# **DISCUSSION PAPERS IN ECONOMICS**

Working Paper No. 09-03

Offshoring, Immigration, and the Native Wage Distribution

William W. Olney

University of Colorado

revised November 2009 revised August 2009 March 2009

**Department of Economics** 



University of Colorado at Boulder Boulder, Colorado 80309

© March 2009 William W. Olney

## O¤shoring, Immigration, and the Native Wage Distribution

William W. Olney<sup>1</sup>

Job Market Paper

September 2009

#### **Abstract**

While workers in developed countries have become increasingly concerned about the impact oxshoring and immigration have on their wages, the available evidence remains mixed. This paper presents a simple model that examines the impact of oxshoring and immigration on wages and tests these predictions using U.S. state-industry level data. According to the model, the productivity exect causes oxshoring to have a more positive impact on low-skilled wages than immigration, but this gap decreases with the workers' skill level. The empirical results con..rm these predictions and thus provide the ..rst evidence of the productivity exect. Furthermore, the impact of oxshoring and immigration on wages dixers depending on the income level of the foreign country, which may explain the mixed results in the literature.

Keywords: o\(\text{pshoring}\), outsourcing, immigration, productivity e\(\text{pect}\), native wages

JEL Codes: F16, F22, J3

# 1 Introduction

try wage di¤erences are captured by the immigrants rather than the domestic ..rms. Thus, comparing the impact of o¤shoring and immigration on the wages of native workers o¤ers a unique opportunity to test for the presence of the productivity e¤ect. Speci..cally, due to the productivity e¤ect, o¤shoring has a more positive impact on low-skilled wages than immigration (Proposition 1), but this gap decreases with the workers' skill level (Proposition 2).

The predicted impact of immigration and oxshoring on the wages of dixerent types of native workers is then tested using a comprehensive U.S. state-industry level dataset. Using state-industry level data is appealing because it introduces a substantial amount of variation, it mitigates many of the mobility concerns associated with city or county level analyses, and it controls for compositional industry adjustments. The results con..rm both predictions of the model. Oxshoring has a positive exect on the wages of low-skilled workers while immigration has a slight negative exect on these wages. However, the impact of oxshoring and immigration on wages converges as the workers' skill level increases.

O¤shoring and immigration are then grouped according to the income level of the foreign country. This focuses attention on the types of o¤shoring and immigration that are best captured by the model, speci...cally the o¤shoring of low-skilled tasks to less-developed countries and the immigration of less-skilled workers from less-developed countries. The results again con...rm both predictions of the model and provide even stronger empirical support for the productivity e¤ect. Again, due to the productivity e¤ect, o¤shoring has a more positive e¤ect on the wages of low-skilled workers than immigration, but as the workers' skill level increases, the e¤ect of o¤shoring and immigration on native wages becomes more similar.

While not the focal point of the model, oxshoring to developed countries and immigration from developed countries are also included in the empirical analysis for comparison purposes. Interestingly, oxshoring to developed countries decreases and

pared. Con‡icting results in the literature typically arise from papers using di¤erent estimation strategies, unit of analyses, or data. However, this paper shows that o¤-

Hansberg (2008), I model o\(\text{pshoring}\) as trade in tasks. The productivity e\(\text{pect}\) arises

O $\bar{z}$ shoring L-tasks to the foreign country and immigration of L-workers to the home state are possible, while the o $\bar{z}$ shoring of H-tasks and the immigration of H-workers are negligible.

where s represents the high-skilled wage and a and a are functions of the relative average costs of the two sets of tasks. The ..rst term on the right-hand side represents the costs paid to domestic low-skilled workers since (1 J) tasks are performed at home with a low-skilled labor needed for each task. The second term on the right-hand side represents the costs of hiring foreign low-skilled workers. Since the costs vary across each task, I integrate from 0 to J. The third term is the costs of hiring native high-skilled workers.

