

# Distributional Approaches to Understand Patterns of Urban Differentiation

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Life and Social Sciences

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Rachata Muneeppeerakul (ASU)

HyeJin Youn (INET, Oxford, SFI)

Deborah Strumsky (SFI)

Kevin Stolarick (MPI)

Vladislav Vysotsky (PDMI, ASU)

# The Big Question

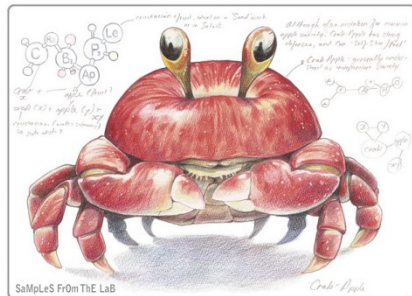
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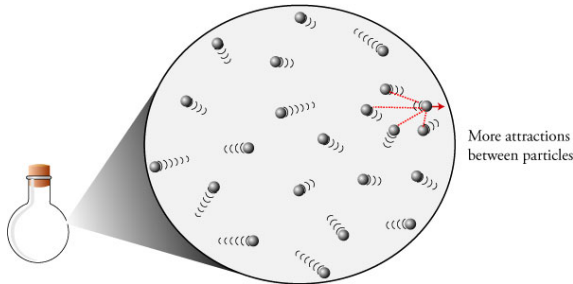
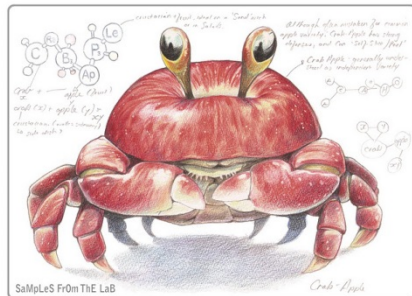
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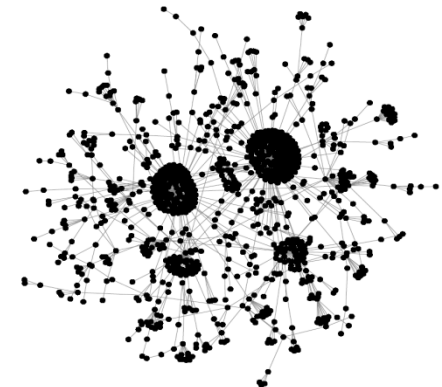
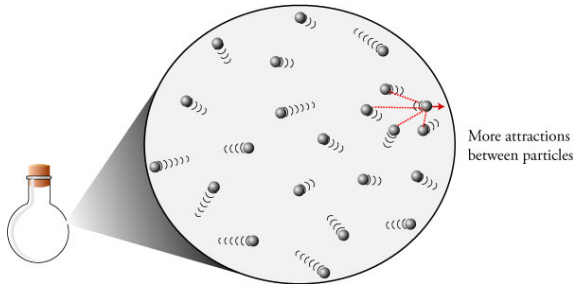
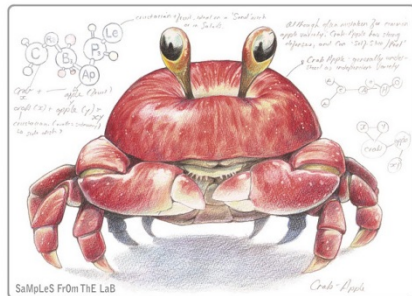
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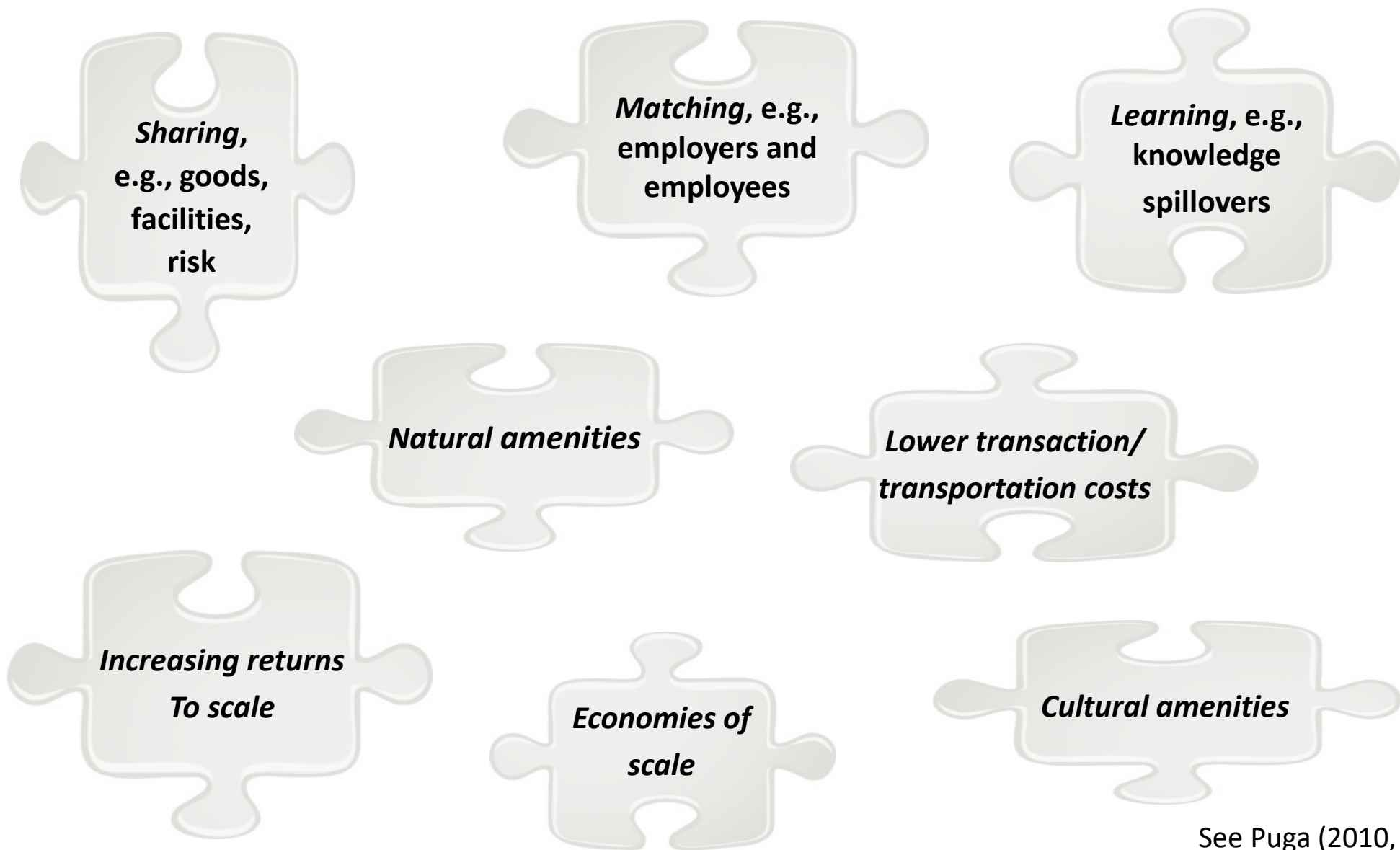
*Diversity of: Shapes, Colors, and Sizes.*



Artist: **Ahmed Farid**. Title: **Urbanization** (Oil on canvas, 2012)

From: <http://almasargallery.com/ahmed-farid-urban-diversity-11-may-30-may-2013-solo-exhibition>

# Diversity and the Puzzle of Cities





# Explaining Productivity Differentials

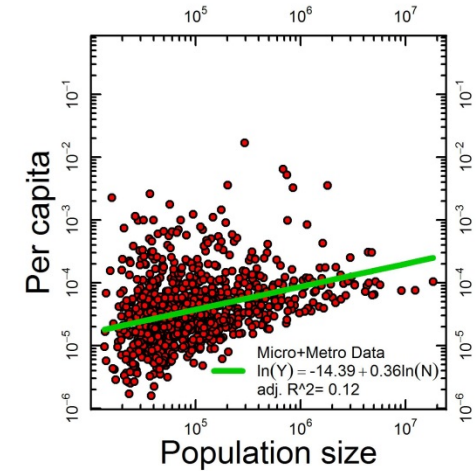
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$$\frac{Y}{N} = A \left( \frac{K}{N} \right)^{\kappa} \left( \frac{L}{N} \right)^{\lambda} \left( \frac{H}{N} \right)^{\eta} \left( \frac{M}{N} \right)^{\mu} \cdots \left( \frac{P}{N} \right)^{\pi}, \quad \kappa + \lambda + \dots + \pi = 1$$

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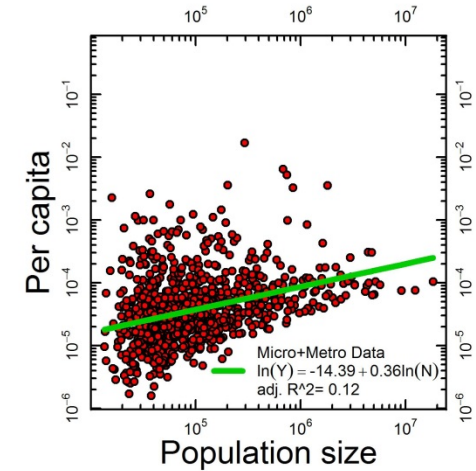
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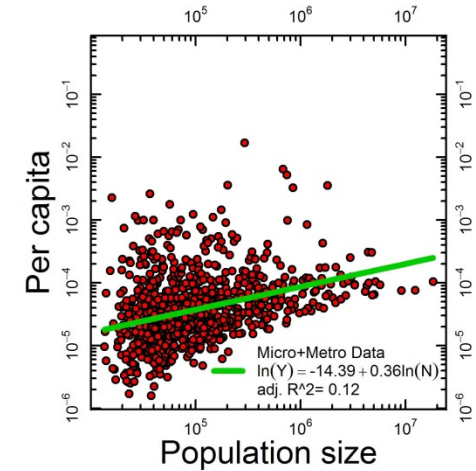
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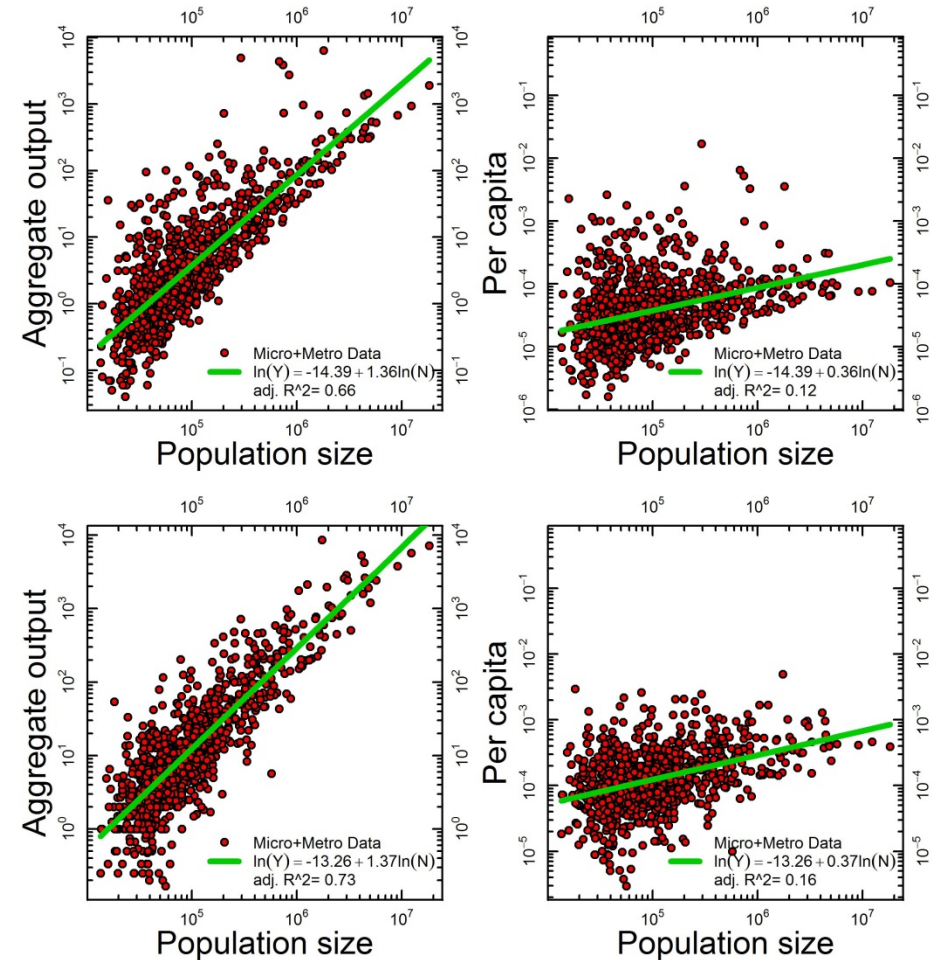


