## Computerization and Immigration: Theory and Evidence from the United States <sup>1</sup>

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## Technology and Polarization

• Labor market polarization likely due to routine-substituting technological innovation (Autor et al.,

2003; Autor & Dorn, 2013)



Changes Employment (Natives+Foreign Born) by Skill Percentile, 1980-2010

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Low and high-educated immigrants as share of the population  $$D_{ata:}$$  IPUMS 1980-2010

# The goal of this paper is to answer **two** simple, yet extremely relevant, **questions** we know little about

- Does technological growth attract migrants?
  - Is it true for both high and low educated (*skilled*)?
- ② Does immigration attenuate or exacerbate the tendency of native job polarization?
  - Can **cross-regional variation** in technology adoption inform us on natives and immigrants job polarization?
  - How the combination of technology and immigration **impact** (native) welfare in the long-run?

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  - How the combination of technology and immigration **impact** (native) welfare in the long-run?

We provide empirical evidence and theoretical support to understand the following:

- In the data, is technology adoption (computer use on the job) associated with immigration inflows (and polarization)?
- We then rationalize these facts in a simple GE model with 3 tasks, exogenous routine-substituting technological change and endogenous immigration
- Finally, we simulate the model equilibrium to provide counterfactual scenarios and back out welfare for natives

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Changes in Foreign-born and Natives' Employment by Skill Percentile 1980-2010

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- immigrant inflows **are associated** with routine-substituting technology adoption
- job polarization **at the low-end** can be mainly attributed to immigrants
- The model simulations predict that **immigration**:
  - contributes to technological progress
  - combined with technology adoption, **induces occupational upgrading**
  - is net welfare enhancing for natives

#### 1 Introduction

- 2 Contributions to the literature
- Immigration and Technology Shocks: Definition and Identification
- 4 Empirical Results
- **5** Model and Simulations

#### 6 Conclusions

## This paper contributes to

An extensive literature on **polarization** and **routine-substituting** (Autor et al., 2003; Goos & Manning, 2007; Autor & Dorn, 2013)

#### Labor supply matters too (Cerina et al., 2017):

- Low-end polarization is mitigated by undocumented migrants (Mandelman & Zlate, 2014)
- We extend AD's framework to endogenous immigration

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Even more literature on **immigration** (Card, 2001; Peri & Sparber, 2009; Ottaviano & Peri, 2012; Dustmann & al., 2015; Lull, 2017) We are the first to **show** that:

- Areas with technological progress attracts low-skilled migrants (Cadena & Kovak, 2016; Jaimovich & Siu, 2017: high-skilled ↑ only)
- *Absent* immigration, polarization, capital accumulation and growth would change

# Technology and immigration

#### 722 self-contained local labor markets: Commuting Zones (CZs)

Change in routine-substituting technology (proxy: PC use), 1980-2010



Change in foreign-born share, 1980-2010



# Identification of technology adoption (I)

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- AD proxy technological change with task-based routine-intensity of CZs

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- $\bullet\,$  Beaudry et al. (2010) use survey firm-level computer adoption for  $\sim\,$  200 city
- *AD* proxy technological change with task-based routine-intensity of CZs

We use an inferred measure of PC adoption for all 722 CZs ( $\sim$  to Autor et al., 2003):

- Industry-level PC use from the CPS as of mid-2000s
- We exploit variation in 1980 local labor markets industrial composition

## Identification of technology adoption (II)

PC use<sub>c,t</sub> = 
$$\sum_{j} \omega_{j,c,1980} * \Delta \frac{\text{PC at work}_{j,US,t}}{empl_{j,US,1980}}$$

where:

• 
$$\Delta \frac{\text{PC at work}_{j,US,1980}}{empl_{j,US,1980}} = \frac{\text{PC at work}_{j,US,2005}}{empl_{j,US,1980}} - \underbrace{\frac{\text{PC at work}_{j,US,1980}}{empl_{j,US,1980}}}_{\sim 0}$$
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$$\Delta \frac{\text{PC at work}_{j,US,t}}{empl_{j,US,1980}} \sim \frac{\text{PC at work}_{j,US,2005}}{empl_{j,US,1980}}$$
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c: CZ; t: survey year; j: industry

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*c*: CZ; *t*: survey year; *j*: industry

We control for generic labor demand shocks (Bartik-style proxy):

• Labor Productivity<sub>c,t</sub> =  $\sum_{j} \omega_{j,c,1980} * \Delta \log(wage)_{j,-c,t}$ 

## Identification of technology adoption (III)

 PC use proxy positively correlates with other measure of RBTC



# US and foreign born migration and PC adoption

 $\frac{\Delta Pop_{c,h,t}}{Pop_{c,1980}} = \alpha + \beta \Delta PC \text{ use}_{c,t} + \gamma \Delta Labor \text{ Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,h,t} \quad (1)$ 

for each skill h, CZ c between 1980 and 2010.

US and foreign born employment

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	Low Skilled	High Skilled			
Panel A: Natives					
PC use	0.219	2.419**			
	(0.314)	(0.641)			
Labor Productivity	-0.790	-4.008*			
	(0.893)	(1.913)			
Obs.	722	722			
R2	0.64	0.55			
Panel B: Foreign Born					
PC use	0.555+	1.038**			
	(0.299)	(0.210)			
Labor Productivity	0.187	$1.028^{+}$			
	(0.562)	(0.556)			
Obs.	722	722			
R2	0.67	0.79			

Note: 722 CZs, 1980-2010. Standard errors (in parentheses) are clustered at the state level. \*\*, \*, + indicate

significance at 1-percent, 5-percent and 10-percent level, respectively.

US and foreign born employme

14/33

Computerization & Immigration

## US and foreign born occupational share and PC adoption

 $\Delta EmplSh_{c,t}^{k} = \alpha + \beta \Delta PC \text{ use}_{c,t} + \gamma \Delta Labor \text{ Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,t}^{k} \quad (2)$ 

for each CZ c, occupation/task group k between 1980 and 2010.

Summary Stats

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	$Manag/Prof\ \overline{Occ}$	Cler/Ret/Prod Occ	Serv/Trans Occ
Panel A: Natives			
PC use	0.558**	-0.646**	0.088
	(0.063)	(0.089)	(0.076)
Labor Productivity	-0.099	0.244	-0.146
	(0.164)	(0.273)	(0.241)
Obs.	722	722	722
R2	0.60	0.73	0.46
Panel B: Foreign Bo	orn		
PC use	0.595*	-1.036**	$0.441^{+}$
	(0.252)	(0.175)	(0.257)
Labor Productivity	-1.620	0.394	1.226
	(0.996)	(0.441)	(0.929)
Obs.	722	722	722
P2	0.51	0.41	0.43

for each CZ c, occupation/task group k between 1980 and 2010.

Note: 722 CZs, 1980-2010. Standard errors (in parentheses) are clustered at the state level. **\*\***, **\***, **+** indicate significance at 1-percent, 5-percent and 10-percent level, respectively.

15/33

Computerization & Immigration

- Similar results using task specialization indexes Tasks
- Additional results on US wages Wages
- Preliminary IV results exploiting early 'PC-adopters' CZs produce consistent results (forthcoming)
- **Pre-trends** indicates no patterns in group-specific migration Migration Pre-Trends

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- Immigrants inflows are associated with PC adoption
   It holds both for low-skilled (*new results*) and high-skilled
- PC adoption also correlated with natives' job polarization
   Immigrants contribute to low-end polarization
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Two sectors, goods and services are complementary in utility:

- CES utility
  - $U = (\rho C_s^{\frac{\sigma-1}{\sigma}} + (1-\rho) C_g^{\frac{\sigma-1}{\sigma}})^{\frac{\sigma}{\sigma-1}}$ , with  $\sigma \in (0,1]$
- Goods can be saved to accumulate capital and human capital