Substituting (1) into (2) yields the following zero-pro...t condition:

(3) 
$$P = (J)wa (w=s) + sa (w=s),$$

where

$$(J) = 1 \quad J + \left( \int_{0}^{J} g(j) dj \right) = g(J).$$

Here the dependence of the factor intensities a and a on the relative average costs is explicitly stated. If J=0, then no tasks are oxshored, (J)=1, and the zero-pro...t condition is of the standard form. Since g'(j)>0, by the ordering of tasks, it can be shown that (J)<1 as long as J>0. Therefore, the costs to the ..rm after oxshoring some tasks are less than if they chose to perform all L-tasks domestically. Finally, an increase in the share of low-skilled tasks that are oxshored (dJ>0) leads to a decrease in ..rms' costs (d-(J)<0). Oxshoring leads to a reduction in ..rms' costs through the extensive margin because more tasks are oxshored and through the intensive margin because it is now cheaper to oxshore the tasks already produced abroad.

 $<sup>\</sup>frac{6 \, \underline{@} \Omega}{\underline{@} J} = \frac{Q}{Q(J)^2} \frac{g(j) \, dj}{g(J)^2} g'(J)$  which is negative when J > 0:

Domestic ...rms reduce their costs by optimally choosing the tasks to o ${\tt mashore}.$ 

supply which consists of native and immigrant workers.

Using the zero pro...t condition and the market clearing conditions, we can examine how an increase in o\(\text{nshoring}\) or an increase in immigration a\(\text{qects}\) domestic wages. Totally di\(\text{qerentiating}\) equation (3), assuming that P is the numeraire, yields\(^8\)

(6) 
$$(\% + ^) + (1)$$

side of (8) is the labor-supply exect of oxshoring. As the cost of oxshoring decreases (d < 0), more L-tasks are oxshored (dJ > 0), and thus some low-skilled workers become unemployed. Due to excess supply, the wage of low-skilled workers declines. Together the ..rst and second terms of equation (8) represent the impact of oxshoring on the wages of low-skilled workers in this model. The third term on the right-hand side of (8) is the labor-supply exect of immigration. The excess supply of low-skilled workers due to immigration reduces the low-skilled wage. From equation (8), the following proposition is immediate:

Proposition 1 Due to the productivity exect, oxshoring has a more positive impact on the wages of low-skilled workers than immigration.

While both oxshoring and immigration generate a labor-supply exect, oxshoring also generates a productivity exect that increases the wages of low-skilled workers. If the productivity exect exceeds the labor-supply exect, then oxshoring will increase the wages of low-skilled workers. Thus, this model generates the seemingly counterintuitive result that oxshoring can bene...t the factor whose tasks are being sent abroad. Immigration, on the other hand, unambiguously decreases the wages of low-skilled labor in this model. Immigration does not generate a productivity exect because the bene...ts of country wage dixerences are captured by the immigrants rather than the domestic ..rm. Unlike oxshoring, immigration does not generate any direct costs savings for domestic ..rms since they pay immigrants and native workers the same market wage.

Using (6) and (7), it is also possible to derive the percent change in the wage of high-skilled workers as a function of changes in oxshoring and immigration:

(9) 
$$\hat{S} = -\frac{dJ}{(1 - J)} + -\frac{dI}{(1 + I)}.$$

Here the labor-supply exect of oxshoring and immigration increases the wages of high-

skilled workers. As is common in a two factor model, an increase in the exective supply of low-skilled labor increases the marginal product and wages of high-skilled workers. Oxshoring does not generate a productivity exect for high-skilled workers because a decrease in the costs of oxshoring (d < 0) reduces the ..rms' costs of performing L-tasks with no direct exect on the costs of performing H-tasks. Thus, oxshoring does not directly impact the productivity of high-skilled workers. Comparing equations (8) and (9) establishes the following proposition:

Proposition 2 Due to the productivity exect, the impact of ox shoring and immigration on wages becomes more similar as the workers' skill level increases.

The labor-supply exects generated by oxshoring and immigration have a negative impact on low-skilled wages and a positive impact on high-skilled wages. However, the productivity exect generated by oxshoring only impacts low-skilled wages since oxshoring axects the costs of performing L-tasks but not H-tasks. Thus, oxshoring and immigration dixer in their impact on low-skilled wages but have a similar impact on high-skilled wages.