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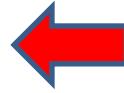
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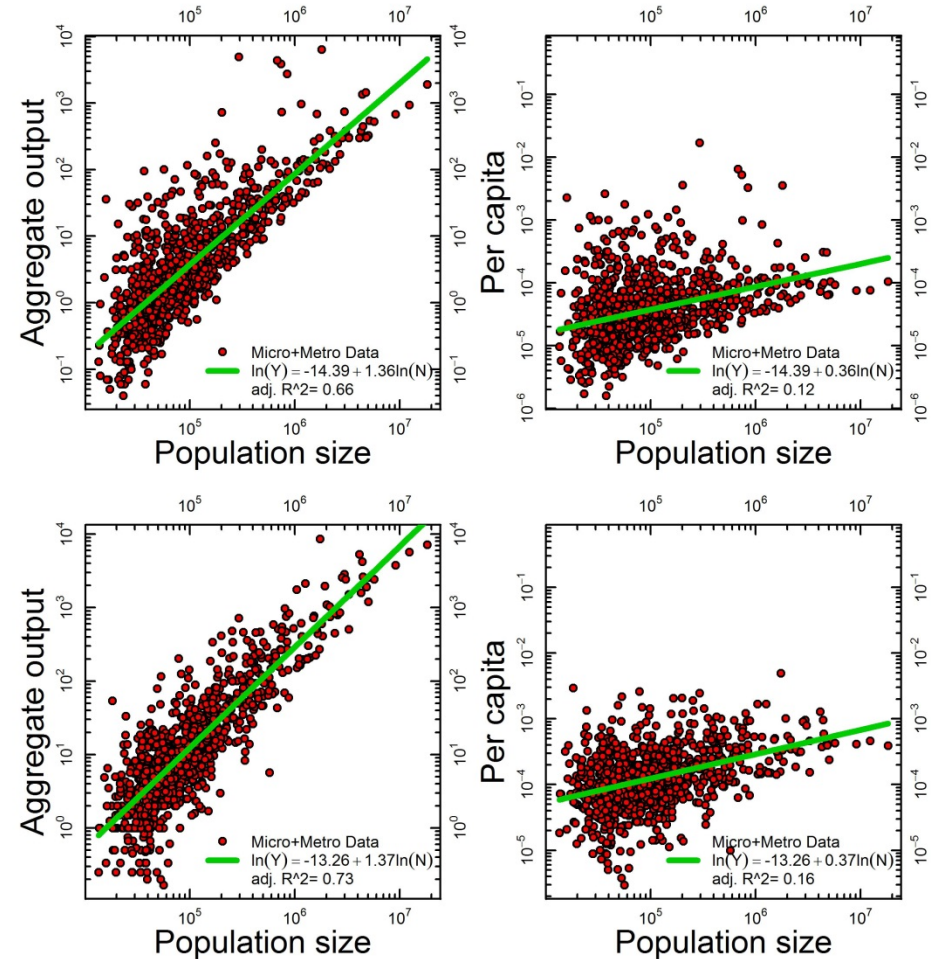
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Simulated data!

- Real populations,
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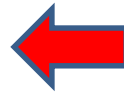
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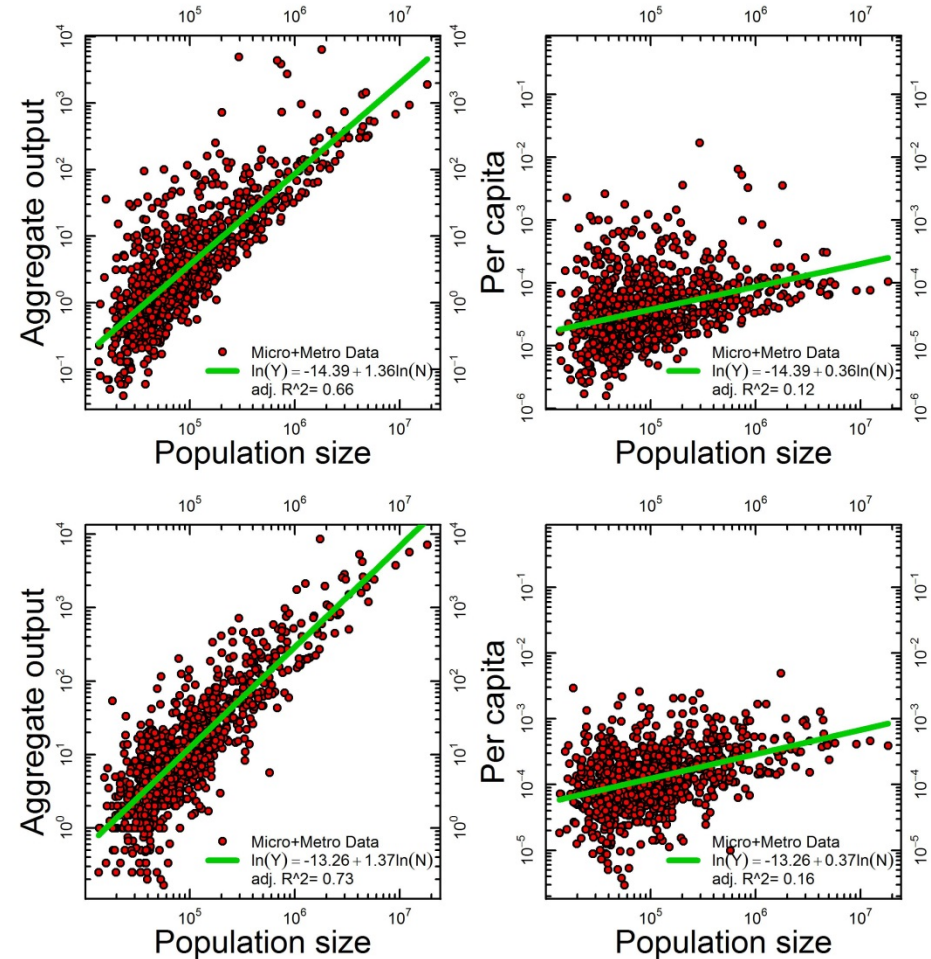
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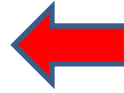




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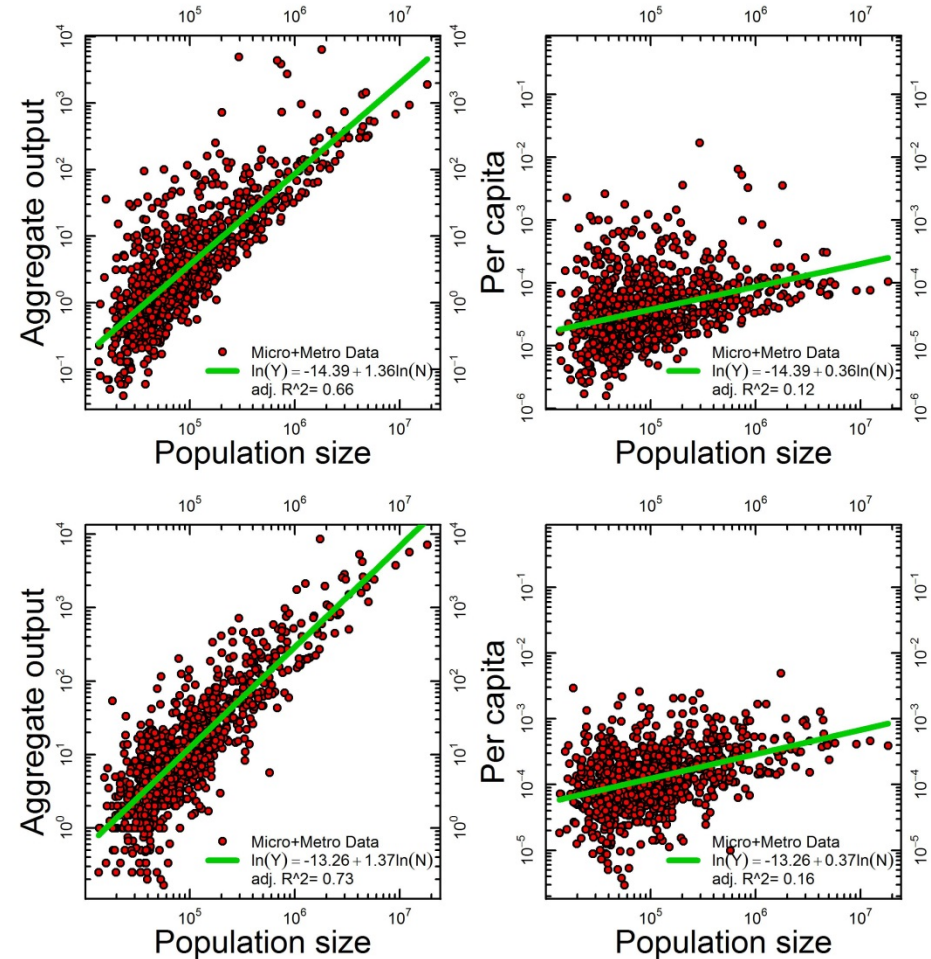
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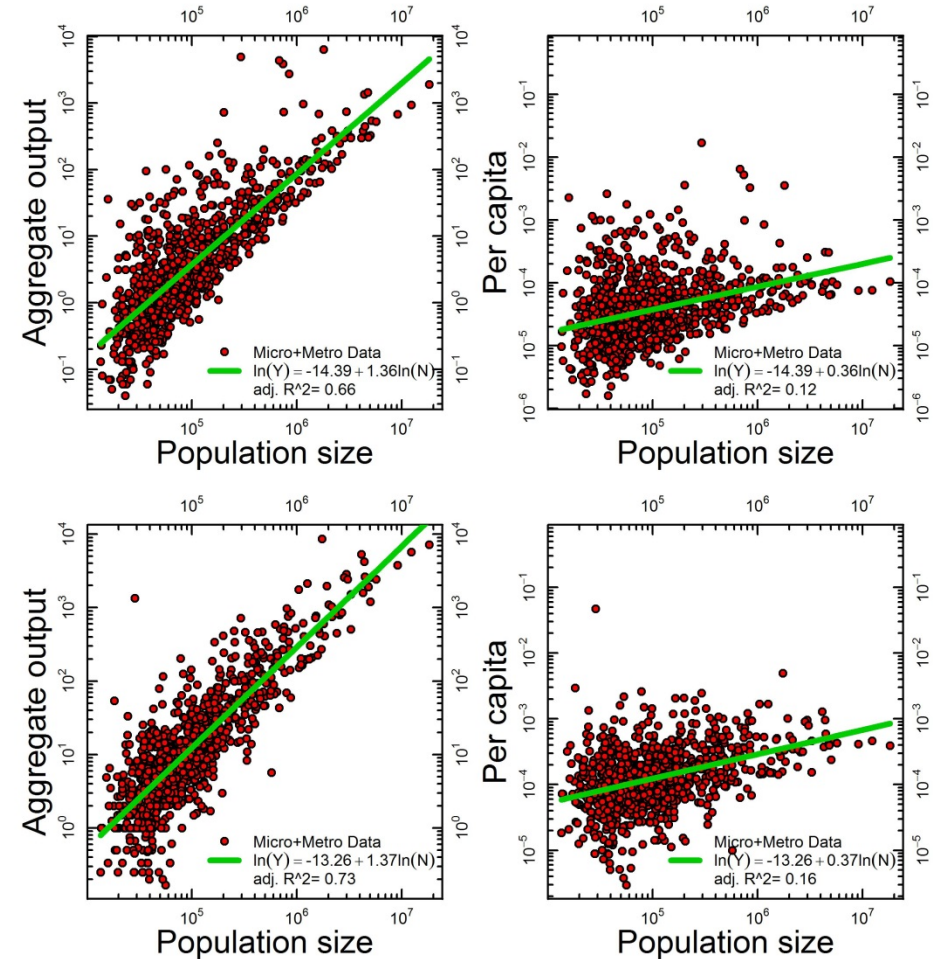
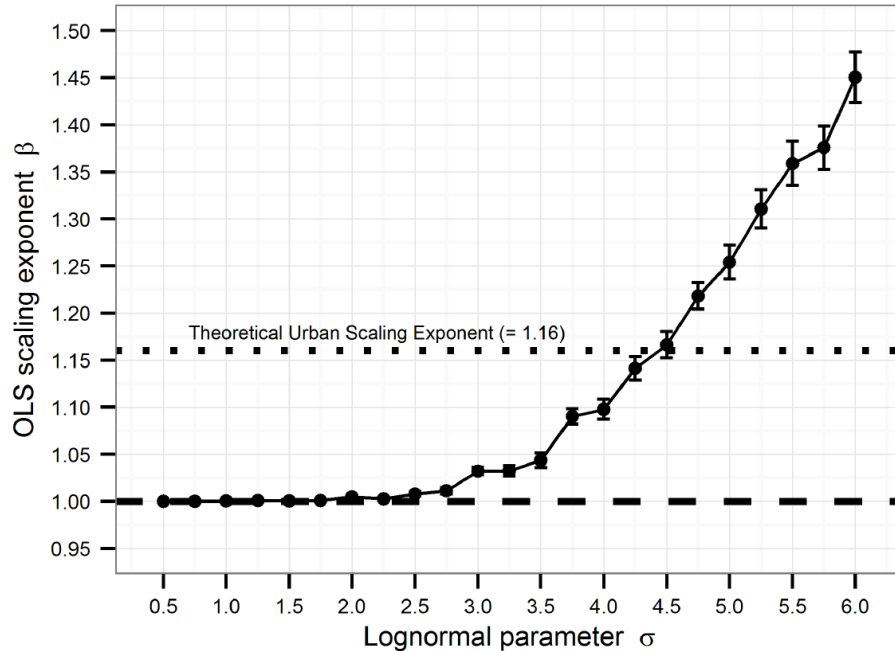
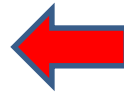