•  $C_g = Y_g - p_k K - p_a L_a$ ;  $C_s = Y_s$ 

- Service production linear ( $Y_s = L_s$ ). Goods production:
  - **O Complementarity between** K and L<sub>a</sub>

• 
$$Y_g = \left[ (\alpha_a L_a)^\beta + X^\beta \right]^{1/\beta}, \ \beta < 0, \ \alpha_a > 1$$

Substitution between K and L<sub>r</sub>

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#### Labor amounts and migration (I)

- Labor ability is ranked: **manual** simpler than **routine**, simpler than **analytical**
- Each worker *i* has  $\eta_i$  amount of routine ability (manual ability stand'd to 1)
- Workers can upgrade to analytical ability (φη<sub>i</sub>, φ > 1) at cost p<sub>a</sub>
- Two thresholds: Equilibrium wages makes workers indifferent between manual and routine and routine and analytical

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$$w_r\eta^* = w_s,$$

$$w_a \phi \hat{\eta} - p_a = w_r \hat{\eta}$$

**Unskilled migration** positively depends on low-skill manual wages (Grogger & Hanson, 2011)

• 
$$mig = \begin{cases} (1+w_s)^{\epsilon} - (k+p_s) & \text{if } (1+w_s)^{\epsilon} - k > p_s \\ 0 & \text{otherwise} \end{cases}$$

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Assuming 
$$\eta \sim f(\eta) = e^{-\eta}$$
, labor amounts are:

• 
$$L_r = \int_{\eta^*}^{\hat{\eta}} \eta e^{-\eta} d\eta$$
  
•  $L_a = \int_{\hat{\eta}}^{\infty} \phi \eta e^{-\eta} d\eta$   
•  $L_m = 1 + \operatorname{mig} - e^{-\eta^*}$ 

#### Partial Equilibrium Intuition

Supply of low-educated migration and service wages: No migration, no tech ( $\varepsilon^s$  given)



#### Partial Equilibrium Intuition

Supply of low-educated migration and service wages:  $\downarrow p_k \ (\epsilon^s \text{ given})$ 



#### Partial Equilibrium Intuition

Supply of low-educated migration and service wages:  $\downarrow p_k$  & demand effect ( $\varepsilon^s$  given)



We depart from Autor and Dorn (2013) in three ways:

- Native workers can upgrade their skills and occupation (by accumulating human capital endogenously):
- **2** Endogenous migration in response to  $p_k \downarrow$
- $\Rightarrow$  We simulate the model to evaluate **counterfactual** scenarios

### We set the parameters as to **match initial labor shares** and **low-skilled migration inflows** in the last 30 years:

- Elasticity of substitution in *production* higher than that in *consumption*  $(\frac{1}{1-\gamma} > \sigma)$ 
  - $\sigma = 0.5$ ,  $\beta = -10$ ,  $\gamma = 0.5$
  - $\rho = 0.025, \ \alpha_a = 7.5$
- Other parameters
  - $\phi = 2$ ,  $p_m = 0.25$  (simulation with  $p_a$  ongoing work)
  - $\epsilon_s = 0.2$
  - ⇒ We simulate the model for a **225% exogenous decline** of  $p_k$  cumulated over 30 years

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#### Model simulations (figures)

1) Technological progress without migration generates employment polarization:

Changes in Native Employment Levels from Higher Computerization



#### Model simulations (figures)

2) Technological progress attracts low-skilled migrants:

Changes in Unskilled Migrants from Higher Computerization



#### Model simulations (figures)

3) Immigration allows natives to upgrade their skills:



3

#### Model simulations (table)

We *target* a 15 p.p. increase in migration, initial occupational shares ( $L_a = 0.27$ ,  $L_r = 0.35$ ,  $L_m = 0.38$ ) and a 225 percent  $\downarrow$  in computer price

