

# Geographic Data Science - Lecture IV

Mapping Data

[Dani Arribas-Bel](#)

# Today

- Mapping data
- MAUP
- Choropleths
  - Definition
  - Classes
- Cartograms
- Conditional maps
- Space-Time mapping

# Data maps

- Abstraction from the purely geographical map
- Representing numerical values within a spatial context

# Data maps

- Abstraction from the purely geographical map
- Representing numerical values within a spatial context

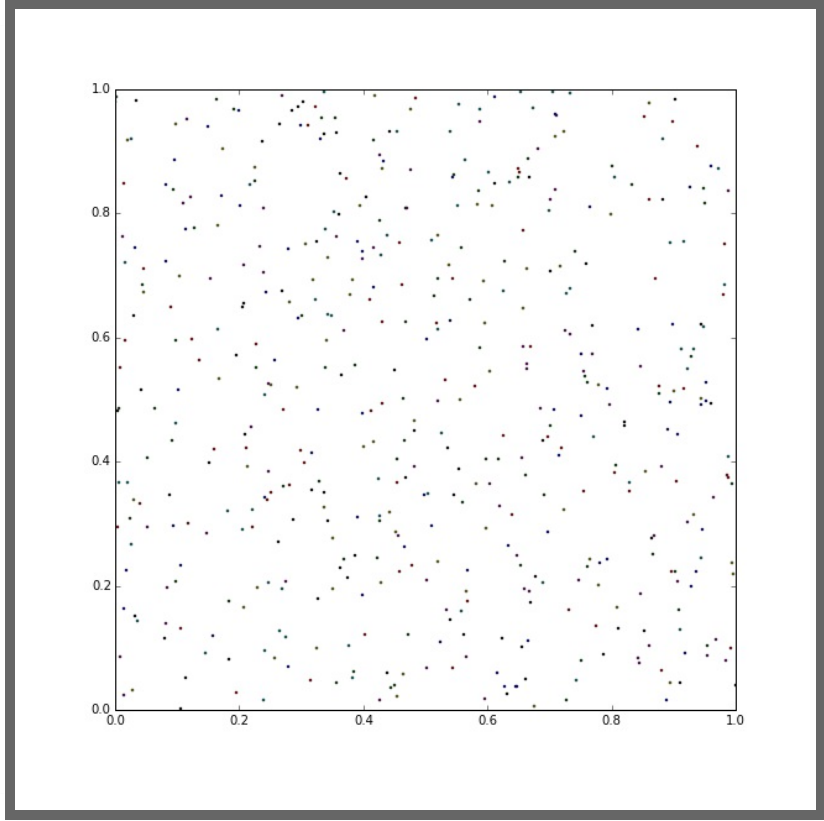
## Mapping data

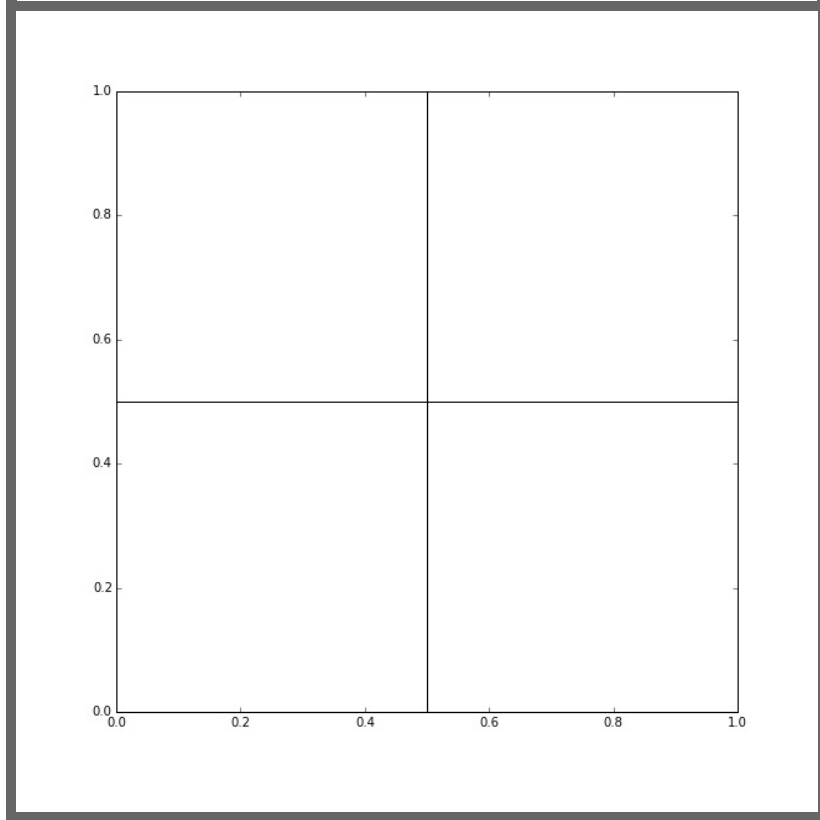
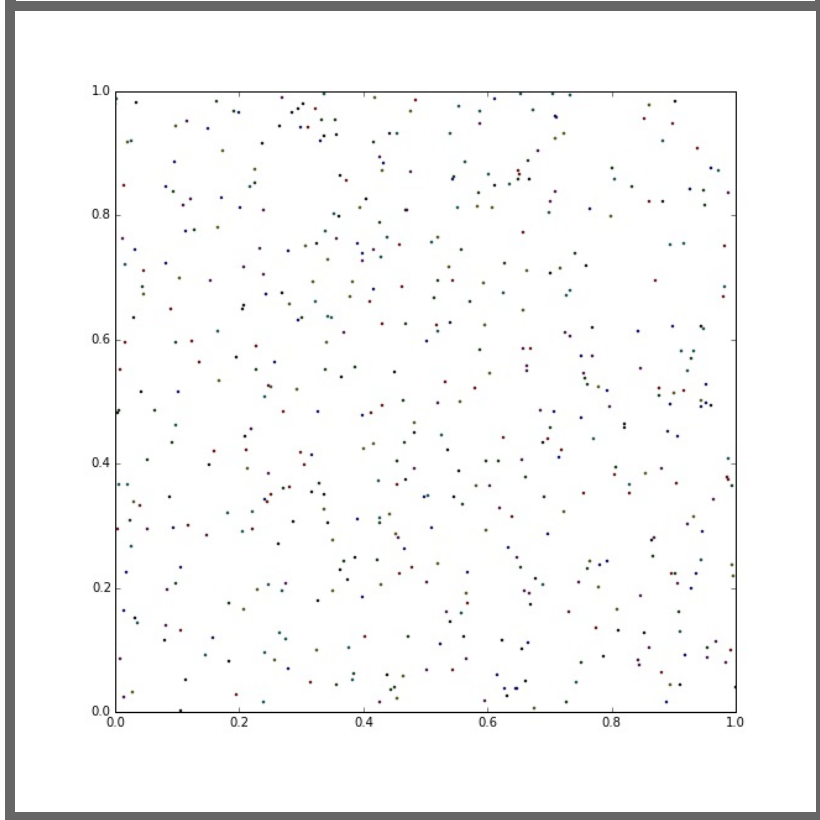
- A geographical approach to statistical visualization
- The spread of data is considered in its geographical dimension

Before we delve into different types of data maps...