# 3 Estimation Strategy

The propositions generated by the model oxer two unique, testable predictions for the productivity exect. Oxeshoring will have a more positive impact on low-skilled wages than immigration (Proposition 1), but this gap decreases with the workers' skill level (Proposition 2). The empirical analysis that follows will test these predictions by estimating the ilwd[b)38(y)-354611(s)8(,)-327(o)10(x)104(s)8(h)11(o)11(r)11(i5295(T)15(h)4611

workers) that are envisioned in the model. Focusing on oxshoring to less-developed countries (i.e. L-tasks) and immigration from less-developed countries (i.e. L-workers) provides a good proxy for these components of interest. Thus, the following equation will be estimated:

(11) W = 
$$_{0}$$
 +  $_{1}$ Off\_lessdev +  $_{2}$ Off\_dev +  $_{3}$ Img\_lessdev +  $_{4}$ Img\_dev +  $_{5}$ X + + + + .

Again the model predicts that  $_1 > _3$  for low wage deciles but that the dixerence between  $_1$  and  $_3$  decreases as the native wage deciles increase.

O¤shoring to less-developed countries takes advantage of low foreign wages by relocating particular low-skilled tasks abroad. This is the type of o¤shoring that is envisioned in the model and entails di¤erent tasks being performed by domestic and foreign low-skilled workers. Since native and foreign workers are complements in the production process, it is more likely that the productivity e¤ect exceeds the labor-supply e¤ect, and thus the impact on low-skilled native wages will be positive. On the other hand, o¤shoring to other developed countries tends to be motivated by the desire to access foreign markets by replicating the production process abroad rather than exporting. While this is not the type of o¤shoring that is discussed in the model, the concepts of the productivity and labor-supply e¤ects are still relevant. This type of o¤shoring consists of similar tasks being performed by domestic and foreign workers. Since foreign workers are substituting for domestic workers, the labor-supply e¤ect likely exceeds the productivity e¤ect, and thus the impact on low-skilled native wages will be negative.9

<sup>&</sup>lt;sup>9</sup>This is consistent with Harrison and McMillan's (2006) ...ndings that vertical foreign a¢ liate employment complements domestic employment whereas horizontal foreign a¢ liate employment substitutes for domestic employment.

Consistent with previous results (Borjas 1995), I ...nd that the skill level of immigrants is strongly correlated with the income level of the foreign source country. Since immigrants from less-developed countries are relatively less skilled, they will compete with less-skilled native workers for jobs. Thus, according to the model, immigration from less-developed countries generates a labor-supply exect that depresses low-skilled wages and increases high-skilled wages. Although the model focuses on less-skilled immigrants, the exects of skilled immigrants from developed countries will be included in the empirical analysis for comparison purposes. If these skilled immigrants bring knowledge and expertise that is not readily available in the domestic labor market, they may raise the wages of all types of native workers.

#### 4 Dre

eign a¢ liates of U.S. ..rms, is obtained from the U.S. Bureau of Economic Analysis (BEA). 12 Given the trade in task model, focusing on foreign a¢ liate employment is preferable to other measures of foreign direct investment such as a¢ liate sales. The BEA provides foreign a¢ liate employment data by year and industry of the foreign a¢ liate. Since o¤shoring data is not available by state, foreign a¢ liate employment is distributed across states based on the share of state GDP to national GDP in that industry. Finally, the share of foreign a¢ liate employment to total employment, including both domestic and foreign employment, is calculated by state, industry, and year. Thus, o¤shoring is de..ned as the following share

offshoring = 
$$\frac{\begin{bmatrix} \frac{P_{-} - sit}{s} & Foreign\_Affiliate\_Empl \end{bmatrix}}{Domestic\_Empl + \begin{bmatrix} \frac{P_{-} - sit}{s} & Foreign\_Affiliate\_Empl \end{bmatrix}} 100,$$

where s indexes states, i indicates industries, and t references years. This measure of oxshoring is consistent with J from the /F2111.955Tf12u9(h)]TJ/Fwd(eiho)11ecasur991(i)6(s)-284(t)8

is preferable to a cross country analysis where it is di¢ cult to control for unobserved factors. Since U.S. states share similar laws, institutions, and cultural characteristics, using states as the unit of analysis limits these confounding factors. Together with the variation in o\(\text{\text{m}}\)shoring and immigration across states (Table 1), this means that the link between these forms of globalization and wages is more easily identi..ed. In addition, state level data mitigates many of the mobility concerns associated with a city or county level study. Thus, states more closely resemble a closed labor market while still o\(\text{\text{m}}\)ering a substantial amount of variation.