Variables	W/ Baseline	No	Driving
	Migration	Migration	Channel w/Migration
$\% \Delta Population$	14.4		
$\Delta L_{manual}^{natives}$	-1.8		
$\Delta L_{routine}^{natives}$	-20.2		
$\Delta L_{analytical}^{natives}$	21.9		
$\Delta W_s$	71.0		
$\%\Delta W_r$	-47.8		
$\%\Delta W_a$	262.3		
$\%\Delta K$	367.4		
$\%\Delta Y_{goods}$	98.8		

Table. Baseline Simulation

Occupation Shares 1980-2010

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Variables	W/ Baseline	No	Driving
	Migration	Migration	Channel w/Migration
$\% \Delta Population$	14.4	0	$\uparrow w_s$ given $\varepsilon_s > 0$
$\Delta L_{manual}^{natives}$	-1.8	4.6	$\downarrow \hat\eta$ , $\eta^*$
$\Delta L_{routine}^{natives}$	-20.2	-22.9	$\downarrow \eta^*$
$\Delta L_{analytical}^{natives}$	21.9	18.3	$\downarrow \hat{\eta}$
$\%\Delta W_s$	71.0	160.1	↑ mig
$\Delta W_r$	-47.8	-51.5	$\uparrow L_r$
$\Delta W_a$	262.3	236.1	L <sub>a</sub> , K complements
$\Delta K$	367.4	333.6	L <sub>a</sub> , K complements
$\%\Delta Y_{goods}$	98.8	84.5	

G. Basso (BdI)

Table. Baseline Simulation

Occupation Shares 1980-2010

#### Extension: High-skilled migrants

**High-skill migrants** supply analytical tasks depending on the level of **analytical wages**:

$$\eta_{s}\left[\left(1+w_{a}\right)^{\epsilon_{a}}-k\right]\geq p_{m}^{a}$$



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Now, we have two sources of analytical labor, possibly competing:

$$L_{a} = \int_{\hat{\eta}}^{\infty} \phi \eta e^{-\eta} d\eta + \int_{\bar{\eta}}^{\infty} \phi \eta_{s} e^{-\eta} d\eta$$
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#### Skilled migration:

- $\varepsilon_a > 0$ s: drop in  $p_k$  raises skilled migration to  $\sim 7.5$  percent of total native skilled population
- Allows for **more capital** accumulation (through production complementarities)
- Favor unskilled migrants inflows (through demand)
- $\Rightarrow$  Although quantitatively different, main results hold

Graphs

Technological progress attracts high-skilled migrants:

Changes in Unskilled Migrants from Higher Computerization



Natives in manufacturing gain (higher earnings and cheaper services):

• 
$$Util_{manuf} = \left(\frac{\rho^{1+\sigma}w_s^{-\sigma} + (1-\rho)^{1+\sigma}}{(1-\rho)^{\sigma} + \rho^{\sigma}w_s^{1-\sigma}}\right) \left(w_r L_r + w_a L_a\right) \left(L_a + L_r\right).$$

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$$Util_{manuf} = \left(\frac{\rho^{1+\sigma}w_s^{-\sigma} + (1-\rho)^{1+\sigma}}{(1-\rho)^{\sigma} + \rho^{\sigma}w_s^{1-\sigma}}\right) \left(w_r L_r + w_a L_a\right) \left(L_a + L_r\right).$$

Natives who work in services lose as lower wages more than offset cheaper services:

• 
$$Util_{serv} = \left(\frac{\rho^{1+\sigma}w_s^{-\sigma} + (1-\rho)^{1+\sigma}}{(1-\rho)^{\sigma} + \rho^{\sigma}w_s^{1-\sigma}}\right)(w_s L_{s,nat}L_{s,nat}).$$

• What is the net effect of immigration?

