared to those in a regression for all male workers (compared with Chiswick and Miller, 2008). Thus, the payoff to years of usual education, as measured by the realized matches (RM) procedure, falls by two to three percentage points when the focus is shifted from all workers to workers with at least a Bachelor's degree, whereas the payoff to years of surplus schooling falls by up to four percentage points.<sup>14</sup>

<sup>&</sup>lt;sup>14</sup> Chiswick and Miller (2008) report estimated effects of the required level of education on earnings of 0.154 for all native-born workers and 0.153 for all foreign-born workers. Their estimates of the effects of surplus years of schooling on earnings were 0.056 for the native born and 0.044 for the foreign born.

Table 5 lists results for the more stringent definition of skilled workers, that is, of workers with a Master's, Professional, or Doctorate degree. These findings show that the payoff to education is 11 percent for the native born and only 5.5 percent for the foreign born. This difference in the payoff to education is comparable to that reported from the analyses based on all workers, but contrasts with the findings for workers with a Bachelor's degree or higher (Table 4), where the payoffs for the native born and foreign born are about the same, at 11 percent. This difference may be due to the relatively high earnings among the native born with a Professional degree, which involves fewer years of schooling than a Doctorate, compared to those with a Doctorate, and their greater numerical importance when the more stringent definition of skilled workers is used.<sup>15</sup>

The payoff to a year of labor market experience (evaluated at 10 years) for native-born workers with a Master's or higher degree is 3.82 percent, about 17 percent higher than the 3.26 percentage point effect for native-born workers with at least a Bachelor's degree. Among the foreign born, however, the payoffs to experience acquired in the country of origin and in the US for the high-skilled group in Table 5 are slightly higher than the payoffs established using the broader definition of skilled immigrants in Table 4. However, the earnings effects associated with very limited English language skills are greater among immigrants with a Master's or higher degree than were reported in

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<sup>&</sup>lt;sup>15</sup> The mean earnings in 1999 for Bachelor's degree, Master's, Professional and Doctorate are \$72,067, \$88,168, \$111,730, and \$82,521 for the adult male native born, and \$65,163, \$78,393, \$92,011, and \$78,650 for the adult male foreign born. Especially for the native born, earnings are very high for those with a Professional degree.

<sup>&</sup>lt;sup>16</sup> The payoff to origin country experience (evaluated at 10 years) is 1.88 percent in the Table 5 estimates compared to 1.62 percent in the Table 4 estimates. The premium to experience (evaluated at 10 years) acquired in the US is 1.26 percent in the Table 5 results, compared to 0.80 in the Table 4 results.

Table 4. This further emphasizes the complementarity between formal schooling and English language proficiency in the immigrant workforce.

#### C. Analyses by Occupation

Are there some occupations where surplus skills can be used more effectively than elsewhere in the economy? This can be captured in the ORU model via a smaller gap between the payoffs to the years of education that are usual for a worker's occupation and to years of education that are considered surplus in the occupation.<sup>17</sup>

The coefficients on the education variables (actual years of schooling, years of usual schooling and years of over-education) for each skill-birthplace group are presented in Appendix B. Sets of simple correlations between the estimated coefficients on the various education variables are presented in Table 6 (Bachelor's degree and above) and Table 7 (Master's degree and above). Figures below the diagonal in each of these tables are for the foreign born, and these are shaded; figures above the diagonal are for the native born. Correlations with the mean level of schooling in the occupation (computed by birthplace) are also provided to illustrate how these payoffs vary with the educational level of the occupation.

Consider the findings for the foreign born with a Bachelor's or higher degree (Table 6). The payoff to actual years of education within the broad occupational category ranges from zero, and very small positive amounts, in a number of occupations to 17.4 percent (Healthcare Practitioners and Technical) (Appendix Table B.1). Education is rewarded more highly in the more skilled occupations. Thus, there is a simple correlation

<sup>&</sup>lt;sup>17</sup> There are 22 Census major non-military occupations. Due to the absence of variation in the usual level of schooling within two of these occupations, the analyses in this sub-section are performed on 20 occupations.

coefficient of 0.72 between the payoff to actual years of education and the mean level of education (as a measure of overall skill) in the occupation. The mean payoff to actual years of education for the 20 occupations is 7.3 percent, which is 3.3 percentage points less than the 10.6 percent reported in the pooled (across occupations) analyses in Table 4.<sup>18</sup> This shows that about one-third of the payoff to schooling among skilled immigrants is due to inter-occupational mobility across the Census major group occupations.

Table 6<sup>(a)</sup>
Correlation Coefficients among Payoffs of Education and Mean Level of Education from Analyses Disaggregated by Occupation, Skilled (Bachelor's or Higher Degree)
25-64 Year Old Males, 2000 US Census

FB\NB <sup>(b)</sup>	EDUC	USUAL	OVER	GAP	MEAN
EDUC	-	0.49*	0.18	0.24	0.84*
USUAL	0.52*	-	0.41	0.46*	0.19
OVER	0.52*	0.15	-	-0.63*	0.08
GAP	0.19	0.85*	-0.40	_	0.08
MEAN	0.72*	0.11	0.31	-0.07	-

Notes: (a) Based on Realized Matches procedure; Shaded cells are correlations for the foreign born.

Source: Appendix B.

The payoff to years of usual education within the broad occupational category are listed in the second column (Appendix Table B.1). There is one negative payoff to usual education—for the Community and Social Services occupation. This is due to the combination of relatively low earnings and high usual level of education for the clergy. Apart from this anomaly, the payoff to usual education ranges from zero (Arts, Design, Entertainment, Sports and Media; Personal Care and Services and Construction and Extraction) to 25.6 percent in Architecture and Engineering among those with a

<sup>(</sup>b) EDUC=payoff to actual years of schooling; USUAL=payoff to usual years of schooling; OVER=payoff to years of surplus schooling; UNDER=earnings penalty to years of under-education; MEAN=mean educational attainment of occupation; GAP=difference between payoff to usual and surplus years of schooling; \* = significant at the 5 percent level.

