

Agglomeration and Spatial Interaction

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Master in Business Analytics

18/04/2020

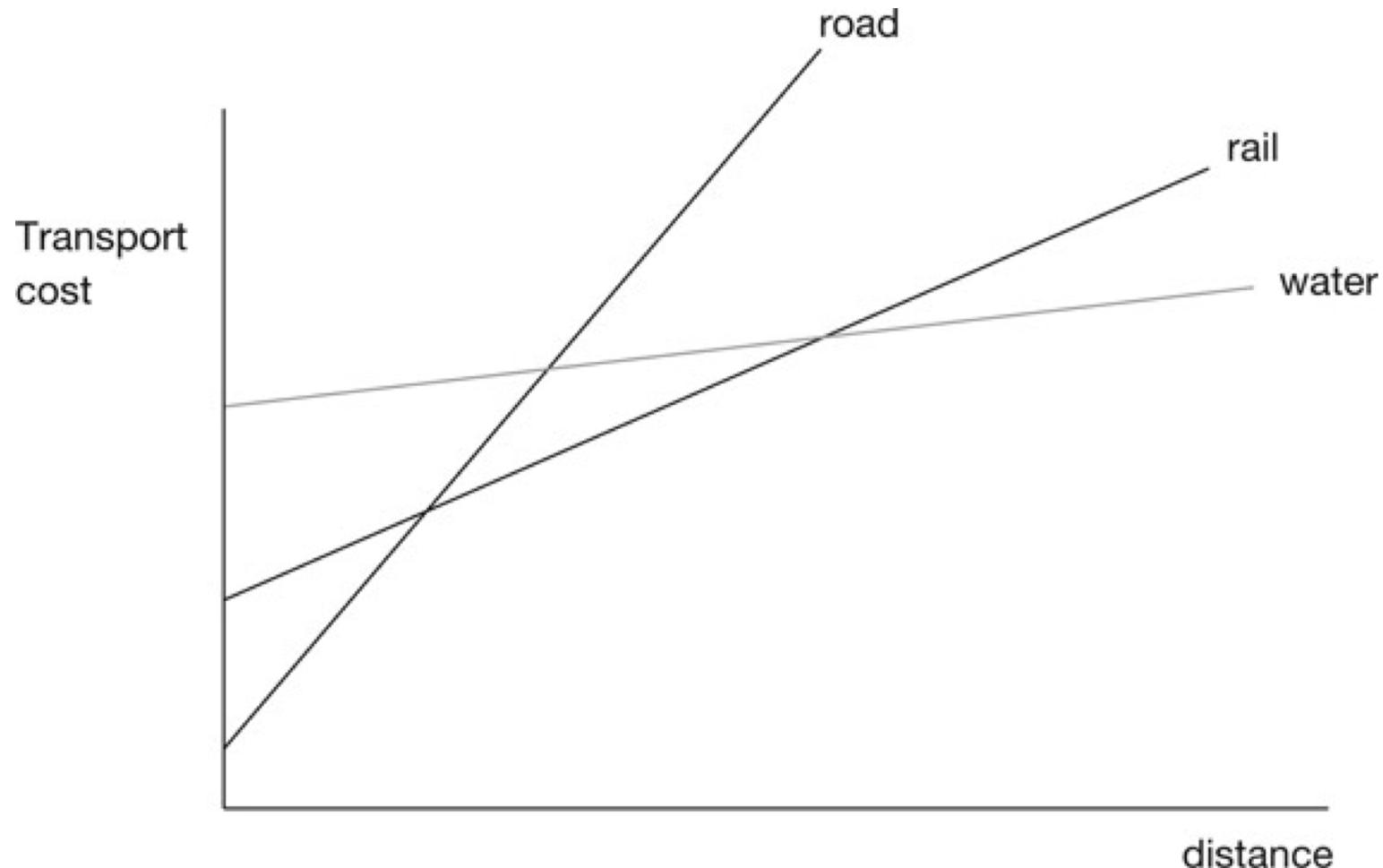
How did we get to live in cities?

- From a nomad to a sedentary society

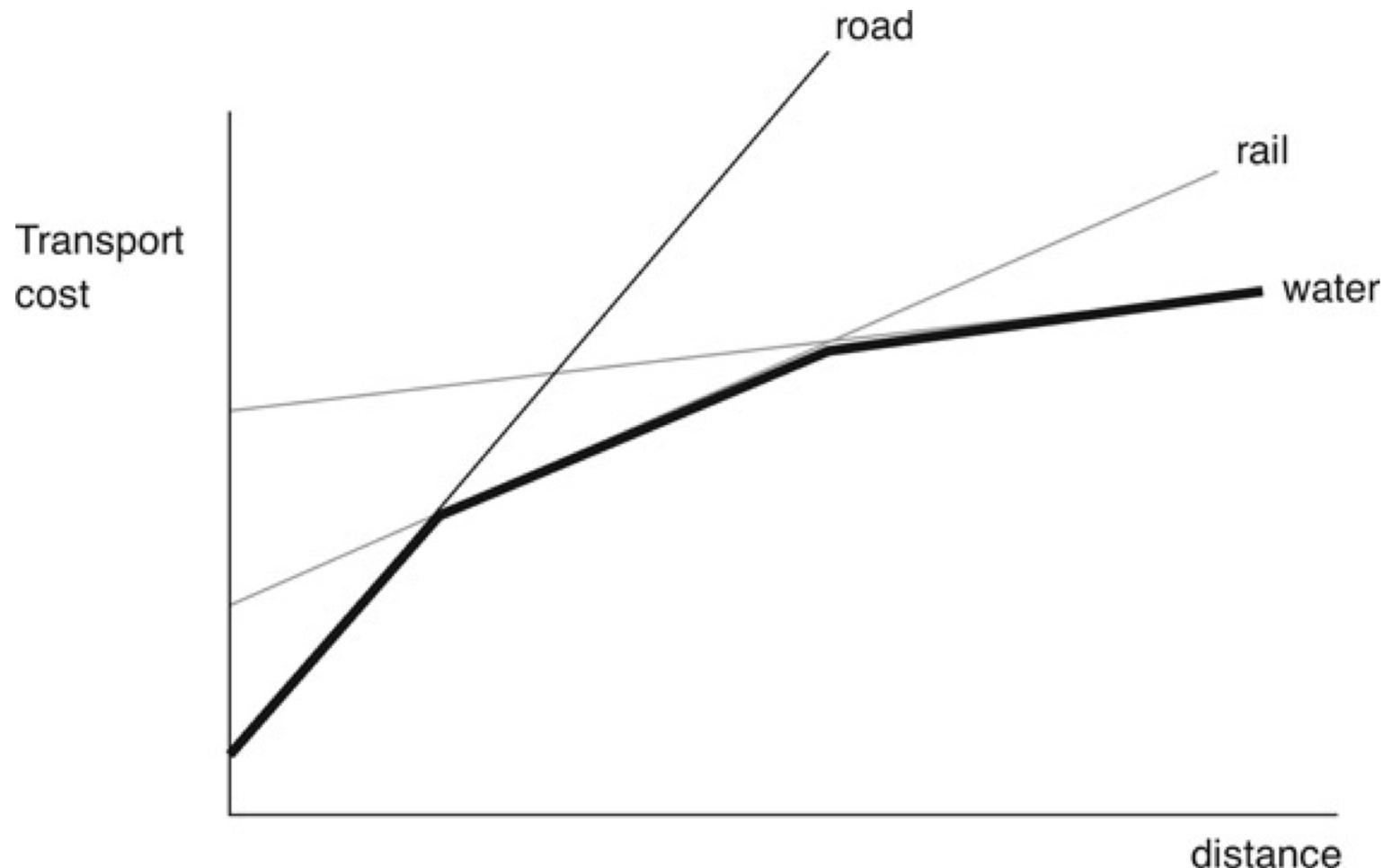


- Ideas of what happened later?