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Migration in the U.S. has been net positive for overall welfare:



Changes in Native Aggregate Utility with Both Types of Migration

- Computerization alone raises welfare by 5.4 percent (blue line)
- Computerization and both types of migration raises welfare by 21.8 percent (green line)
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- attracts low-skilled immigrants through higher service wages
  - We document **unskilled migration** response which complements existing work (Moretti, 2013; Cadena & Kovak, 2016): due to an increase demand for manual tasks (Mazzolari & Ragusa, 2013)
  - High-skilled migration response comes at no surprise
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- Further hastens technological progress
- Induces occupational upgrading among natives in the long run: natives join more routine & analytical occupations (i.e., balance back unbalanced growth)
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### Appendix

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# US and foreign born employment and PC adoption

 $\frac{\Delta Empl_{c,h,t}}{Empl_{c,1980}} = \alpha + \beta \Delta PC \text{ use}_{c,t} + \gamma \Delta Labor \text{ Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,h,t} \quad (4)$ 

for each skill h, CZ c between 1980 and 2010.

# US and foreign born employment and PC adoption

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for each skill h, CZ c between 1980 and 2010.

	Low Skilled	High Skilled
Panel A: Natives		
PC use	0.115	2.047**
	(0.304)	(0.635)
Labor Productivity	-0.681	$-3.745^{+}$
	(0.896)	(2.080)
Obs.	722	722
R2	0.59	0.52
Panel B: Foreign B	orn	
PC use	0.416	1.008**
	(0.301)	(0.199)
Labor Productivity	0.247	$1.078^{+}$
	(0.549)	(0.539)
Obs.	722	722
R2	0.66	0.79

Note: 722 CZs, 1980-2010. Standard errors (in parentheses) are clustered at the state level. \*\*, \*, + indicate

significance at 1-percent 5-percent and 10-percent level respectively

G. Basso (BdI)

Computerization & Immigration

# Observed Occupational Share Changes

#### Table. Occupational Employment Shares

	Managers/prof/	Clerical/sales/	Services/transp/
	tech	operators	construct
Prevalently	Analytical/Cognitive	Routine	Manual
Panel A: Natives			
1980	0.276	0.409	0.315
2010	0.405	0.304	0.292
Delta	0.129	-0.106	-0.023
Panel B: Foreign born			
1980	0.241	0.420	0.339
2010	0.294	0.250	0.456
Delta	0.053	-0.169	0.116

Back to Regressions

Back to Simulation

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We construct measures of task supply based on the DOT indexes of **Manual**, **Routine** and **Analytical Task** intensity (Peri & Sparber, 2009; Autor & Dorn, 2013)

	Analytical/	Routine	Manual/
	Cognitive		Communication
Managers/prof/tech	0.807	0.343	0.478
Clerical/sales/operators	0.415	0.664	0.358
Services/construct/transp	0.322	0.451	0.737
Average Specialization	0.493	0.505	0.517
% of Total	32%	34%	34%

Occupations and Task Index in 1980

### Natives and Foreign-born Task Specialization Indexes

	Analytical	Manual	Routine
Panel A. Natives		All	
1980	0.321	0.339	0.340
2010	0.370	0.331	0.299
Delta %	15.26	-2.36	-12.06
Panel B. Foreign-born		All	
1980	0.292	0.353	0.355
2010	0.313	0.367	0.319
Delta %	7.19	3.97	-10.14

## US and foreign born task specialization and PC adoption

 $\Delta EmplSh_{c,h,t}^{k} = \alpha + \beta \Delta PC \text{ use}_{c,t} + \gamma \Delta Labor \text{ Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,h,t}^{k}$ (5)

for each skill h, CZ c, occupation/task group k between 1980 and 2010.

