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Ken Clark and Jo Lindley*

**Immigrant Labour Market Assimilation and Arrival Effects:
Evidence from the Labour Force Survey**

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* Corresponding author

Department of Economics
University of Sheffield
9 Mappin Street
Sheffield
S1 4DT
United Kingdom
www.shef.ac.uk/economics

Abstract

We estimate models of earnings and employment outcomes for a sample of white and non-white male immigrants drawn from the Labour Force Survey between 1993 and 2002. Two hypotheses are investigated: (i) whether immigrant outcomes assimilate towards those of natives and (ii) whether labour market conditions at time of entry to the UK labour market have a permanent impact on outcomes. We find positive earnings assimilation for all immigrant groups and strong employment assimilation for those immigrants who complete their education in the UK. We find negative employment assimilation for South Asian immigrants who completed their education overseas. There is some evidence of unemployment rates at time of entry to the labour market causing permanently lower earnings for non-white immigrants.

Keywords: immigrants, assimilation, earnings, employment.

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Ken Clark is a Senior Lecturer in the School of Economic Studies, University of Manchester, UK.

1. Introduction

The labour market performance of immigrants is central to political and public discourse on immigration policy in the UK. In 2001 around 8.3 percent of the UK population were born abroad and the Treasury has estimated that net migration contributes 0.5% to the economic growth rate.¹ Recognising the contribution that immigrants make to the economy, the government has endorsed future controlled and selective immigration. Equally, in response to perceived public concerns about the scale of immigration and the motivation of immigrants, the government has emphasised that immigrants should not be dependent on the state. In the Prime Minister's view, "All those who come here to work and study must be able to support themselves"² How immigrants fare in the labour market is important both for their ability to support themselves and for their contribution to the wider economy, hence in this paper we analyse the earnings and employment outcomes of immigrants observed in the UK labour market over the period 1993-2002.

We focus on two key hypotheses from the literature. The first is that, after arrival in the host country, immigrant labour market outcomes will adjust towards those of non-immigrant or native workers. This view is often known as the assimilation hypothesis and has received much attention from economists³. Assimilation is thought to take place through human capital enhancement: immigrants acquire skills that are specific to the destination country, including knowledge of the labour market and language proficiency, allowing them to improve their labour market outcomes relative to natives. The longer the process of

¹ The population figures were taken from the 2001 Census available from the Office for National Statistics website at <http://www.statistics.gov.uk>.

² Prime Minister's speech to the Confederation of British Industry, April 27th, 2004. The full text is available at <http://www.pm.gov.uk/output/Page5708.asp>.

³ Chiswick (1978) and Borjas (1985) are classic references for the US while Bell (1997) examines the UK. Antecol et al. (2003) is a recent example which takes a cross-country perspective examining Australia, Canada and the US.

assimilation takes, the less successful any cohort of immigrants is likely to be at any given time since arrival.⁴

The second hypothesis we examine is the view that current labour market outcomes for immigrant workers are influenced by labour market conditions when they arrived in the UK. Labour economists often argue that early experiences of unemployment can permanently increase an individual worker's risk of unemployment and reduce their future earnings. This is called the 'scarring hypothesis' (see Arulamapalam et al. (2001) for a recent symposium) and may be relevant for immigrants arriving in a foreign labour market. Scarring can occur for a number of reasons. On the supply side, unemployment spells lead to a loss of firm-specific and general human capital. On the demand side, where information is incomplete, employers may use past unemployment events as a signal of low productivity. This latter mechanism may be particularly important for immigrants if employers are relatively ignorant of the qualifications and skills of workers arriving from overseas. The tendency of immigrants to cluster in particular geographic areas may also lead us to observe effects consistent with scarring if those areas are persistently depressed and the compensating benefits of co-ethnic proximity restrict geographic mobility.

We look for evidence of assimilation and scarring effects using a sample of native and immigrant workers from the UK's *Labour Force Survey* (LFS). The labour market outcomes that we focus on are real weekly earnings and employment and we divide our sample of immigrants along two dimensions. First, to account for well-documented racial differences in

⁴ We use the term "assimilation" purely to refer to convergence in labour market outcomes – adjustment" or "adaptation" would be equally descriptive.

labour market outcomes, we examine white and non-white immigrants separately. There is considerable evidence that non-whites receive differential treatment in the UK labour market (Blackaby et al. (2002) is a recent example) and separating the distinct contributions of immigrant status and ethnicity is important. Second, and more unusually, we compare immigrants who arrive in the UK to enter the labour market, having completed their education at some time in the past, with those who arrive to complete their education in the UK and subsequently enter the labour market. We call this first group “labour market entrants” and the latter group “education entrants”. Note that the group of education entrants includes foreign-born children who arrive with their parents as well as adults who arrive to undertake education in the UK.

Clearly, compared to labour market entrants, those who enter education are affected differently by the two labour market hypotheses outlined above. Kossoudji (1989) makes the important distinction between labour market assimilation and pre-labour market assimilation. For the education entrants, assimilation consists of labour market assimilation (time spent after leaving full-time education) and pre-labour market assimilation (in the UK education system). Most investigators of the assimilation hypothesis exclude those who arrive as children or with incomplete education from the estimation sample; they therefore focus on labour market assimilation. We explore whether, given their earlier exposure to the language and culture of the UK, such education entrants have outcomes which are closer to their native counterparts than to those immigrants who enter the labour market directly. Education entrants represent around one half of all immigrants in our sample - excluding them risks neglecting a potentially important aspect of the immigrant experience.

Our work builds on previous UK studies which have used cross-section survey data to paint a picture of immigrant labour market performance. In an early paper Chiswick (1980) used a single cross section of the *General Household Survey* (GHS) and found that white immigrants earned as much as their native counterparts but that there was, other things equal, a 25% earnings penalty for non-white immigrants. He found no statistically significant role for years since migration, controlling for other things. Shields and Wheatley Price (1998) also examined earnings and used LFS data from 1992-94. Like Chiswick they found earnings differences between white and non-white immigrants. They also emphasised the differential returns to human capital acquired in the home country compared to the host country, with UK human capital generally better rewarded in the UK labour market. Using the same data Wheatley Price (2001) examined the unemployment experience of immigrants and found that more recent immigrants had higher unemployment rates than previous arrivals.

None of these studies attempts to separate the effects on labour market outcomes of changes in the quality of immigrant cohorts from that of years since migration, however this is a requirement of testing the assimilation hypothesis. In this sense our work is closer to Bell (1997) and Dustmann et al. (2003) each of which used pooled cross section data to create a 'synthetic panel' of immigrant and native workers. Bell (1997), using GHS data from 1973-92, found positive assimilation for non-white immigrants and negative assimilation (dis-assimilation) for white immigrants. That is to say, whites were predicted to arrive with higher earnings than natives but this advantage eroded through time. Dustmann et al. (2003) using LFS data from 1992-2000 distinguished immigrants by ethnicity and by region of origin. Wages were broadly predicted to rise with time in the UK for non-white immigrants (positive assimilation) and for whites from the British Commonwealth. There was strong evidence of dis-assimilation for white immigrants from Ireland and Europe. Dustmann et al. also

examined other labour market outcomes including employment rates where they found that non-white immigrants assimilate towards native levels from an initially inferior position.

Compared to previous work the innovative features of our research are the following. First, we use a larger sample of more recent data. Second, and as far as we are aware for the first time using UK data, we investigate the impact of arrival year effects on immigrant earnings and employment. Third, we make the (it turns out) important distinction between those immigrants who arrive with their education complete and those who enter the education system. Finally we employ a semi-parametric estimator, which places fewer restrictions on the estimated assimilation profiles than previous work.

Amongst our key findings are:

- positive earnings assimilation for virtually all immigrant groups;
- strong employment assimilation for those immigrants who completed their education in the UK;
- negative employment assimilation for South Asian immigrants who completed their education overseas, and
- some evidence of unemployment rates at time of entry to the labour market causing lower earnings for non-white immigrants.

The remainder of the paper proceeds as follows. Section 1 gives an overview of the data while section 2 describes the econometric methods. Section 3 discusses the results and section 4 concludes.

2. The Data

The data are drawn from the *Labour Force Survey* (LFS), conducted by the Office for National Statistics (ONS), and represent pooled cross-sections over the period 1993-2002. The LFS collects information on earnings, employment and socio-economic characteristics such as age and years of schooling. Further details on the sampling methodology and questionnaires are available from the ONS.⁵

Our first labour market outcome of interest is real gross weekly pay in main job and we analyse male, full-time workers aged between 16 and 65 at the time of interview⁶. Our second labour market outcome is whether the survey respondent was employed for pay at the time of the interview. In all the results employment rates are expressed relative to a denominator comprising the employed and the unemployed; in other words, the self-employed and inactive are excluded from the analysis. Overall after excluding observations with missing data and trimming outliers we have a sample of 148,528 native and 9,454 immigrant men.

An important component of the analysis is the distinction between those who enter the UK having completed full-time education (labour market entrants) and those who have yet to complete (education entrants). This requires dividing the sample based on information about the year in which individuals left full time education and their year of arrival in the UK. We make the assumption that education is obtained in a continuous block before (potential) labour market experience is accrued. This is the standard assumption in the human capital

⁵ <http://www.ons.gov.uk>

⁶ All earnings data were deflated to a common year. We also re-estimated the models using hourly wages and obtained qualitatively similar results.

literature.⁷ It is also worth noting that we adopt another standard convention of human capital studies: since we do not observe panel data or work histories, labour market ‘experience’ is in fact potential experience.