Transportation Costs



Transportation Costs



Transportation Costs

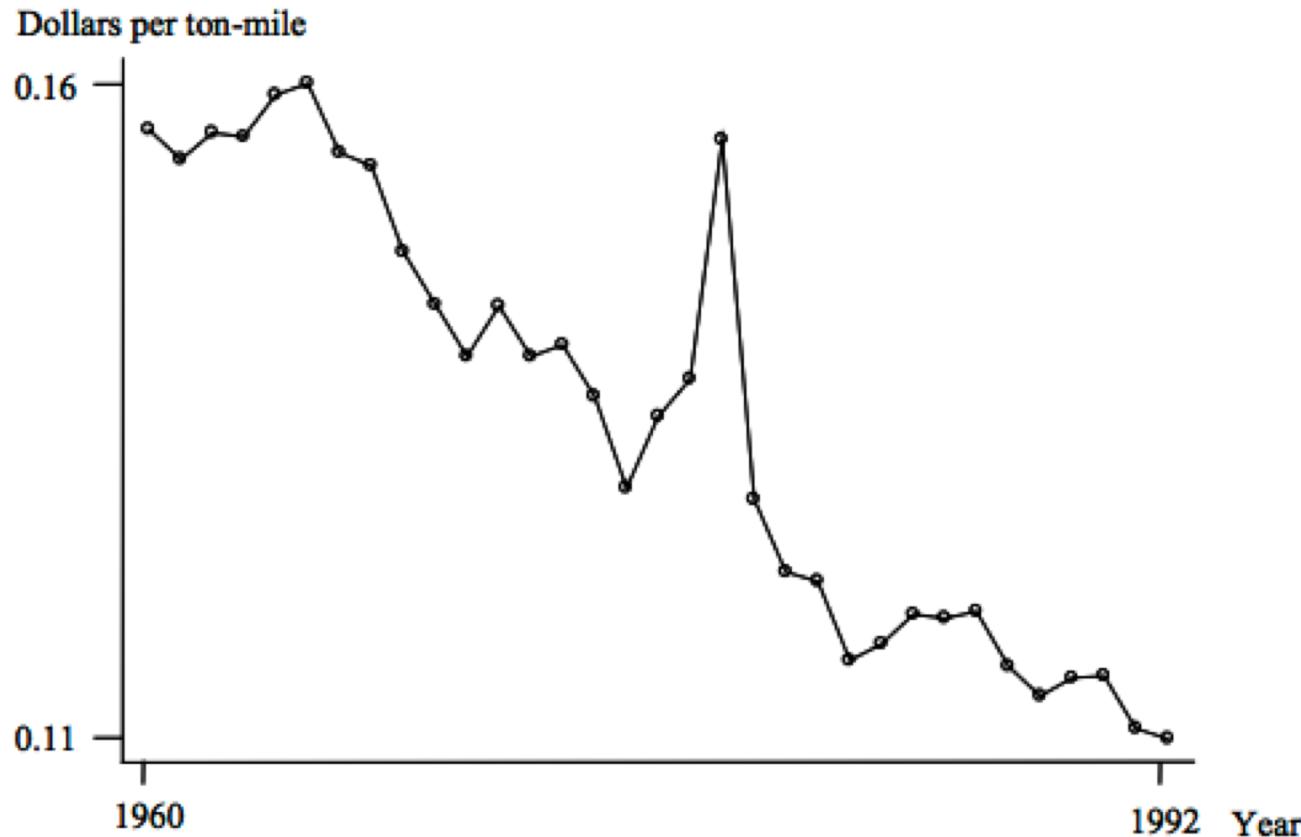


Fig. 5. Revenue per ton-mile, all modes together

Source: Bureau of Transportation Statistics Annual Reports

Transportation Costs

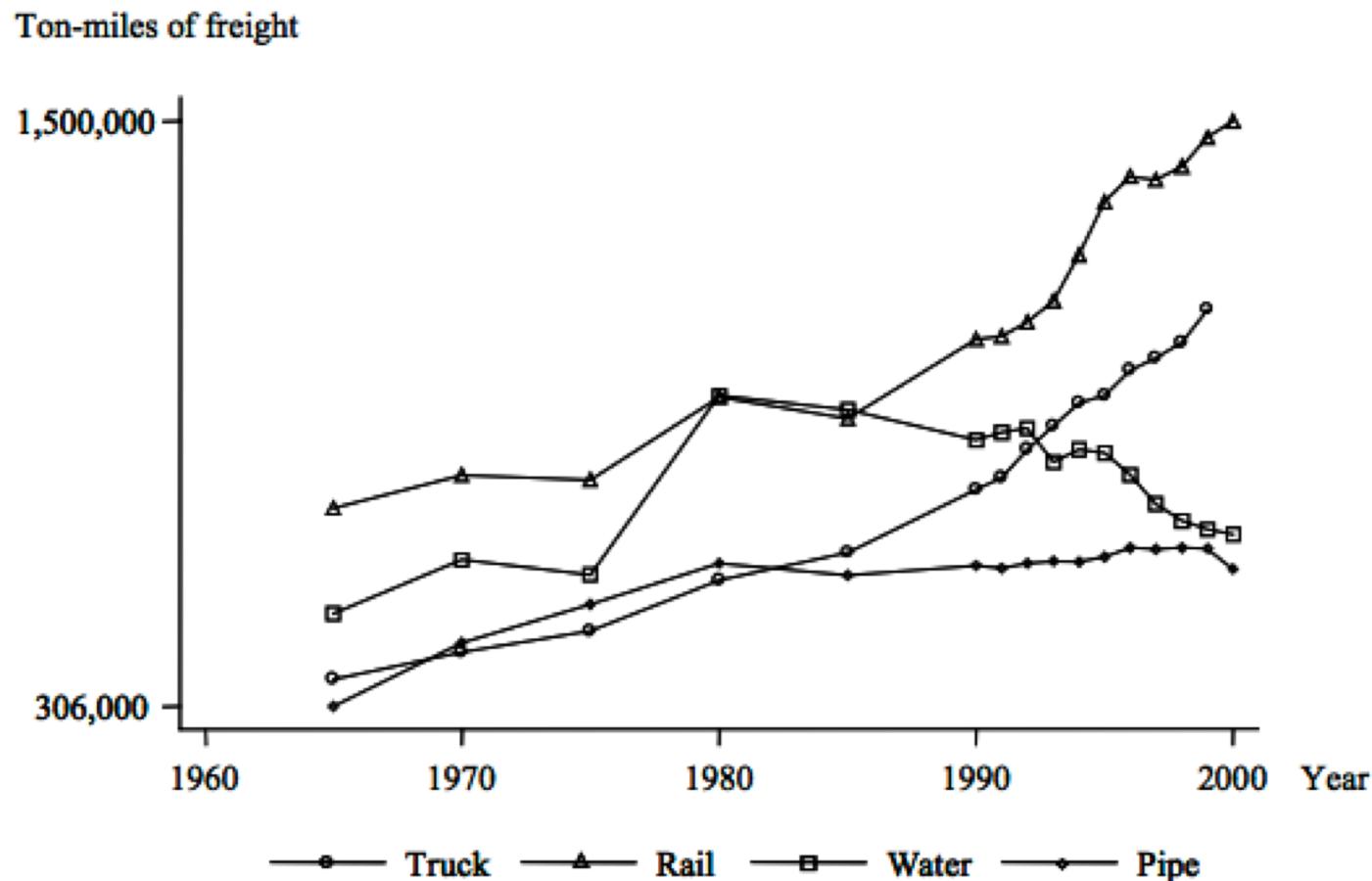


Fig. 6. Ton-miles of freight over time

Source: Bureau of Transportation Statistics Annual Reports

trade

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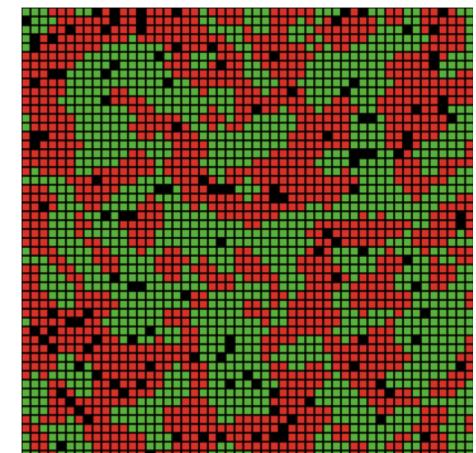
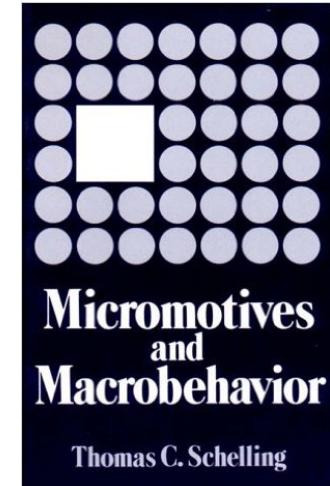


Learning Objectives

1. Agglomeration Economies
2. Agglomeration as a human condition
3. Types of Agglomeration Economies
4. Empirical Evidence
5. Markets and Spatial Markets

The Schelling Model

- A 1978 model of spatial segregation
 - Book: Micro-motives and Macro-behaviors
- Micro-motive: People move if they are unhappy
- Macro-behavior: Spatial segregation
- Have you seen any other patterns?