<sup>&</sup>lt;sup>18</sup> All means in this section are weighted by the number of workers in the occupation.

Bachelor's or higher level of schooling. The payoff to usual education is positively correlated across occupations with the payoff to actual years of education (r = 0.52). However, there is no association between the payoff to usual education and the mean level of education in the occupation (r = 0.11). The mean payoff to usual years of education across the 20 occupations is 14.9 percent, which is of the same order of magnitude as the 14.0 percent reported in Table 4. The usual education variable takes into account movements, within the sample analyzed, to occupations where the worker's schooling is at the usual level. Thus, the fact that there is little change in the payoffs to usual schooling when the Census major group occupations are held constant suggests that the payoff to matching mainly occurs within the Census major group occupations, rather than across these occupations. Schooling may be used to qualify for a higher status occupation, but there is a sorting/matching process within these occupations that is very important to the earnings determination process.

The payoff to years of over-education range from zero (in eight occupations) to over 15 percent (Education, Training and Library, and Healthcare Support). The mean payoff to years of over-education is 5.2 percent, which compares favorably with the 4.6 percent for the analyses across occupations in Table 4.

The absence of a pattern to the ways the payoffs to years of surplus education and usual education change across occupations shows up clearly when the gap between these payoffs is linked to the mean level of schooling: the simple correlation coefficient is -0.07. That is, surplus schooling is not used effectively in high-skilled occupations, as is also the case in less-skilled occupations.

Similar patterns are evident for the native born, for the highly skilled groups, and for when the Worker Self-Assessment (WSA) procedure is used to construct the usual level of schooling for each occupation (Chiswick and Miller 2009b). This reinforces the conclusion that there is minimal evidence that some sections of the economy are immune from the ineffective use of surplus schooling. Whether this conclusion carries across to all levels of schooling is considered in the next section.

Table 7<sup>(a)</sup>

Correlation Coefficients among Payoffs of Education and Mean Level of Education from Analyses Disaggregated by Occupation, Highly-Skilled (Master's or Higher Degree) 25-64 Year Old Males, 2000 US Census

FB\NB <sup>(b)</sup>	EDUC	USUAL	OVER	GAP	MEAN
EDUC	-	0.49*	0.77*	-0.12	0.73*
USUAL	0.54*	-	0.18	0.73*	0.47*
OVER	0.92*	0.46*	-	-0.54*	0.29
GAP	-0.19	0.68*	-0.34	-	0.20
MEAN	0.59*	0.07	0.33	-0.20	-

Notes: (a) Based on Realized Matches procedure; Shaded cells are correlations for the foreign born.

(b) See Table 5.

Source: Appendix B.

#### D. <u>Analyses by Level of Education</u>

Vahey's (2000) theoretical estimating equation includes dichotomous variables for each level of over-education and under-education for a given level of usual education. In other words, for a usual level of education of a Bachelor's degree, for example, workers who hold a Master's degree would be represented by one dichotomous over-education variable, those who hold Professional degrees by a separate dichotomous over-education variable, and workers who hold a Doctorate by a further separate dichotomous over-education variable. Similarly for under-educated workers, and also for the other usual levels of education. In some instances, however, this flexible approach would

result in very small samples in specific over-education and under-education groups. Indeed, for this reason Vahey (2000) considered only one variable for over-education and one for under-education at each usual level of education.

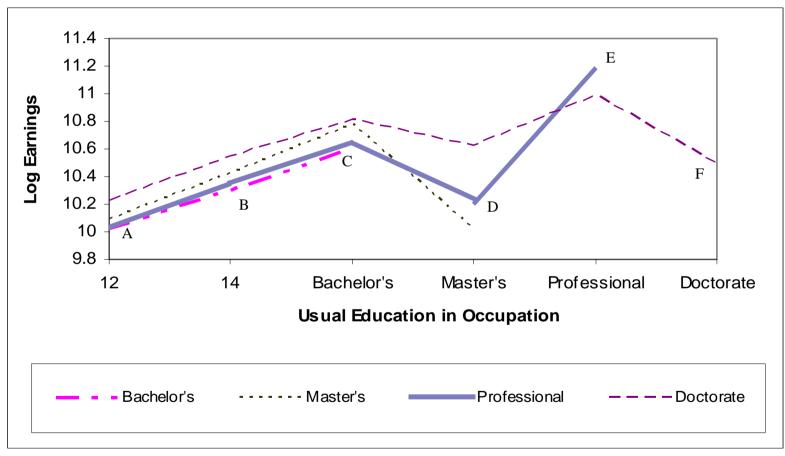
In the current analysis, however, the maximum detail on the extent of overeducation is incorporated into the estimating equation. This follows from the aim of the section, which is to assess whether the difficulties in using surplus education are equally prevalent across all levels of education. These analyses are undertaken only for the sample of skilled workers with at least a Bachelor's degree.

Given the array of findings from this approach, a graphical presentation of the main results will be used. Figure 3 presents the relevant findings for the foreign born, and Figure 4 provides comparable results for the native born using the realized matches approach.

Figures 3 and 4 have the natural logarithm of earnings on the vertical axis, and the usual level of education in the occupation on the horizontal axis. Earnings by usual level of education profiles are presented for each of four actual levels of education: Bachelor's degree, Master's degree, Professional qualifications and Doctorates. The first line to consider is the short line for workers with a Bachelor's degree that truncates at a usual level of education in the occupation of a Bachelor's degree, and which has the letters A-B-C positioned on it.