Table 1 provides sample means and standard deviations for some key variables by immigrant and ethnic status (white or non-white). We also further divide our white and non-white samples into labour market entrants and education entrants. The latter of course will have some UK education and may have some foreign education, but have no foreign labour market experience. Labour market entrants, by contrast, will have no UK schooling but may have foreign schooling and foreign experience⁸. Native born men, white and non-white, are included for comparative purposes.

Comparing mean earnings, immigrants fare better than natives on average and whites better than non-whites. White labour market entrants earn more on average than white education entrants, although the reverse is true for non-whites. Comparing employment rates, relative to white natives, all immigrants exhibit lower employment rates. Non-whites tend to underperform whites in all cases and for non-whites, education entrants do substantially better than labour market entrants.

⁷ Of course one could easily imagine an immigrant working either in the origin or destination country for some period before undertaking education in the destination country. Without more detailed panel or life history data it is very difficult to ascertain whether this is the case for any sample member. We can, however, examine the age at which individuals left full time education; if this is implausibly high then the assumption of a single continuous period of education may well be flawed. In the *LFS* data, the proportion of such workers was relatively small hence we proceed to make the standard assumption.

⁸ We focus on years of schooling due to the difficulty in the *LFS* of comparing qualifications obtained abroad with those obtained in the UK.

For natives, mean potential labour market experience is considerably larger for whites than non-whites (non-whites are younger on average), whilst mean years of schooling are less for whites compared to non-whites. For immigrants, white labour market entrants have less UK labour market experience than their non-white counterparts, although their years of foreign experience and schooling are the same. For education entrants whites have more UK potential experience, more years of UK schooling and less years of foreign schooling compared to non-whites. Not surprisingly, immigrants who arrived with their education complete were older on arrival than those with education incomplete. Although for the latter whites were younger on arrival than non-whites.

3. Modelling framework

Our investigation of immigrant labour market outcomes is based on the following econometric model:

$$Z_i = f(Y_i) + \gamma C_i + \delta S_i + \mathbf{x}_i \beta + \varepsilon_i \quad i = 1, \dots, n \quad (1)$$

In equation (1), Z represents a measure of labour market status, Y is years since migration, C is immigrant cohort, S is survey year (year in which the individual was observed), \mathbf{x} is a vector of other explanatory variables including human capital and ε is an error term.

Two measures of labour market status (Z) are used - real weekly wages in logarithmic form and a discrete dependent variable taking the value 1 if the individual is employed and the value 0 if they are unemployed. We follow the recent literature, particularly Dustmann and Fabbri (2003) and Antecol et al. (2003), in two regards. First, given the difficulty of finding identifying exclusion restrictions, we do not attempt to correct for sample selection bias in either employment or earnings models. Clearly this will affect the interpretation of our results

if it is thought that selection bias is a problem. Second, in order to make computation of the semi-parametric estimates more tractable, we use a linear probability model, rather than a probit or logit, to analyse employment status. There turns out to be little difference in the estimated marginal effects of the explanatory variables if a probit model is employed instead.

The years since migration variable Y will capture assimilation effects - how immigrant earnings change with length of residence in the host country. The specification of the function $f(Y)$ is discussed in the next sub-section. C is the immigrant cohort to which an individual belongs (thought of here as year of arrival) and captures otherwise unobserved differences in immigrant cohort quality over time. It has been argued that cohort quality changes have been important in explaining immigrant earnings performance in the US and UK. For example, Borjas (1985) suggests that a secular decline in the quality of immigrant cohorts to the US explains the relatively poor performance of some immigrant groups while Bell (1997) using UK data emphasises how the different national origin mix of immigrant waves has affected the overall picture of immigrant earnings. We model C using dummy variables for decade of arrival but, since cohort effects are not central to our work, we do not discuss the results in detail. It turns out that there are no clear, statistically significant, patterns in the cohort dummies in the estimated models.

In order to identify cohort and assimilation effects separately it is necessary to have observations at different points in time. Panel data would be ideal however, like most studies of immigrant earnings, we have to make do with pooled cross section data, sometimes called the 'synthetic panel' approach. The variable S reflects when the individual was observed and captures the effect of secular trends on immigrant outcomes.

The vector x contains other worker characteristics including human capital. We distinguish between human capital (education and potential experience) obtained in the UK and that obtained before arrival in the UK. It also contains marital status, region of residence and, where appropriate, industry of employment.

For both labour market outcome measures we estimate separate equations for the following four groups: (i) white labour market entrants, (ii) non-white labour market entrants (iii) white education entrants (iv) non-white education entrants. It is worth noting that most previous studies of immigrant assimilation do not estimate separate regression models for immigrants and natives but rather pool the two groups of workers and allow certain coefficients to vary by immigrant status. Two additional models for white natives and non-white natives are also estimated for comparison purposes.

3.1 Modelling Assimilation

Not all of the parameters of equation (1) can be estimated since there is perfect multicollinearity: $S \equiv C + Y$. In line with previous studies of immigrant assimilation we adopt the normalisation of fixing the coefficient on S (δ - the secular wage growth effect) and estimating the effects of C and Y freely. An estimate of δ can be obtained from the sample of native workers thus the constraint is equivalent to assuming that the period effect is equal for natives and immigrants.

With respect to the specification of the function $f(Y)$, most studies impose a non-linear functional form – a polynomial – in Y (Bell, 1997; Dustmann *et al.*, 2003; Barth *et al.*, 2004),

or divide Y into categories and use dummy variables to represent the categories (Antecol et al., 2003). Since the shape of f is key to the measurement of assimilation we adopt a slightly different approach, which imposes somewhat less structure on the model. Specifically we estimate a semi-parametric version of (1) using a partially linear model (Yatchew, 2003).

Consider rewriting equation (1) as:

$$Z_i = \mathbf{w}_i \xi + f(Y_i) + \varepsilon_i \quad i = 1, \dots, n \quad (2)$$

where the vector \mathbf{w} includes C , S and \mathbf{x} from equation (1). The function f is assumed simply to be some smooth function of years since migration. The data are ordered by Y and quasi-differenced according to the formula: $\{\mathbf{w}_i - \mathbf{w}_{i-1}\}/\sqrt{2}$. Consider the estimated regression on differenced data

$$\hat{\xi}_D = (\mathbf{W}'_D \mathbf{W}_D)^{-1} \mathbf{W}'_D \mathbf{Z}_D \quad (3)$$

where \mathbf{W}_D is a matrix of quasi-differenced individual observations on the explanatory variables (excluding Y) and \mathbf{Z}_D is the equivalent for the dependent variable. Yatchew (2003) shows that

$$Z_i - \mathbf{w}_i \hat{\xi}_D \approx f(Y_i) + \varepsilon_i \quad (4)$$

and that kernel regression methods applied to the ordered pairs $\{Z_i - w_i \hat{\xi}_D, Y_i\}$ yield a consistent semi-parametric estimator of the function f . In the empirical application, the non-parametric estimation was done using a Nadaraya-Watson kernel density estimator implemented using a modified version of the Stata module *kernreg1*. We used a Gaussian kernel and began from a bandwidth chosen according to the formulae in StataCorp. (2001, p. 167). The bandwidth was then adjusted (invariably upwards) to give an appropriate degree of smoothing. The results were not particularly sensitive to choice of kernel function and were qualitatively similar to results obtained using other smoothing techniques.⁹

In terms of the amount of structure imposed on the data, the semi-parametric estimator can be thought of as lying somewhere between a polynomial in Y and modelling each year since migration with a dummy variable. The former imposes a smooth shape on the function but is restrictive in the sense that it requires symmetry around the function's turning points while the latter imposes no smoothness on the function but may, in a finite sample, be susceptible to sampling error.

3.2 Modelling Arrival Effects

To investigate the impact of economic conditions at time of arrival to the UK we replace the cohort dummies in (1) with two variables. The first is the male unemployment rate for the UK in the year of entry to the labour market while the second is the rate of GDP growth. The unemployment rate has been used in a number of studies including Chiswick et al. (1997) and

⁹ One further issue with the semi-parametric approach arises from the quasi-discrete nature of the variable Y which is measured as whole years since migration. Since the data are to be sorted by Y , multiple different sort orders are possible. To overcome this problem we took averages over a large number of sorts of the data. Experimentation suggested that estimates settled down sufficiently after 40 replications of the quasi-differenced regression in equation (3)

Chiswick and Miller (2002) for the US, and Aslund and Rooth (2003) for Sweden. Conceptually this captures the essence of the scarring hypothesis. We have also included the growth rate to investigate whether more general economic conditions at arrival have any impact on future earnings and employment opportunities (Stewart and Hyclak (1984) do this for the US). The ‘macro’ variables pertaining to the year in which the immigrant entered the labour market are entered into the regression model: this is year of arrival for labour market entrants and year left full-time education for education entrants. Following Chiswick et al. (1997), we also experimented with entering an average unemployment or growth rate based on a 3-year moving average centred on the year of entry to the labour market plus one.