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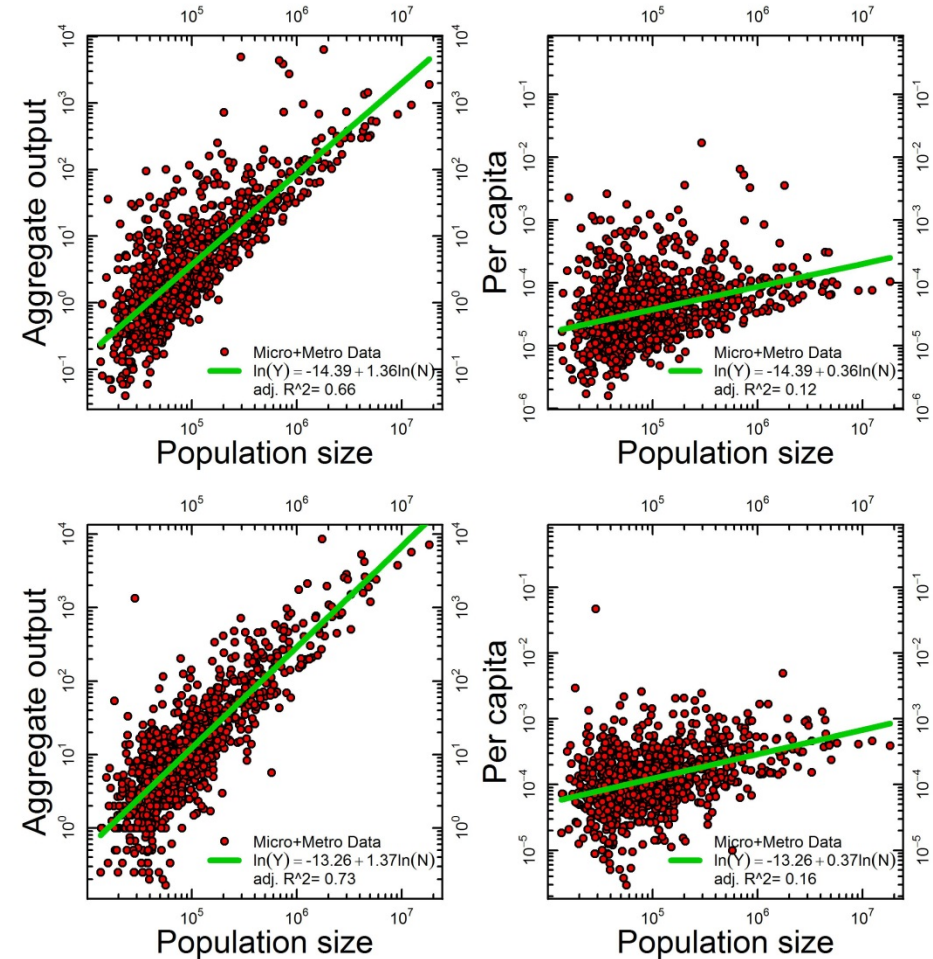
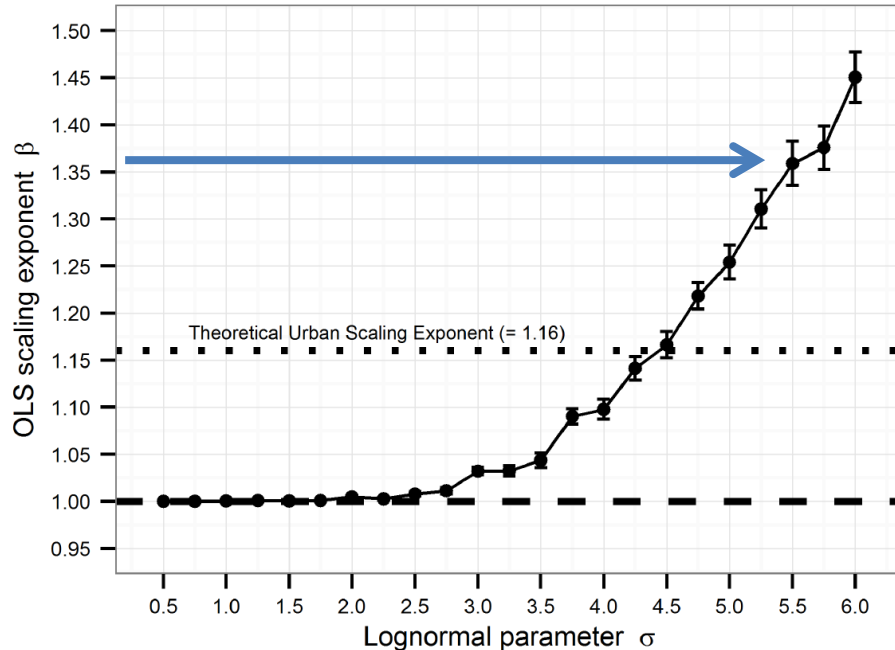
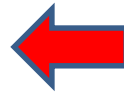
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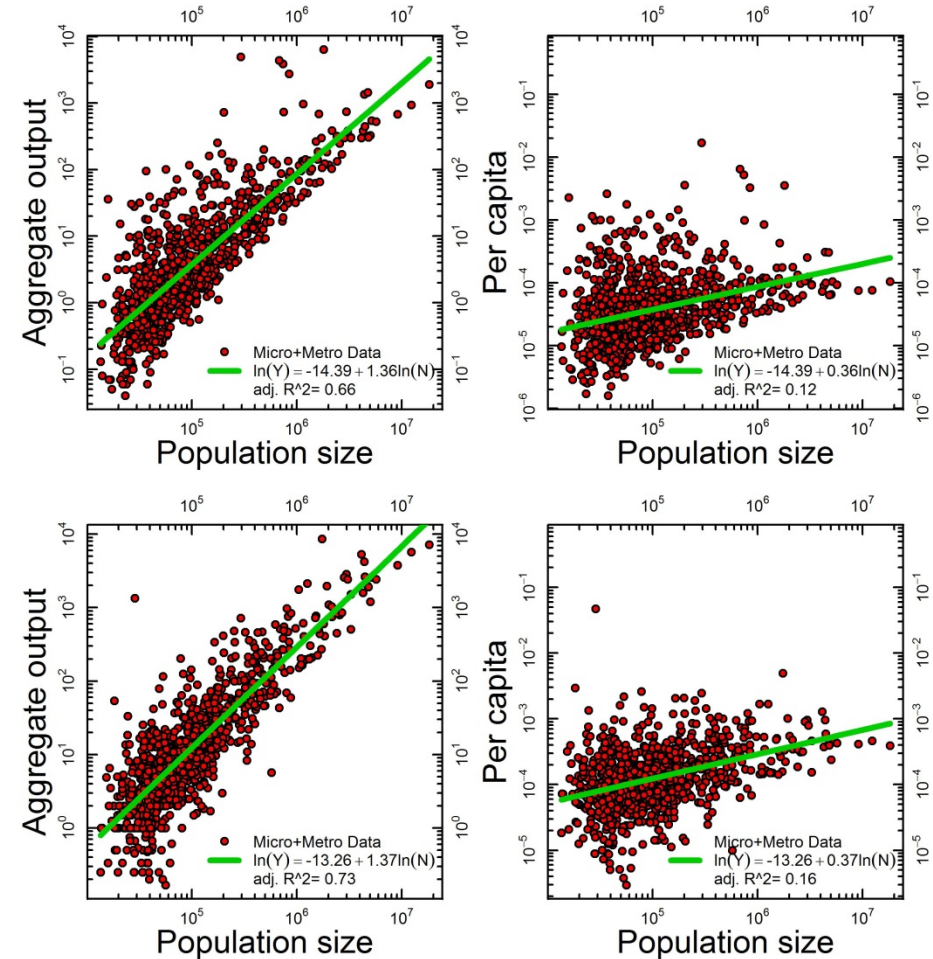
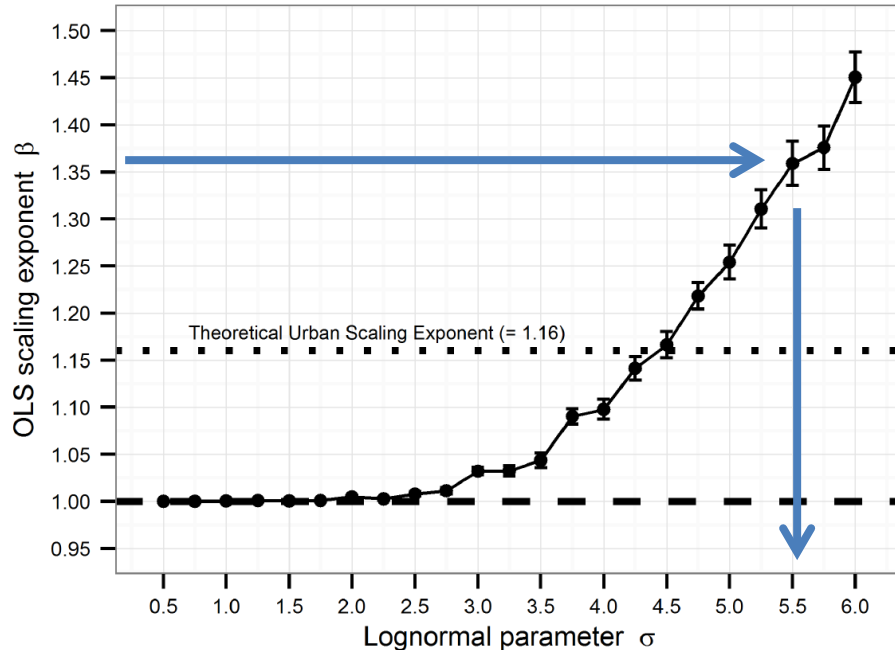
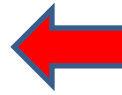
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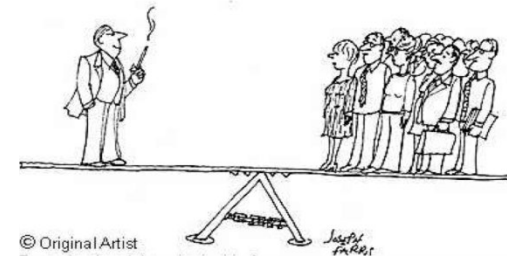


The interplay between city **sizes** and their internal **heterogeneity** determines:

- how we should **measure** productivity,
- and the **statistical properties** of the **aggregate output**.

# Facing the Heterogeneity

Are cities **finite**  
systems that **violate**  
the *Law of Large*  
*Numbers*?



Gomez-Lievano, Vysotsky, and Lobo (2014, *in preparation*)



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- Sveikauskas (1975); Ó hUallacháin (1999); Bettencourt et al. (2007a, 2007b, 2010); Bettencourt (2013).

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## Knowledge is spatially localized

- E.g., Jaffe et al. (1993)
- “tacit knowledge”
- People move with *less difficulty* than their ideas
  - (See Breschi & Lissoni, 2001a,b, 2004, 2009; Klepper & Sleeper, 2005; Klepper, 2010; Hausmann, Neffke, Otto, 2013)

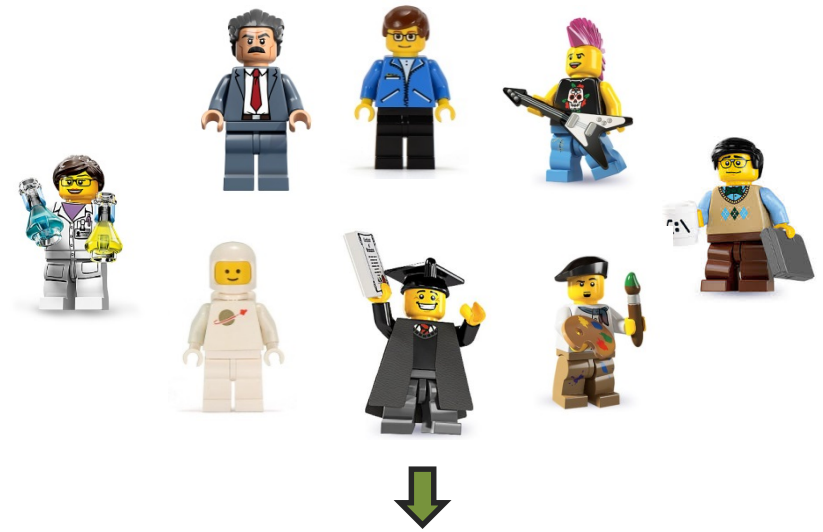
Conceptually, the economic output of an entire city (a *system-wide* property) should not be understood using the average properties of individuals (e.g., years of schooling)

# What citizens *do*

- Creative Class (Florida, 2004)
- Inventors

# Obstacles to Economic Development

- Are the counts of 'creatives' and inventors ***constrained*** in urban areas?