Homophily = Love for the similar

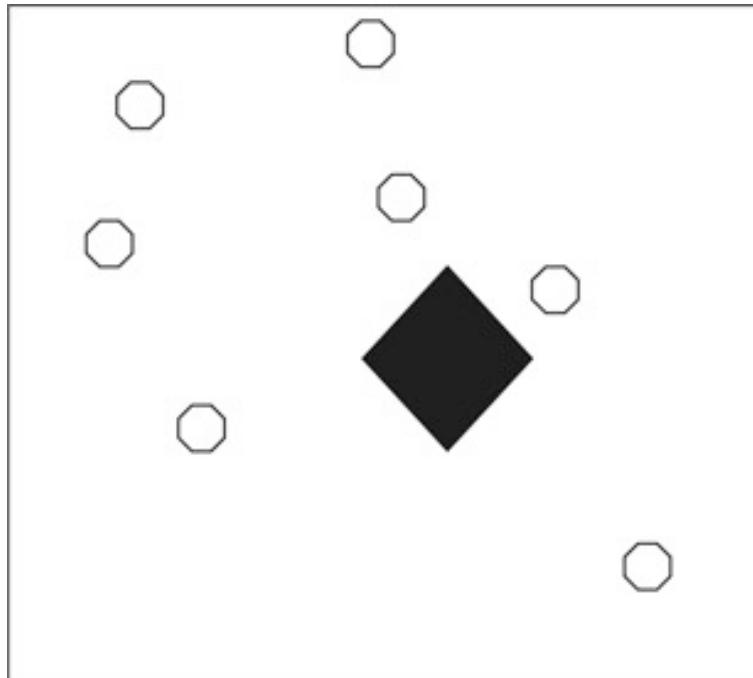
- Some spatial configurations we see come from individual preferences
 - How much these preferences reveal more than just social constructs?
- Is the Schilling model telling us something about economics as well?
- Is there a version of this in the economic world?

Agglomeration Economies

- Can happen at:
 - the social level & the economic level
- Why?: Benefits of agglomeration
- Economies of Scale: *Internal*
 - Def.: Advantages from the size of production per firm
 - division of labor / factory system → Pins
 - Capital Indivisibility → farming
 - Statistical scale Economies → shoe sizes

“if production is at a larger scale it will be concentrated at fewer points in the landscape”

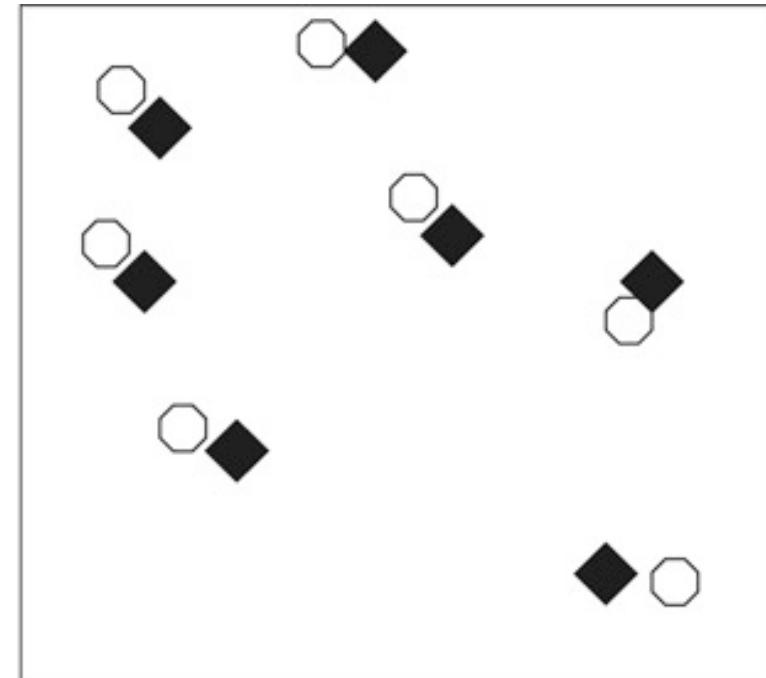
Agglomeration Economies



Minimum production cost

○ town

◆ factory



Minimum transportation cost

The Amazon logo, featuring the word "amazon" in white lowercase letters above a yellow smile icon.

der online. Pick

2

COME IN WE'RE
OPEN
AND WAITING

107

amazon.com/islav

6

7

a



ORE



Agglomeration Economies

- Economics of Scale: *External*
 - *Def.*: Advantages from the size of an industry or groups of firms
 - *Main benefit*: combined scale of production in an industry
 - Industrial complexes (collaboration)
 - The Akron tire cluster:
 - Does anyone knows about this?

Types of Agglomeration Economies

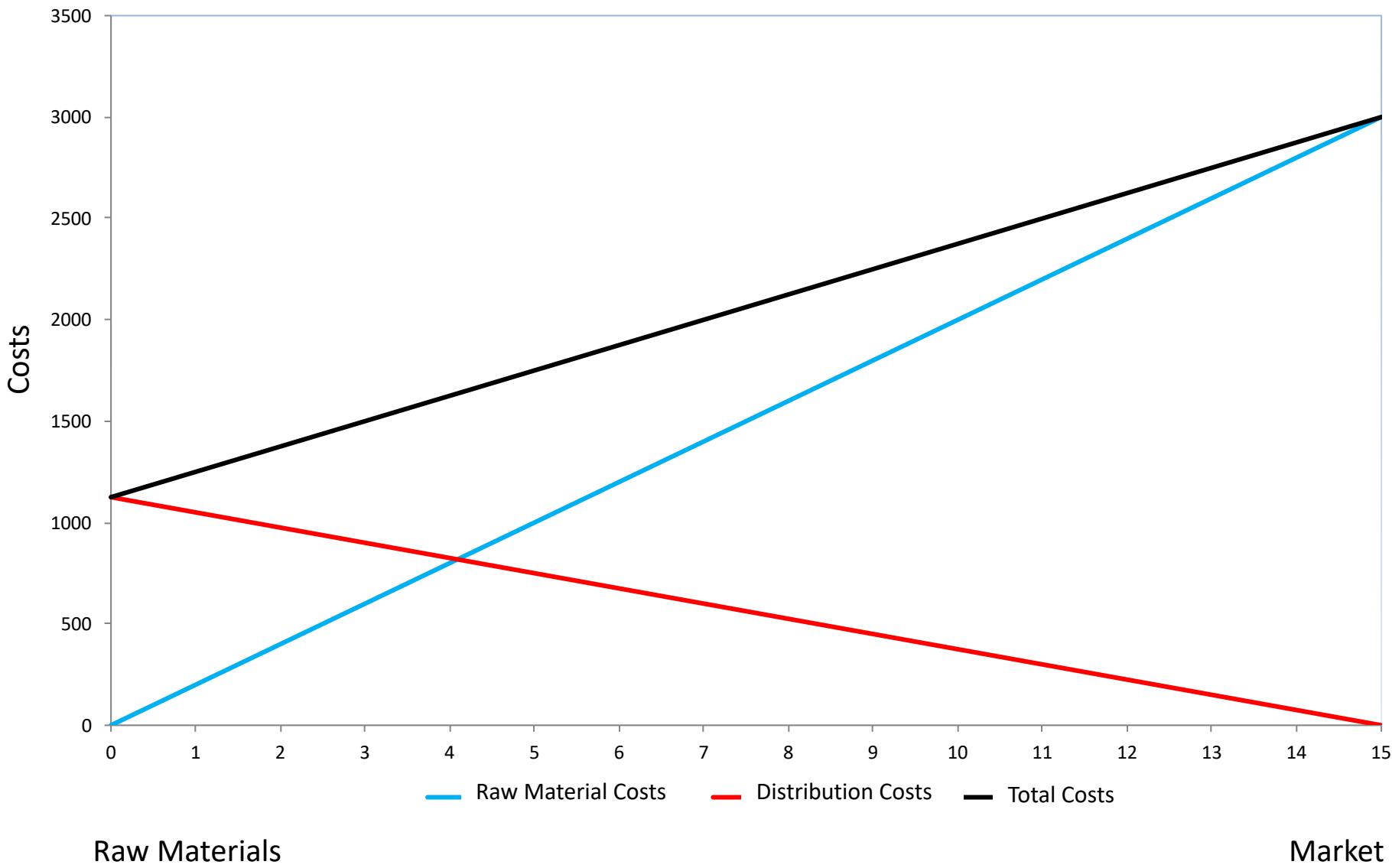
- Urbanization Economies
 - Sewer pipes
 - Juxtaposition economies
 - Diversity/Creative Class
- Localization Economies
 - Labor sorting/skills
- Natural advantage
 - Natural and localization economies are not mutually exclusive