4. Results

4.1 Assimilation

To discuss the results of our regressions pertaining to the assimilation hypothesis, we note first that the variable Y in equation (1), representing years since migration, has a quite different interpretation depending on whether an immigrant is a labour market entrant or an education entrant. For those who enter the labour market, Y is identical to years of potential UK labour market experience and whether or not such an immigrant assimilates towards the labour market status of a similar native is a function of how labour market outcomes depend on UK experience. For those who complete their education in the UK, years since migration is some combination of years in the UK education system plus years of UK potential labour market experience. Hence assimilation for this group will depend on their labour market returns to UK education and the returns to UK experience.

Thus a good place to start understanding how immigrant outcomes adjust with length of time in the UK is to examine returns to human capital for immigrant and native groups. Table 2 provides estimates of the returns to immigrant and native human capital in the UK labour market over the period 1993-2002.

Panel (a) of Table 2 contains the results for the log of weekly earnings while panel (b) is the equivalent for employment status. In each table we have reported the estimated coefficients and standard errors for UK schooling and foreign schooling. For both labour market outcomes both types of schooling have a positive and statistically significant impact. There are, however, differences in the estimated returns to an additional year's education depending on where that education was obtained and to which sub-group the individual belongs. In the earnings models an additional year of UK schooling benefits natives more than immigrants and, for both outcomes, non-white natives have higher rates of return than whites. For those immigrants for whom we observe both UK and foreign schooling (education entrants), it is, perhaps unsurprisingly, the UK variety which offers the higher earnings and employment return.

We also report, in Table 2, estimates of the returns to UK and foreign potential experience. These are based on a quadratic specification of experience and represent an estimate of the marginal return to an additional year's experience calculated at the mean level of (UK or foreign) experience for the sub-sample in question¹⁰. Considering earnings first, the return to an additional year of UK experience is estimated at around 1-2% for virtually every group and

¹⁰ These are estimated as $\gamma = \beta_1 + 2\beta_2 \overline{\text{exp}}$ where β_1 is the coefficient on experience and β_2 the coefficient on experience squared and $\overline{\text{exp}}$ is the mean value of experience. The reported standard error and significance level is for a test of the null hypothesis that γ is equal to zero.

is highly significant. The exception is non-white natives where the return is estimated to be around 4%. Similar to the findings of Shields and Wheatley Price (1998), foreign experience is a statistically significant determinant of earnings only for the whites in our sample. This can be interpreted in at least two ways: white immigrants to the UK are more likely to come from a developed country labour market where acquired human capital will have value in other developed countries. On the other hand, differences in returns to human capital between racial groups are often seen as evidence of labour market discrimination and the differences seen here between white and non-white immigrants may reflect such employer attitudes.

The results for the employment probability regressions in panel (b) have many broad similarities with those for earnings. The main difference is for those immigrants who arrived in the UK to enter the labour market for whom UK experience is statistically insignificant and foreign experience has a significant negative coefficient. This may reflect that immigrants with large amounts of foreign experience will tend to be relatively old; studies have found higher unemployment incidence among older workers (e.g. Nickell, 1980).

However we should be careful about interpreting the marginal returns to UK experience as indicative of how time in the UK affects the labour market status of immigrants relative to natives. In particular, the results in table 2 might suggest that the experience-earnings profiles for white and non-white natives are very different. As Figure 1 demonstrates this is not the case.

Figure 1(a) plots the earnings-experience profiles for white and non-white native workers while Figure 1(b) plots the employment-experience profiles for the same groups. Two

estimates of this profile are produced for each group giving a total of 4 lines on each graph. The first estimate is based on the OLS regression results reported in Table 2 and shows the fitted quadratic in experience. The second is a semi-parametric estimate based on the partially linear model introduced in section 3. Two important points are worth making on the basis of Figure 1. First, consider again the apparent difference between white and non-white natives in the marginal returns to labour market experience shown in Table 4. Figure 1 reveals that, for earnings at least, the profiles for these two groups are very similar. The large difference in marginal returns is because non-whites have much less UK experience on average (see Table 1) and hence their average return to experience is computed at a steeper part of the profile. The second thing to note is that the quadratic specification of experience can impose too much structure on the predicted profiles. This is most apparent for the non-whites in Figure 1(b) where the semi-parametric estimate is much flatter than the quadratic curve.¹¹ Together these suggest that we should look flexibly at the whole range of UK experience when analysing how immigrant outcomes vary with time in the UK labour market.

4.1.1 Labour Market Entrants

We now move on to explicitly consider the assimilation of immigrant groups. We look first at the labour market progress of those immigrants who arrived in the UK and entered the labour market (i.e. those whose education was complete).

The discussion of assimilation requires a baseline: to what are immigrants supposed to assimilate? Much of the existing literature measures assimilation by conducting the following

¹¹ In fact for some of the immigrant groups the differences between quadratic and semi-parametric results were even more pronounced.

simulation exercise: imagine suitably similar native and immigrant workers entering the UK labour market at some particular age. Progressively increment their age and observe what happens to their relative labour market success. If the paths converge, this is evidence of assimilation; if they diverge this is evidence of dis-assimilation. An analogy might be to two escalators, one native, one immigrant, and the objective is to see which escalator is rising fastest. A problem with this approach is that it runs the risk of obscuring the performance of immigrants behind that of natives. That is to say, if native earnings are rising quickly with experience we may conclude that there is dis-assimilation, even though immigrant workers also experience rising earnings as their time spent in the UK increases. From the perspective of a policy maker, or indeed that of a new or potential immigrant, it may be more useful to measure immigrant outcomes relative to a fixed, “average” native worker. The question is then: how long does it take an immigrant to become like the average native. This latter approach, which we follow here, isolates the importance of returns to immigrant human capital in the host labour market.

We adopt this approach by examining the predicted weekly earnings and employment probabilities for an immigrant who arrived in the UK having completed the sample average amounts of education and experience in their own country. For both natives and immigrants, we consider a married male, living in the South East, working in non-manufacturing who otherwise has sample average characteristics. To abstract from secular wage and employment changes we use a comparison year of 1997 for immigrants and natives. We then allow our typical immigrant’s predicted earnings and employment probabilities to evolve in line with the semi-parametric estimate of their experience profile. This exercise is conducted separately for whites and non-whites.

Figure 2 shows the results for immigrant labour market entrants. Panel (a) refers to (log) gross weekly earnings while panel (b) refers to the employment probability. The horizontal lines are the predicted earnings and employment probabilities for the white and non-white native comparison individuals.

The earnings profiles in Figure 2(a) for whites and non-whites have a very similar shape rising gradually to a peak at around 30 years after migration. There is therefore evidence of positive assimilation relative to the fixed baseline of the average native worker. Between arrival and 30 years since migration real white immigrant earnings are estimated to rise by 0.21 log points while for non-whites the equivalent figure is around 0.22 log points. The big difference between white and non-white immigrants is in the intercept of the earnings profile rather than its slope with white immigrants earning 0.31 log points more on arrival than non-whites. The gap between a newly arrived white immigrant and the comparison white native is 0.02 log points compared to 0.161 for non-whites.

As the figure demonstrates, white immigrants earn more than non-white natives on arrival and overtake the comparison white native worker after 4 years since migration. Our non-white immigrant take something approaching 20 years to achieve parity with the comparison non-white native and never reaches the earnings level of the white native. Thus while earnings assimilation, in the sense of higher earnings growth with UK experience, takes place for both white and non-white immigrants, there is a substantial ethnic differential between white and non-white immigrants that is not significantly eroded as time in the UK increases.

Figure 2(b) undertakes the same exercise for the fitted employment probabilities. Here a quite different picture emerges. Non-white immigrant employment probabilities begin at a higher level than their native comparator and decline from around 18 years since migration. Such immigrants therefore experience what could be described as dis-assimilation. For the white immigrants, employment probabilities begin at a comparatively low level – a white immigrant has an employment probability on arrival very similar to a non-white immigrant – but gradually rise with years since migration overtaking the white native comparator after some 28 years.

To provide further insight into the ethnic differences in labour market outcomes, Figure 3 breaks down the assimilation results by individual ethnic group for four of the largest ethnic minority groups in the UK. The figure shows, relative to a native individual of the same ethnic group, the immigrant advantage in earnings or employment plotted as a function of years since migration. These results are based on the semi-parametric assimilation profiles estimated separately for each non-white group.

For earnings, most non-white immigrants (excluding Indians) demonstrate positive assimilation. Black African immigrants enter the labour market on slightly lower earnings than African natives, their earnings then overtake those for natives fairly quickly and then diverge away before starting to converge again after 39 years. Pakistani and Caribbean immigrants start on much lower earnings than their native counterparts (around 0.3 and 0.4 percentage log points lower respectively) but then earnings assimilate towards natives, overtaking after 13 (20) years for Pakistanis (Caribbeans). Finally Indian immigrants appear to start on slightly higher earnings than Indian natives, although earnings fall below those for

natives after only a couple of years. Immigrant Indian earnings remain fairly close to natives thereafter.

