

# Does Immigrant Diversity Make Urban Workers More Productive?

Thomas Kemeny – University of Southampton  
Abigail Cooke – University at Buffalo, SUNY

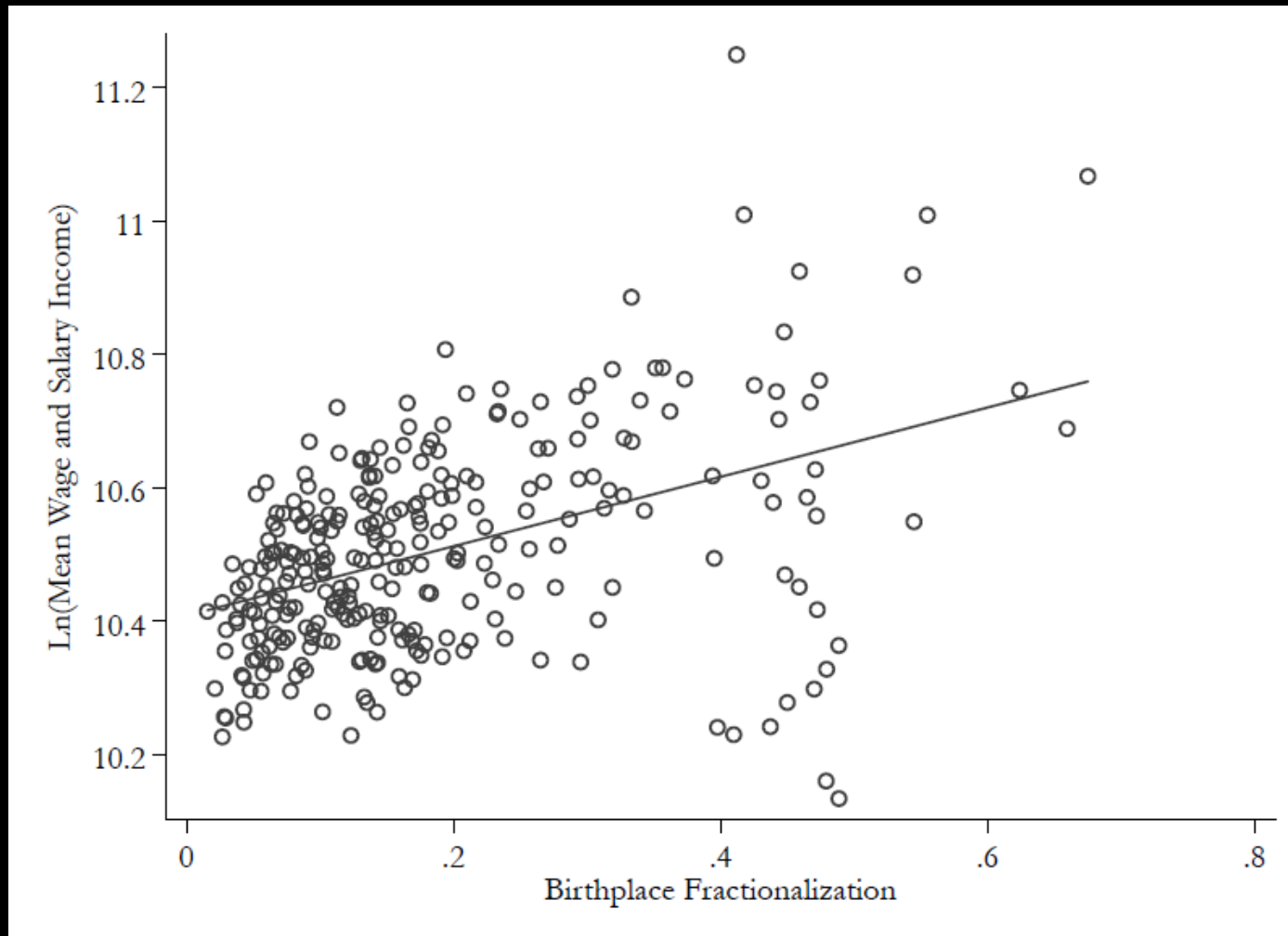
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# Disclaimer & Acknowledgements

- Disclaimer:
  - Any opinions and conclusions expressed herein are those of the authors and do not necessarily represent the views of the U.S. Census Bureau. All results have been reviewed to ensure that no confidential information is disclosed.
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# Diversity and Productivity

U.S. Metropolitan Wages and Birthplace Fractionalization, 2007



Data: American Community Survey 1% Public-Use Sample

# Diversity and Productivity

- Jane Jacobs-style theory on interactions leading to idea cross-pollination
- Spatial EQ flavored studies relating diversity to wages and rents
- Largely positive and robust association with wages (and rents)
  - Ottaviano and Peri, 2006, Nathan 2011, Kemeny 2012, Trax et al. 2012, Bellini et al. 2013, Bakens et al. 2013, Longhi 2013, Ager and Brückner 2013; Suedekum et al. 2014.

# Four Big Open Questions

1. Sorting on unobservables
  - i.e. Zuckerberg vs Not-Zuckerberg
2. Longitudinal dynamics
3. City versus Workplace effects
4. Reverse causality

## Research Question

Do changes in city- and  
workplace-diversity  
influence worker  
productivity?