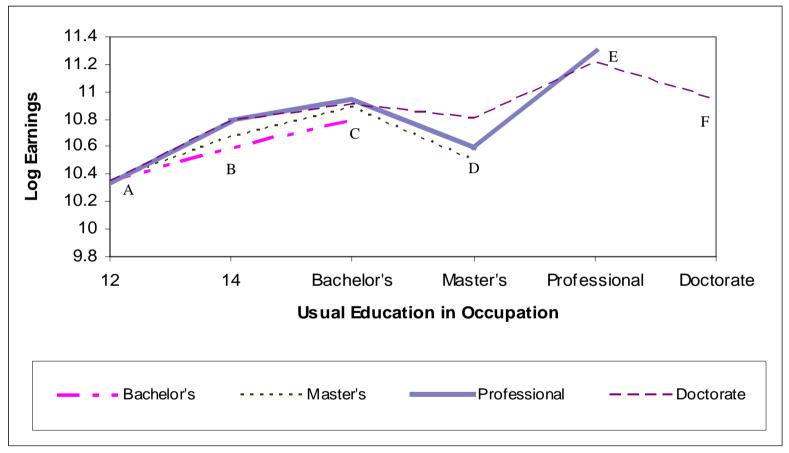
<sup>&</sup>lt;sup>19</sup> In this presentation, Doctorates are ranked above Professional qualifications, based on the typical years of formal schooling. If post-qualification training as residents/interns by physicians is considered formal schooling rather than on-the-job training, Professional qualifications might be ranked above Doctorates. This alternative ranking would reduce or remove the anomaly associated with the comparison of points E and F.

Figure 3
Results from Flexible Specification of ORU Model in Equation 2 for Foreign Born, based on Realized Matches Procedure



Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

Figure 4
Results from Flexible Specification of ORU Model in Equation 2 for Native Born, based on Realized Matches Procedure



Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

If workers with a Bachelor's degree are employed in an occupation where the usual level of education is a Bachelor's degree, they will be correctly matched in terms of educational attainment. There are represented in Figure 3 by the point C. If workers with a Bachelor's degree are employed in an occupation where the usual level of education is 12 or 14 years then they will be over-educated. Workers in these situations are represented in Figure 3 by the points A and B, respectively. The highest earnings among workers with a Bachelor's degree occur when these workers are correctly matched, that is, they are working in occupations in which the usual level of education is a Bachelor's degree (point C). The over-educated workers earn considerably less than the correctly matched workers (21 percentage points lower earnings if working in an occupation where the usual level of schooling is 14 years, at point B, and 45 percentage points less if working in an occupation where the usual level of schooling is 12 years, at point A). The fact that points A and B are lower than point C shows that, among holders of Bachelor's degree, years of surplus education are not used as effectively in the labor market as are years of correctly matched education.

Now consider the earnings by usual level of education profile for individuals who possess a Master's degree. This is the dotted line that truncates at point D. Across the usual education levels of 12 years to a Bachelor's degree, where workers with a Master's degree would be over-educated, this profile is a little above the profile for workers who possess a Bachelor's degree, and is essentially parallel. There is thus some advantage to having a Master's degree rather than a Bachelor's degree if over-educated. Note, however, that the higher qualification does not greatly assist in overcoming the difficulties degree-qualified workers have in getting adequate reward for their schooling

if they are working in an occupation that results in them being classified as overeducated.

The foreign-born men with a Master's degree who are correctly matched to the usual educational requirements of their job earn less than workers who have Master's degrees and who work in jobs that require only a Bachelor's degree. The Master's degree appears to offer access to a particular set of occupations that are relatively poorly paid (school teachers, social workers, *etc.*). This may explain why only nine percent of the foreign-born workers with a Master's degree are correctly matched in terms of levels of education.

Foreign-born workers with either Professional qualifications or Doctorates (the lines truncating with the letters E and F, respectively) earn amounts similar to workers with either a Bachelor's degree or a Master's degree when working in occupations where the usual level of education is from 12 years of education to a Bachelor's degree. Compared to when working in occupations where a Bachelor's degree is usual, if they work in an occupation where a Master's degree is usual they earn less. They earn more, however, than workers with a Master's degree who work in an occupation where the usual level of education is a Master's degree. These slightly higher earnings are the modest rewards to the surplus years of education.

Workers with a Professional degree who are correctly matched to the usual educational requirements of their jobs have very high earnings (point E) whereas workers with a Doctorate who are correctly matched (point F) have much more modest salaries. It is noted that those with Doctorates working in occupations where the usual level of education is a Professional qualification actually earn more than their

Again, this evidence shows that earnings follow the usual level of education in the occupation rather than the actual years of education for the individual. It is where you work that governs your relative success in the labor market rather than simply your years of education, although it is years of education that, in part, influences where you work.

Thus, these analyses show that if a skilled worker works in an occupation that requires between 12 years of education and a Bachelor's degree, any surplus years of schooling will be used ineffectively, and the extent of ineffectiveness is largely invariant to the actual level of schooling. In the small group of occupations with usual levels of schooling greater than a Bachelor's degree, the pattern of earnings effects is irregular. But they clearly support the view of earnings being more strongly related to the usual level of education for the job than to the individual's actual years of education.

For the native-born high-skilled workers, information on the earnings rewards to over-education and correctly matched education by the level of schooling is presented in Figure 4. The earnings by usual level of education profiles for each of the levels of schooling, Bachelor's, Master's, Professional and Doctorates, for the native born are largely the same as those discussed for the foreign born. Thus, the ineffective use of surplus years of schooling that occurs at each level of schooling is not a foreign-born phenomenon: it is a labor market phenomenon.

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<sup>&</sup>lt;sup>20</sup> To ascertain if the relatively poor earnings outcome for Doctorates was simply linked to either low salaries in the education sector or misreporting of weeks worked in that sector, a dichotomous variable for employment in the education industry was included in the model. This variable was associated with coefficients of -0.154 among the native born, and -0.192 among the foreign born. This change in the specification was associated with a four (native born) to eight (foreign born) percentage point improvement in the *ceteris paribus* earnings of Doctorates compared to workers who hold Bachelor's degrees, but little change in the relative standing of workers with Professional qualifications and Doctorates.