For employment, there is a clear dichotomy between Black and South Asian immigrants in terms of how they enter the labour market relative to their native counterparts. Caribbean and Black African immigrants start off worse than natives (around 0.21 and 0.08 log points lower respectively) but employment propensities overtake those for natives after 10-13 years, diverging thereafter. For Caribbeans the employment advantage peaks after about 20 years and then converges towards natives again, whereas Black African employment appears to diverge exponentially. Indian and Pakistani immigrants start off doing better than natives (around 0.12 and 0.06 log points higher respectively) but then employment assimilates towards natives falling below that for natives after around 32 (24) years for Indians (Pakistanis).

Figure 2 demonstrated that non-white immigrant average earnings begin below those for non-white natives, but then their earnings increase. Figure 3 shows that this is an average of three different types of behaviour: Black Africans who start pretty close to natives and then increase their earnings; Caribbeans and Pakistanis who start well below their counterparts but gradually assimilate and finally Indians whose earnings remain fairly close to natives. For employment, the non-white dis-assimilation observed in Figure 2 can be mainly attributed to the South Asian groups. Pakistani and Indians start off at an advantage but then deteriorate, whereas Black Africans and Caribbeans start off disadvantaged and then improve.

On the whole it is the Asian immigrants who do worse (with the exception of Pakistanis for earnings differences). Caribbeans and Black African immigrants fare much better than South

Asians, relative to natives, over the long term. The choice of comparator is crucial here, however. Blackaby et al. (1998, 2002) study the employment and earnings of non-white ethnic minorities (immigrant and native-born) in the UK and find that, compared to the majority white community, it is the Black (Pakistani) group who do worst in terms of employment (earnings).

4.1.1 Entrants to Education

We now turn to examine the labour market assimilation of those immigrants who arrived in the UK to enter the education system, either as adults or as children. Here the correspondence between years since migration and UK experience is broken and this needs to be accounted for when examining assimilation profiles. To illustrate the results we consider four individual ‘types’ similar to those typical individuals used in the preceding sub-section but differentiated by their age at arrival in the UK and their UK educational attainment. Specifically the four types are:

- Type I: arrived aged 5, leaves education aged 16
- Type II: arrived aged 5, leaves education aged 21
- Type III: arrived aged 16, leaves education aged 18
- Type IV: Arrived aged 18, leaves education aged 22.

The results of this exercise are shown in Figure 3, which has four panels. Consider panel (a) which examines earnings for white immigrants who entered education on arrival. The upper horizontal line is the level of (log) earnings for the white native comparison individual as described in the preceding sub-section and the lower horizontal line is the same for the non-white native comparison individual. Years since migration are measured along the horizontal axis and predicted earnings profiles are plotted for each of the four types described above.

These do not begin at zero years since migration as education entrants do not join the labour market until after their UK education is complete and this time period varies by type. Thus we can see that the individual who arrives aged 5 and leaves school at 16 has lower earnings on entry to the labour market than any of the other groups while the immigrants with more education have higher initial earnings. The most striking feature of this graph is the strong earnings growth that takes place over the first 15 or so years of UK labour market experience. From labour market entry to the peak of the profile is a log difference of 0.82.

Panel (b) plots fitted earnings for non-whites on the same basis. A quite different picture emerges with much slower earnings growth over the range of UK experience. From labour market entry to the peak is a log difference of 0.18 suggesting much lower earnings assimilation. What dominates the differences in earnings in this picture are differences in entry-level earnings. The 'best educated' non-white immigrants (types II and IV) are predicted to earn more than the comparison white native individual on entry to the labour market. The contrast with the 'best educated' white immigrants (the same types) in panel (a) is marked suggesting that non-white immigrants can achieve considerable progress in the labour market through investment in UK education.

Panels (c) and (d) complete the picture by plotting the evolution of employment probabilities for white and non-white immigrants respectively using the four immigrant types outlined above. For both whites and non-whites there is strong growth in employment probabilities over the first 10 years after arrival. For whites this flattens out while for non-whites it gradually declines again through time.

4.2 Arrival Year Effects

Tables 3 and 4 report the results of parametric least squares estimates of equation (1) where we replace the cohort dummies with variables reflecting the state of the labour market and wider economy in the immigrant's first year in the British labour market. For immigrants who arrived with their education complete this is their year of arrival to the UK. For immigrants who arrived with their education incomplete this is the year that they left full time education. We estimated six models for each labour market outcome. Year of entry unemployment and growth rates were considered singly and jointly then the same procedure was carried out for the three-period moving averages (centred on the year after entry) of the aggregate variables.

For both earnings and employment many of the estimated coefficients are insignificantly different from zero. The principal exception is for the earnings of non-white labour market entrants where significant negative coefficients on unemployment are found in all four models where unemployment appears. The estimated coefficients range from -0.017 to -0.014 . To give some idea of the economic significance of these results, a coefficient of -0.015 implies that a one percentage point increase in the unemployment rate on entry to the UK labour market is associated with a reduction in earnings of around 1.5%. Over the period during which members of our sample arrived in the UK the male unemployment rate varied between 1% (1943) and 22% (1932). Unemployment changes in the initial years of labour market experience could therefore have a quantitatively significant impact on the long-term earnings potential of non-white immigrants.

The only other significant coefficients in Table 3 are negative signs on GDP growth rates for white labour market entrants in two models. These are counter-intuitive: a one percentage point higher growth rate is associated with a reduction in earnings of around 0.9%. UK growth rates over the period were almost always in the range -5% to 5% .

Turning now to the linear probability models for employment in Table 6 we observe that for non-white labour market entrants, only growth in GDP is statistically significant. Non-white labour market entrants who arrived in a period of economic growth enjoy an employment premium. This is not true for their white counterparts, since all the arrival effects are statistically insignificant. For the education entrant immigrants, all the labour market entry variables are statistically insignificant for non-whites, although whites appear to enjoy an employment premium for high arrival year unemployment rates. Again this is a counter-intuitive result.

In fact these “perverse” results are not out of line with previous research: the existing literature on arrival year effects and the scarring hypothesis is characterised by somewhat mixed results. Chiswick et al. (1997) find that arrival year unemployment rates exhibit significantly positive coefficients in one specification of an individual employment equation with insignificant coefficients in other specifications. MacDonald and Worswick (1998) find a positive impact of initial unemployment on earnings using Australian data. Stewart and Hyclak (1984) and Chiswick and Miller (2002), both for the US, obtain results more in line with the scarring hypothesis.

5. Discussion and Conclusions

In this paper we use Labour Force Survey data to document the impact of arrival year economic conditions and assimilation on the labour market outcomes of immigrants to the UK. The innovative features of our work include the separate analysis of immigrants who arrive in the UK to enter the labour market from those who enter education, the investigation of arrival year economic conditions on labour market outcomes for immigrants and the use of a semi-parametric method to estimate assimilation profiles. Below we summarise and discuss our key results.

Earnings for white and non-white immigrants, whether labour market entrants or entrants to education, rise during the early years of their stay in the UK. The levels of earnings and the rate of change differ substantially across immigrant ethnicity and type, nevertheless, compared to a baseline or typical native, there is evidence of positive earnings assimilation in the UK labour market. Annual rates of earnings growth over the first 15 years in the UK, based on the semi-parametric profiles in Figure 2(a) and computed to be net of secular wage growth and inflation, are around 1.0% for non-whites and 1.1% for whites. Assimilation is particularly strong for white entrants to education once they reach the labour market: based on the profiles in Figure 3(a), real weekly earnings are predicted to double in around 11 years. This compares to over 40 years for non-white immigrants.

Earnings levels for non-whites fall short of those for whites, whether for immigrant or native workers, and this is a recurring theme of the paper, emerging in the raw data and the regression estimates. If we consider the gap between the earnings of an immigrant beginning their labour market career and a comparison, typical, native worker, the initial gap is around

0.15 log points larger for non-whites and takes 16 years longer to close. Ethnic wage gaps are recognised as an enduring feature of the UK labour market and our results suggest that Pakistani and Caribbean immigrants who enter the labour market directly encounter a particular disadvantage, although they assimilate and overtake natives after 15-20 years. Compared to their native counterparts, Black African and Pakistani workers fare better than Caribbeans and Indians.

The one optimistic note for non-white immigrants concerns those who entered the UK prior to the completion of their education. While assimilation profiles are less steep for non-white education entrants compared to whites, such immigrants do experience strong returns to their UK education as evinced by the entry-level earnings displayed in Figure 3(b). Sufficient UK schooling can completely wipe out the disadvantage faced by non-white immigrants as they enter the labour market. Our measure of schooling (years in education) is somewhat crude, due to data limitations, and future work could profitably establish what types of educational investment underlie these high returns for non-white immigrants.

Our results pertaining to employment assimilation were more mixed. Amongst labour market entrants, whites exhibited positive assimilation with employment rates rising with time since migration. This was also the case for white and non-white education entrants. For non-white labour market entrants, however, the relationship was negative with a pronounced fall off in predicted employment rates around 16 years after migration. This observation can mainly be attributed to South Asian immigrants, since Black African and Caribbean immigrants fare much better than their native counterparts in the long-run.