Second, this analysis incorporates 14 2-digit NAICS industries which range from manufacturing to professional services to ...nance (Table 2). Due to data constraints, many previous studies focus just on manufacturing industries (Feenstra and Hanson 1999, Harrison and McMillan 2006, Amiti and Wei 2009). However, manufacturing represents only 13% of total U.S. GDP in 2008.<sup>14</sup> Unlike these previous studies which focus on a small component of the U.S. economy, this analysis examines how oxshoring and immigration axect wages in a wide variety of industries. Furthermore, by focusing on highly aggregated NAICS industries, mobility across industries is less problematic.

Incorporating 14 industries into this analysis not only provides an additional source of variation but it also controls for the compositional mix of industries within states. It is possible that an in‡ux of immigrants or an increase in o¤shoring could lead to a change in industry composition within a state. Speci..cally, a labor supply shock can be fully absorbed through a change in industry mix without any change in factor returns. By using a state-industry-year unit of observation, this analysis controls for the changing compositional mix of industries within states. Finally, the years included in this analysis span exogenous shocks to both o¤shoring and immigration caused by China joining the World Trade Organization in 2001 and changes

<sup>&</sup>lt;sup>14</sup>Gross Domestic Product by Industry Accounts (BEA).

to immigration policy following 9/11.

Table 1 presents the median wage, immigration, and oxshoring by state. While the state ..xed exects will capture much of this variation, Table 1 provides insight into the states that are most susceptible to oxshoring and immigration. There is substantial variation across states, with the median wage ranging from \$23,721 in Montana to \$41,595 in Connecticut, immigration ‡uctuating from 1.6% in West Virginia to 34.3% in California, and oxshoring varying from 3.2% in Montana to 9.0% in Indiana. Figure 1 plots average immigration and oxshoring by state. Not surprisingly, the urban coastal states of California, New York, and New Jersey have high shares of oxshoring and immigration while the rural isolated states such as Montana and North Dakota have low shares of both. Florida and Nevada have high shares of immigration

on state, industry, and year ..xed exects. The residuals from these regressions will be

brackets. We see that globalization leads to an increase in wages of all types of native workers, thus contradicting many of the fears of American workers. A protectionist policy that limited oxshoring, immigration, and inshoring would unambiguously decrease the wages of native workers. While Table 3 demonstrates that these forms of globalization, on the whole, bene...t native workers, the model predicts that oxshoring and immigration should dixer in their impact on the wages of native workers. Next, the aggregate exect of oxshoring and immigration on native wages is examined, while the subsequent section focuses on the types of oxshoring and immigration that are most similar to those considered in the model.

importance of controlling for the income level of the foreign country.

#### results in Table 4.

While the model focuses on the oxshoring of low-skilled tasks and the immigration of low-skilled workers, I include oxshoring to developed countries and immigration from developed countries in the regressions in Table 5 for comparison purposes. Oxshoring to other developed countries entails replicating the production process abroad in order to access foreign markets and avoid transport costs. This results in foreign workers substituting for domestic labor and explains the negative coe¢ cients on

results in Table 6 are consistent in sign, magnitude, and signi...cance level to those reported in the baseline results in Table 5.

Second, local wages are unlikely to be a driving force in the state location decision of immigrants. Non-economic factors such as family and friends, distance from home country, and weather are typically found to be important determinants of immigrant location decisions. The migration of residents in response to wages is more problematic at a more ...nely disaggregated geographic level (i.e. cities or counties) or across more ...nely disaggregated industries (i.e. 6-digit NAICS). However, for the sake of argument, suppose immigrants did choose states and industries solely based on which paid a relatively higher wage. Then there would be a spurious positive correlation between immigration and wages. The fact that the Immigration (Less Dev) coet cients in Table 5 are signi...cantly negative implies that either this positive bias is negligible or the impact of immigration on domestic wages is even more negative

rather than a productivity exect as this paper proposes.