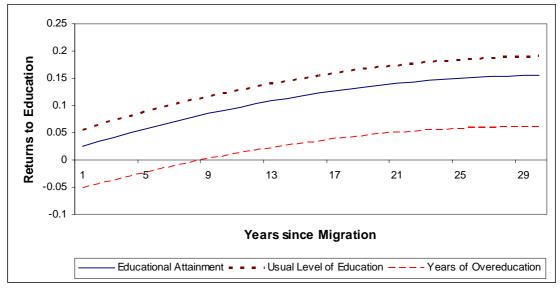
## E. Analysis of Effect of Education by Duration of Residence

The analysis can be extended by asking whether the effect of education on earnings varies systematically with duration of residence in the United States. To answer this question, the education variables in the standard and ORU equations are interacted with the variables for duration and duration squared. These interaction terms are highly statistically significant (regression equations available on request). Based on these regression equations, Figures 5 and 6, respectively, plot the partial effects of education on earnings with respect to years since migration for immigrants with at least a Bachelor's degree and at least a Master's degree. The effects of education on earnings for educational attainment (standard analysis) and for usual level and years of over-education (ORU analysis) show that the partial effects increase, but at a decreasing rate, with duration in the US. That is, the effect of schooling on earnings becomes greater with a longer residence.

Note, however, that the partial effects are systematically higher for the usual level of education than for the respondent's actual level of schooling. Most dramatic, however, is the consistently very low effect on earnings of years of over-education throughout the range of years since migration. Indeed, the effect of over-education on earnings is in fact negative until about 9 years in the US for those with at least a Bachelor's degree, and is negative until about 20 years duration for those with at least a Master's degree.

Figure 5

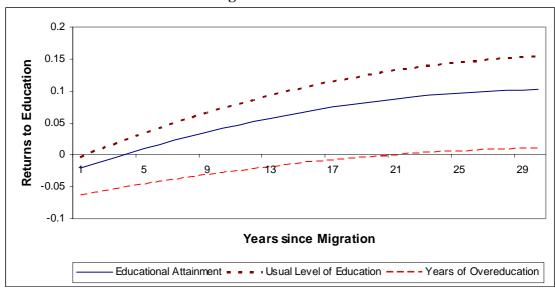
Partial Effects of Education on Earnings By Duration in the United States,
Bachelor's Degree and Higher Levels of Education,
Foreign-Born Adult Males



<u>Source</u>: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

Figure 6

Partial Effects of Education on Earnings By Duration in the United States,
Master's Degree and Higher Levels of Education,
Foreign-Born Adult Males



<u>Source</u>: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

## VII. SUMMARY AND CONCLUSION

This paper is concerned with the extent and consequences for earnings of the mismatch of the educational attainment (measured by formal schooling) and the occupation of employment among high-skilled adult male immigrants in the United States. The "Over-Required-Under" education decomposition methodology is employed with mismatches identified based on a "realized matches" approach (modal educational level in the occupation). The empirical analyses focus on the foreign born, but for comparative purposes parallel analyses are conducted for the native born. The empirical analysis is conducted using the 2000 US Census of Population, 5 percent microdata file. High-skilled immigrants are identified as those with at least a Bachelor's degree, and a higher bar, at least a Master's degree, is the threshold for parallel analyses.

While there has been a long history of high-skilled migration to the United States, the provisions of the 1990 Immigration Act increased the number of permanent and temporary visas to the United States for these workers. In 2007, of the nearly 1.1 legal immigrants, 162,000 received a permanent resident visa under an employment-based category, of whom about half were the spouses and minor children of the principal applicants.

Educational mismatches refer to the difference in the educational attainment of a worker and the usual or typical (modal) level of education of those working in the occupation. These mismatches can arise from several causes, including occupational skill upgrading for younger cohorts of workers, and unmeasured worker productivity characteristics that may be positive or negative. In addition, among immigrants mismatches may arise from the limited international transferability of skills and

selectivity in migration, in addition to labor market discrimination against immigrants. It would be expected that among both the native born and the foreign born there would be "mismatches," that is both "over-educated" and "under-educated" workers.

The empirical analysis necessitates the identification of a required or usual level of education in each occupation. A "realized matches" approach is used, where this refers to what actually occurs in the labor market, as reflected by the modal level of schooling in each of the approximately 500 occupations identified in the 2000 Census.

Over all educational levels, among adult men, there are greater mismatches among the immigrants than among the native born. Among the immigrants, only 26 percent were correctly matched (compared to 40 percent for the native born), with 45 percent under-educated and 29 percent over-educated (in contrast to 26 percent and 33 percent, respectively, for the native born). Among the high-skilled workers, however, 63 percent of immigrants with a Bachelor's degree or higher were over-educated (50 percent for the native born), while among those with a Master's degree or higher, fully 79 percent were over-educated (70 percent for the native born). Nearly all of the rest were correctly matched – their own education was equal to that of the modal education in their occupation. Except for the immigrants who arrived in the 1990's, the extent of correct matching of education and occupation increases with the duration in the US. Thus, among the highly educated, particularly among immigrants, there is a high degree of "surplus" education. But is this surplus education wasted?

When the analysis of earnings is limited to those with a Bachelor's degree or more schooling, earnings increase by about 11 percent per year of schooling among the foreign born and the native born. If years of education is divided into years of usual (modal) education in the worker's occupation and years of over-education, the effect on earnings of usual education is about 13 percent for both nativity groups, whereas the effect of years of over-education is about 2 percent per year for both groups. Thus, whereas the return to years of required or usual schooling is high, the returns to education in excess of what is needed for one's occupation are extremely low.