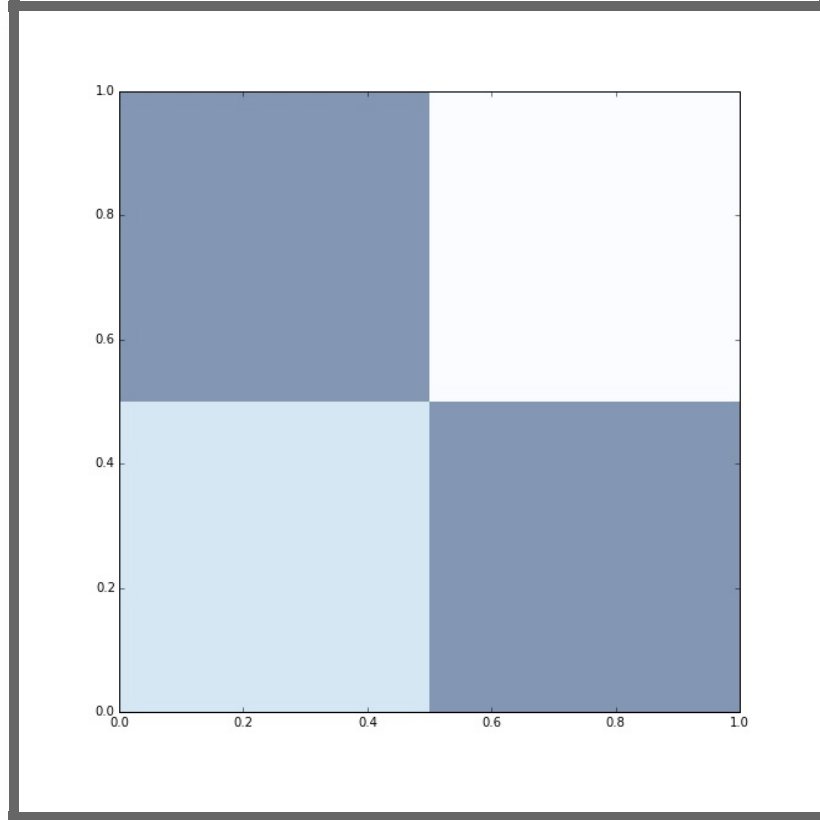
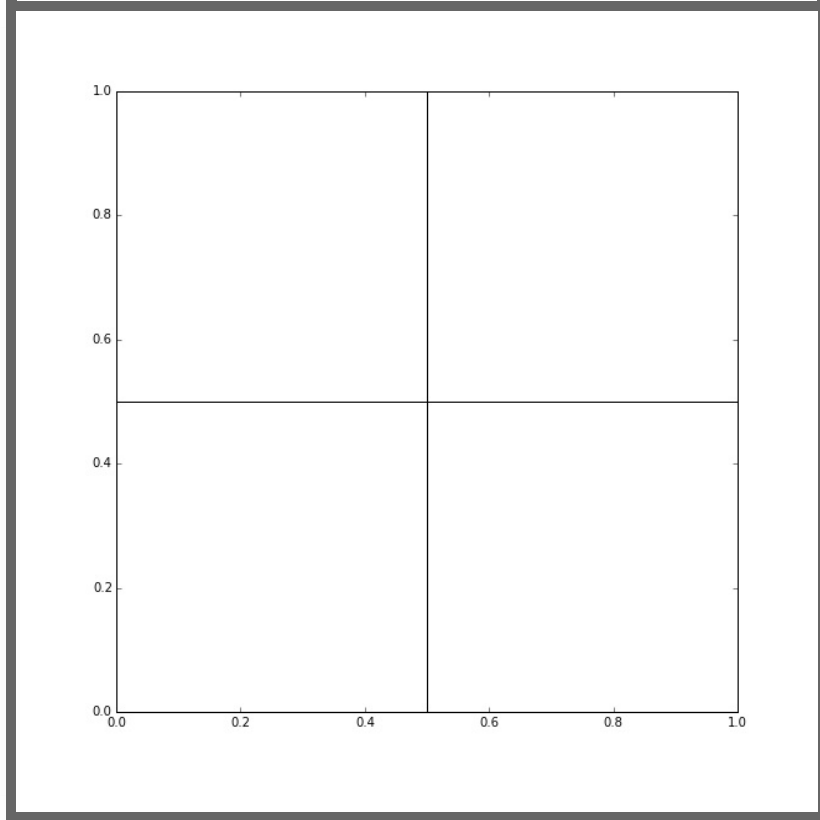
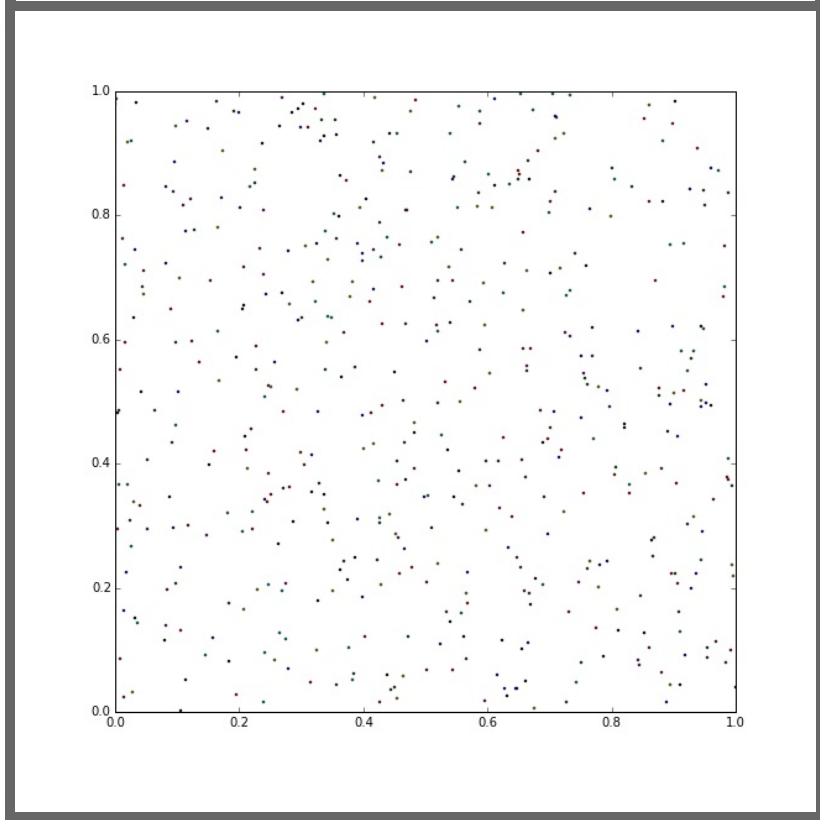
MAUP