## US and foreign born task specialization and PC adoption

 $\Delta EmplSh_{c,h,t}^{k} = \alpha + \beta \Delta PC \text{ use}_{c,t} + \gamma \Delta Labor \text{ Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,h,t}^{k}$ (5)

	Analytical Task	Routine Task	Manual Task
Panel A: Natives			
PC use	0.142**	-0.255**	0.113**
	(0.036)	(0.025)	(0.016)
Labor Productivity	-0.075	0.093	-0.018
	(0.074)	(0.064)	(0.044)
Obs.	722	722	722
R2	0.52	0.74	0.57
Panel B: Foreign B	orn		
PC use	0.078	-0.208**	$0.130^{+}$
	(0.104)	(0.068)	(0.066)
Labor Productivity	-0.318	0.049	0.269
	(0.385)	(0.148)	(0.284)
Obs.	722	722	722
R2	0.54	0.40	0.43

for each skill h, CZ c, occupation/task group k between 1980 and 2010.

Note: 722 CZs, 1980-2010. Standard errors (in parentheses) are clustered at the state level. **\*\***, **\***, **+** indicate significance at 1-percent, 5-percent and 10-percent level, respectively.

33 / 33

Computerization & Immigration

# 1950-1980 migration and 1980-2010 PCs adoption

 $\frac{\Delta Pop_{c,h,t}}{Pop_{c,1950}} = \alpha + \beta \Delta PC \text{ use}_{c,1980-2010} + \gamma \Delta Labor \text{ Productivity}_{c,1980-2010} + \phi_s + \Delta \varepsilon_{s,h,h}^k$ (6)

for each skill h, CZ c, occupation/task group k between 1950 and 1980.

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## 1950-1980 migration and 1980-2010 PCs adoption

$$\frac{\Delta Pop_{c,h,t}}{Pop_{c,1950}} = \alpha + \beta \Delta PC \text{ use}_{c,1980-2010} + \gamma \Delta Labor \text{ Productivity}_{c,1980-2010} + \phi_s + \Delta \varepsilon_{s,h,s}^k$$
(6)

for each skill h, CZ c, occupation/task group k between 1950 and 1980.

	Low Skilled	High Skilled			
Panel A: Natives					
PC use	4.565	3.321			
	(4.946)	(3.789)			
Labor Productivity	-4.648	-3.541			
	(8.080)	(7.774)			
Obs.	722	722			
R2	0.48	0.51			
Panel B: Foreign Born					
PC use	0.458	0.426			
	(0.761)	(0.386)			
Labor Productivity	-0.385	-0.479			
	(1.125)	(0.766)			
Obs.	722	722			
R2	0.52	0.54			

Note: 722 CZs, 1950-1980. Standard errors (in parentheses) are clustered at the state level. \*\*, \*, + indicate

significance at 1-percent, 5-percent and 10-percent level, respectively.

33 / 33

Computerization & Immigration

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 $\Delta \log(w)_{c,t}^{k} = \alpha + \beta \Delta PC \text{ use}_{c,t} + \gamma \Delta \text{Labor Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,t}^{k} \quad (7)$ for each skill *h*, CZ *c*, occupation/task group *k* between 1980 and 2010.



$$\Delta \log(w)_{c,t}^{k} = \alpha + \beta \Delta \mathsf{PC} \text{ use}_{c,t} + \gamma \Delta \mathsf{Labor Productivity}_{c,t} + \phi_s + \Delta \varepsilon_{s,t}^{k} \quad (7)$$

for each skill h, CZ c, occupation/task group k between 1980 and 2010.

	Manag/Prof Occ	Cler/Ret/Prod Occ	Serv/Trans Occ
PC use	0.046**	-0.045**	0.006
	(0.004)	(0.006)	(0.005)
Labor Productivity	-0.000	$0.031^{+}$	-0.002
	(0.012)	(0.018)	(0.016)
Obs.	722	722	722
R2	0.66	0.73	0.46

Note: 722 CZs, 1980-2010. Standard errors (in parentheses) are clustered at the state level. **\*\***, **\***, **+** indicate significance at 1-percent, 5-percent and 10-percent level, respectively.

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# Skilled migration: Employment

Routine and Analytical Employment w/out and w/Skilled Migration



Back

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