Resource Oriented vs. Market Oriented

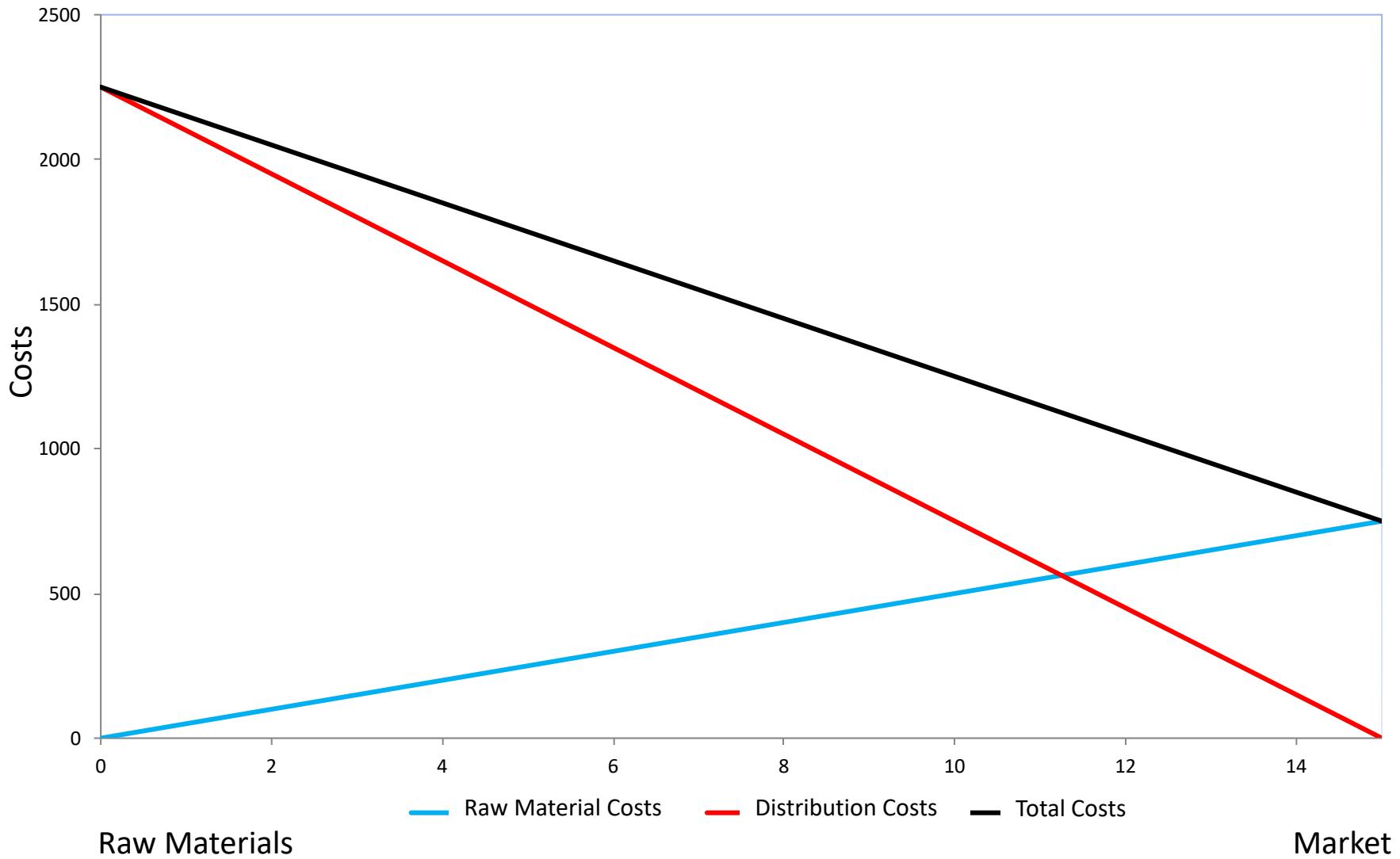
- *Type of Firms:* Transportation Costs are the main factor to decide a location
- *Objective:* Minimize Total Costs
- *Type of Cost:* It doesn't matter the brut cost, but the “monetary weight”
 - Weight Input * Trans. Cost
 - Weight Product * Trans. Cost

	Resource Oriented	Market Oriented
Type of Activity	Weight loss	Weigh gaining
High costs in	Raw Materials (RM)	Distribution (D)
Costs differences	RM > D	D > M
Distance to the market	High	Low

Resource Oriented Firms



Market Oriented Firms

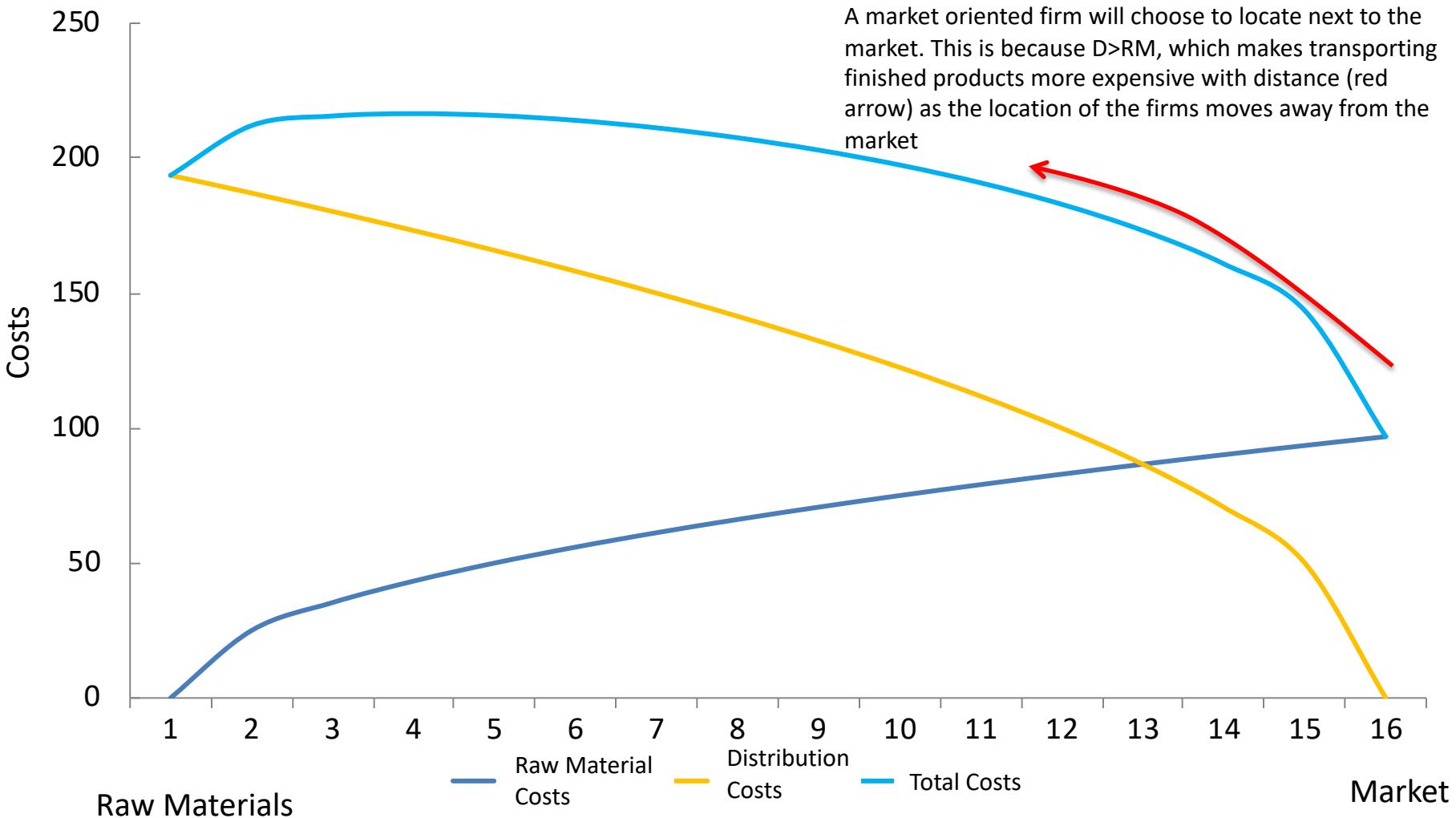


Economies of Scale in Transportation

- The medium costs *decrease* with distance
 - Bureaucracy fixed costs
 - Efficiency of new technologies
- Examples?
- *This reinforce the tendency of businesses to locate next to:*
 - *Inputs -> Resource oriented firms because RM>D*
 - *These firms will prefer transporting finished products (Distribution costs) rather than raw materials (RM)*
 - *Markets -> Market oriented firms because RM<D*
 - *These firms will prefer transporting raw materials instead of finished products*