To address these concerns, I include the average educational attainment of the native population as a control in all the regressions presented in this paper. This will capture changes in the average skill level of native employees and thus any compositional shifts in employment will be controlled. The results indicate that native educational attainment is an important control variable. However, there is still an important relationship between oxshoring, immigration, and wages which is not driven

# 7 Conclusion

Americans have become increasingly concerned about the impact oxshoring and immi-

certain components of oxshoring and immigration can depress the wages of speci..c types of native workers. Policy makers, whose goal is to increase the wages of native

# References

- Amiti, Mary and Shang-Jin Wei. 2009. "Service O¤shoring and Productivity: Evidence from the US." World Economy, 32(2): 203-20.
- Antràs, Pol. 2003. "Firms, Contracts, and Trade Structure." Quarterly Journal of Economics, 118(4): 1375-418.
- Bartel, Ann P. 1989. "Where Do the New U.S. Immigrants Live?" Journal of Labor Economics, 7(4): 371-91.

Feenstra, Robert C. and Gordon H. Hanson. 1999. "The Impact of Outsourcing and High-Technology Capital on Wages: Estimates for the United States, 1979-1990." The Quarterly Journal of Economics

Public Use Microdata Series: Version 4.0. Minneapolis, MN: Minnesota Population Center.

Slaughter, Matthew J. 2000. "Production Transfer within Multinational Enterprises and American Wages." Journal of International Economics, 50(2): 449-72.

# FIGURE 1 IMMIGRATION AND OFFSHORING BY STATE

State average of the share of employees that are foreign born and the share of employees that work abroad weighted by the sample size.

TABLE 2 INDUSTRY AVERAGES

Industry	Median Wage	Immigration	Offshoring
Agriculture, Forestry, Fishing, Hunting, and Mining	\$31,256	16.5	4.7
Utilities	\$48,742	5.4	9.4
Construction	\$33,957	14.9	0.3
Manufacturing	\$38,097	14.2	21.2
Wholesale Trade	\$36,740	12.6	10.6
Retail Trade	\$24,030	10.7	3.4
Transportation and Warehousing	\$37,735	11.4	2.6
Information	\$41,728	10.2	7.9
Finance and Insurance	\$38,889	9.6	3.5
Real Estate, Rental, and Leasing	\$31,663	12.6	0.9
Professional, Scientific, Techinical Services and Management	\$46,766	12.8	3.6
Administration and Waste Services	\$24,730	18.4	4.9
Health Care and Social Assistance	\$28,324	11.6	0.1
Accomodations and Food Services	\$15,433	22.7	3.8

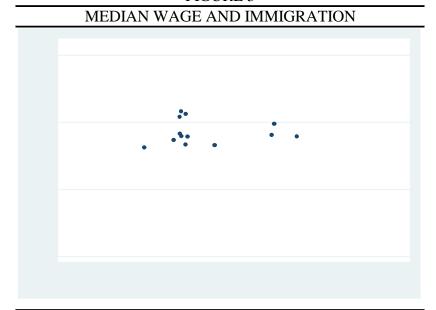
Industry average of the median native wage, the share of employees that are foreign born, and the share of employees that work abroad weighted by the sample size.

FIGURE 2



The residuals from regressing the ln native median wage on state, industry, and year fixed effects are plotted against the residuals from regressing offshoring on state, industry, and year fixed effects.

FIGURE 3



The residuals from regressing the ln median native wage on state, industry, and year fixed effects are plotted against the residuals from regressing immigration on state, industry, and year fixed effects.