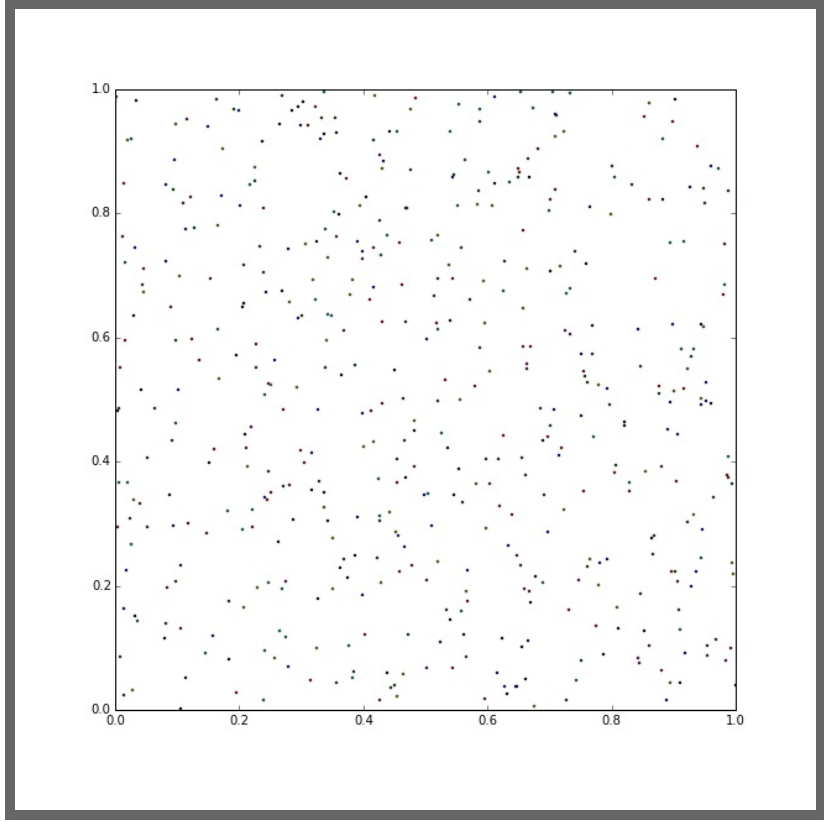
# Modifiable Areal Unit Problem

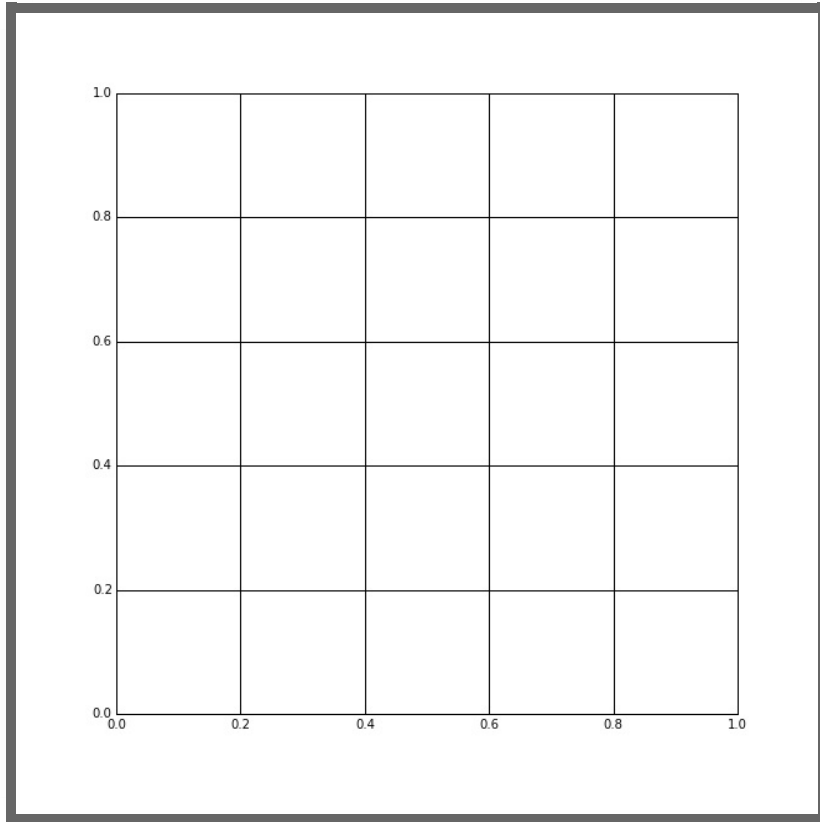
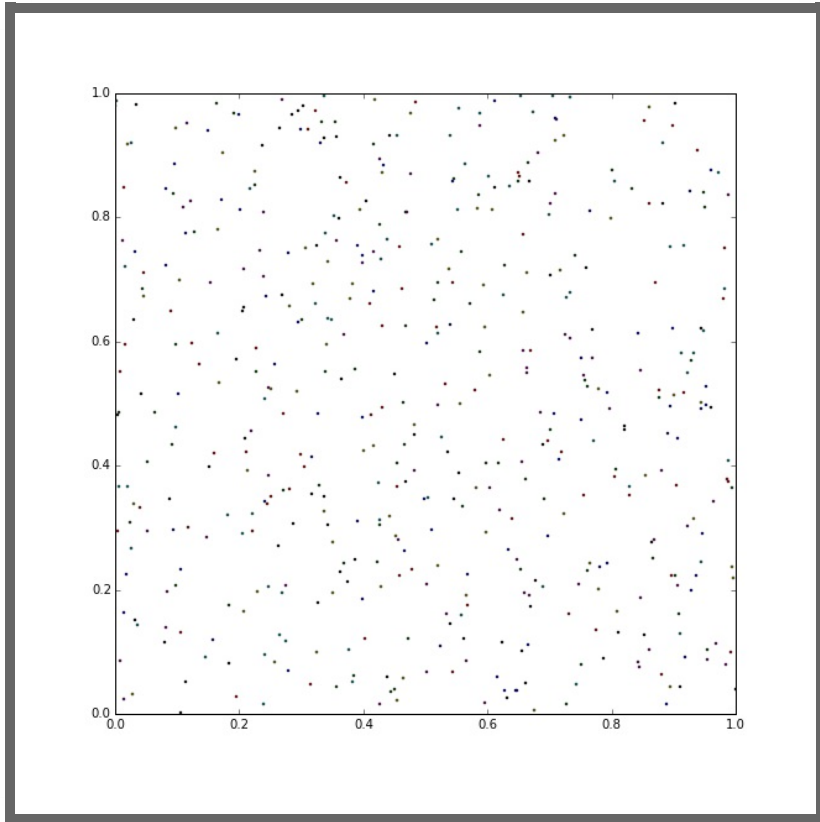


