# Agglomeration Economies Urban Economics

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March 8, 2022

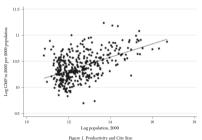
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# **Agglomeration Economies**

- ▶ Why do we see such a remarkable clustering of human activity in a small number of urban areas?
- ▶ Spatial Eq. Model: cities may form because some places have innate advantages in productivity, housing supply or amenities.
- Or it may be because clusters of people endogenously increase productivity, housing supply or amenities (agglomeration effects)
  - Example: Los Angeles. In its early history, prosperous retirees came to enjoy the climate (an innate amenity). Also: restaurants and theather endogenously emerged with the infux of population.
  - ▶ But: if cities were driven by amenities, then real wages should be lower in big urban areas. This is not true. People require a wage premium to locate in big cities.
  - ► Can cities be driven by innate advantage in supplying housing, or because density makes it easier to build? No. It is more expensive to build vertically than horizontally. Housing supply is more expensive in bigger areas.

# **Agglomeration Economies**

► Then cities exist because they are areas with high levels of productivity, which might occur because people come to places that are innately more productive or because density itself enhances productivity because of agglomeration economies



Notes: Units of observation are Metropolitan Statistical Areas under the 2006 definitions. Population is from the Census, as described in the Data Appendix. Gross Metropolitan Product is from the Bureau of Economic

The regression line is  $\log GMP \ per \ capita = 0.13 \ [0.01] \times \log \ population + 8.8 \ [0.1].$  $R^2 = 0.25 \ and N = 363.$ 

# **Evidence of Agglomeration Economies**

- ► Three strategies to identify agglomeration economies
  - 1 Show there is too much spatial concentration to be random (Duranton and Overman, 2005)
  - 2 Compare productivity over space (Greenstone, 2010)
  - 3 Compare wages and rents across space (Quantitative Spatial Models, Ahlfeldt et al, 2015)

# **Spatial Concentration**

Extremes of Localization and Dispersion



(c) Other Agricultural and Forestry Machinery (SIC2932)

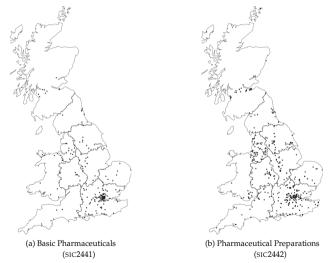


(d) Machinery for Textile, Apparel and Leather Production (SIC2954)



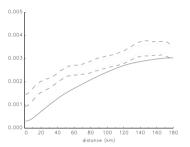
# **Spatial Concentration**

#### **Ambiguous Cases**

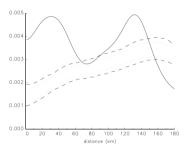


## **Spatial Concentration**

#### K Density Estimates



(c) Other Agricultural and Forestry Machinery (SIC2932)



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#### Most Localized

sic92	Industry	$\Gamma$ or $\Psi$		
Most localised				
2214	Publishing of Sound Recordings	0.470		
1711	Preparation and Spinning of Cotton-type Fibres	0.411		
2231	Reproduction of Sound Recordings	0.403		
1760	Manufacture of Knitted and Crocheted Fabrics	0.321		
1713	Preparation and Spinning of Worsted-type Fibres	0.319		
2861	Manufacture of Cutlery	0.314		
1771	Manufacture of Knitted and Crocheted Hosiery	0.290		
1810	Manufacture of Leather Clothes	0.203		
1822	Manufacture of Other Outerwear	0.181		
2211	Publishing of Books	0.178		

# Most Dispersed

	U				
Most dispersed					
1520	Processing and Preserving of Fish and Fish Products	0.200			
3511	Building and Repairing of Ships	0.113			
1581	Manufacture of Bread, Fresh Pastry Goods and Cakes	0.094			
2010	Saw Milling and Planing of Wood, Impregnation of Wood	0.082			
2932	Other Agricultural and Forestry Machinery	0.067			
1551	Operation of Dairies and Cheese Making	0.064			
1752	Manufacture of Cordage, Rope, Twine and Netting	0.062			
3615	Manufacture of Mattresses	0.050			
1571	Manufacture of Prepared Feeds for Farm Animals	0.049			
2030	Manufacture of Builders' Carpentry and Joinery	0.047			

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- Can you think on an ideal experiment?

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- What is the model?
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  - ▶ Places without new plants are not a valid control group

- ▶ Regular feature in the corporate real estate journal Site Selection Stories about the location choice of large new plants
- ► Gradual narrowing down of potential counties to 2 or 3 finalists
- ▶ The 1 or 2 losers in the shortlist provide a control group
  - Almost as attractive as the winning county
  - Yet, they did not receive the treatment

TABLE 1
THE MILLION DOLLAR PLANT SAMPLE

	(1)
Sample MDP openings: <sup>a</sup>	
Across all industries	47
Within same two-digit SIC	16
Across all industries:	
Number of loser counties per winner county:	
1	31
2+	16
Reported year - matched year:b	
-2 to $-1$	20
0	15
1 to 3	12
Reported year of MDP location:	
1981–85	11
1986–89	18
1990-93	18
MDP characteristics, 5 years after opening:	
Output (\$1,000s)	452,801
•	(901,690)
Output, relative to county output 1 year prior	.086
	(.109)
Hours of labor (1,000s)	2,986
	(6,789)

► Plant-level regression

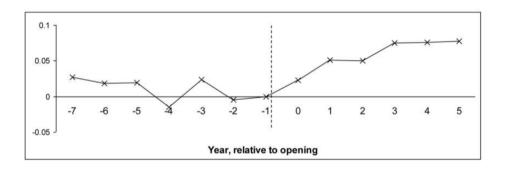
$$log(Y) = log(A) + \beta_1 \log(L) + \beta_2 \log(K_B) + \beta_3 \log(K_E) + \beta_4 \log(M)$$
 (1)

where

$$log(A) = \delta_1 Winner + \delta_2 Post + \delta_3 Winner \times Post$$
 (2)

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Difference: Winners - Losers



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TABLE 9
CHANGES IN COUNTIES' NUMBER OF PLANTS, TOTAL OUTPUT, AND SKILL-ADJUSTED
WAGES FOLLOWING AN MDP OPENING

	A. Census of Manufactures		B. CENSUS OF POPULATION	
	Dependent Variable: Log(Plants) (1)	Dependent Variable: Log(Total Output) (2)	Dependent Variable: Log(Wage) (3)	
Difference-in-				
difference	.1255**	.1454	.0268*	
	(.0550)	(.0900)	(.0139)	
$R^2$	.9984	.9931	.3623	
Observations	209	209	1,057,999	

# **Further Readings**

- ▶ Duranton, G., & Overman, H. G. (2005). Testing for localization using micro-geographic data. The Review of Economic Studies, 72(4), 1077-1106.
- ▶ Glaeser, E. L., & Gottlieb, J. D. (2009). The wealth of cities: Agglomeration economies and spatial equilibrium in the United States. Journal of economic literature, 47(4), 983-1028.
- ▶ Glaeser, E. (2008). Cities, agglomeration, and spatial equilibrium. OUP Oxford.
- ▶ Greenstone, M., Hornbeck, R., & Moretti, E. (2010). Identifying agglomeration spillovers: Evidence from winners and losers of large plant openings. Journal of Political Economy, 118(3), 536-598.
- Ponzetto, G. (2012) Agglomeration and Transport Costs. Mimeo