	In(Wage 10th%)	In(Wage 20th%)	In(Wage 30th%)	In(Wage 40th%)	ln(Wage 50th%)	ln(Wage 60th%)	In(Wage 70th%)	In(Wage 80th%)
Globalization	0.002***	0.002***	0.001***	0.001***	0.002***	0.002***	0.002***	0.002***
	[0.001]	[0.001]	[0.000]	[0.000]	[0.000]	[0.000]	[0.000]	[0.000]
Age	0.004	0.004	0.004	0.003	0.002	0.002	0.003	0.003
	[0.004]	[0.003]	[0.002]	[0.002]	[0.002]	[0.002]	[0.002]	[0.002]
Education	0.166***	0.170***	0.180***	0.187***	0.198***	0.207***	0.216***	0.237***
	[0.013]	[0.009]	[0.008]	[0.007]	[0.007]	[0.007]	[0.007]	[0.011]
Male	0.010***	0.010***	***600.0	0.010***	0.010***	0.010***	0.010***	0.010***
	[0.001]	[0.001]	[0.000]	[0.000]	[0.000]	[0.000]	[0.001]	[0.001]
Black	0.000	0.001*	0.002***	0.002***	0.002***	0.002***	0.001**	0.000
	[0.001]	[0.001]	[0.000]	[0.000]	[0.000]	[0.000]	[0.000]	[0.000]
Asian	0.014**	0.022***	0.017***	0.017***	0.016***	0.019***	0.020***	0.021***
	[0.006]	[0.004]	[0.004]	[0.004]	[0.004]	[0.004]	[0.004]	[0.005]
Hispanic	***900.0-	-0.005***	-0.004***	-0.004***	-0.004***	-0.004***	-0.004***	-0.003***
	[0.002]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]
Married	0.000	0.000	0.000	0.001	0.001**	0.002**	0.002***	0.002**
	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]
Single	****00.0-	***900'0-	-0.005***	-0.005***	-0.004***	-0.004***	-0.003***	-0.002**
	[0.002]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]	[0.001]
Observations	4032	4032	4032	4032	4032	4032	4032	4032
R-squared	0.91	0.95	96.0	0.96	96:0	96.0	0.96	0.95

		-
		_
		_
		-

	ln(Wage 10th%)	ln(Wage 20th%)	In(Wage 30th%)	ln(Wage 40th%)	ln(Wage 50th%)	ln(Wage 60th%)	In(Wage 70th%)	ln(Wage 80th%)
Offshoring (Less Dev)	0.041***	0.033***	0.025***	0.019***	0.017***	0.014***	0.009**	0.002
	[0.007]	[0.005]	[0.004]	[0.004]	[0.004]	[0.004]	[0.004]	[0.005]
Offshoring (Dev)	-0.025***	-0.019***	-0.015***	-0.010***	-0.009***	-0.005*	-0.001	0.003
	[0.006]	[0.004]	[0.003]	[0.003]	[0.003]	[0.003]	[0.003]	[0.004]
Immigratio[0.005]								

	In(Wage 10th%)	) In(Wage 20th%)	In(Wage 30th%)	In(Wage 40th%)		In(Wage 60th%)	In(Wage 70th%)	In(Wage 80th%)	
Offshoring (Less Dev)	0.039***	0.032***	0.027***	0.021***		0.016***	0.007	0.006	
	[0.009]	[0.007]	[0.006]	[0.005]	[0.005]	[0.005]	[0.005] [0.005]	[0.005]	
Offshoring (Dev.)	***CCU U-	***000-	-0.015***	-0.011***		-0.006	0000	-0.001	

## A Model Appendix

#### A.1 Deriving Equation (6):

Total dimerentiating equation (3), assuming that P is the numeraire, yields:

$$0 = d wa + dw a + da w + dsa + da s$$

or:

$$0 = ^{\wedge} + \hat{w} + \hat{a} + \hat{s} + \hat{a}$$

where and are the cost shares of low-skilled and high-skilled labor (and + = 1). Since pro...t maximizing ..rms have minimized costs,  $\hat{a} = 0$  by the envelope theorem. Thus:

(6) 
$$0 = (^ + \hat{w}) + (1)$$

### A.2 Deriving Equation (7):

Totally dimerentiating the ratio of (4) to (5) gives:

or:

$$\frac{\bot}{H}$$
 (â â)  $\left(\frac{\Omega}{-\Omega}\right)\left(\hat{W} + \hat{S}\right) = \frac{(1+)}{(1-)}\left(\hat{L} + \frac{1}{(1+)} + \hat{H} + \frac{1}{(1-)}\right)$ 

The ..rst terms on each side cancel following from the ratio of (4) to (5) and since the native factor supplies are ..xed then  $\hat{L} = \hat{H} = 0$ . Therefore:

(â â) 
$$\left(\frac{\Omega}{}\right)$$
  $\left(\$$   $\$$   $\$$   $\$$   $\$$   $\$$   $\$$ 

or:

(7) 
$$\left(\$ \quad \mathring{\mathbb{Q}} \quad ^{\wedge}\right) = \frac{1}{(1+1)} + \frac{1}{(1-1)}$$

where the elasticity of substitution is de..ned as:

$$=\frac{\frac{a_H}{a_L} \frac{a_H}{a_L}}{\left(\frac{w}{s}\right)\left(\frac{w}{s}\right)} = \frac{(\hat{H} - \hat{L})(\Omega)(\hat{H} - \hat{\Omega} - \hat{M})}{(\hat{H} - \hat{L})} = (\hat{A} \hat{A})(\hat{W} = \hat{S})$$

## A.3 Deriving Equation (8):

Rearranging equation (7) as follows:

$$\$ = \frac{1}{(1-)} + \frac{1}{(1+)} + \$ + \$ + \$$$

and plugging this into equation (6) yields:

$$(\hat{W} + \hat{D}) + (\hat{D} + \hat{D}) = 0$$

or:

(8) 
$$\hat{W} = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \end{pmatrix}$$

### A.4 Deriving Equation (9):

Rearranging equation (7) as follows:

$$\hat{W} = \frac{1}{(1-1)} + \hat{S}^{-1}$$

and plugging this into equation (6) yields:

$$\left[\begin{array}{cc} \frac{1}{(1-)} & \frac{1}{(1+)} + \$ \right] + (1) & )\$ = 0$$

or:

(9) 
$$\$ = \frac{\bot}{(1-)} + \frac{\bot}{(1+)}$$

# B Data Appendix

#### B.1 Data Sources

Individual level data was obtained from the 2000 1% Census sample and the 2001-2005 American Community Survey (ACS) via IPUMS. The 2000 1% sample was preferable to the 2000 ACS because it was approximately seven times the size (the 2000 ACS) and the 2000 ACS because it was approximately seven times the size (the 2000 ACS).

observations by state in these industries. Thus, the analysis includes 14 NAICS industries. Finally, available Census and BEA data restricts the sample to the years 2000-2005.

#### B.3 De..nition of Developed

The countries with the highest 2006 GDP per capita according to the World Development Indicators database (World Bank, April 11, 2008) were Canada, Denmark, Finland, Iceland, Norway, Sweden, UK, Ireland, Belgium, France, Luxembourg, Netherlands, Switzerland, Italy, Austria, Germany, Japan, and Australia (not including San Marino or the U.S.). Immigrants that were born in these 18 countries were assigned to the Developed group, while those immigrants born in the remaining countries were assigned to the Less-Developed group. O¤shoring to developed countries includes foreign a¢ liate employment in Europe, Canada, Australia, and Japan, while o¤shoring to less-developed countries consists of the remaining foreign a¢ liate employment. Unfortunately data limitations do not allow "Europe" to be broken into individual countries that correspond to those included in the immigrant de..nition. However, of the total foreign a¢ liate employment in Europe, 85% is going to the 14 European countries included in the immigrant Developed group.

### B.4 Missing Values

Due to con...dentiality concerns, the BEA withholds some industry-country speci...c foreign a¢ liate employment numbers. There are no missing values for total foreign af...liate employment, but when constructing o¤shoring to developed and less-developed countries, this issue needs to be addressed. Data for these 18 missing values are ...lled with the industry-country average across years. The majority of the time this average falls within the employment range indicated by the BEA for that employment cell; when it does not, I replace the missing value with the midpoint of this range instead.