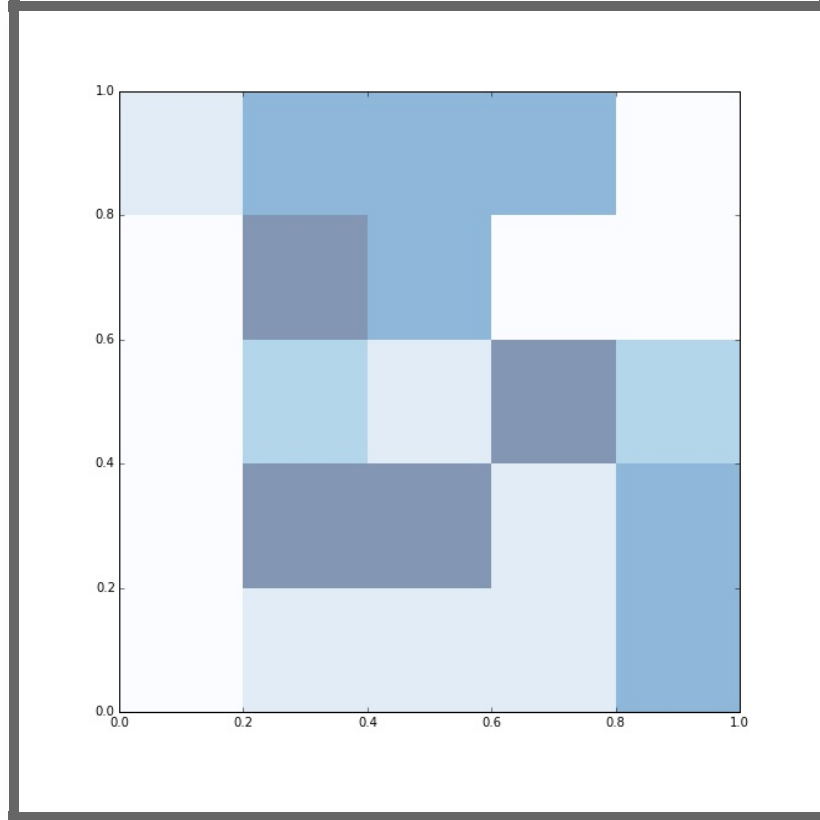
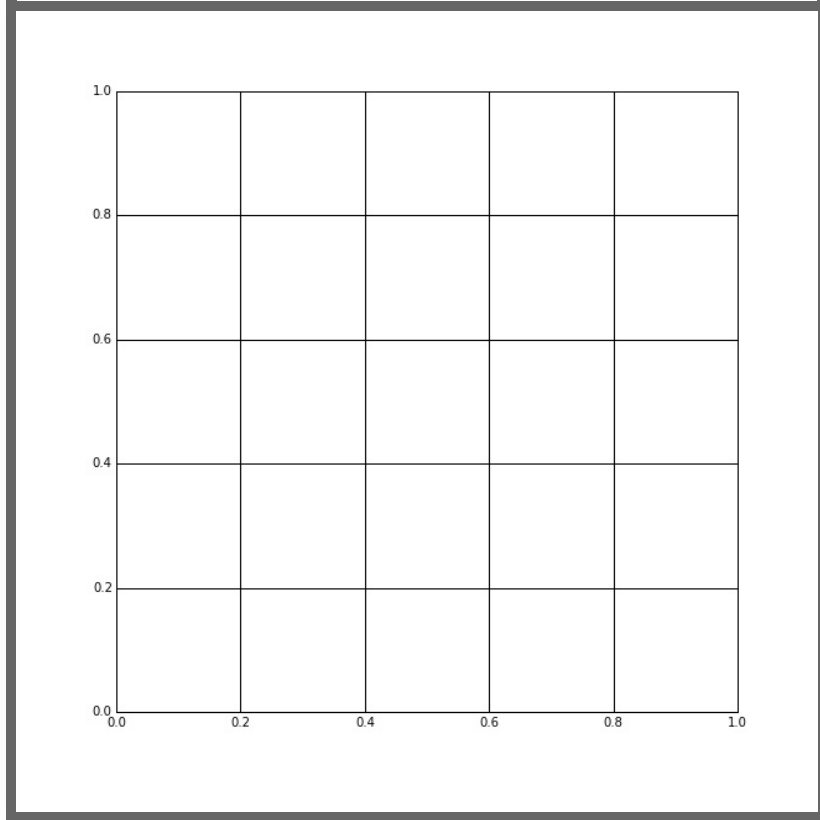
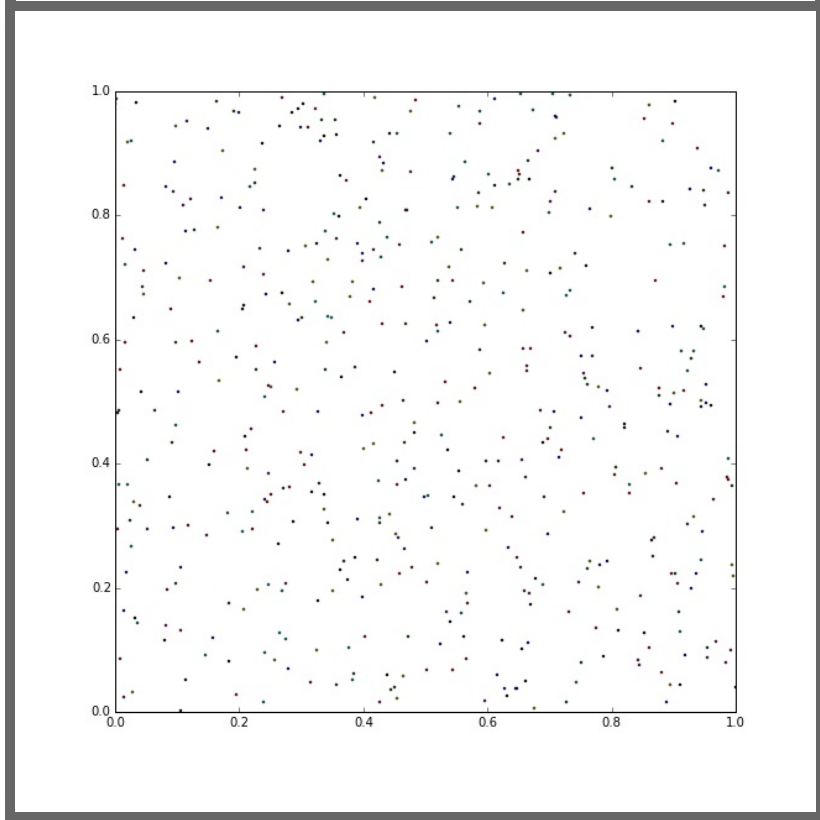