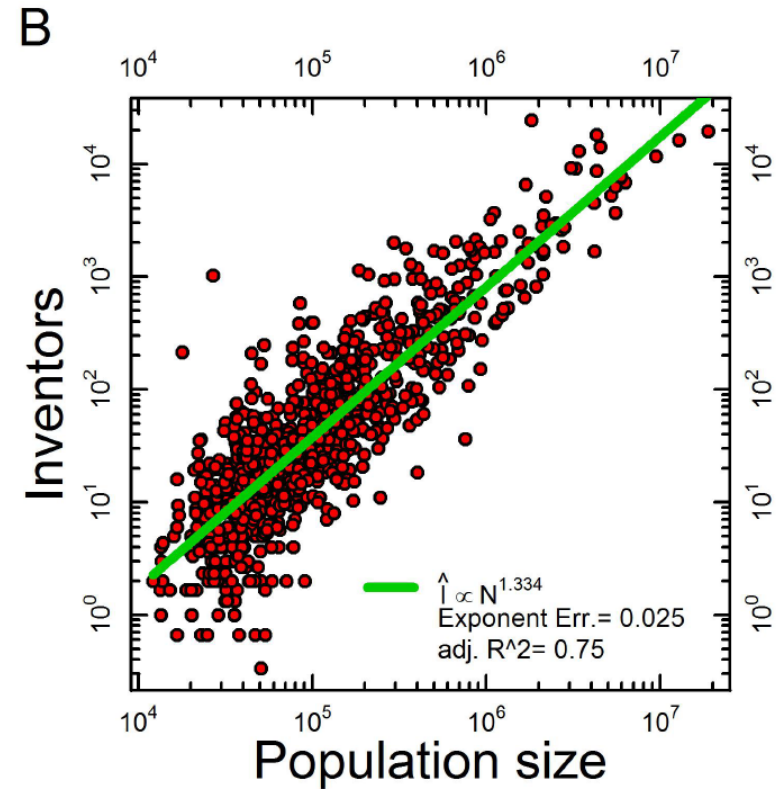
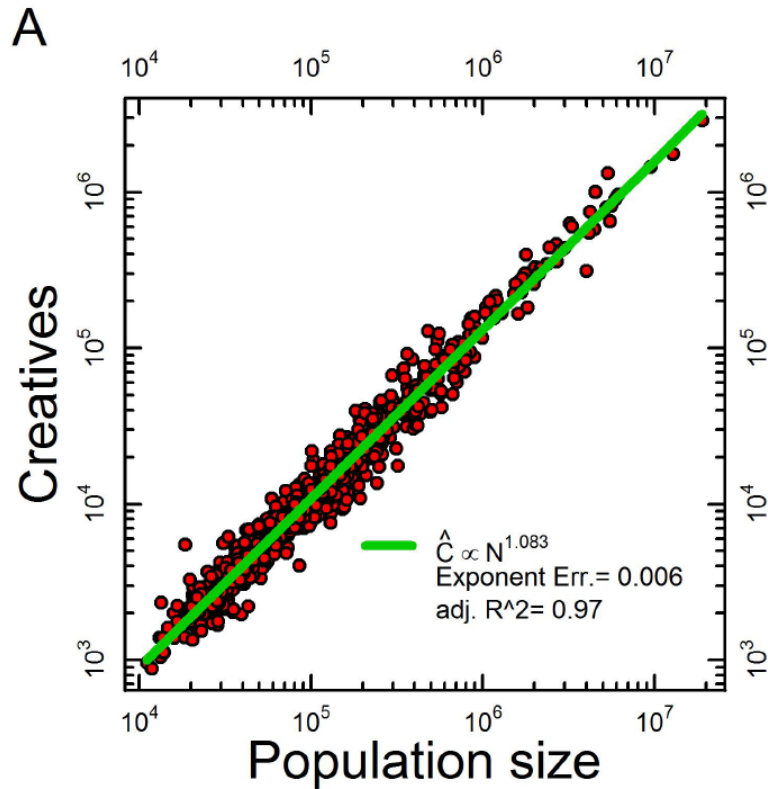
Gomez-Lievano, Bettencourt, Stolarick, Strumsky, Lobo  
(2014, *in submission*)

# Assumption

The ability of a city to “learn” is constrained by its **capacity to attract** creative and inventive individuals (Florida, 1995, *Futures*).



# More than 70% explained by urban population

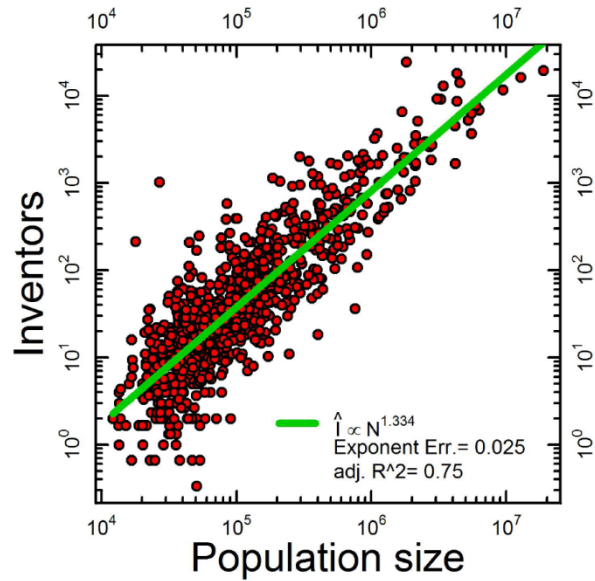


$$E(Y|N) = Y_0 N^\beta$$

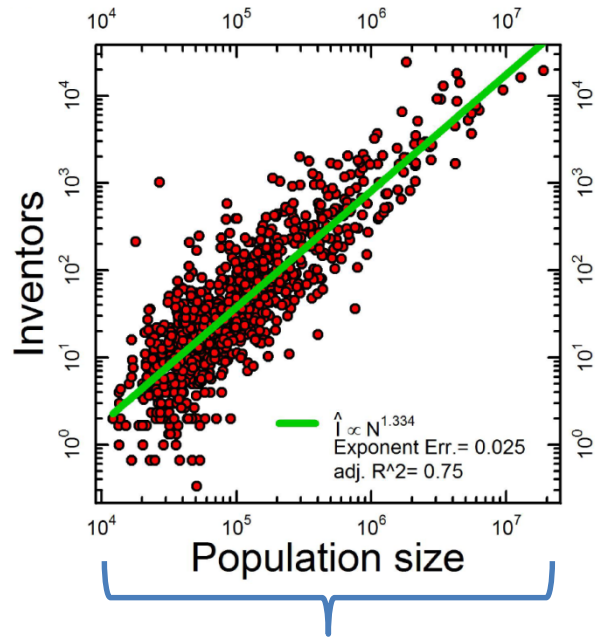
(Rauch, 1993; Glaeser & Saiz, 2004; Bettencourt et al., 2007)

Methodology & Results:  $P(Y,N)=P(Y|N)P(N)$

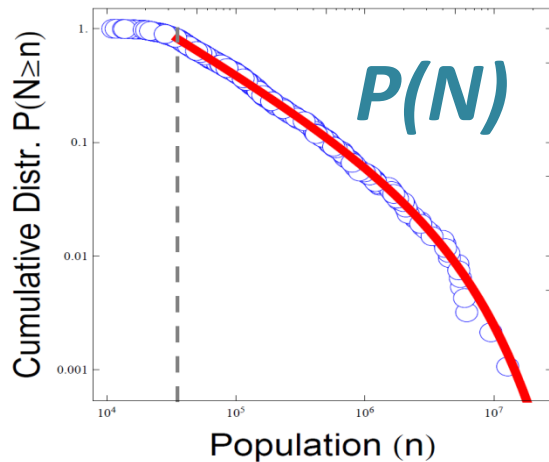
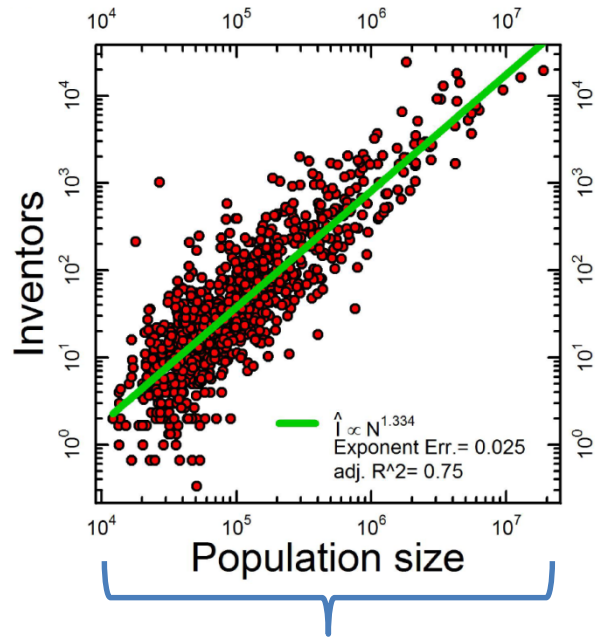
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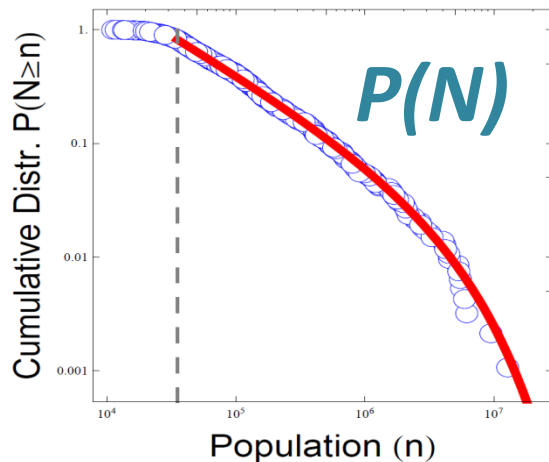
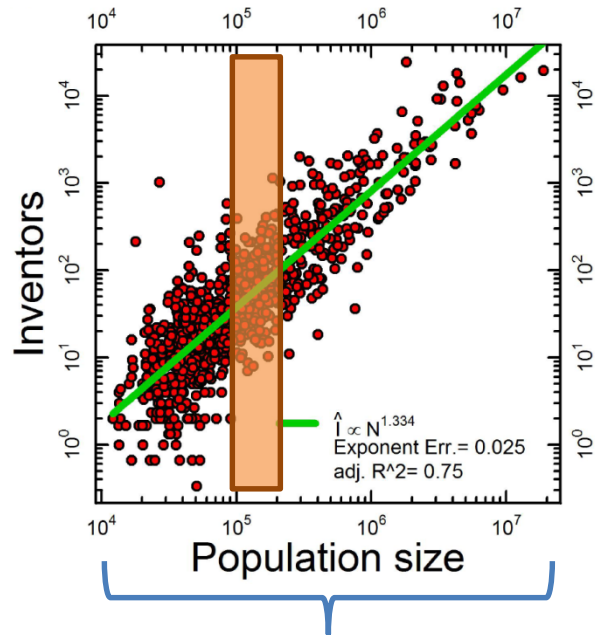