When the high-skilled are limited to those with educational credentials beyond the Bachelor's degree (*i.e.*, only those with Master's, Professional, and Doctorate degrees), the educational coefficients are less stable, but yet they tell a similar story – high rates of return to years of usual education, with extremely low returns (negative for the immigrants) for years of over-education.

When separate analyses are conducted by broad occupational categories, similar patterns emerge – in general a high return to years of usual education and very little or no returns to surplus or over-education. When analyses are performed by educational level, the most striking feature is that perhaps the group with the greatest number of years in school (those who receive a Doctorate) tend to have relatively low earnings compared to those whose highest degree is a Professional degree. The return to surplus education is actually negative in the first 10 to 20 years in the US among high-skilled immigrants. After that it becomes positive but remains very small.

The very low return to years of over-education indicates that, among the highly educated, educational attainments beyond what is required for the occupation in which one works are not productive. This is true for the native born as well as for the foreign born. It means that earnings are far more influenced by the educational norms in one's occupation, as distinct from one's own educational level.

The private and social losses in economic welfare are substantial from workers (whether native or foreign immigrant) not being employed in the higher level occupations that might make better use of their educational credentials.

This has implications for immigration and absorption policy. It suggests that, in general, employment-based immigrant visas are more likely to attract the high-skilled immigrants who are more likely to be working in jobs better matched to their educational attainment than would be the case for family or refugee visa recipients. An employer sponsored targeted employment approach might result in a better matching than one that does not require an employment sponsor. Even a point system for rationing visas, as used in Canada and Australia, might be improved by adding a set of points for prearranged employment, or for having a particular set of skills in high demand.<sup>21</sup>

Moreover, proactive efforts to facilitate the adjustment of high-skilled new immigrants in the labor market might be most productive. These could include specific programs for: English-language training, obtaining a US certification or occupational license, or merely teaching new immigrants how to navigate the job search process in one's occupation in the US.

The analyses reported above focus on high-skilled male immigrants, and are based on a single cross-section of data, the 2000 Census. They demonstrate that including demand-side considerations (*i.e.*, the usual level of education in the occupation) in addition to the usual supply-side factors (the actual educational attainment of workers)

List".

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<sup>&</sup>lt;sup>21</sup> The points test used in Australia at present, for example, allocates points for "specific employment", which is defined as employment, for at least three of the four years immediately before application for a visa, in an occupation on a government determined "skilled occupation list". Further points are allocated for "Occupation in demand (and job offer)", where the occupations in demand are from a government determined "Migration Occupations in Demand

has considerable merit. Many extensions to this research are possible, including undertaking separate analyses for high-skilled females, analysis of low- and medium-skilled workers, and the use of data from earlier censuses in synthetic cohort models. Disaggregating the analysis by age at migration could permit differences in the effects of schooling acquired in the US and schooling acquired abroad to be assessed. Conducting separate analyses for each of the major foreign-born groups in the US, and relating the estimates of the realized matches specification to characteristics of the immigrants' countries of origin, such as the internationally standardized scores from the Programme for International Student Assessment (PISA), may provide a means of determining the influence of the quality of schooling acquired abroad on the estimated effects of surplus schooling. The sensitivity of the estimates to the specification of the empirical model (e.g., more extensive controls for region of residence, race, birthplace, as well as interaction effects between the more important regressors) might also be considered. These extensions form the basis of our current research program.

#### REFERENCES

- Chiswick, Barry R., (1999). "Are Immigrants Favorably Self-Selected?", *American Economic Review*, Vol. 89, No. 2, pp.181-185.
- Chiswick, Barry R., (2008). "Are Immigrants Favorably Self-Selected? An Economic Analysis", Chapter 3 in Caroline B. Brettell and James F. Hollifield, *Migration Theory: Talking Across Disciplines*, 2<sup>nd</sup> Edition, New York: Routledge, pp.63-82.
- Chiswick, Barry R. and Miller, Paul W., (2008). "Why is the Payoff to Schooling Smaller for Immigrants?", *Labour Economics*, Vol. 15, No. 6, pp.1317-1340.
- Chiswick, Barry R. and Miller, Paul W., (2009a). "The International Transferability of Immigrants' Human Capital Skills", *Economics of Education Review*, Vol. 28, No. 2, pp.162-169.
- Chiswick, Barry R. and Miller, Paul W., (2009b). "Does the Choice of Reference Levels of Education in the ORU Earnings Equation Matter?", Xerox, University of Illinois at Chicago.
- Cohen-Goldner, Sarit and Weiss, Yoram, (2009). "The Impact of High-Skilled Russian Immigrants on the Israeli Labor Market", paper presented at the AEI Conference on High-Skilled Immigration in a Globalized Labor Market, Washington DC, April.
- Cohn, Elchanan and Khan, Shahina, P., (1995). "The Wage Effects of Overschooling Revisited", *Labour Economics*, Vol. 2, pp.67-76.
- Duncan, Greg J. and Hoffman, Saul D., (1981), "The Incidence of Wage Effects of Overeducation", *Economics of Education Review*, Vol. 1, No. 1, pp.75-86.
- Ferrie, Joseph, (2009). "A Historical Perspective on High-Skilled Immigrants in the US, 1820-1920", paper presented at the AEI Conference on High-Skilled Immigration in a Globalized Labor Market, Washington DC, April.
- Hartog, Joop, (2000). "Over-education and Earnings: Where Are We, Where Should We Go?", *Economics of Education Review*, Vol. 19, No. 2, pp.131-147.
- McDonald, James Ted, Warmon, Casey and Worswick, Christopher, (2009). "Earnings, Occupation, and Schooling Decisions of Immigrants with Medical Degrees: Evidence for Canada and the United States", paper presented at the AEI Conference on High-Skilled Immigration in a Globalized Labor Market, Washington DC, April.
- Remennick, Larissa, (2008). "Survival of the Fittest: Russian Immigrant Teachers in Israeli Schools", Paper presented at "The Conference on Understanding

- Contemporary Migration: Challenges to Society", Kibbutz Tsuba, Israel, November 19, 2008.
- Rubb, Stephen, (2003). "Post-College Schooling, Overeducation, and Hourly Earnings in the United States", *Education Economics*, Vol. 11, No. 1, pp.53-72.
- Rumberger, Russell W., (1987). "The Impact of Surplus Schooling on Productivity and Earnings", *The Journal of Human Resources*, Vol. 22, No. 1, pp.24-49.
- Vahey, Shaun P. (2000). "The Great Canadian Training Robbery: Evidence on the Returns to Educational Mismatch", *Economics of Education Review*, Vol. 19, pp. 219-227.
- Verdugo, Richard R. and Verdugo, Naomi T., (1989), "The Impact of Surplus Schooling on Earnings: Some Additional Findings", *The Journal of Human Resources*, Vol. 26, No. 4, pp.629-643.