Market Oriented Firms

With Economics of Scale in Transportation

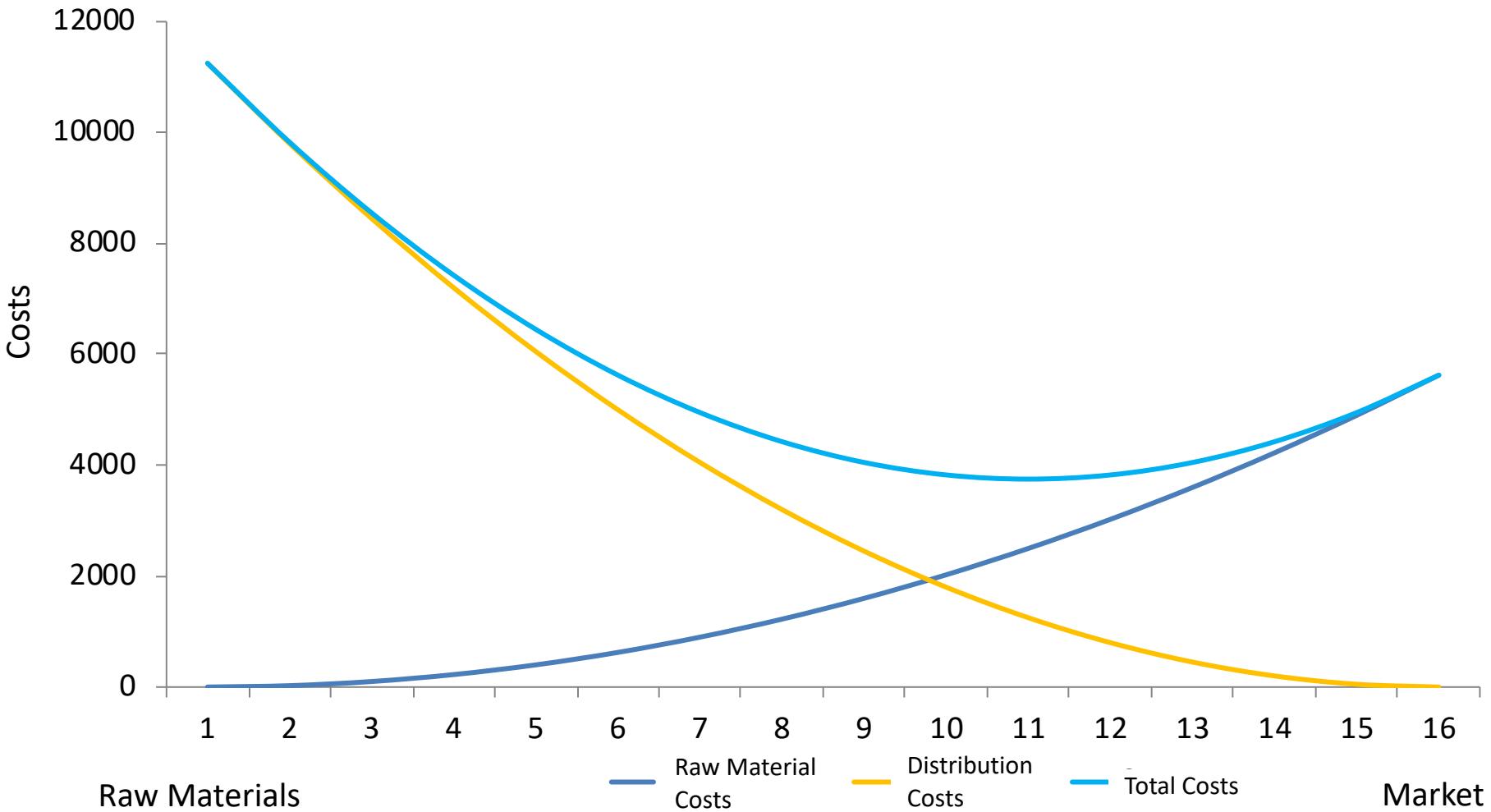


Diseconomies of Scale in Transportation

- The medium costs *increase* with distance
 - Bureaucracy costs increase (border, tolls, etc.)
 - Changes in means of transportation
- Examples?
- *This reinforce the tendency of businesses to locate intermediate to inputs or markets*
 - *This is because as costs increase with distance, shipping long distances of one or the other makes them expensive regardless of their initial costs*
 - *As a result, the firm will prefer to locate where the monetary weight from transporting raw materials equals the one of distributing a product RM = D*

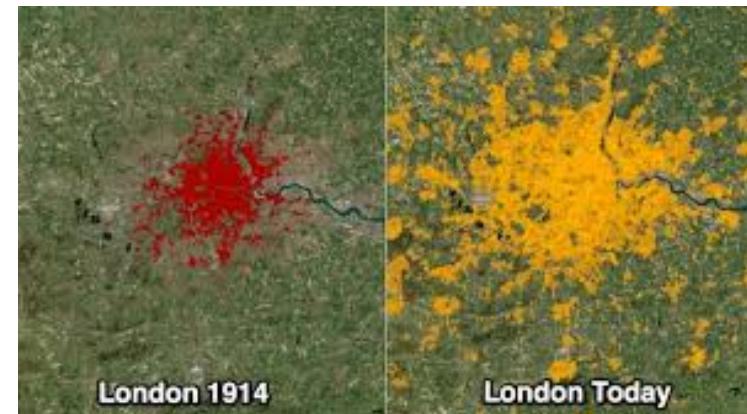
Market Oriented Firms

With diseconomies of Scale in Transportation



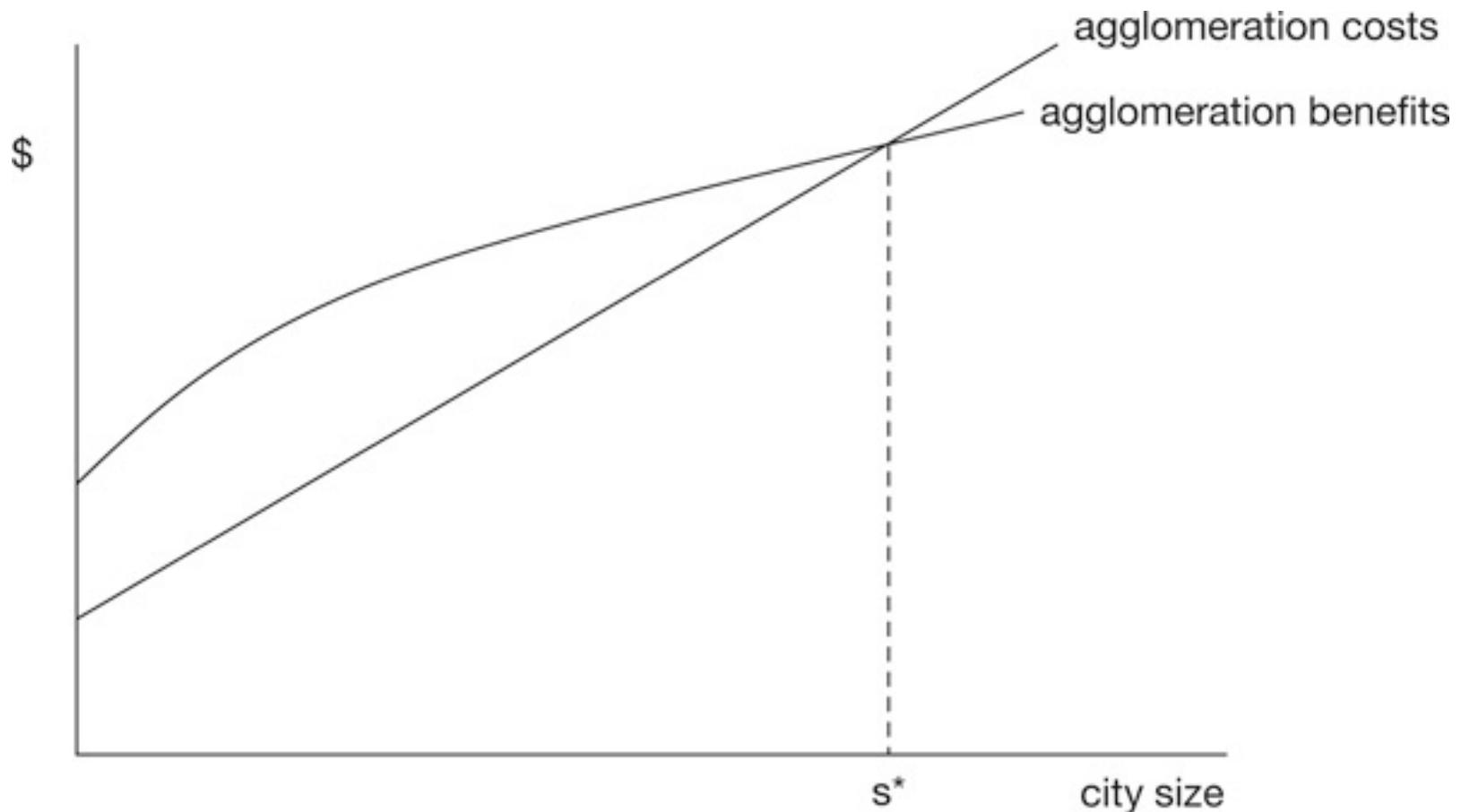
Empirical Evidence

- Cities are the best example of agglomeration economies



Chicago's Expansion??