# Data

- Longitudinal Employer Household Dynamics (LEHD)
  - Confidential US Census Bureau data
  - Based on Unemployment Insurance data
    - Captures link between worker and employer
  - Supplemented with other Census data
  - Quarterly data for 90% of US workforce for 30 eligible states
  - Data for 1992 to 2008

# Analytical Strategy

- Build standard birthplace diversity measures using all workers in establishments and cities
- Analysis: Focus on 'stayers'
  - Spells for workers who stay at same establishment for at least 2 years
  - We build a panel dataset capturing the longest such spell per worker
- Analytical Sample
  - 181 million records
    - 36.4 million workers
    - 1.3 million establishments (seinunits)
    - 232 metro areas



# Our Empirical Approach

$$\ln(w)_{ipjt} = d_{jt}\beta + d_{pjt}\gamma + X'_{ipjt}\delta + E'_{pjt}\theta + C'_{jt} + \mu_{it} + \eta_t + \nu_{ipjt}$$

- $\ln(w)_{ipjt}$  – individual's wages
- $d_{jt}$  – city specific immigrant diversity
- $d_{pjt}$  – workplace specific immigrant diversity
- $X'$ ,  $E'$ ,  $C'$  – Worker, Workplace, City characteristics
- $\mu_{ipj}$  – individual-workplace-city fixed effect
- $\eta_t$  – year fixed effect

# Identification: Do We Need Rents?

- Following the education spillovers literature (Acemoglu & Angrist 2001, Moretti 2004):
  - wages ought to be sufficient
    - If a firm serves a national market, then its prices reflect competition from non-local firms
    - Thus, if it pays higher wages, it *must* be due to higher worker productivity... Otherwise it would be forced to move

# Results 1 – Main FE Estimates

Dependent Variable = Log of Annual Earnings

N=181.1m → 36.4m individuals

	(1) City only	(2) Workplace only	(3) Both
City-level			
<b>Diversity</b>	<b>0.353***</b>		<b>0.324***</b>
Mean education	-0.115***	-0.120***	-0.114***
Employment	0.000***	0.000**	0.000***
Workplace-level			
<b>Diversity</b>		<b>0.073***</b>	<b>0.066***</b>
Employment	0.000*	0.000	0.000*

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ ; Year FE in all models; SE clustered by estab

# Results 2 – Additional FE Estimates

	(4) Large workplaces (N=164.7m)	(5) 1-unit firms (N=89m)	(6) White native men (N=65.2m)
City-level			
<b>Diversity</b>	<b>0.313***</b>	<b>0.488***</b>	<b>0.434***</b>
Mean education	-0.111***	-0.112***	-0.135***
Employment	0.000**	0.000**	0.000***
Workplace-level			
<b>Diversity</b>	<b>0.073***</b>	<b>0.076***</b>	<b>0.073***</b>
Employment	0.000*	0.000	0.000**

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ ; Year FE in all models; SE clustered by estab

# Results 3 – FE Tradables Only

	(7) Manufacturing (N=33.2m)	(8) Info Tech (N=6.3m)
City-level		
<b>Diversity</b>	<b>0.498***</b>	<b>0.671**</b>
Mean education	-0.163***	-0.167
Employment	0.000	-0.000
Workplace-level		
<b>Diversity</b>	<b>0.102***</b>	<b>0.123***</b>
Employment	0.000	0.000***

\*  $p < 0.10$ , \*\*  $p < 0.05$ , \*\*\*  $p < 0.01$ ; Year FE in all models; SE clustered by estab

# Results 4: Other Specs

- Approx 80 other models run so far
  - Other measures of immigrant diversity
  - Other kinds of diversity (e.g. race, age, education)
  - Additional controls on changes in city- or estab-level human capital among immigrants
  - Combinations of these variations
  - ....all the same result positive and significant coefficient on both city- and establishment diversity

# Results: Magnitudes

- $+\Delta 1\text{SD}$  diversity city  $\rightarrow$  wages  
 $+\Delta 4\%$
- $+\Delta 1\text{SD}$  diversity estab  $\rightarrow$  wages  
 $+\Delta 1.3\%$

# Discussion

- 3 out of 4 identified issues addressed
  1. Unobserved heterogeneity
  2. Dynamics: 18-year panel
  3. Scale issues: confirm Trax et al (2012)
  4. *Instruments: Coming soon*



Thank you

Tom: [t.e.kemeny@soton.ac.uk](mailto:t.e.kemeny@soton.ac.uk)

Abigail: [amcooke@buffalo.edu](mailto:amcooke@buffalo.edu)

# LEHD data sets

- U2W
  - most likely workplace for multi-unit employers – mode of 10 imputes
- ICF
  - workers' birthplace
- EHF
  - longest job spell
- ECF-SEINUNIT and SEIN
  - workplace location, size, industry

# Measures/Sample Construction

- Annual CBSA Diversity: Fractionalization

$$Fractionalization_j = 1 - \sum_{r=1}^R s_{rj}^2$$

- Annual Workplace Diversity: Fractionalization
  - Weighted by quarters worked
- Identifying “stayers”
  - Longest job-spell > 2 calendar years
  - Drop workers with multiple simultaneous jobs
  - Drop
    - employers < 10 employees,
    - workers earning < 5<sup>th</sup> p

# Descriptive Statistics

	Mean	Standard Deviation
Individual Characteristics		
Log Annual Earnings	10.47	0.633
US Born	0.848	0.359
Spell Duration	4.970	3.302
Establishment Characteristics		
Fractionalization	0.210	0.205
Employment	63	273
City Characteristics		
Fractionalization	0.160	0.126