# APPENDIX A DEFINITIONS OF VARIABLES

The variables used in the statistical analyses are defined below.

**Data Source:** 2000 Census of Population, Public Use Microdata Sample, 5 percent sample of the foreign born, and 0.15 percent random sample of the native born.

**Definition of Population:** Native-born and foreign-born employed men aged twenty-five to sixty-four years.

Dependent Variables	_
Income in 1999	Natural logarithm of earnings in 1999 (where earnings are defined as gross earnings from all sources).

## **Explanatory Variables**

Years of Education

This variable records the total years of full-time equivalent education. It has been constructed from the Census data on educational attainment by assigning the following values to the Census categories: completed less than fifth grade (2 years); completed fifth or sixth grade (5.5); completed seventh or eighth grade (7.5); completed ninth grade (9); completed tenth grade (10); completed 11th grade (11); completed 12th grade, no diploma (11.5); completed high school (12); attended college for less than one year (12.5); attended college for more than one year or completed college (14); Bachelor's degree (16); Master's degree (17.5); Professional degree (18.5); Doctorate (20). Further discussion of these years-of-schooling equivalents is presented in the text.

## Note:

- (a) As with other census data, the values for educational attainment are self-reported responses. While academic degrees may have required different years of schooling for immigrants educated in some countries of origin, US values are used in the analysis.
- (b) Sensitivity tests were performed using 18, 20 and 21 years of schooling for the Master's, Professional and Doctorate degrees, respectively. The findings are essentially invariant with respect to this alternative set of years of schooling. Vocational and technical training for a specific trade that does not require an advanced degree beyond the Bachelor's are excluded from professional degrees.

Usual Level of Education

This variable records the required years of education. It is constructed using the modal level of education of the native-born workers in the respondent's occupation of employment based on the *Realized Matches* procedure.

Years of Overeducation

The over-education variable equals the difference between the person's actual years of education and the years of education required for the person's job where this computation is positive. Otherwise, it is set equal to zero.

Weeks worked in 1999

This is a continuous variable for the numbers of weeks the individual worked in 1999.

Experience

Age - Years of Education - 6.

Location

The two location variables record residence in a metropolitan area or in the Southern States. The states included in the latter are: Alabama, Arkansas, Delaware, District of Columbia, Florida, Georgia, Kentucky, Louisiana, Maryland, Mississippi, Missouri, North Carolina, Oklahoma, South Carolina, Tennessee, Texas,

Virginia, West Virginia.

Marital Status

This is a binary variable that distinguishes individuals who are married, spouse present (equal to 1) from all other marital states.

Veteran

This is a binary variable set equal to one for someone who had served in the US Armed Forces, and set equal to zero otherwise.

Race

This is a dichotomous variable that distinguishes between individuals who are Black and all other races.

English Language **Proficiency** 

Three dichotomous variables (speaks English very well; well; not well or not at all) are used to record the English

language proficiency of the respondents.

Years Since Migration

This is computed from the year the foreign-born person

came to the United States to stay.

Citizenship

This is a dichotomous variable set equal to one for

foreign born who hold an US citizenship.

Means and standard deviations for these variables are presented below.

Appendix Table A-1
Means and Standard Deviations of Variables in Earnings Equation, for Workers
Aged 25-64 with a Bachelor's or Higher Degree

Variable	Native Born	Foreign Born
Log Income	10.800	10.650
	(0.98)	(1.03)
Educational	16.744	17.046
Attainment	(1.10)	(1.29)
Usual Level of	15.335	15.121
Education	(1.93)	(2.10)
Years of Over-	1.447	1.960
education	(1.72)	(1.93)
Experience	20.008	18.332
	(10.07)	(9.94)
Experience Squared	501.808	434.95
	(426.05)	(410.70)
Log Weeks Worked	3.855	3.812
	(0.34)	(0.40)
Married	0.704	0.695
	(0.46)	(0.46)
South	0.330	0.272
	(0.47)	(0.45)
Metropolitan	0.890	0.971
	(0.31)	(0.17)
Veteran of US	0.192	0.045
Armed Forces	(0.39)	(0.21)
Black	0.053	0.076
	(0.22)	(0.27)
English Very Well	0.046	0.537
	(0.21)	(0.50)
English Well	0.005	0.196
	(0.07)	(0.40)
English Not	0.004	0.060
Well/Not at All	(0.06)	(0.24)
Years Since	(a)	16.683
Migration (YSM)		(11.82)
YSM Squared	(a)	418.13
		(526.92)
Citizen	(a)	0.514
		(0.50)
Sample Size	100,885	100,968

APPENDIX B
Table B.1<sup>(a)</sup>
Selected Estimates from Standard and ORU Models of Earnings by Occupation,
Skilled (Bachelor's or Higher Degree) 25-64 Year Old Foreign-Born Males, 2000 US
Census