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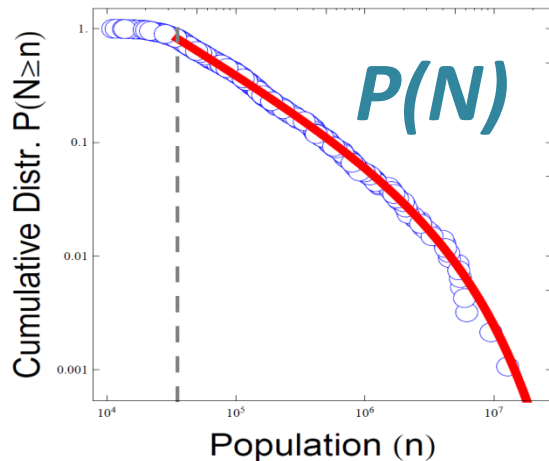
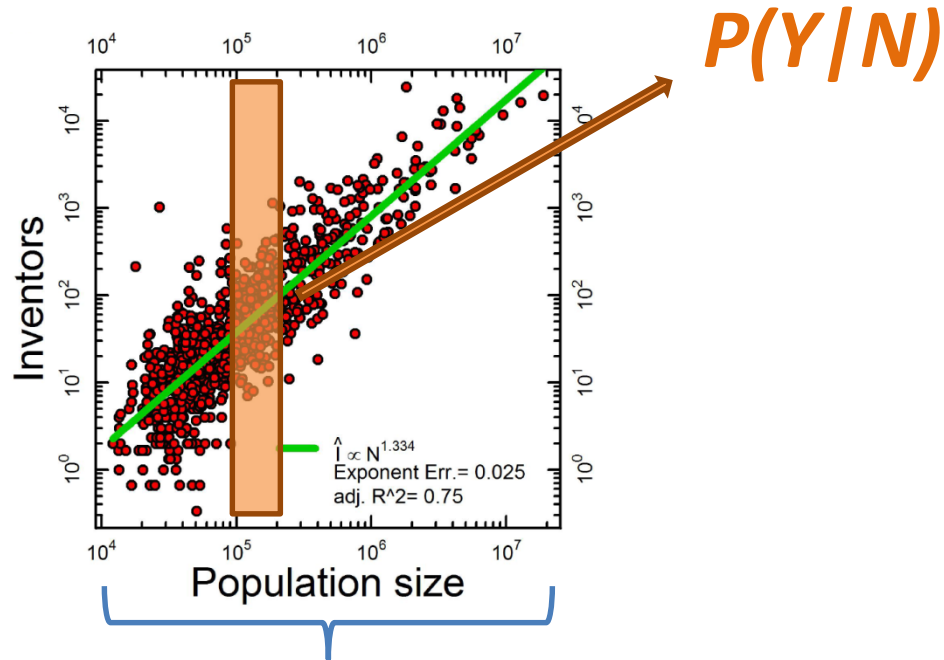
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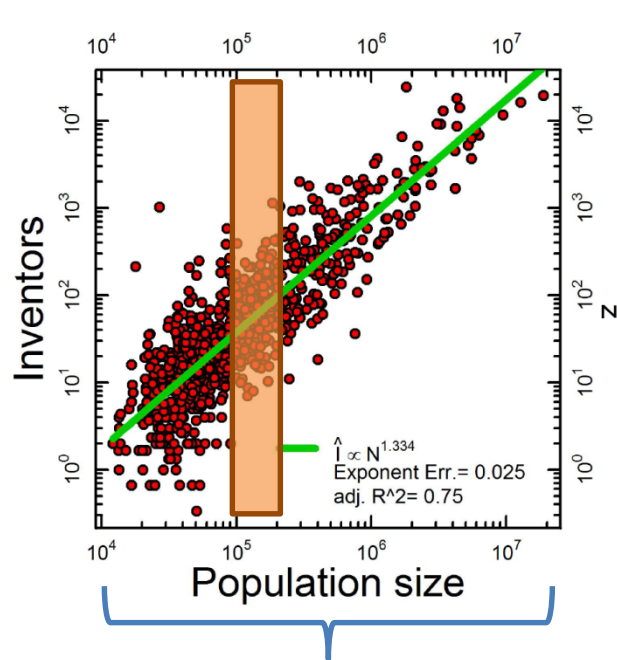
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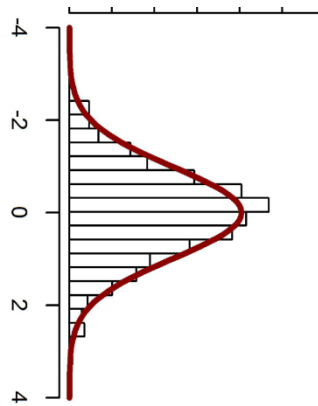


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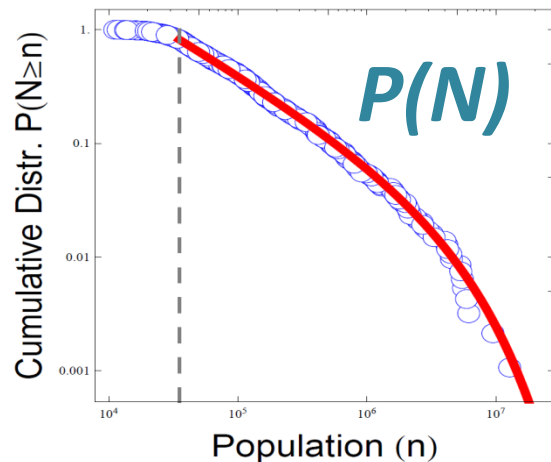
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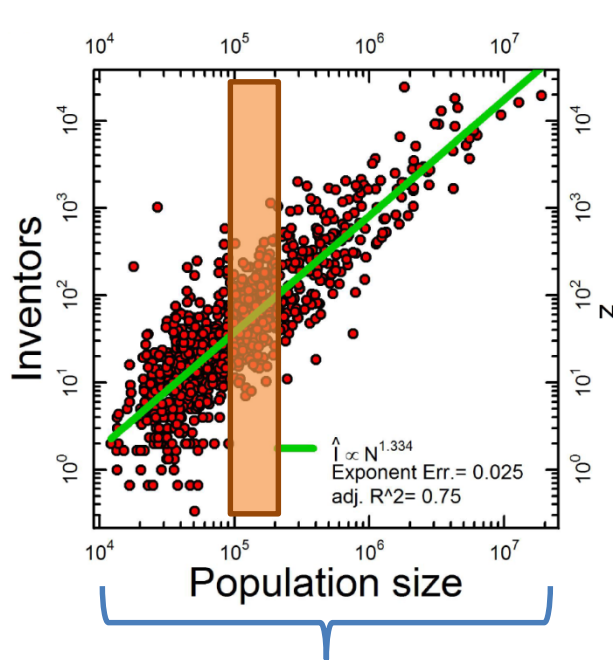
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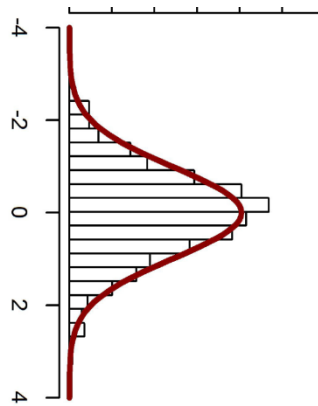
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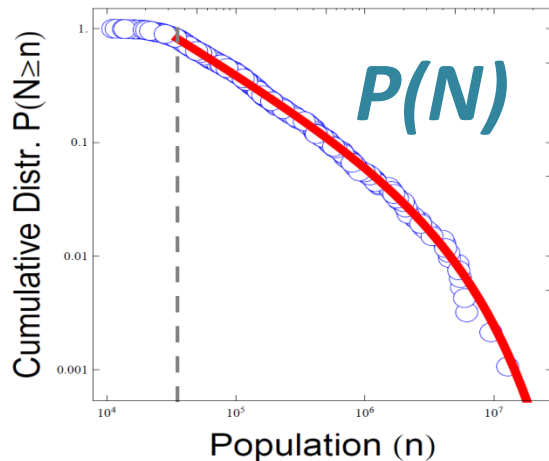
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For a similar study regarding homicides in Brazil, Colombia, and Mexico:

**Gomez-Lievano, Youn, Bettencourt (2012, *PLoS ONE*)**

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$$X_1 * X_2 * X_3 * \dots = "X_1 \text{ AND } X_2 \text{ AND } X_3 \text{ AND } \dots"$$



# Why is this important?

- Whereas **normally** distributed RVs often arise from random *additive* processes

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- Jaynes, 2003; Frank, 2009; Frank and Smith, 2011

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**“Urban dynamics, as can be seen, are the ultimate noisy  
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– Storper et al. (2012, *JRS*, p. 4)



- “Size”, “Heterogeneity” and “Structure”
  - *Policies and Distributions*



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THANKS!



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existing set of skills in the city,  
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generate **lognormal** distributions in **output**?

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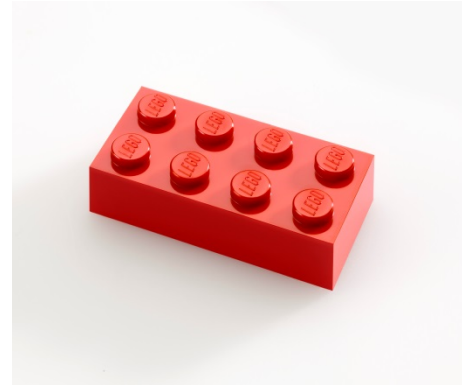
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I expand H&H's model in **two** ways:

1. I will model a **process of acquisition** of skills (pbytes).
2. I will model *what* cities produce, as a **step** to know ***how much*** they produce.

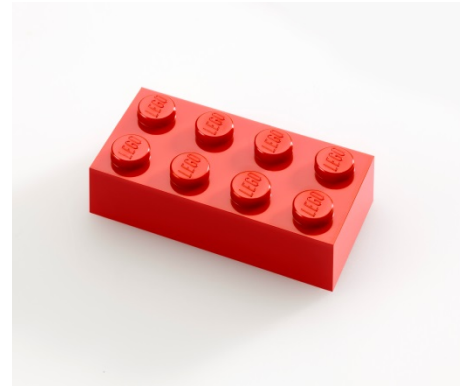


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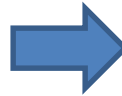
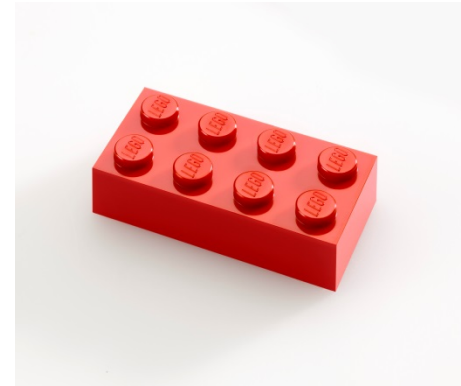


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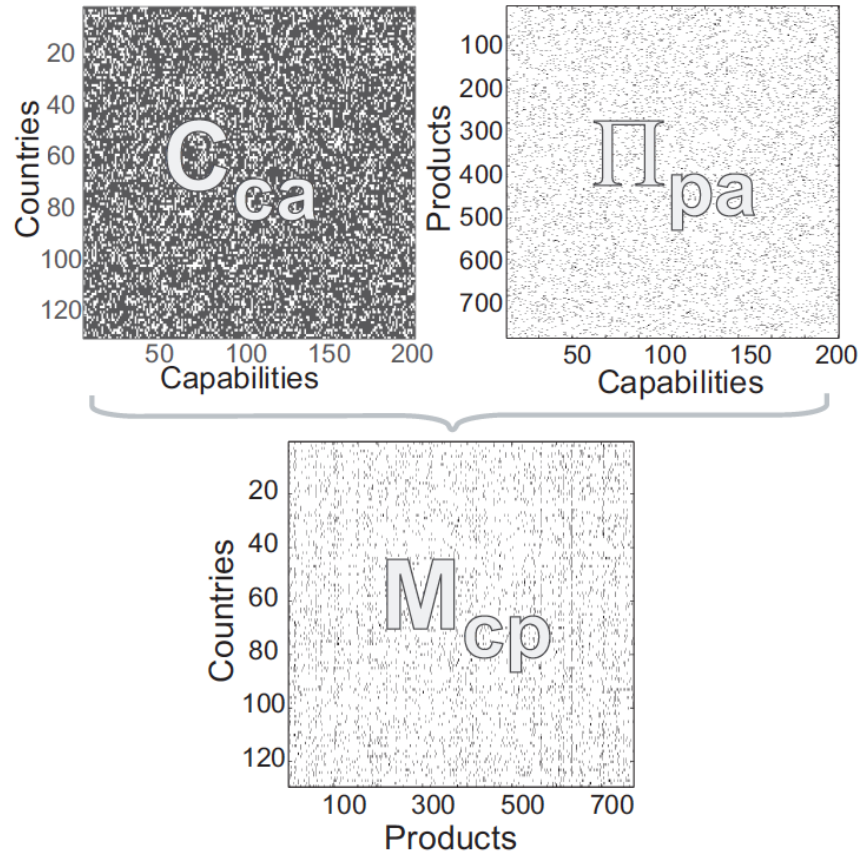
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Lego analogy inspired by:  
TEDxBoston - César A. Hidalgo - Global Product Space

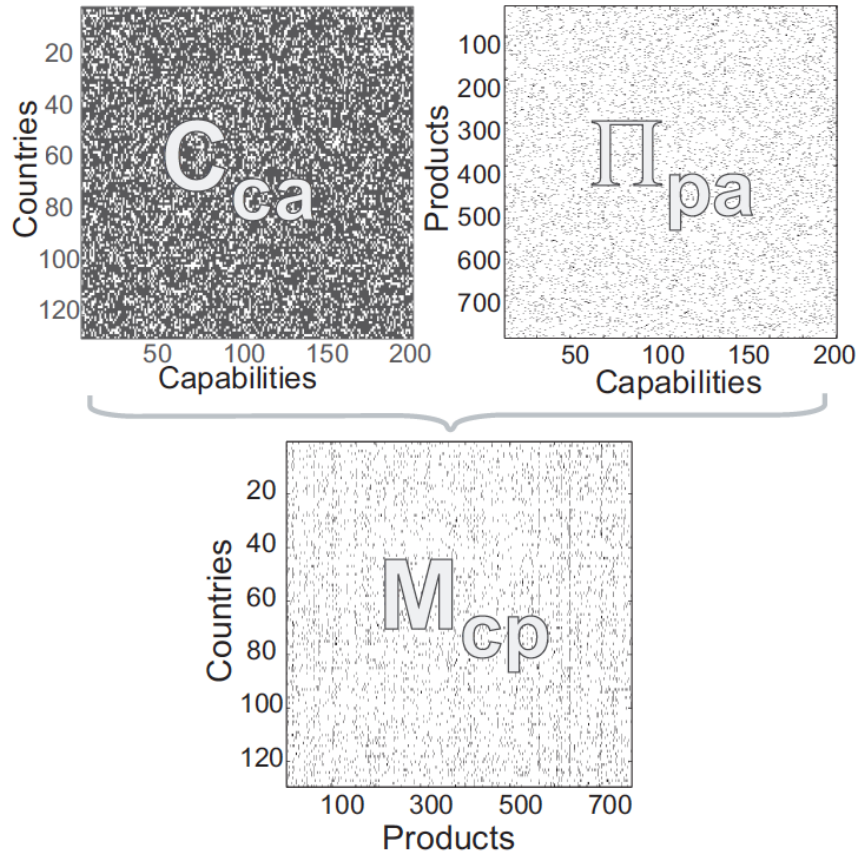


# The Model



From: Hidalgo, C. A. and Hausmann, R. (2009).  
The building blocks of economic complexity. PNAS,  
106(25):10570-10575.