Is agglomeration always good?



SPATIAL STATISTICS

PART 1

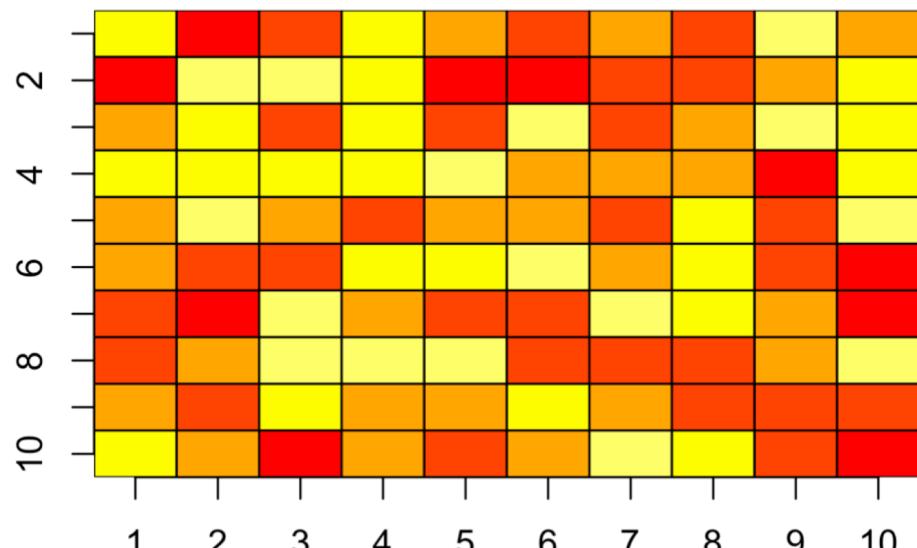
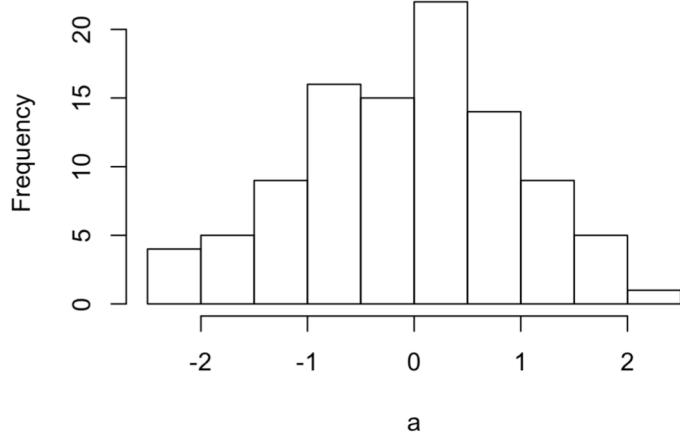
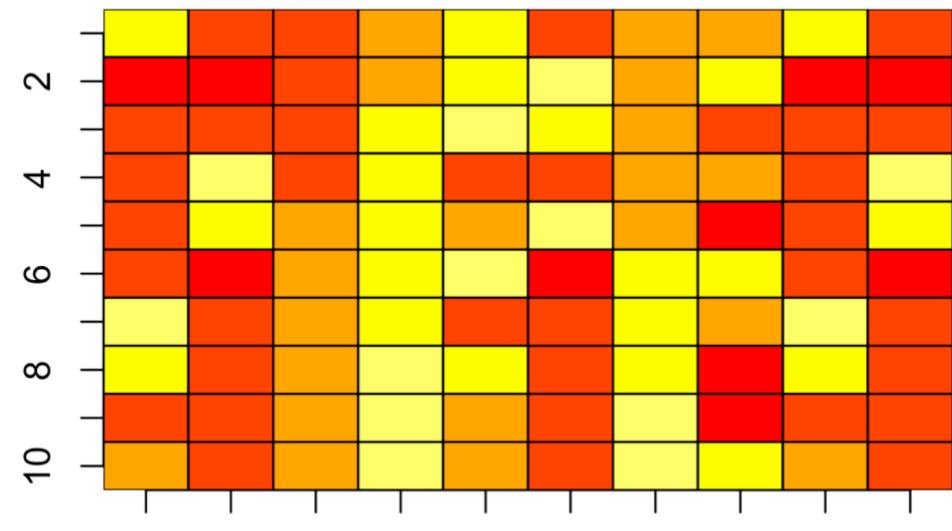
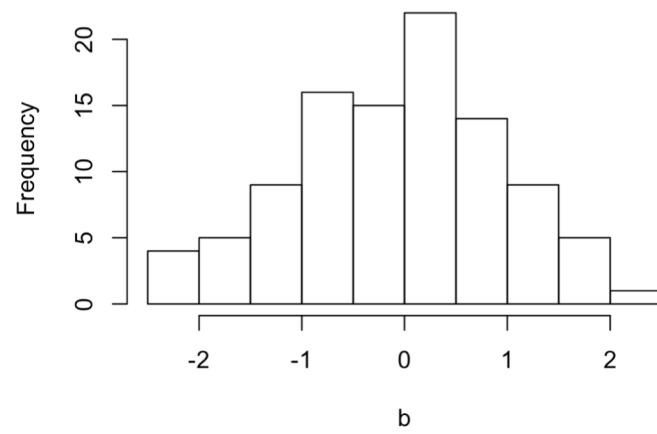
Slides inspired on Luc Anselyn's Lectures on Spatial Data, Spatial Analytics, and Spatial Data Science
Watch Luc's video on: <https://youtu.be/MmCYeJ27DsA>

- Exploratory Spatial Data Analysis (ESDA)
 - Extended EDA to space
 - More than just mapping
 - Mapping – take a variable and plot it (results)
 - ESDA Definition
 - Understand that spatial information as a key component of the data exploration process
 - Focuses on spatial patterns
 - ESDA Activities
 - Describe spatial distributions – dynamic maps
 - Identify atypical observations – spatial outliers
 - Discover patterns of spatial dependence and spatial heterogeneity
 - Spatial clusters, hot/cold spots
 - Spatial structured breaks
 - Spatial aggregation – Functional Labor Market areas

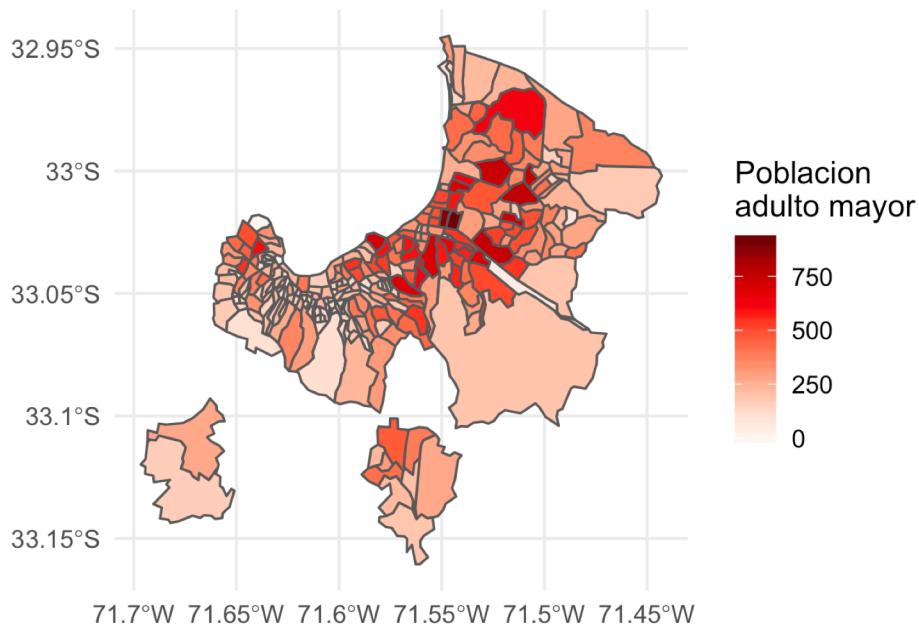
Types of spatial association

- Spatial randomness (null)
- Spatial autocorrelation
 - Clustering pattern vs. Finding Clusters
 - Positive
 - Negative
- Common issues
 - Scale of Aggregation - MAUP
- Statistics
- Spatial weights