Occupation	Educational Attainment	Usual Education	Over- education	Mean of Education	Sample Size
Management, Business and	0.078	0.175	0.050	16.914	15,175
Financial Operations	(11.93)	(21.56)	(7.64)		
Business and Financial	0.099	0.149	0.098	16.762	7,171
Operations	(8.81)	(7.15)	(8.69)		
Professional and Related	0.038 (6.47)	0.146 (13.27)	0.028 (4.76)	16.925	11,360
Architecture and	0.074	0.256	0.064	16.992	9,231
Engineering	(14.32)	(21.21)	(12.27)		
Life, Physical, and Social	0.054	0.039	0.047	18.464	3,133
Science	(7.07)	(4.92)	(5.17)		
Community and Social	0.004	-0.129	0.020	17.200	2,097
Services	(0.40)	(6.66)	(1.45)		
Legal	0.132	0.173	0.079	18.095	1,360
	(5.78)	(6.82)	(2.08)		
Education, Training, and	0.157	0.140	0.169	18.095	6,769
Library	(28.57)	(13.07)	(22.53)		
Arts, Design, Entertain.,	0.031	0.047	0.031	16.724	3,071
Sports, and Media	(1.62)	(1.51)	(1.63)		
Healthcare Practitioner and	0.174	0.224	0.016	18.023	9,384
Technical	(22.92)	(30.99)	(1.52)		
Healthcare Support	0.156	0.209	0.152	16.978	495
	(5.14)	(3.63)	(4.99)		
Protective Service	0.016	0.159	0.002	16.515	1,088
	(0.56)	(5.28)	(0.06)		
Food Preparation	-0.011	0.109	-0.014	16.549	1,901
	(0.61)	(3.45)	(0.75)		
Personal Care and Service	0.004	0.064	0.001	16.541	730
	(0.12)	(1.78)	(0.05)		
Sales and Related	0.051	0.155	0.045	16.545	9,578
	(4.42)	(12.19)	(3.96)		
Office and Administrative	0.053	0.172	0.053	16.546	5,506
Support	(5.06)	(11.81)	(4.99)		
Construction and Extraction	-0.010	-0.005	-0.010	16.622	2,413
	(0.57)	(0.14)	(0.57)		
Installation, Maintenance,	0.034	0.062	0.036	16.524	2,230
and Repair	(1.82)	(2.34)	(1.93)		
Production, Transport. and	0.069	0.092	0.069	16.583	3,824
Material Moving	(5.57)	(2.07)	(5.55)		
Transportation and Material	-0.005	0.130	-0.005	16.583	3,123
Moving	(0.32)	(5.95)	(0.34)		

Notes: Based on Realized Matches Procedure; heteroskedasticity-consistent 't' statistics in parentheses. Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

Table B.2<sup>(a)</sup>
Selected Estimates from Standard and ORU Models of Earnings by Occupation,
Skilled (Bachelor's or Higher Degree) 25-64 Year Old Native-Born Males, 2000 US
Census

		Census			
Occupation	Educational Attainment	Usual Education	Over- education	Mean of Education	Sample Size
Management, Business and	0.065	0.133	0.017	16.627	19,193
Financial Operations	(9.27)	(16.51)	(2.34)		
Business and Financial	0.080	0.157	0.077	16.486	9,722
Operations	(6.69)	(9.40)	(6.42)		
Professional and Related	0.058	0.132	0.045	16.461	5,182
	(5.67)	(9.23)	(4.41)		
Architecture and	0.044	0.233	0.378	16.497	6,060
Engineering	(4.94)	(15.06)	(4.22)		
Life, Physical, and Social	0.087	0.080	0.093	17.469	2,398
Science	(9.57)	(8.50)	(8.64)		
Community and Social	0.061	-0.043	0.068	17.061	3,439
Services	(6.43)	(3.10)	(4.96)		
Legal	0.155	0.148	0.020	18.312	4,498
	(7.92)	(6.40)	(0.57)		
Education, Training, and	0.100	0.071	0.098	17.295	9,878
Library	(21.71)	(7.50)	(16.03)		
Arts, Design, Entertain.,	-0.029	0.008	-0.030	16.438	3,635
Sports, and Media	(1.12)	(0.19)	(1.16)		
Healthcare Practitioner and	0.230	0.263	0.081	17.898	6,299
Technical	(25.71)	(30.56)	(5.95)		
Healthcare Support	0.233	0.073	0.228	16.776	226
**	(3.90)	(0.76)	(3.87)		
Protective Service	0.007	0.115	0.012	16.269	2,499
	(0.34)	(4.64)	(0.58)		
Food Preparation	-0.018	0.121	-0.013	16.279	731
•	(0.46)	(2.35)	(0.34)		
Personal Care and Service	0.074	0.083	0.074	16.423	742
	(1.63)	(1.71)	(1.63)		
Sales and Related	0.024	0.127	0.022	16.287	11,704
	(1.49)	(7.80)	(1.36)		
Office and Administrative	0.048	0.160	0.041	16.349	5,299
Support	(3.19)	(9.08)	(2.74)		
Construction and Extraction	0.013	-0.021	0.012	16.316	2,183
	(0.57)	(0.34)	(0.53)		•
Installation, Maintenance,	-0.050	-0.073	-0.049	16.268	1,439
and Repair	(1.43)	(1.90)	(1.42)		•
Production, Transport. and	0.023	-0.079	0.024	16.341	2,307
Material Moving	(0.97)	(0.63)	(1.01)		*
Transportation and Material	-0.008	0.178	-0.012	16.299	2,406
Moving	(0.28)	(6.64)	(0.47)		,

Notes: Based on Realized Matches Procedure; heteroskedasticity-consistent 't' statistics in parentheses. Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