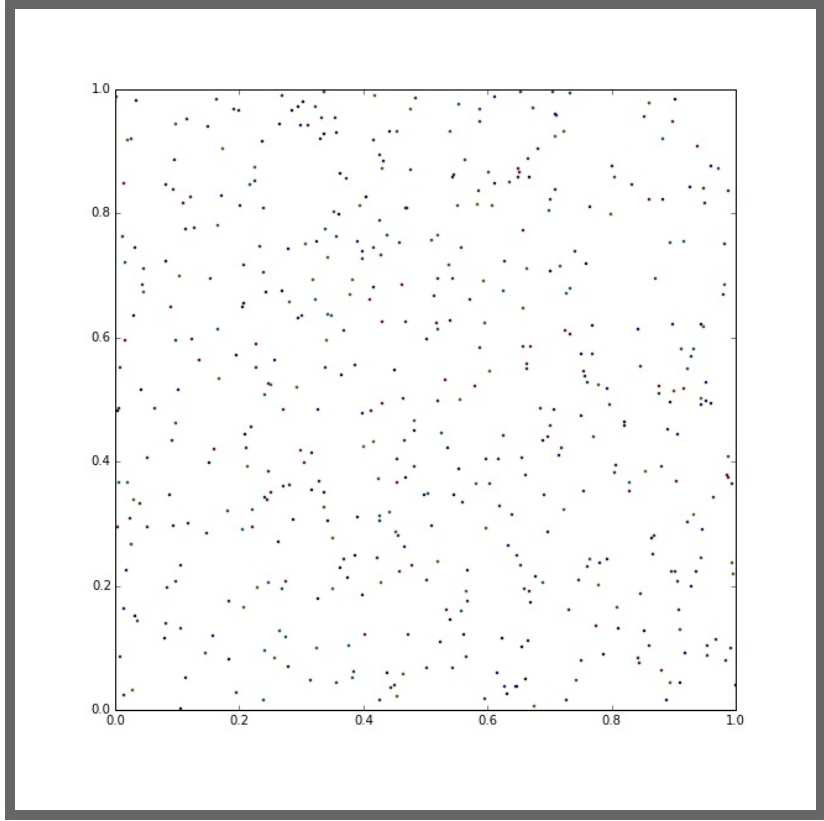


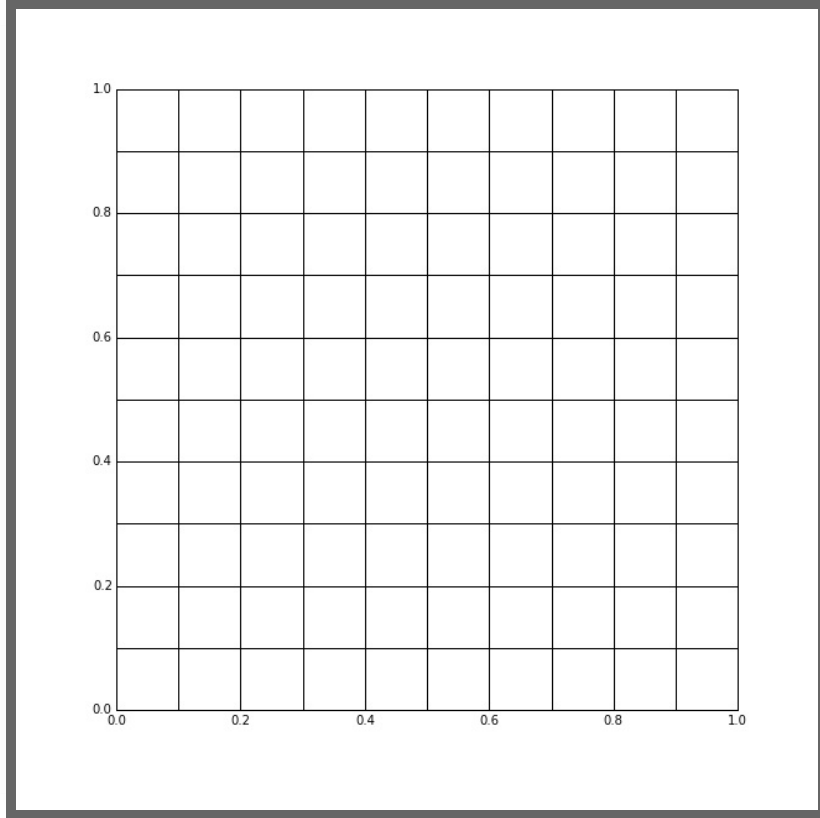
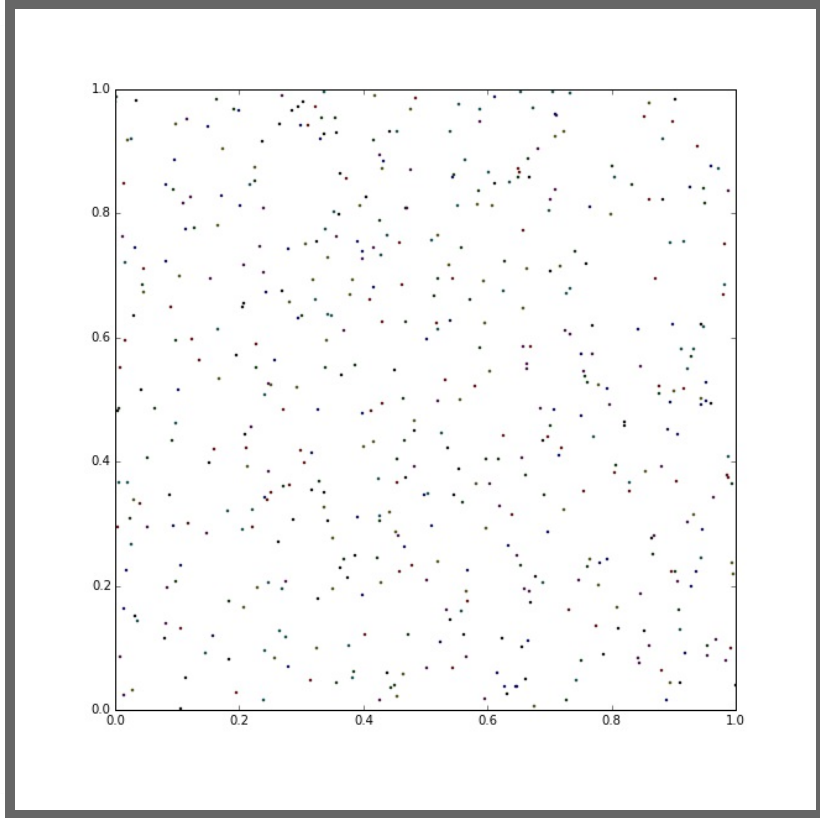