Dis-assimilation for particular immigrant groups has been found in the UK before: Bell (1997) observed earnings dis-assimilation for white immigrants and postulated that selective re-migration was the explanation. The issue of re-migration is difficult to address in the absence of detailed administrative records or panel data with sufficiently large immigrant samples, however Rendall and Ball (2004) have suggested that as many as half of all recent immigrants to the UK emigrate again within 5 years. However, they also find that the countries previously sending many non-white immigrants to the UK (the Indian sub-continent and Commonwealth Caribbean) have significantly lower rates of re-migration than countries, which are the source of white immigrants. It therefore seems unlikely that selective re-migration is wholly responsible for the patterns observed here and we posit an alternative explanation. Many non-white immigrants to the UK were recruited directly to public sector employment in the 1950s and 1960s. We would expect such immigrants to have very low unemployment risk on arrival and in the next few years. As time goes by this group would experience shocks and an 'equilibrium' rate of unemployment for those individuals given their skills and market opportunities would be established. Such a view would be consistent with some employment dis-assimilation as we observe.

As a final comment on our assimilation results it is worth pointing to some important differences between our results and those of previous studies. Both Bell (1997) and Dustmann et al. (2003) find positive earnings assimilation profiles for non-whites and negative profiles for some white groups. However these authors compute their assimilation profiles differently to us - comparing immigrants and natives who are both on rising earnings profiles - and they use different comparison groups of natives. Furthermore they do not estimate separate regression models for natives and immigrants. Borjas (1994) notes how alternative methodologies have led to conflicting estimates of the extent of assimilation for

Mexican workers in the US hence estimates of assimilation must be interpreted in the light of the modelling decisions made by the researchers in question. We have focussed on immigrants' returns to UK experience and UK schooling on the basis that these are the fundamental drivers of immigrant labour market outcomes.

The second main hypothesis that we investigate is whether aggregate economic conditions at the time immigrants enter the labour market can have a permanent impact on their labour market success. There is some evidence that non-white immigrants who arrive in the UK at times of high unemployment and immediately enter the labour market suffer an earnings penalty compared to those who arrive in years of lower unemployment. Arguably, non-white labour market entrants might be expected to be more susceptible to the scarring effect of unemployment than other types of immigrant if employers have less information about their qualifications and backgrounds than they would about white immigrants who originate in countries with education systems and labour markets similar to the UK. For other immigrant groups, and in models where we use the growth rate as the indicator of aggregate economic conditions, the results are either insignificant or, counter intuitively, suggest that a poorer macroeconomic environment has a permanent, positive impact on immigrant outcomes. There is no consensus in the empirical literature on the size, sign or significance of immigrant arrival effects, however this may be due to data limitations: panel data for the UK have been used to provide convincing evidence of unemployment scarring for native workers (Arulampalam, 2001). It is probably asking a lot of our data, using a single aggregate annual unemployment or growth rate as an indicator of individual employment risk, to uncover scarring effects. In other countries, where administrative records allow large samples of immigrants to be analysed, important linkages between aggregate or local labour market

conditions and immigrant outcomes have been observed (Aslund and Rooth, 2003; Barth et al., 2004).

One further caveat which should be mentioned is the absence of data on English language ability in the Labour Force Survey. Using other data sources Dustmann and Fabbri (2003) and Lindley (2002) demonstrate the important association between language proficiency and labour market success. Differences in language ability between ethnic groups may go some way to explaining the ethnic differences we observe.

From a policy perspective assimilation matters if immigrants are expected by the host country to, in the Prime Minister's words, "support themselves". Evidence of a significant and persistent failure of immigrant labour market outcomes to approach those of natives could be used to bolster estimates of the economic cost of immigration. There is little evidence in our results that, taken as a whole, immigrants in the UK labour market systematically fail to reach high levels of success. Clearly there are caveats to this. First, we only observe those immigrants who make it to the labour market and do not observe non-participants or those operating in the shadow economy. Second, patterns of immigration are constantly changing in response to international developments and host country policy changes, hence we should be wary of extrapolating from what previous immigrant cohorts experienced to the performance of future cohorts. Nevertheless, the large differences in outcomes that we do observe are related as much to ethnicity rather than immigrant status *per se*, and it would seem to be here that appropriate policy measures would be most effective.

**Table 1. Sample Means of Key Variables by Immigration and Ethnic Status.
QLFS 1993-2002**

	Natives		Immigrants: Labour Market Entrants		Immigrants: Education Entrants		Total
	Whites	Non- Whites	Whites	Non- Whites	Whites	Non- Whites	
Mean Gross Weekly Pay	376.28 (204.18)	342.66 (191.55)	454.99 (264.13)	363.36 (234.78)	434.25 (240.52)	390.89 (223.19)	378.23 (206.971)
Employment Rate	90.43 (0.2898)	75.13 (0.4323)	89.69 (0.3041)	79.39 (0.4045)	89.97 (0.3005)	81.77 (0.3862)	90.14 (0.2981)
Arrival Age	-	-	26.81 (8.000)	26.45 (7.232)	6.03 (6.446)	11.33 (6.453)	17.88* (11.758)
UK Experience	21.51 (12.65)	10.07 (8.158)	14.03 (13.363)	15.99 (11.991)	18.39 (11.861)	15.85 (9.877)	20.97 (12.70)
Foreign Experience	-	-	7.87 (7.260)	7.43 (6.585)	-	-	4.01* (6.304)
UK Schooling	13.02 (2.45)	14.29 (2.860)	-	-	11.24 (4.890)	8.06 (5.315)	12.46 (3.572)
Foreign Schooling	-	-	14.98 (4.027)	14.99 (3.811)	3.42 (5.407)	7.22 (5.807)	10.24* (7.035)
N	204338	3382	4046	4115	4356	3000	223237
N for employed and positive wage	146719	1809	2481	2185	3054	1734	157982

Notes: Standard deviations in parentheses.

** For the sample of immigrants only.*

Table 2. Returns to Human Capital: OLS Regression Results**(a) Log Gross Weekly Earnings**

	Native Born		Immigrants			
	White	Non-White	Labour Market Entrants		Education Entrants	
White			Non-White	White	Non-White	White
UK Schooling	0.083** (0.000)	0.090** (0.004)	-	-	0.072** (0.003)	0.061** (0.004)
Foreign Schooling	-	-	0.066** (0.003)	0.060** (0.004)	0.063** (0.003)	0.050** (0.003)
UK Experience	0.014** (0.000)	0.041** (0.002)	0.010** (0.003)	0.011** (0.003)	0.018** (0.002)	0.007** (0.003)
Foreign Experience	-	-	0.016** (0.002)	0.003 (0.003)	-	-
N	146719	1809	2481	2185	3054	1734
R ²	0.35	0.37	0.23	0.21	0.33	0.30

(b) Employment Status

	Native Born		Immigrants			
			Labour Market Entrants		Education Entrants	
	White	Non-White	White	Non-White	White	Non-White
UK Schooling	0.010** (0.000)	0.026** (0.003)	-	-	0.010** (0.002)	0.012** (0.002)
Foreign Schooling	-	-	0.010** (0.001)	0.011** (0.002)	0.007** (0.001)	0.010** (0.002)
UK Experience	0.0007** (0.00006)	0.020** (0.001)	0.002 (0.001)	-0.002 (0.002)	0.003** (0.001)	0.004** (0.001)
Foreign Experience	-	-	-0.002** (0.001)	-0.003** (0.001)	-	-
N	204338	3382	4046	4115	4356	3000

Notes:

1. The returns to experience are based on a quadratic specification and are computed at the sample mean of the experience variable (foreign or UK) for the relevant group. The coefficients are based on separate regressions for each group containing the following additional explanatory variables: a time trend (survey year), immigrant cohort dummy variables, marital status, regional dummy variables, a manufacturing dummy and, for the non-white regressions, ethnic group dummy variables. For the immigrant equations the coefficient on the time trend variable is constrained to be equal to that for the white native group.
2. The table reports estimated coefficients and standard errors. ** indicates significance at the 5% level or lower while * indicates significance at between the 10% and 5% level.

Table 3. Arrival Year Effects: Earnings

	Immigrants			
	Labour Market Entrants		Education Entrants	
	White	Non-White	White	Non-White
Model 1				
Unemployment Rate	0.0063 (0.0047)	-0.0142** (0.0063)	0.0033 (0.0052)	0.0014 (0.0051)
GDP Growth	-0.0089** (0.0032)	0.0050** (0.0022)	-0.0034* (0.0018)	-0.0013 (0.0037)
Model 2				
Unemployment Rate	0.0056 (0.0049)	-0.0139** (0.0062)	0.0027 (0.0050)	0.0013 (0.0051)
Model 3				
GDP Growth	-0.0085** (0.0035)	0.0045* (0.0025)	-0.0032* (0.0018)	-0.0012 (0.0037)
Model 4				
Unemployment Rate 3 Period MA	0.0056 (0.0054)	-0.0167** (0.0065)	0.0046 (0.0058)	-0.0010 (0.0053)
GDP Growth Rate 3 Period MA	-0.0034 (0.0063)	-0.0003 (0.0068)	-0.0078 (0.0048)	-0.0046 (0.0073)
Model 5				
Unemployment Rate 3 Period MA	0.0059 (0.0054)	-0.0166** (0.0065)	0.0043 (0.0058)	-0.0013 (0.0053)
Model 6				
GDP Growth Rate 3 Period MA	-0.0042 (0.0062)	0.0015 (0.0066)	-0.0076 (0.0050)	-0.0047 (0.0072)
N	2481	2185	3054	1734

Note:

1. The table contains estimated coefficients and standard errors based on a parametric (quadratic) specification of equation (1) with cohort dummies replaced by the aggregate-level variables.
2. * indicates statistical significance at between 5 and 10% while ** indicates significance at 5 % or lower.