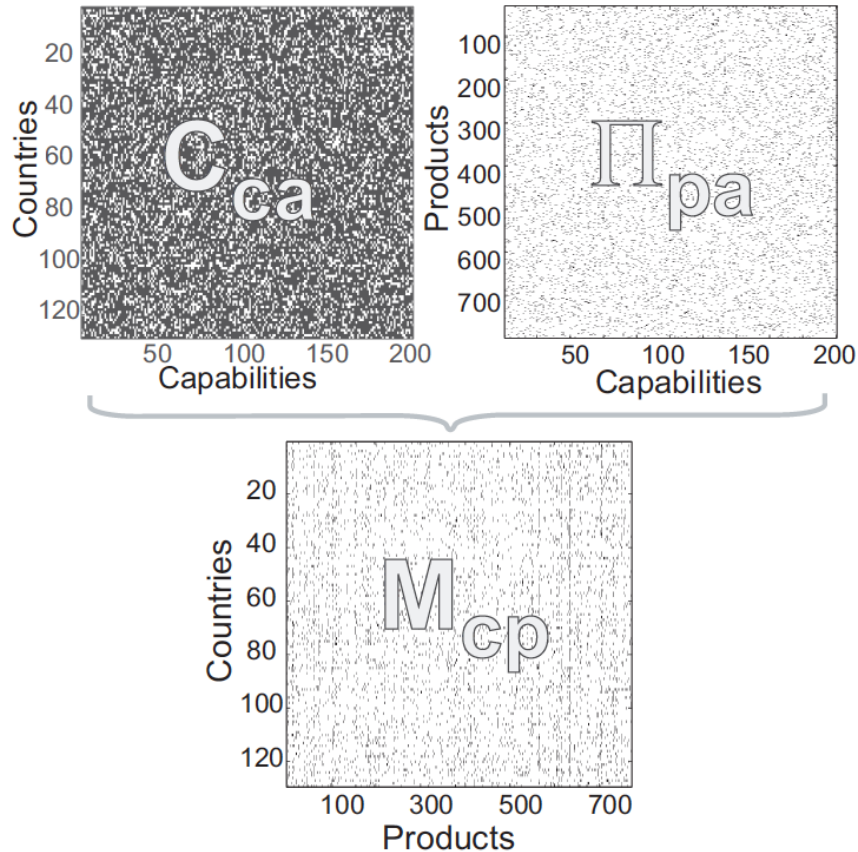
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$$X^t = C^t \odot P$$

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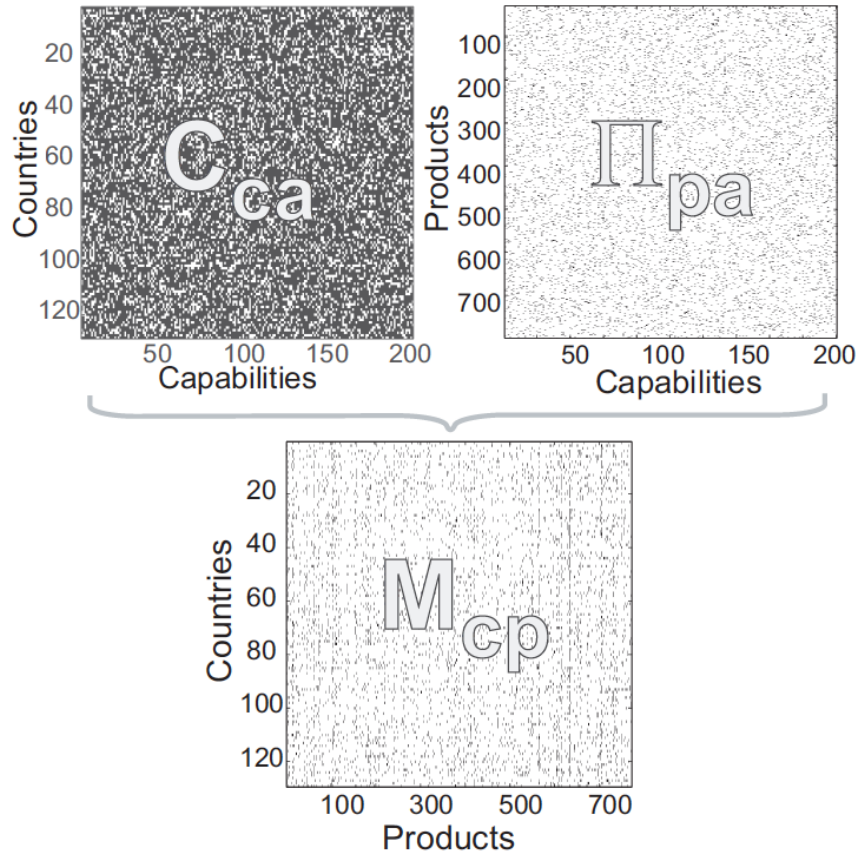


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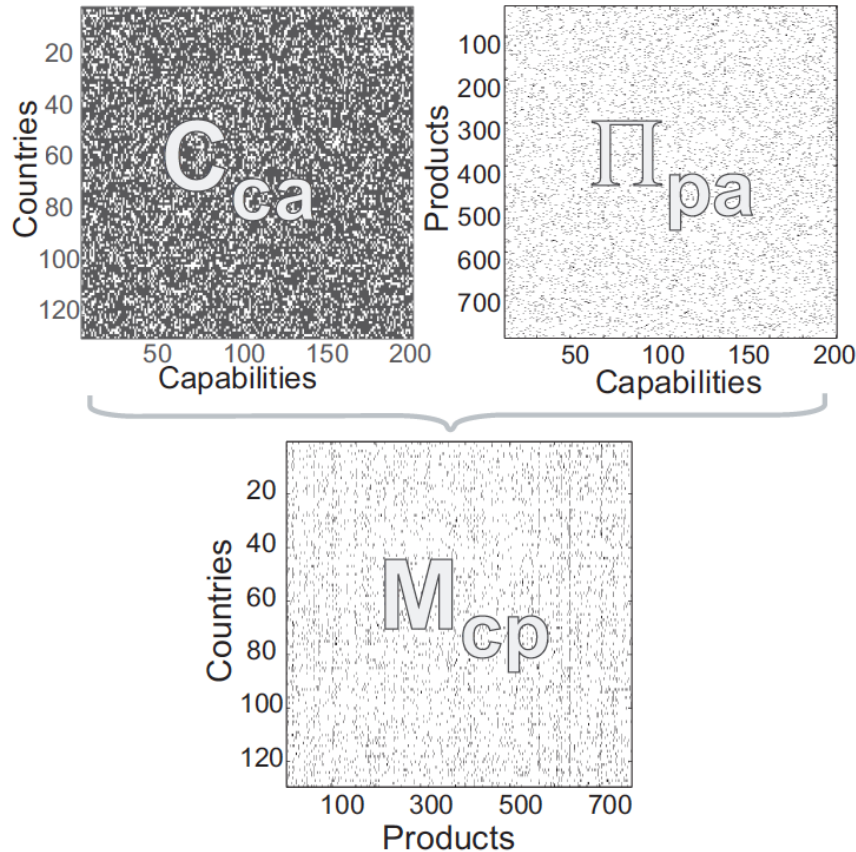
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# Model (currently):

1. The matrix  **$C$**  starts empty.
2. The matrix  **$P$**  is Bernoulli-filled (with prob.  $q$ ), and fixed.
3. For each city (i.e., row) in  **$C$** ,
  - i. a random skill is added,
  - ii. and lost with a certain probability.
4. We calculate the matrix  **$X$**  of output

# Will the model reproduced data?

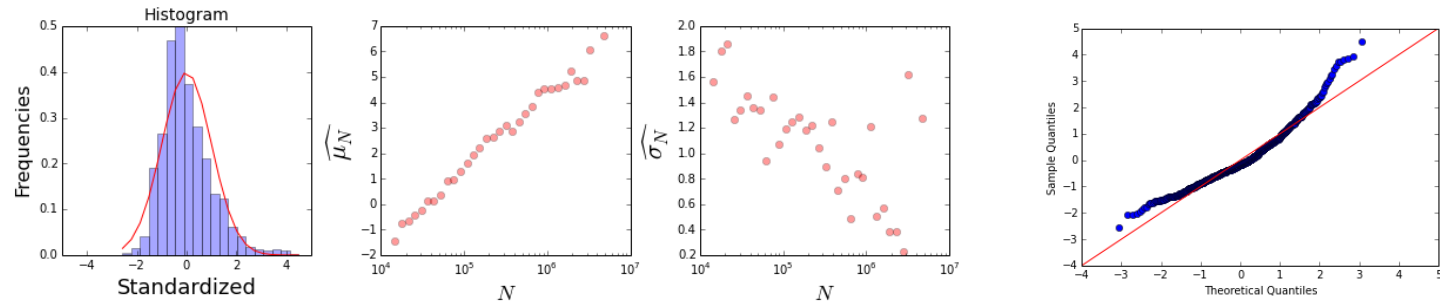
$$\begin{aligned} y_c^t &= \sum_{p=1}^{N_p} X_{cp}^t \\ pop_c &= \sum_{a=1}^{N_a} C_{ca} \end{aligned} \quad \left. \vphantom{\begin{aligned} y_c^t &= \sum_{p=1}^{N_p} X_{cp}^t \\ pop_c &= \sum_{a=1}^{N_a} C_{ca} \end{aligned}} \right\} \Pr(y_c^t \mid pop_c) \sim \text{Lognormal}$$



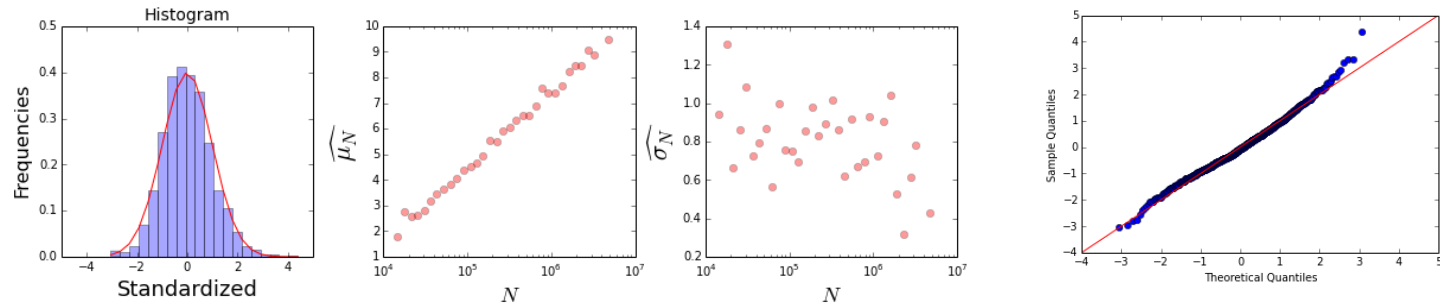


# Profiles of $P(Y|N)$

Null Model:



Inventors:



H&H Model

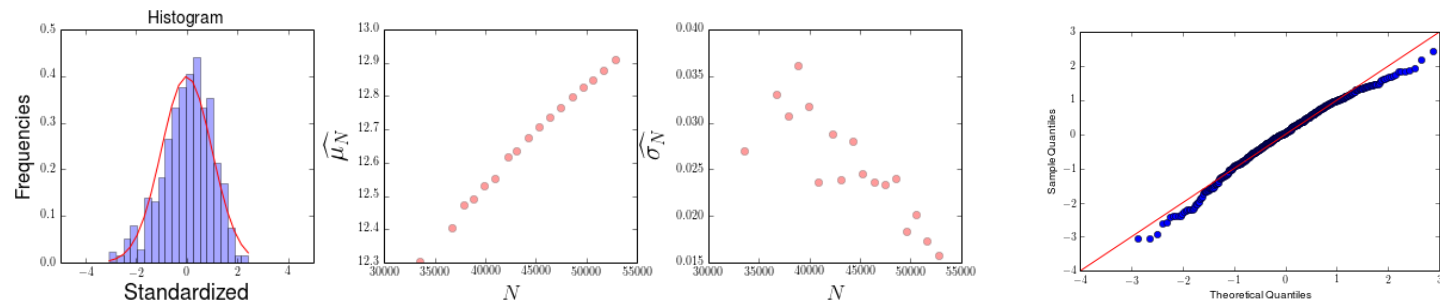




Table 1: US urban areas, sorted by population size, that have counts, in inventors or creatives, that are outside a  $z = 3.07$  *sigma* interval around the log-meann, i.e.,  $y \notin [e^{\mu(n)-z\sigma(n)}, e^{\mu(n)+z\sigma(n)}]$ . The numbers shown in the Population and Inventors columns are estimated averages from 2008-2010, although all values representing counts have been rounded to the nearest integer. The value  $z = 3.07$  corresponds to a log-deviation such that  $1 - \Phi(z) < 1/938$ . The random variables and their values are for inventors  $I$  and  $i$ , for creatives  $C$  and  $c$ , respectively