Table B.3<sup>(a)</sup>
Selected Estimates from Standard and ORU Models of Earnings by Occupation, Highly Skilled (Master's or Higher Degree) 25-64 Year Old Foreign-Born Males, 2000 US Census

		o OS Census			
Occupation	Educational Attainment	Usual Education	Over- education	Mean of Education	Sample Size
Management, Business and Financial Operations	0.033 (2.63)	0.149 (8.90)	0.026 (2.09)	18.010	6,883
Business and Financial Operations	0.022 (0.81)	0.142 (3.16)	0.020 (0.75)	17.881	2,912
Professional and Related	-0.012 (1.16)	0.079 (4.37)	-0.020 (1.80)	17.866	5,677
Architecture and Engineering	0.056 (6.14)	0.276 (9.79)	0.055 (6.07)	18.037	4,478
Life, Physical, and Social Science	0.044 (3.61)	0.005 (0.34)	0.019 (1.26)	19.122	2,461
Community and Social Services	-0.008 (0.44)	-0.156 (4.81)	0.002 (0.13)	18.064	1,242
Legal	0.093 (2.04)	0.153 (2.85)	0.068 (1.20)	18.547	1,115
Education, Training, and Library	0.209 (25.22)	0.207 (14.20)	0.209 (22.02)	18.777	5,065
Arts, Design, Entertain., Sports, and Media	-0.010 (0.26)	0.064 (0.99)	-0.011 (0.28)	17.942	1,150
Healthcare Practitioner and Technical	0.029 (1.42)	0.156 (7.63)	-0.051 (2.44)	18.605	7,310
Healthcare Support	0.089 (1.24)	0.127 (1.44)	0.091 (1.28)	18.419	201
Protective Service	-0.104 (1.42)	0.116 (1.57)	-0.110 (1.64)	17.968	284
Food Preparation	0.011 (0.21)	0.196 (2.88)	0.004 (0.08)	18.067	487
Personal Care and Service	0.034 (0.50)	0.085 (1.16)	0.038 (0.57)	18.150	184
Sales and Related	-0.028 (1.01)	0.106 (3.61)	-0.016 (0.60)	17.867	2,807
Office and Administrative Support	0.032 (1.20)	0.182 (5.17)	0.052 (1.95)	17.971	1,517
Construction and Extraction	-0.125 (2.33)	-0.223 (2.66)	-0.129 (2.38)	18.016	751
Installation, Maintenance, and Repair	-0.073 (1.42)	-0.040 (0.62)	-0.070 (1.36)	17.938	611
Production, Transport. and Material Moving	0.059 (1.98)	0.008 (0.10)	0.061 (2.06)	18.026	1,089
Transportation and Material Moving	0.076 (1.85)	0.281 (6.70)	0.073 (2.13)	17.980	911

Notes: Based on Realized Matches Procedure; heteroskedasticity-consistent 't' statistics in parentheses. Source: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the population.

Table B.4<sup>(a)</sup>
Selected Estimates from Standard and ORU Models of Earnings by Occupation, Highly Skilled (Master's or Higher Degree) 25-64 Year Old Native-Born Males

Occupation	Educational Attainment	Usual Education	Over- education	Mean of Education	Sample Size
Management, Business and Financial Operations	0.003 (0.19)	0.059 (3.24)	-0.001 (0.04)	17.810	6,680
Business and Financial		0.072		17 727	2.725
Operations	-0.020 (0.60)	(1.41)	-0.021 (0.61)	17.737	2,735
Professional and Related	0.018	0.054	0.012	17.806	1,337
Froiessional and Related	(0.75)	(1.79)	(0.52)	17.800	1,557
Architecture and Engineering	-0.008	0.155	-0.006	17.761	1,691
Architecture and Engineering	(0.34)	(3.76)	(0.25)	17.701	1,071
Life, Physical, and Social	0.096	0.093	0.107	18.588	1,328
Science	(6.16)	(6.04)	(5.98)	10.500	1,320
Community and Social	0.013	-0.068	0.022	17.890	1,976
Services	(0.74)	(2.81)	(1.22)		,
Legal	0.058	0.093	-0.018	18.546	4,100
S	(1.47)	(2.20)	(0.43)		,
Education, Training, and	0.110	0.087	0.125	18.165	5,770
Library	(15.16)	(7.84)	(13.62)		
Arts, Design, Entertain.,	-0.109	-0.188	-0.111	17.808	902
Sports, and Media	(1.54)	(1.92)	(1.57)		
Healthcare Practitioner and	0.123	0.205	0.011	18.541	4,714
Technical	(4.26)	(7.27)	(0.37)		
Healthcare Support	0.451	-0.130	0.365	18.241	78
	(2.61)	(0.59)	(2.32)		
Protective Service	-0.001	0.074	-0.007	17.696	397
	(0.02)	(0.98)	(0.11)		
Food Preparation	-0.278	-0.200	-0.272	17.926	102
	(2.62)	(1.35)	(2.56)		
Personal Care and Service	0.010	0.046	0.023	17.846	174
	(0.08)	(0.34)	(0.18)		
Sales and Related	-0.071	0.094	-0.043	17.714	1,964
	(1.39)	(1.80)	(0.87)	4==04	1.001
Office and Administrative	-0.009	0.106	-0.002	17.791	1,034
Support	(0.23)	(2.23)	(0.05)	17.004	271
Construction and Extraction	0.127	-0.021	0.125 (2.18)	17.884	371
Installation Maintenance and	(2.27) -0.134	(0.07)	-0.136	17 996	200
Installation, Maintenance, and Repair	-0.134 (1.42)	-0.085 (0.79)	-0.136 (1.43)	17.826	208
Production, Transport. and	0.028	0.127	0.023	17.834	414
Material Moving	(0.42)	(1.50)	(0.35)	17.034	414
Transportation and Material	-0.130	0.112	-0.062	17.744	414
Moving	(1.50)	(1.28)	(0.75)	1/./44	717

<u>Notes</u>: Based on Realized Matches Procedure; heteroskedasticity-consistent 't' statistics in parentheses. <u>Source</u>: US Census of Population, 2000, Public Use Microdata Sample, 5 percent sample of the Population.