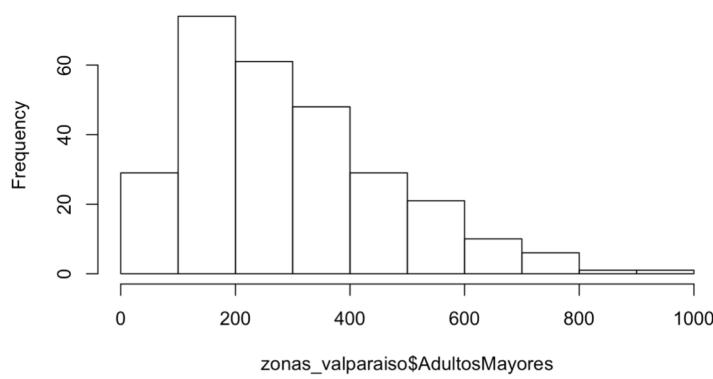
Spatial Randomness?

A**Histogram of a****B****Histogram of b**

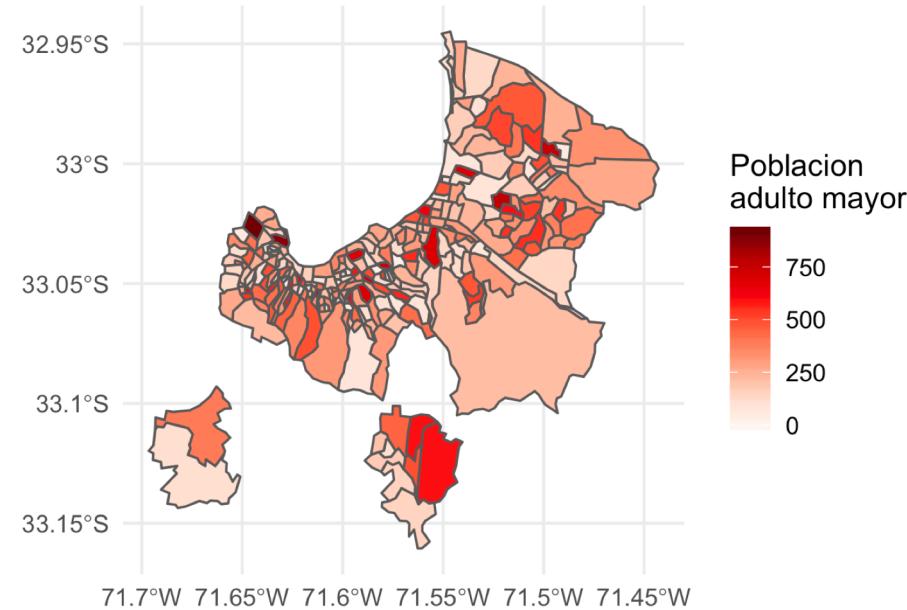
Poblacion de 65 años y más
Valparaíso y Viña del Mar



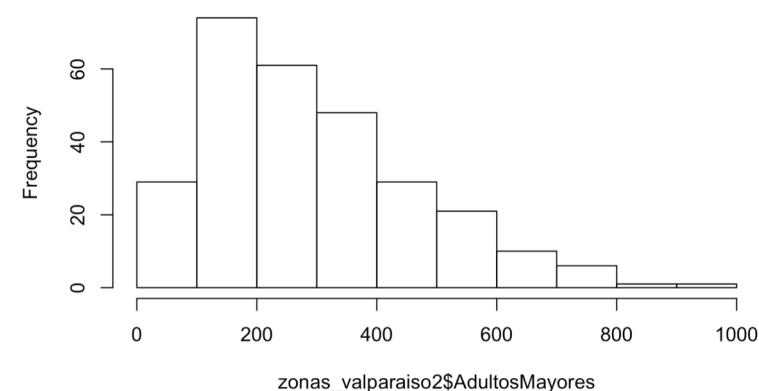
Histograma Adultos Mayores Viña-Valpo



Poblacion de 65 años y más
Valparaíso y Viña del Mar



Histograma Adultos Mayores Viña-Valpo

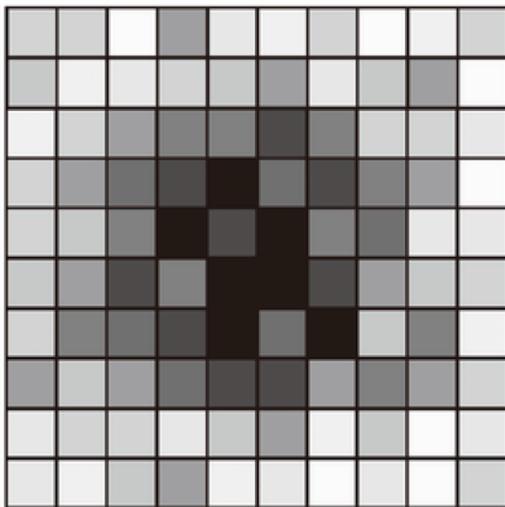


Autocorrelación Espacial Positiva

- Agrupamiento general de valores de una variable
 - valores similares estan más cerca que valores menos similares
 - First Law of Geography – Tobbler
 - Valores similares pueden ser altos (high → hot spots) o bajos (low → cold spots)
 - Dificultad para confiar en la percepción humana

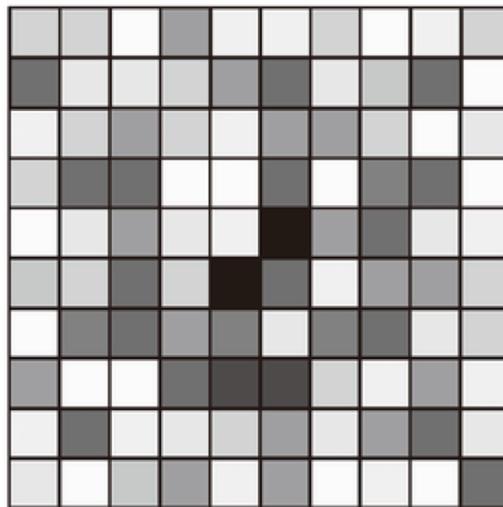
Spatial patterns of correlation

(a)



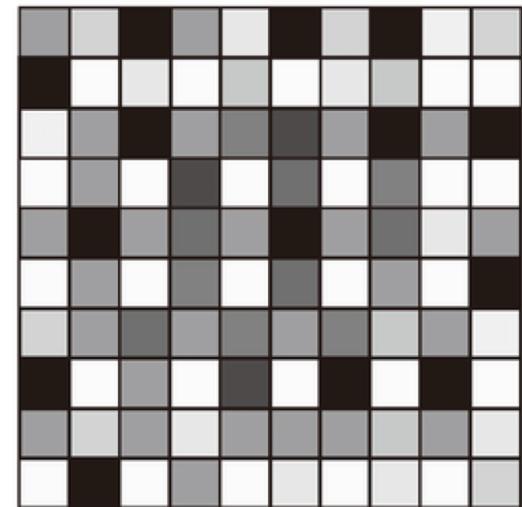
Positive
Autocorrelation

(b)



Spatial Randomness

(c)