Name of Urban Area	Population ( $n$ )	Inventors ( $i$ )	Most probable $I$	$P(I \geq i n)$	Creatives ( $c$ )	Most probable $C$	$P(C \geq c n)$
Los Alamos, NM (Micro)	17,899	213	1	0.	5,502	1,697	0.
Mountain Home, ID (Micro)	26,926	1,027	2	0.	2,801	2,751	0.562
Clewiston, FL (Micro)	39,109	8	4	0.626	1,976	3,848	0.999
Clovis, NM (Micro)	47,009	1	5	0.999	4,504	4,461	0.577
Eagle Pass, TX (Micro)	53,392	1	7	1.	4,328	5,144	0.834
Palm Coast, FL (Metro)	94,755	44	15	0.426	4,040	8,958	1.
Lake Havasu City-Kingman, AZ (Metro)	200,447	43	43	0.82	10,940	21,603	0.999
Merced, CA (Metro)	253,198	37	59	0.924	13,650	27,633	0.999
Ocala, FL (Metro)	330,780	69	87	0.873	17,620	37,218	1.
Durham-Chapel Hill, NC (Metro)	498,511	1,692	155	0.028	128,900	57,900	0.001
McAllen-Edinburg-Mission, TX (Metro)	758,064	36	279	0.999	60,400	87,582	0.963
San Jose-Sunnyvale-Santa Clara, CA (Metro)	1,818,864	24,531	952	0.	396,820	240,275	0.033



# Example

<b>C</b>	a1	a2	a3	a4	a5				
city 1	1	1	0	0	1	}			
city 2	0	1	0	1	1				
<b>P</b>	a1	a2	a3	a4	a5		<b>M</b>	p1	p2
p1	0	1	0	0	1		city 1	1	0
p2	0	0	0	1	0		city 2	1	1

- City 1 has skills {1,2,5}
- City 2 has skills {2,4,5}
- Product 1 requires {2,5}
- Product 2 requires {4}
- City 1 only produces Product 1
- City 2 produces Both



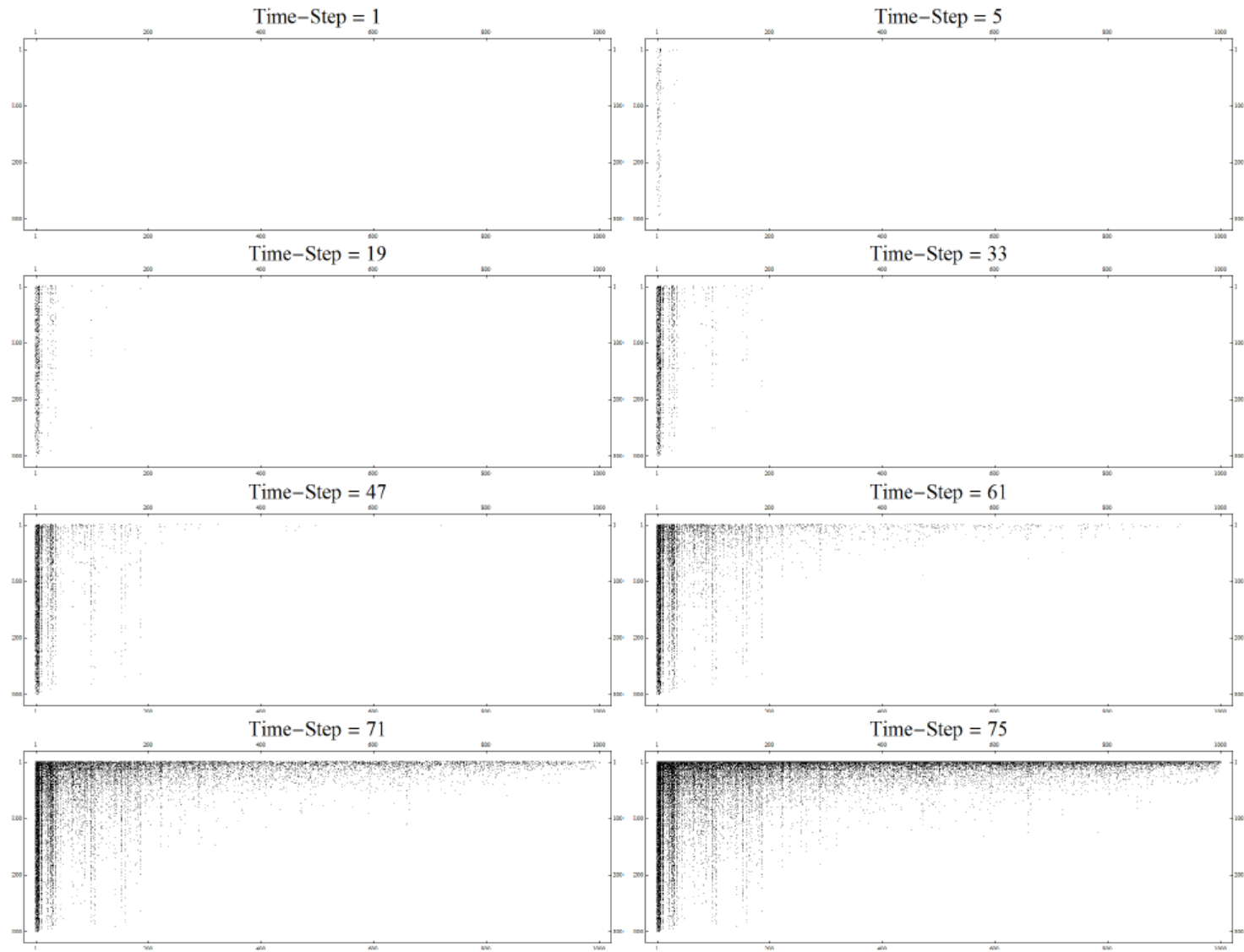


Figure 5: **Model 1: Evolution of the  $M^t$  matrix of cities versus products.** The model assumes a fixed set of products uniquely defined by their set of capabilities. In each time-step, each city acquires a new random capability previously missing. If the capability allows the city to produce a new product, the capability is fixed, otherwise it is lost (until it acquires it again by chance in the future). Here,  $N_c = 300$ ,  $N_p = 1000$ ,  $N_a = 50$ ,  $r = 0$ , and  $q = 0.13$ . The  $M_{cp}$  elements are sorted in the same order throughout all steps from most diversified city to the least, and from the most ubiquitous product to the least, as they appear in the last time-step.



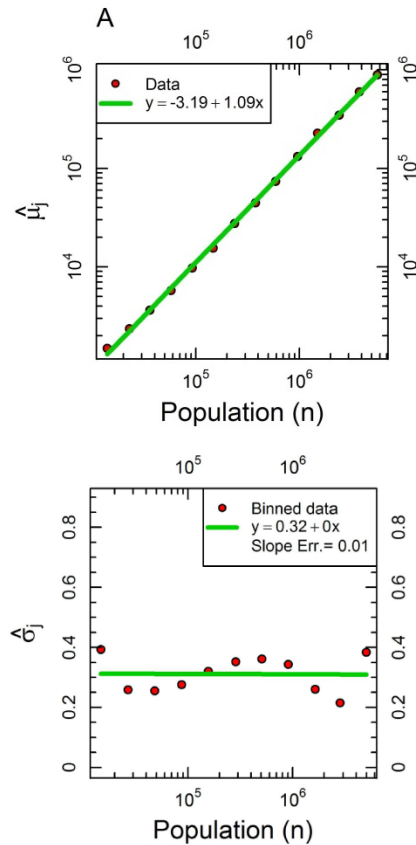
The following relationship has to hold if we want a confidence  $(1-\alpha)$  that  $Y/N$  is close to the mean of the process

$$(\alpha\epsilon^2)N^r \geq c(r) \left( N\mathbb{E}[y_1^r] + N^{r/2}\sqrt{\text{Var}[y_1]} \right)$$

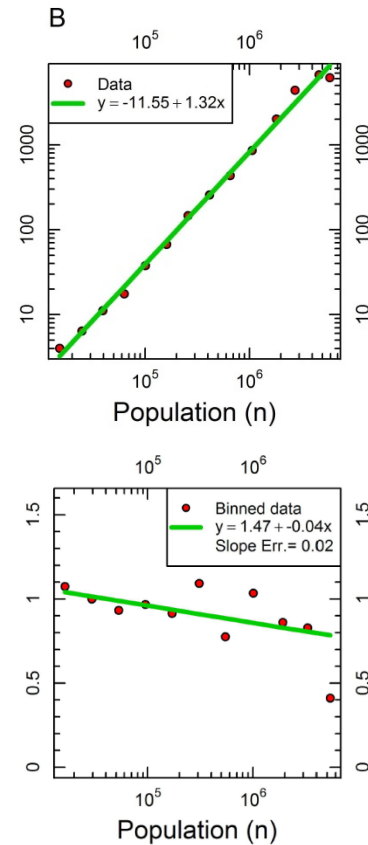


# Population size dependence

Creatives

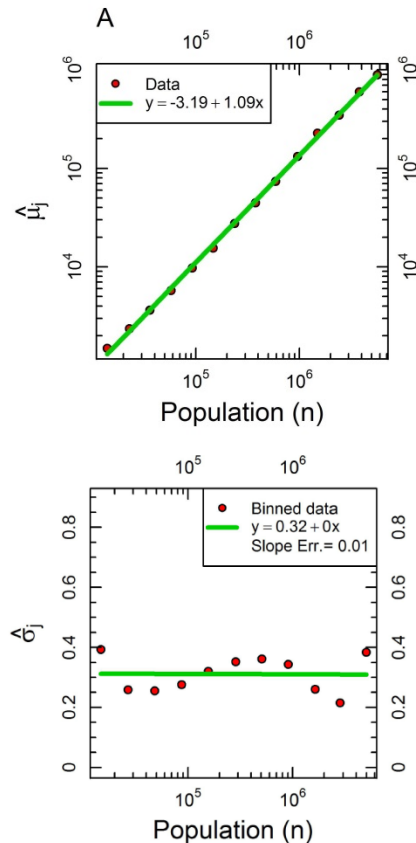


Inventors

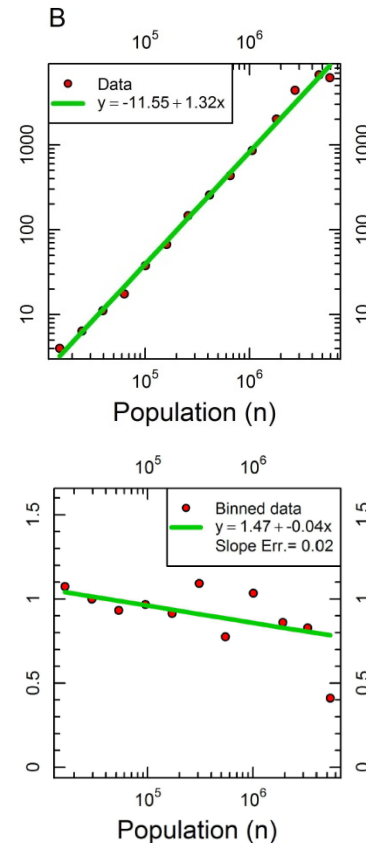


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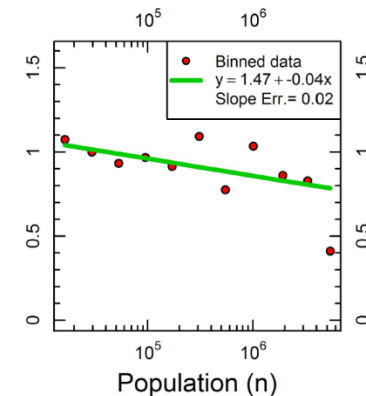
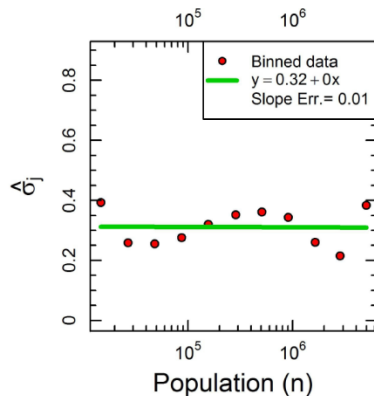
Creatives