Table 4. Arrival Year Effects: Employment

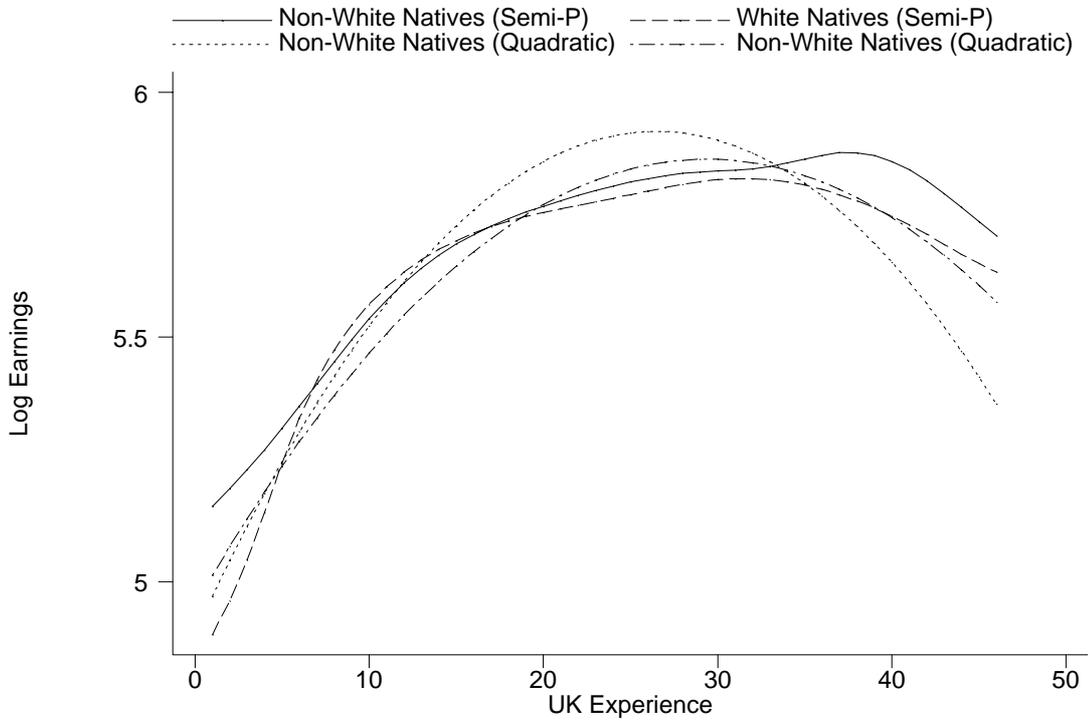
	Immigrants			
	Labour Market Entrants		Education Entrants	
	White	Non-White	White	Non-White
Model 1				
Unemployment Rate	0.0037 (0.0026)	-0.0029 (0.0036)	0.0077** (0.0030)	0.0067* (0.0037)
GDP Growth	-0.0016 (0.0017)	0.0063** (0.0024)	-0.0012 (0.0018)	0.0003 (0.0028)
Model 2				
Unemployment Rate	0.0036 (0.0026)	-0.0024 (0.0038)	0.0076** (0.0029)	0.0068* (0.0036)
Model 3				
GDP Growth	-0.0015 (0.0017)	0.0062** (0.0025)	-0.0008 (0.0019)	0.0006 (0.0029)
Model 4				
Unemployment Rate 3 Period MA	0.0027 (0.0028)	-0.0066 (0.0040)	0.0075** (0.0033)	0.0068* (0.0039)
GDP Growth Rate 3 Period MA	-0.0030 (0.0028)	0.0056 (0.0038)	-0.0062 (0.0049)	-0.0072 (0.0063)
Model 5				
Unemployment Rate 3 Period MA	0.0032 (0.0028)	-0.0072* (0.0042)	0.0076** (0.0034)	0.0069* (0.0040)
Model 6				
GDP Growth Rate 3 Period MA	-0.0035 (0.0028)	0.0064 (0.0039)	-0.0062 (0.0053)	-0.0073 (0.0069)
N	4046	4115	4356	3000

Note:

1. The table contains estimated coefficients and standard errors based on a parametric (quadratic) specification of equation (1) with cohort dummies replaced by the aggregate-level variables.
2. * indicates statistical significance at between 5 and 10% while ** indicates significance at 5 % or lower.

Figure 1. Experience Profiles: Native Born

(a) Earnings



(b) Employment

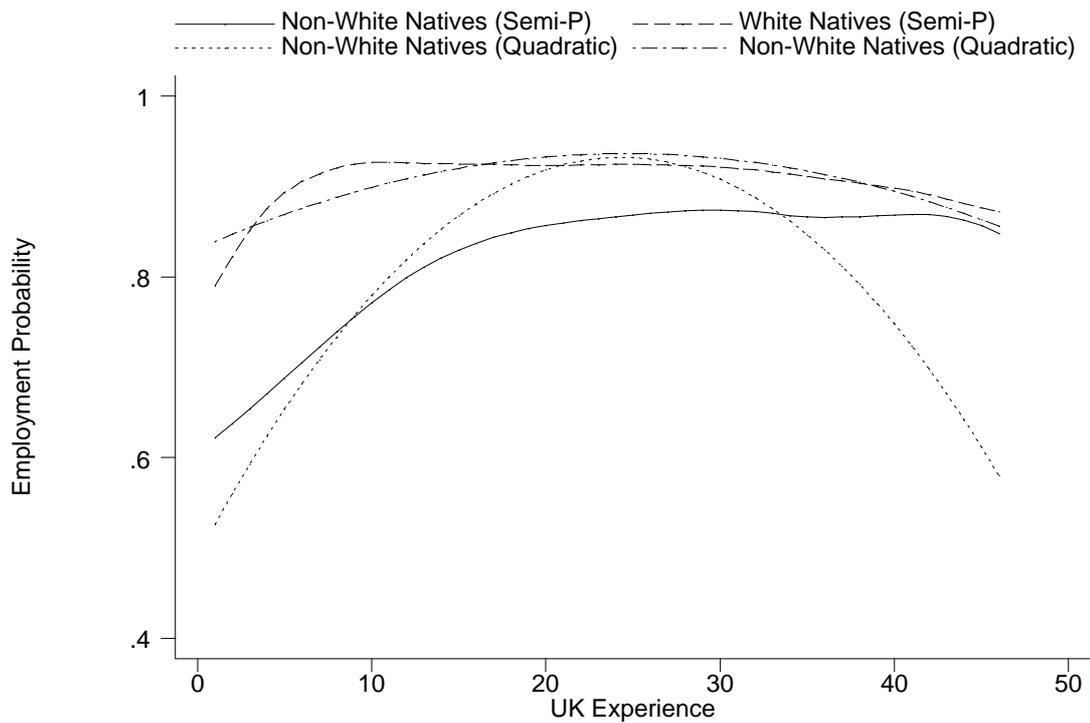
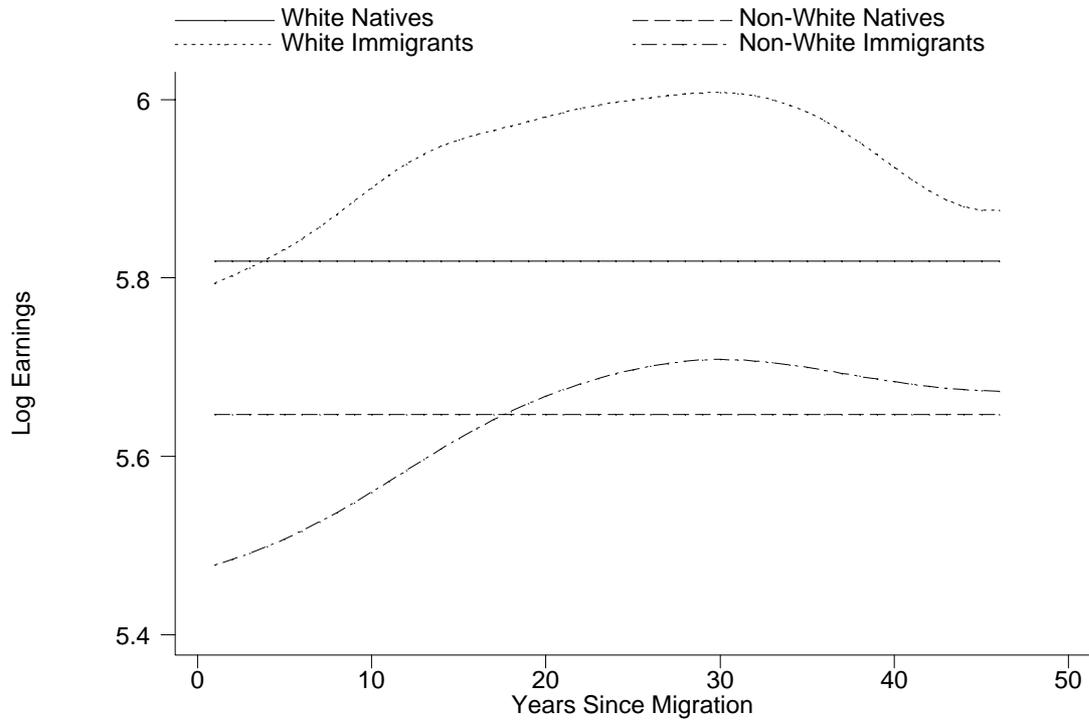