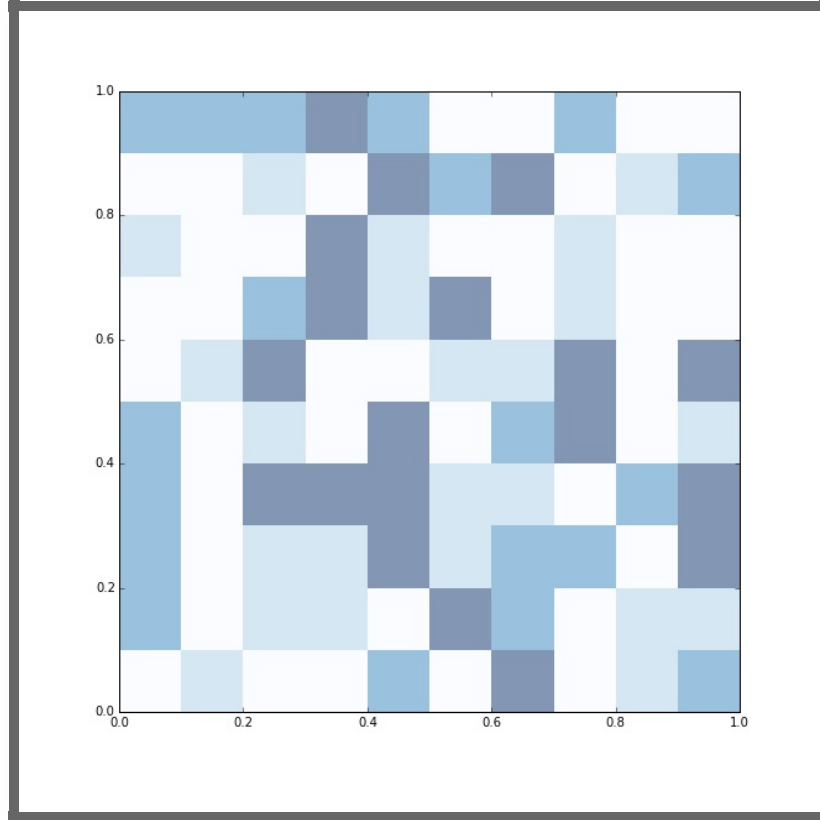
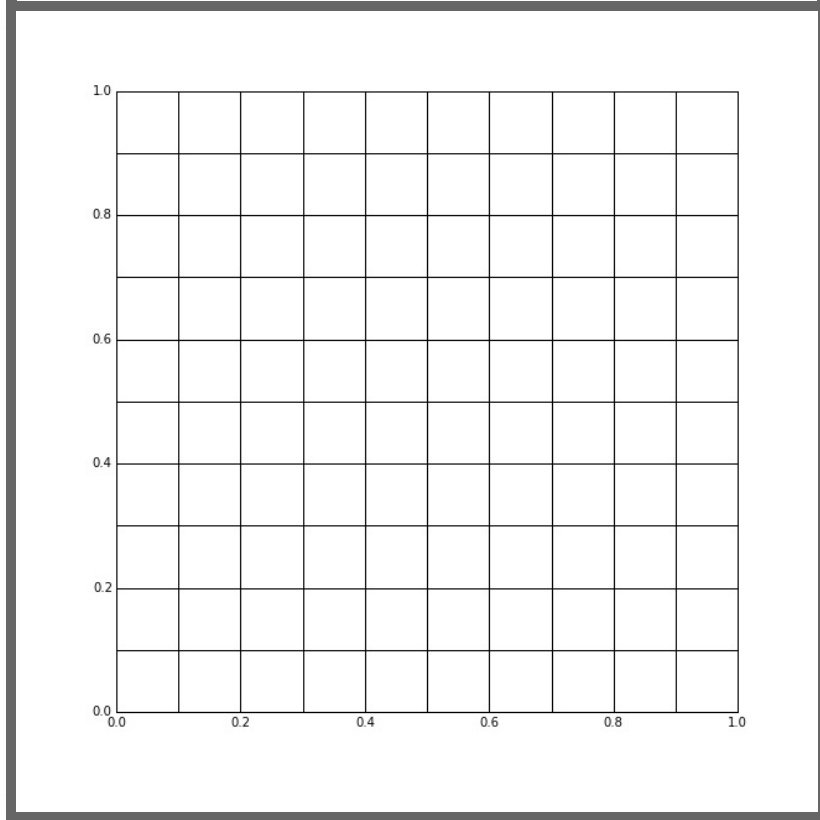
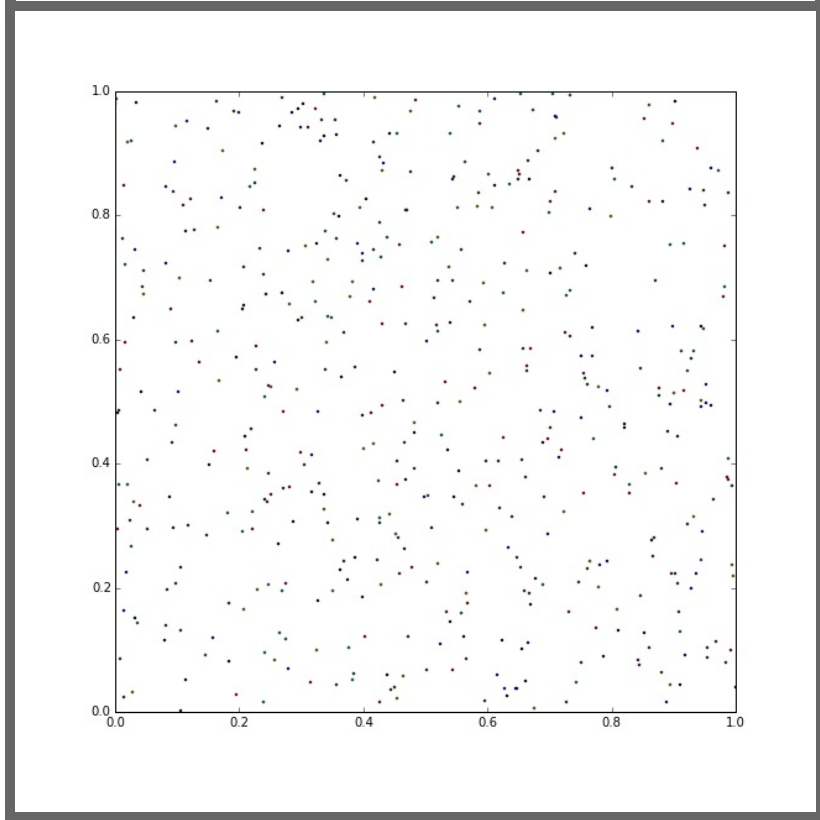