Inventors

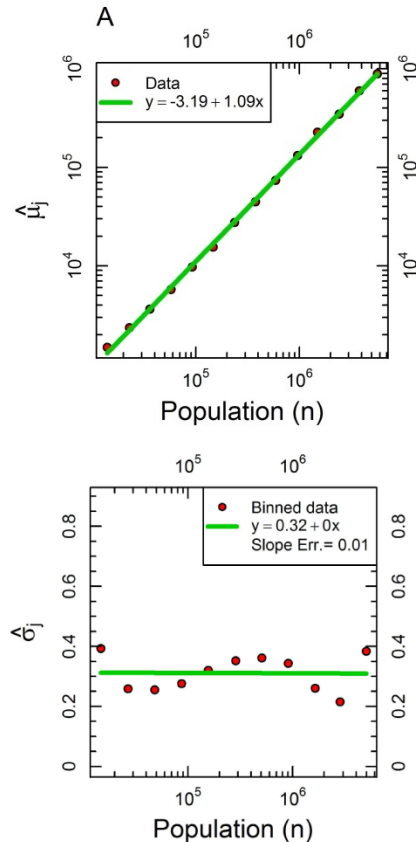


- Note this way we have completely characterized

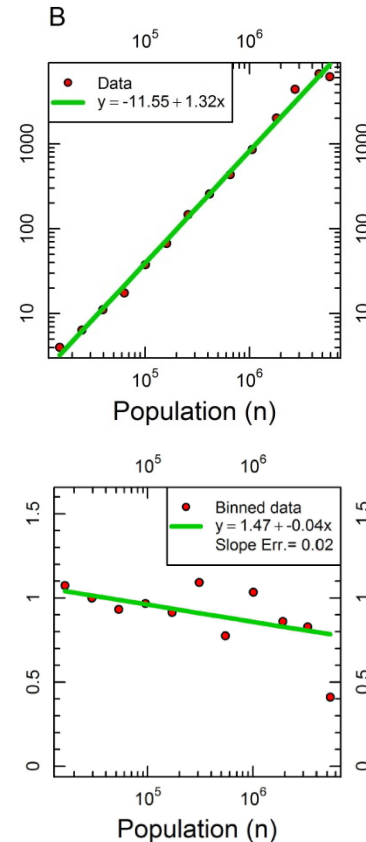


# Population size dependence

Creatives



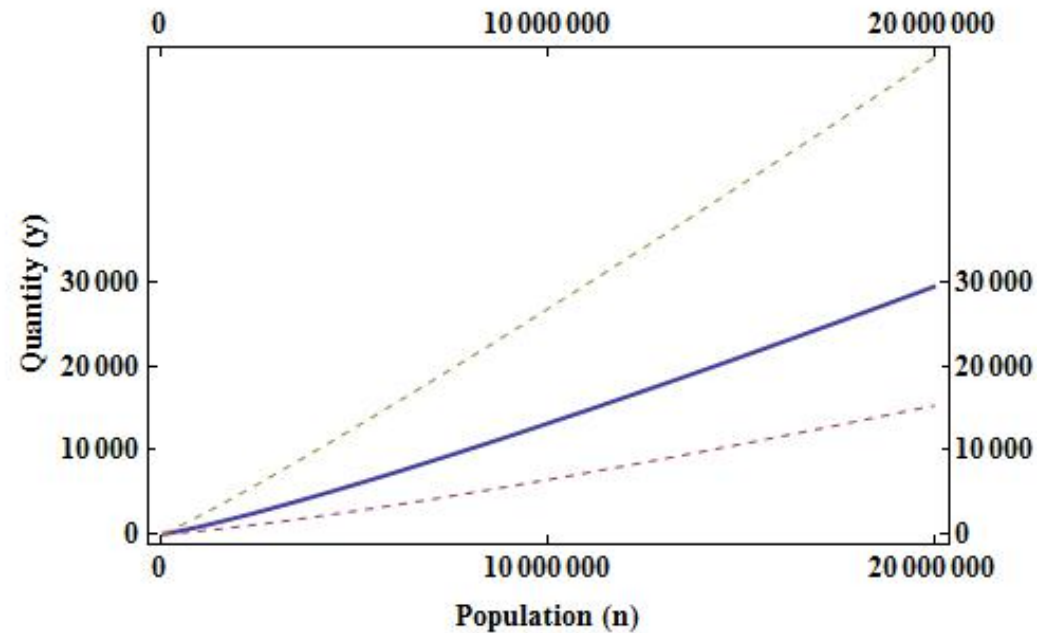
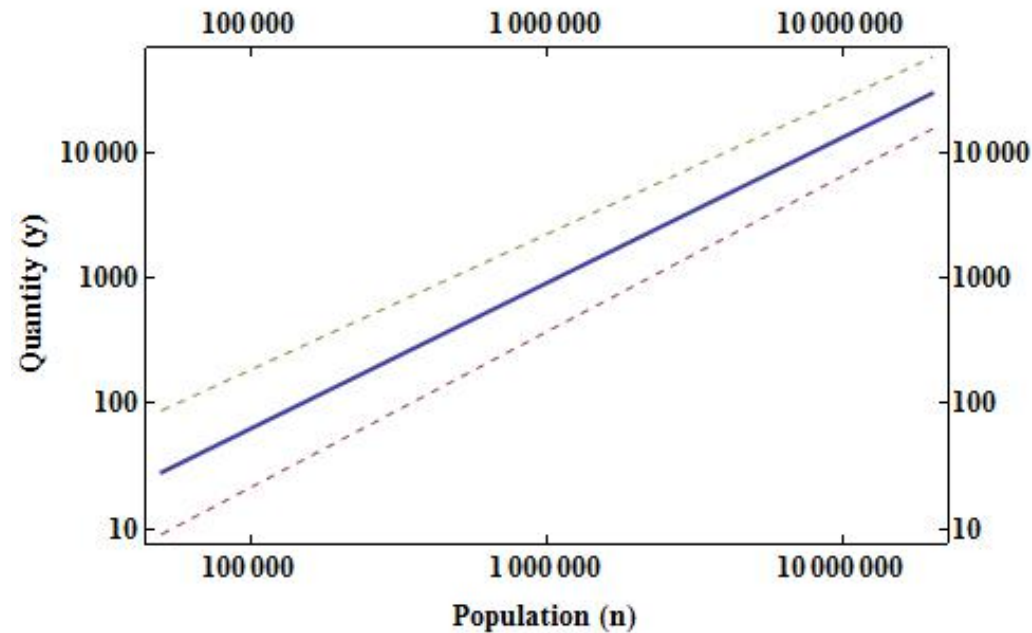
Inventors



- Note this way we have completely characterized  $P(Y,N)=P(Y/N)P(N)$



# *Logarithmic* convergence vs. *Absolute* divergence of cities





# Urban Economies and Occupation Space: Can They Get “There” from “Here”?

**Rachata Muneeppeerakul<sup>1,2\*</sup>, José Lobo<sup>1</sup>, Shade T. Shuttters<sup>1,3</sup>, Andrés Gómez-Liévano<sup>2,4</sup>, Murad R. Qubbaj<sup>1</sup>**

**1** School of Sustainability, Arizona State University, Tempe, Arizona, United States of America, **2** Mathematical, Computational, and Modeling Sciences Center, Arizona State University, Tempe, Arizona, United States of America, **3** Center for Social Dynamics and Complexity, Arizona State University, Tempe, Arizona, United States of America, **4** School of Human Evolution and Social Change, Arizona State University, Tempe, Arizona, United States of America

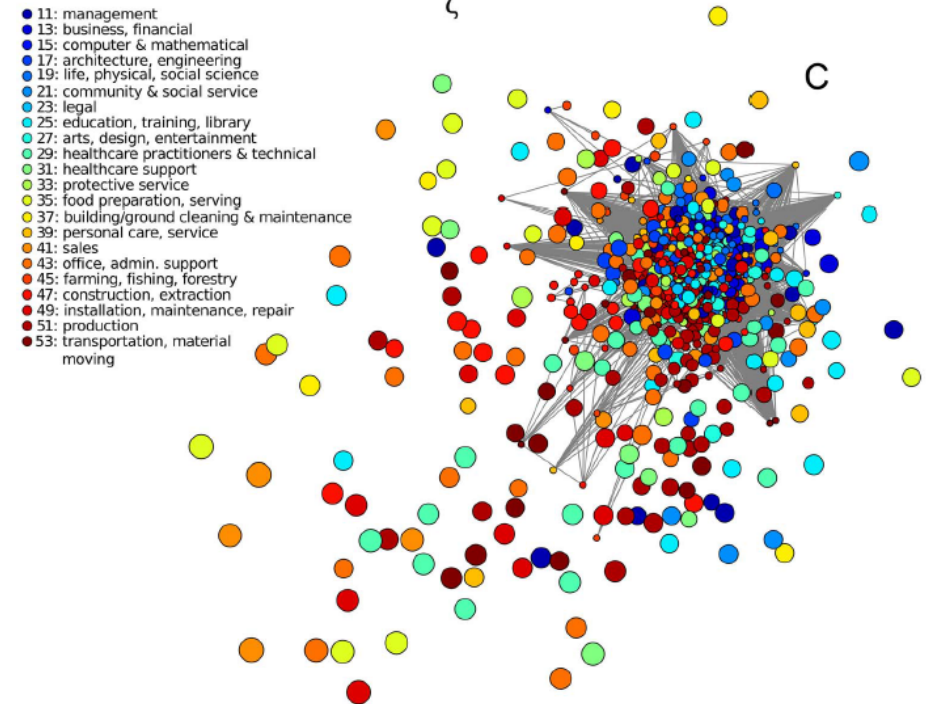
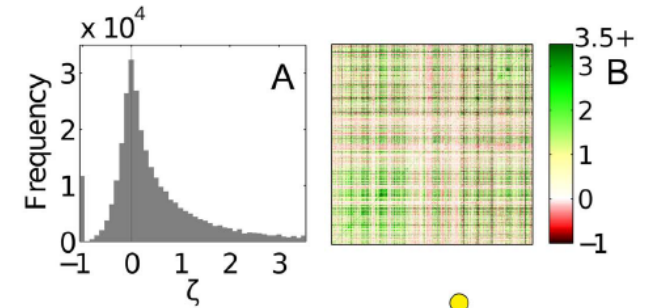
## Abstract

Much of the socioeconomic life in the United States occurs in its urban areas. While an urban economy is defined to a large extent by its network of occupational specializations, an examination of this important network is absent from the considerable body of work on the determinants of urban economic performance. Here we develop a structure-based analysis addressing how the network of interdependencies among occupational specializations affects the ease with which urban economies can transform themselves. While most occupational specializations exhibit positive relationships between one another, many exhibit negative ones, and the balance between the two partially explains the productivity of an urban economy. The current set of occupational specializations of an urban economy and its location in the occupation space constrain its future development paths. Important tradeoffs exist between different alternatives for altering an occupational specialization pattern, both at a single occupation and an entire occupational portfolio levels.

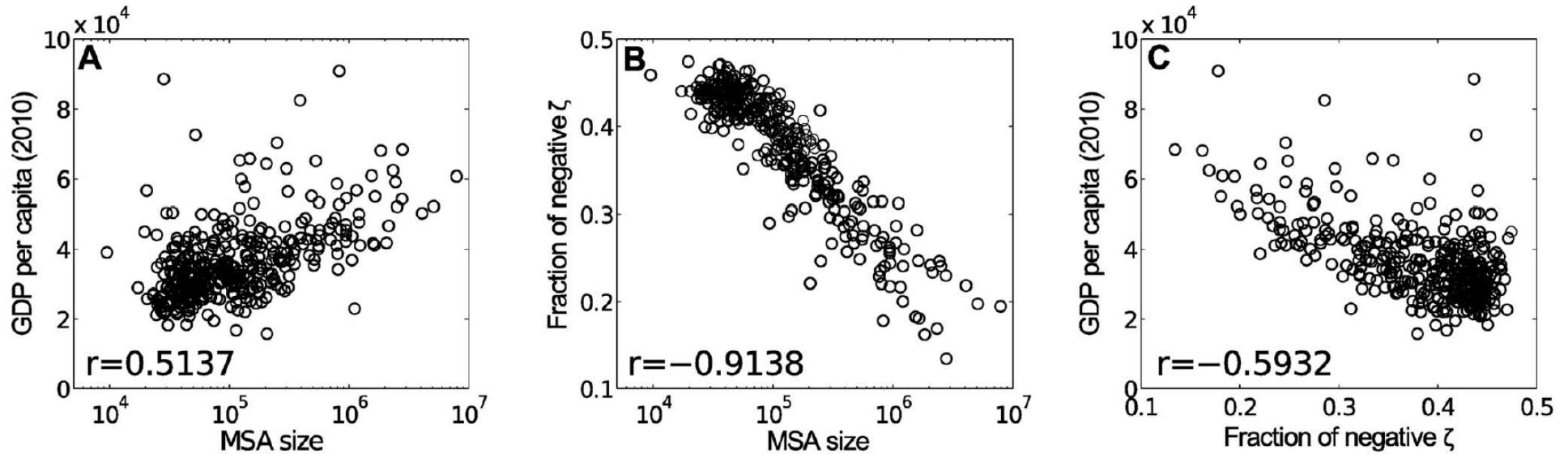
# Occupational “interdependencies”

- Definition:

$$\zeta_{ij} = \frac{P[LQ_i^{(M)} > 1, LQ_j^{(M)} > 1]}{P[LQ_i^{(M')} > 1]P[LQ_j^{(M'')} > 1]} - 1$$



# “Harmony” between occupations



# Scaling relationships

(*sensu* Stat. Mech.)

$$X_1|N \quad \sim \quad \mathcal{LN}(\mu_1, \sigma_1^2)$$

$$X_2|N \quad \sim \quad \mathcal{LN}(\mu_2, \sigma_2^2)$$

$$X_2 = f(X_1)$$

$$\Rightarrow X_2 = aX_1^\beta, \beta = \frac{\sigma_2}{\sigma_1}$$

$$X_1 \quad \sim \quad \mathcal{P}(\alpha)$$

$$\Rightarrow X_2 \sim \mathcal{P}(\tau), \tau = \frac{\alpha}{\beta}$$