Figure 2. Assimilation: Labour Market Entrants

(a) Earnings



(b) Employment

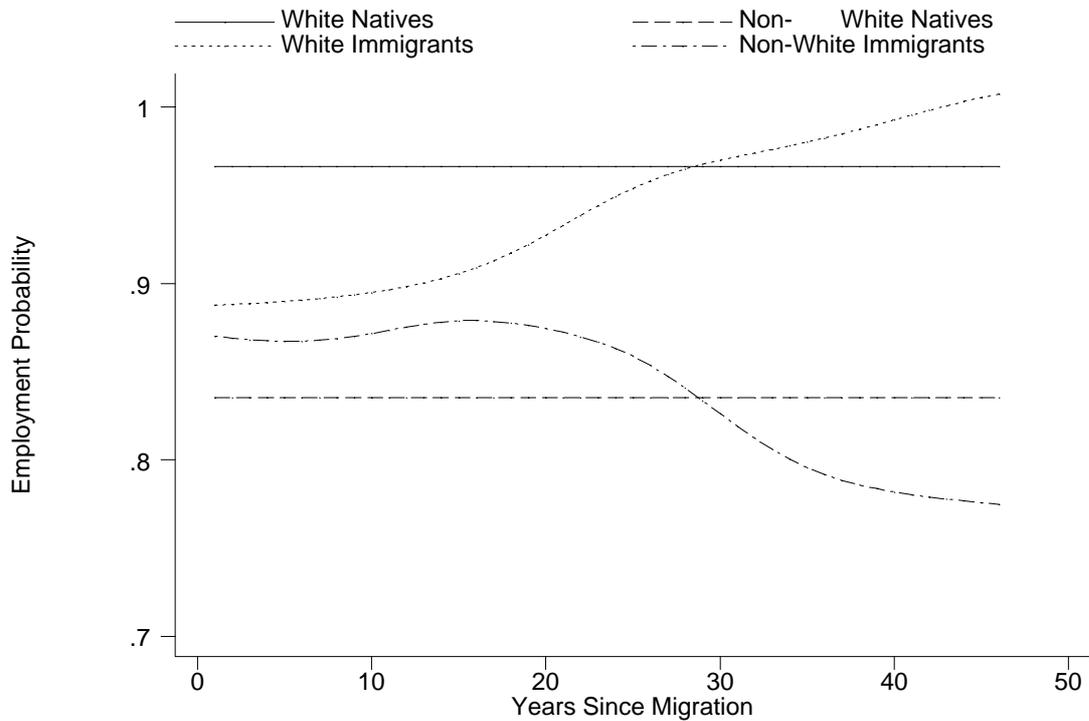
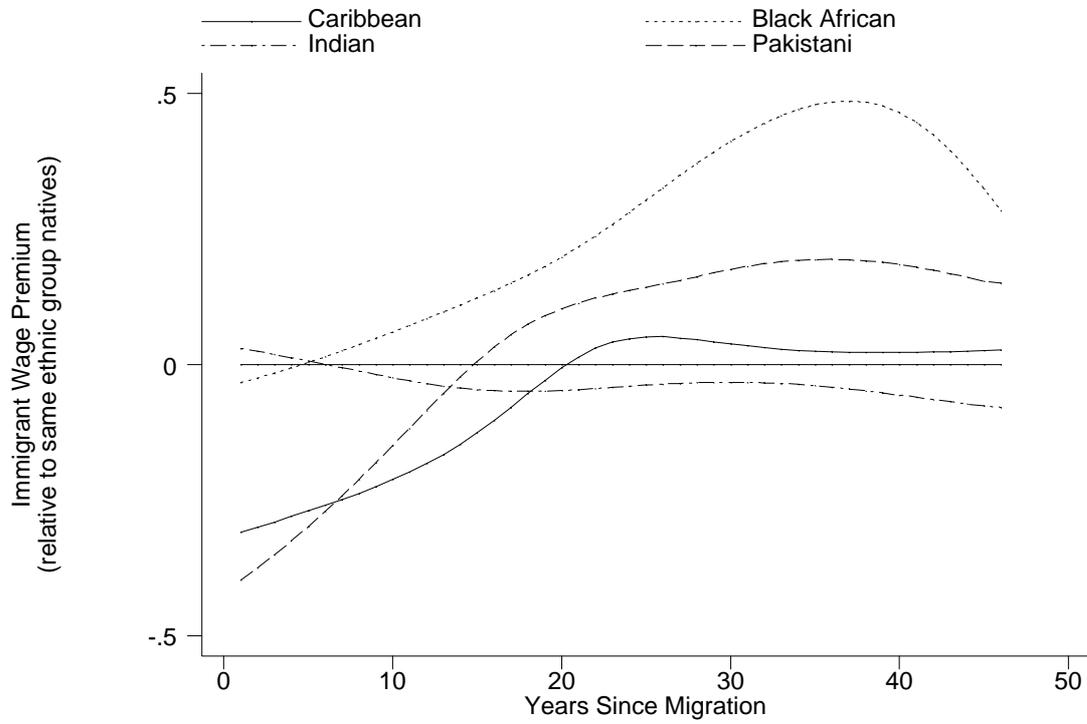