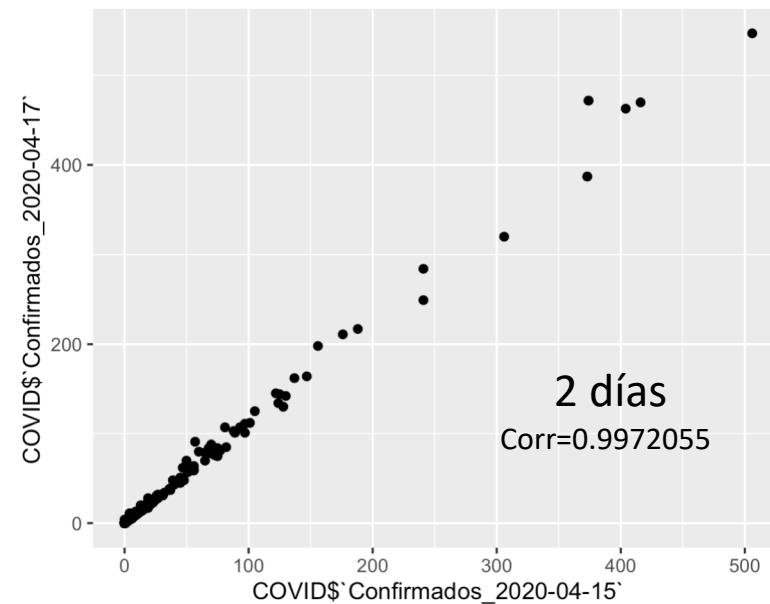
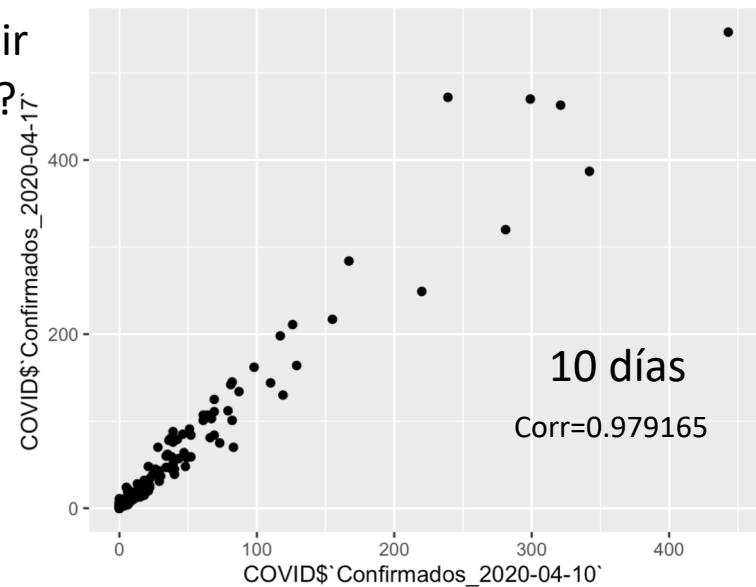
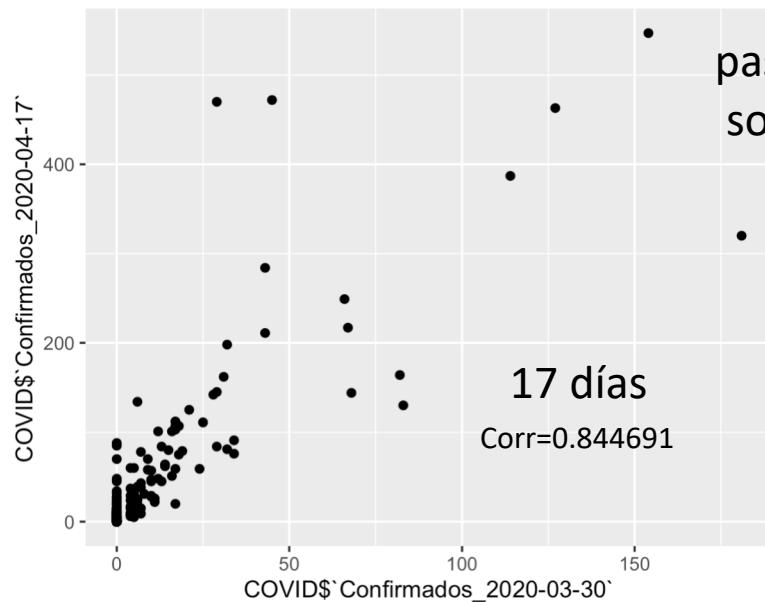


Negative
Autocorrelation

- Statistic
 - value that summarizes the distribution
 - Calculated from the data
 - Compared to a reference distribution
 - How far is the statistics from the null hypothesis?
- Spatial autocorrelation Statistic
 - Attribute similarity
 - Based on the values of the data
 - Autocorrelation $\text{corr}(y_i, y_j)$, with i, j as different locations
 - Locational similarity
 - Based on the location of the data

Time correlation

¿Qué tanto el
pasado puede decir
sobre el presente?



Spatial ‘auto’-correlation?

Attribute similarity

- Time: correlation of a variable with a realization of the same variable at a different time
- Space: Correlation of a variable with itself, considering cases at different locations
 - Main idea: is there any correlation between values of a variable considering their locational similarity?
- Similarity measures
 - Cross-product:
 - y_i, y_j . (linear) larger values are closer → large product
- Dissimilarity measures:
 - Squared differences: $(y_i - y_j)^2$
 - Absolute differences: $|y_i - y_j|$
 - the smaller the more similar

Spatial 'auto'-correlation?

Locational similarity

- Time: correlation of a variable with a realization of the same variable at a different time
- Space: Correlation of a variable with itself, considering cases at different locations
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Spatial ‘auto’-correlation?

Statistic - spatial weights

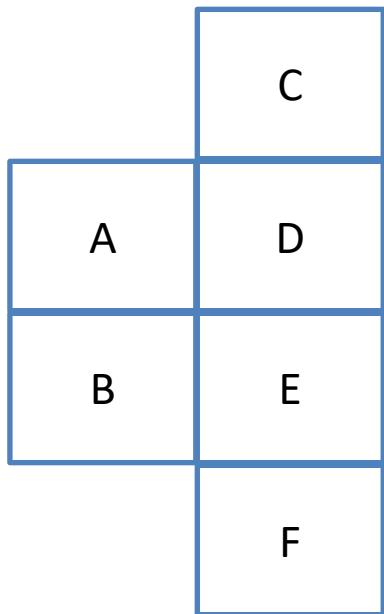
- Spatial weights
 - Let W be an n by n matrix indicating the level of association between n spatial observations w_{ij}
 - Let $f(y_i, y_j)$ be an attribute similarity measure between i and j
- Then:
 - Statistic: $\sum_{ij} f(y_i, y_j) * w_{ij}$

Spatial 'auto'-correlation?

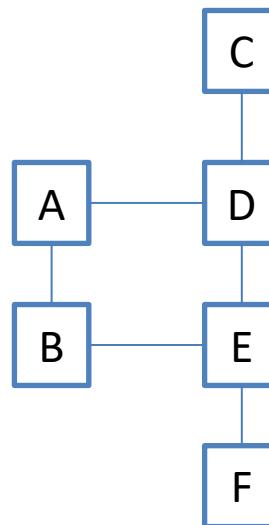
Statistic - spatial weights

- Capturing interaction
 - n spatial units, can be interacting with all n observations
 - This is an n^2 problem, but we only have n observations
 - Classic solution
 - Let's impose a structure of how interaction works → spatial structure
 - Exclude some interactions
 - Capture in a single parameter → spatial autocorrelation coefficient
- Spatial weights
 - Let W be an n by n matrix indicating the level of association between n spatial observations w_{ij}
 - Let $f(y_i, y_j)$ be an attribute similarity measure between i and j
 - Then:
 - Statistic: $\sum_{ij} f(y_i, y_j) * w_{ij}$

Spatial structure

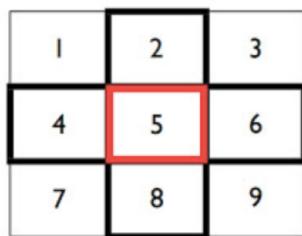


Imposed Spatial linking structure

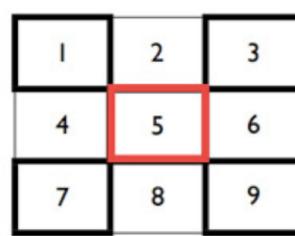


Spatial Weight Matrix

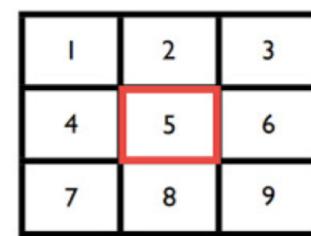
$$W = \begin{pmatrix} 0 & 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$



rook contiguity - edges only
2, 4, 6, 8 are neighbors of 5



bishop contiguity - corners only
1, 3, 7, 9 are neighbors of 5



queen contiguity - edges and corners
5 has eight neighbors

Standardizing weights

- Row-Standardized weights
 - $w'_{ij} = w_{ij} / \sum_j w_{ij}$
 - Row-sum equals to 1
 - Spatial lag = ave. of the neighbors
 - Makes analysis more comparable

$$W = \begin{pmatrix} 0 & 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix} \quad W' = \begin{pmatrix} 0 & 1/2 & 0 & 1/2 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 1/2 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1/3 & 0 & 1/3 & 0 & 1/3 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$