Figure 3. Assimilation Profiles by Ethnic Group

(a) Earnings



(b) Employment

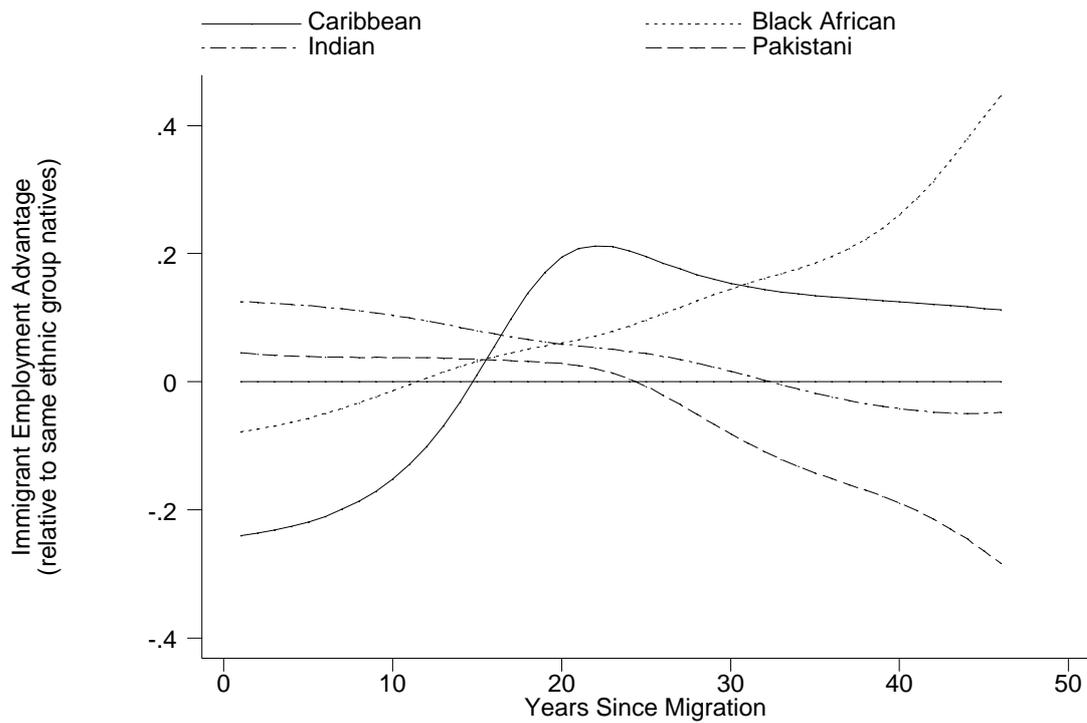
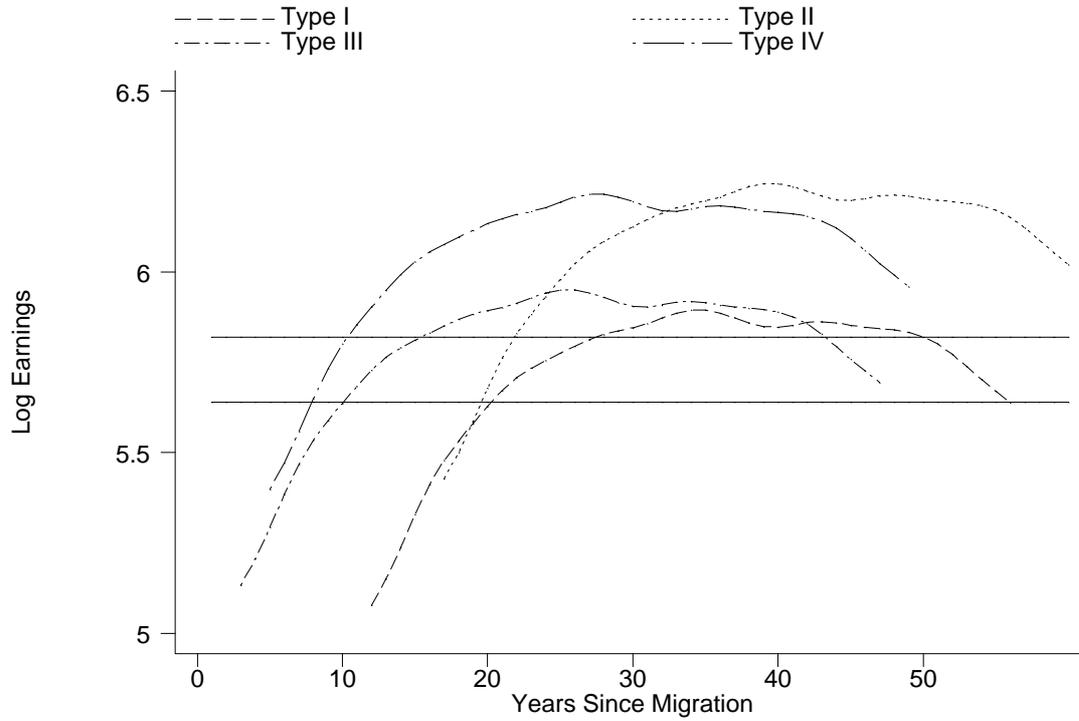
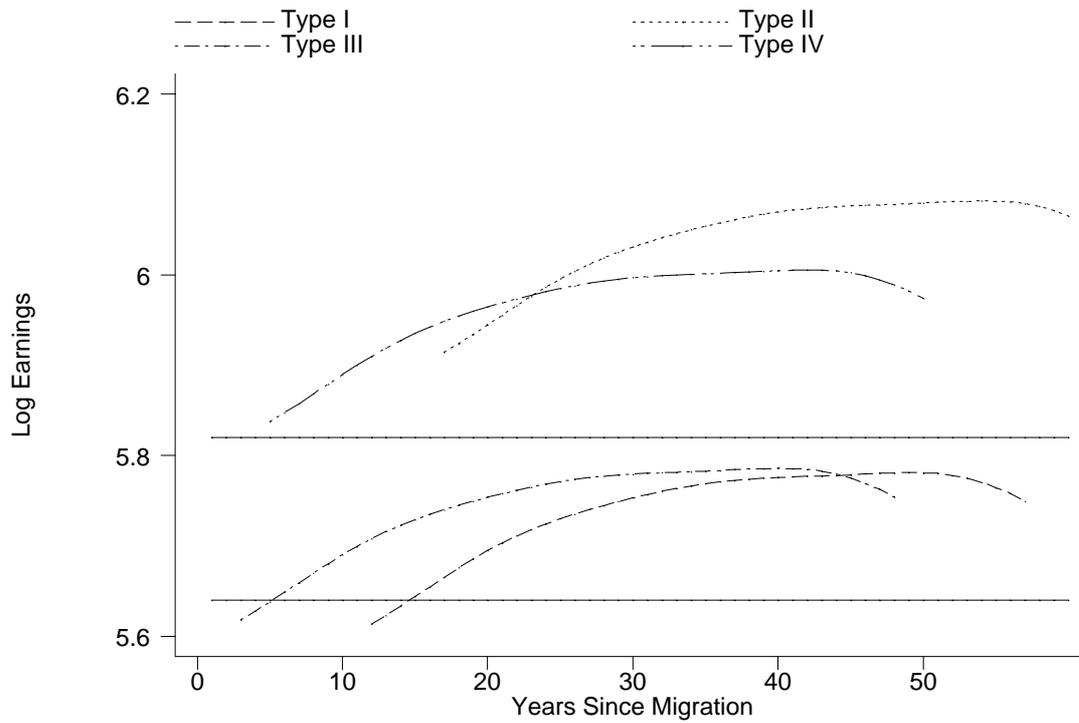


Figure 4. Assimilation: Education Entrants

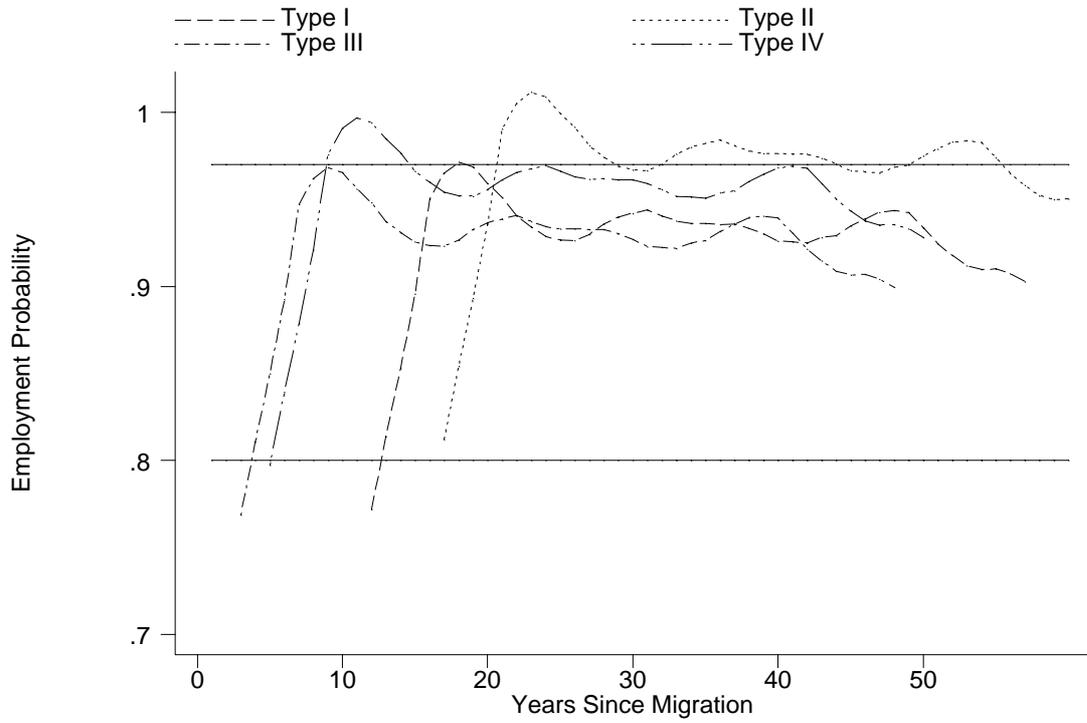
(a) White Earnings



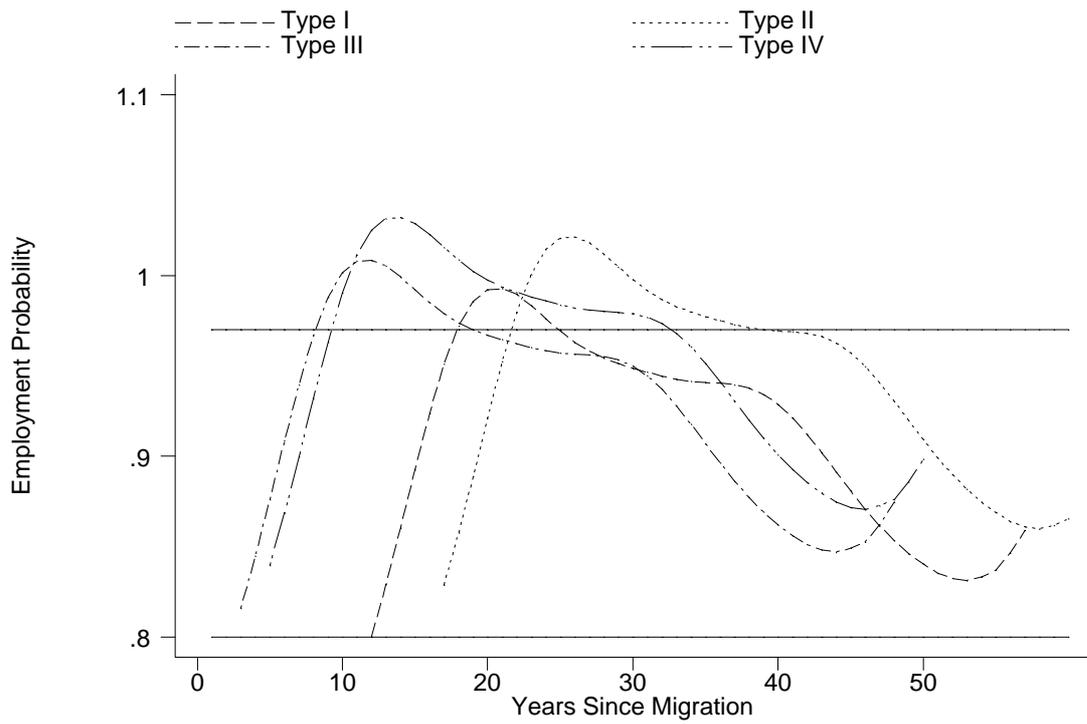
(b) Non-White Earnings



(c) White Employment



(d) Non-White Employment



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