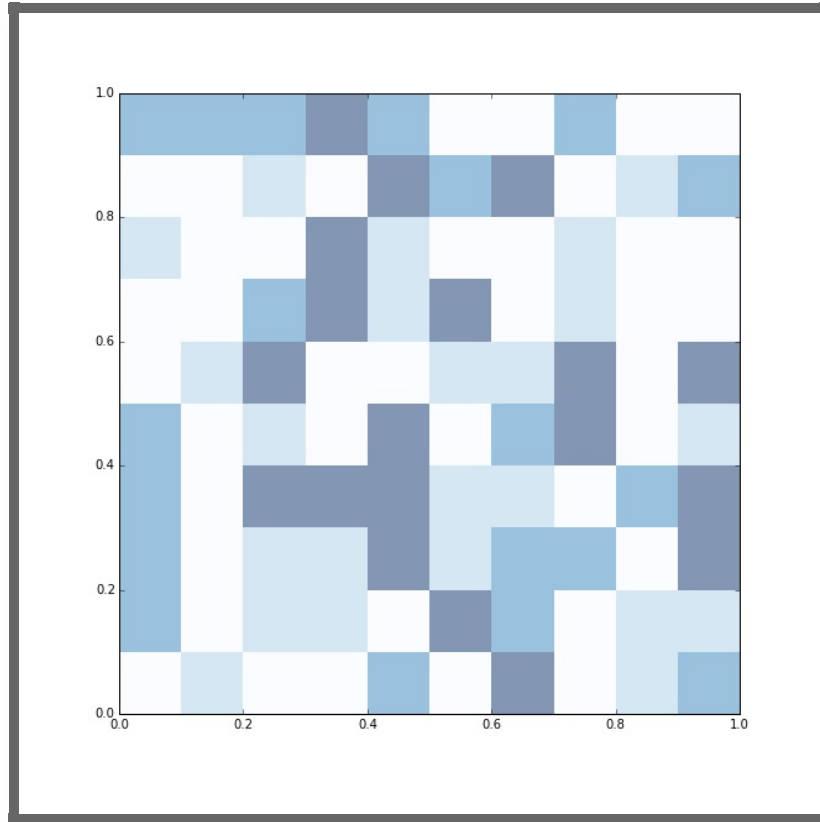
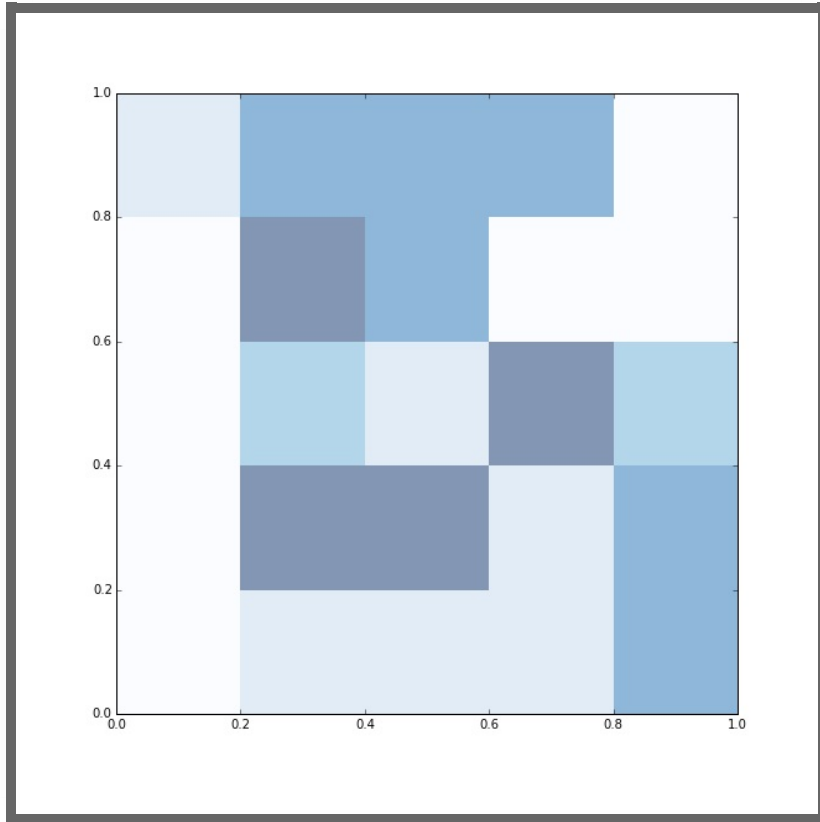
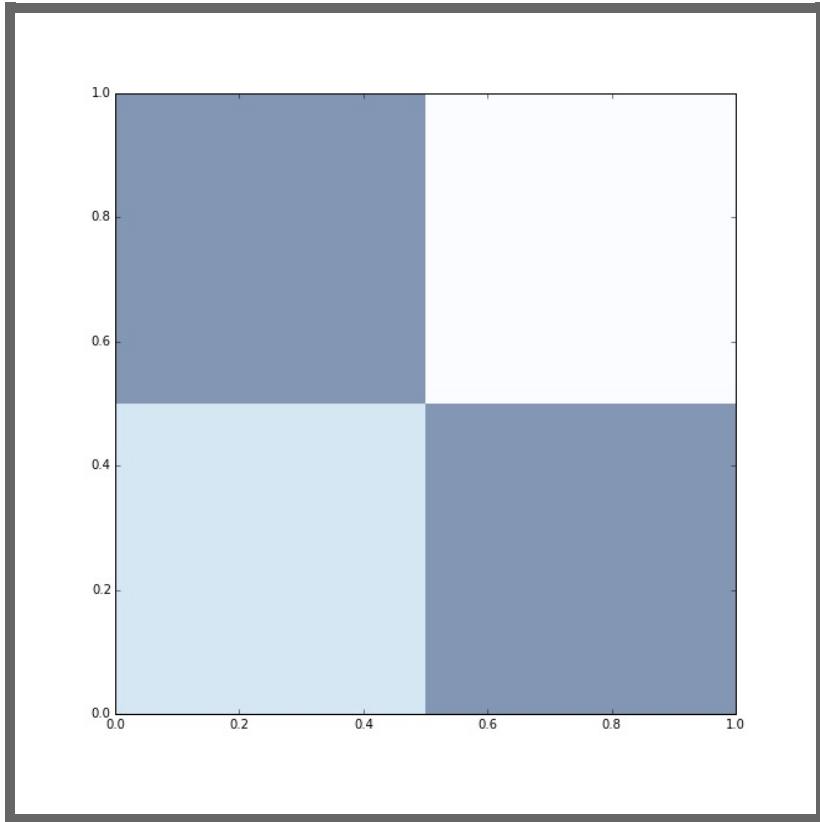














# MAUP

**Scale and delineation mismatch** between:

- Underlying process (e.g. individuals, firms, shops)
- Unit of measurement (e.g. neighborhoods, regions, etc.)

In some cases, it can **seriously mislead** analysis on aggregated data

# MAUP

**Scale and delineation mismatch** between:

- Underlying process (e.g. individuals, firms, shops)
- Unit of measurement (e.g. neighborhoods, regions, etc.)

In some cases, it can **seriously mislead** analysis on aggregated data

Always keep **MAUP** in mind when exploring aggregated data!!!

# Choropleths

# Choropleths

*Thematic map in which values of a variable are encoded using a color gradient of some sort*

- Counterpart of the histogram
- **Values are classified into specific colors: value --> bin**
- **Information loss as a trade off for simplicity**

# Classification choices

- Colors
- Bins
- Algorithm:

# Classification choices

- Colors --> in alignment with the goal of the map
- Bins
- Algorithm:

# Classification choices

- Colors --> in alignment with the goal of the map
- Bins --> How many?
- Algorithm:

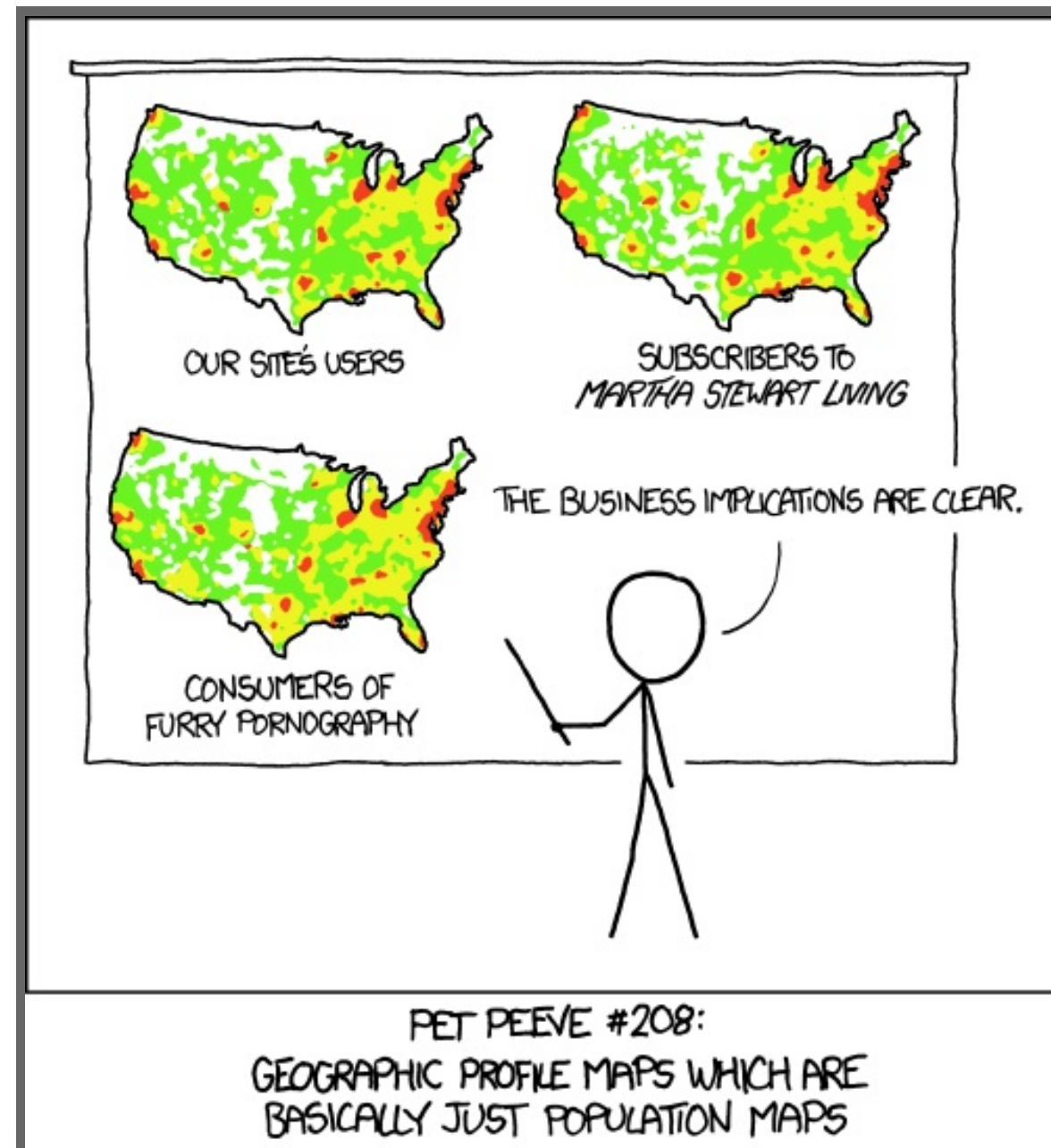
# Classification choices

- Colors --> in alignment with the goal of the map
- Bins --> How many?
- Algorithm:
  - Unique values
  - Equal interval
  - Qua / Quintiles (equal count)
  - Fisher-Jenks
  - ...



# Beware standarization!!!

[Source]



# Color schemes

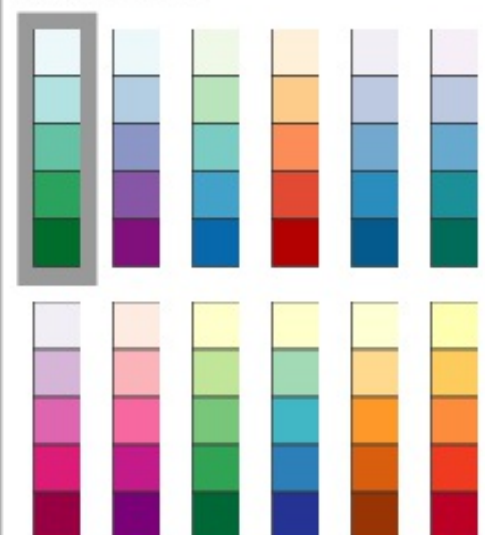


Number of data classes: 3 ▼

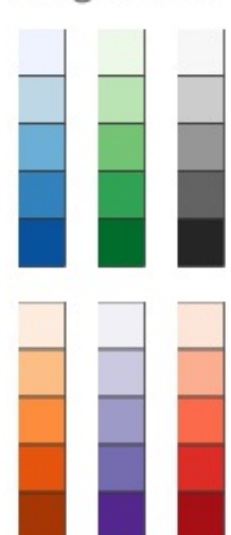
Nature of your data:  sequential  diverging  qualitative

Pick a color scheme:

Multi-hue:



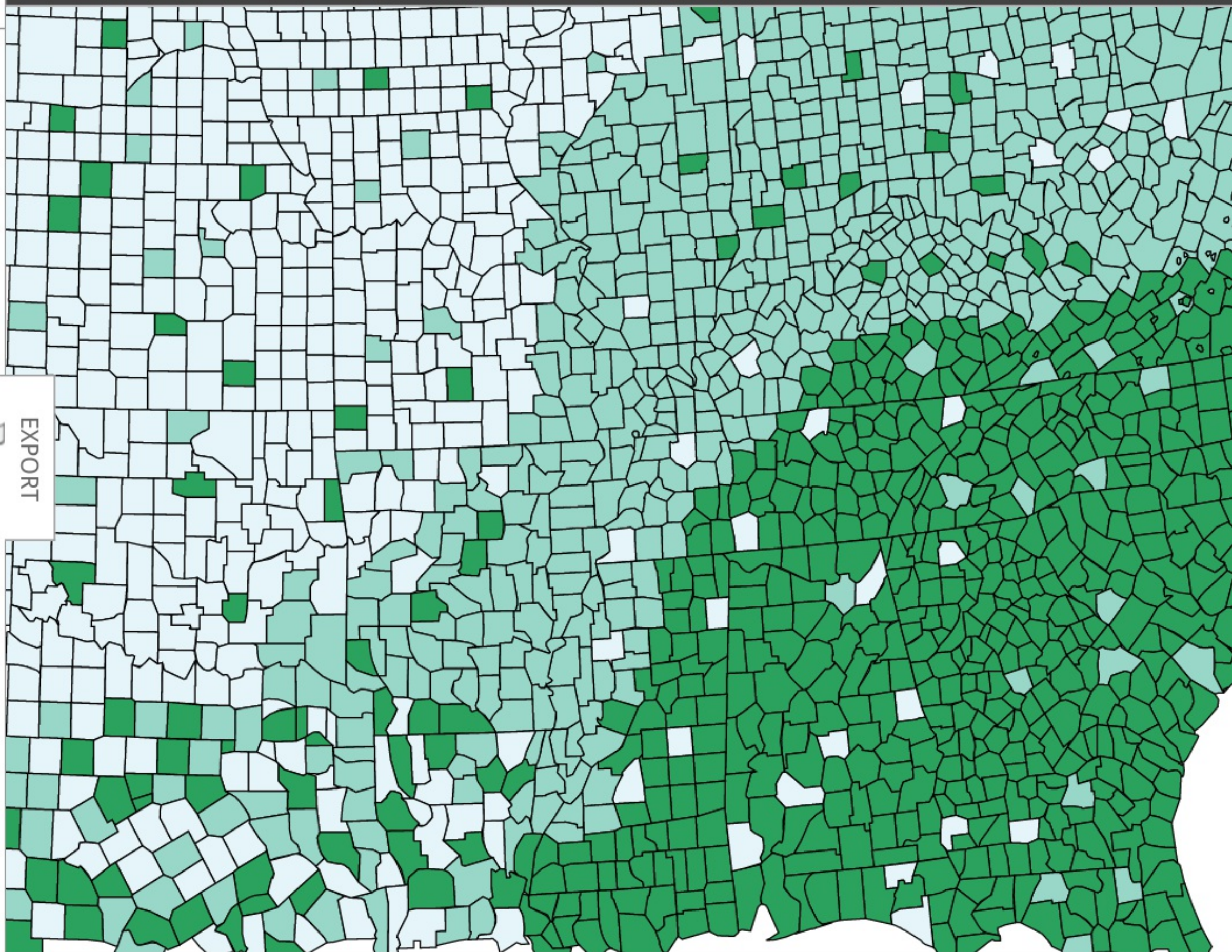
Single hue:



[how to use](#) | [updates](#) | [downloads](#) | [credits](#)

# COLORBREWER 2.0

color advice for cartography



Only show:

- colorblind safe
- print friendly
- photocopy safe

Context:

- roads
- cities
- borders

Background:

- solid color
- terrain

color transparency

[Source](#)

3-class BqGn

HEX ▼

- #e5f5f9
- #99d8c9
- #2ca25f

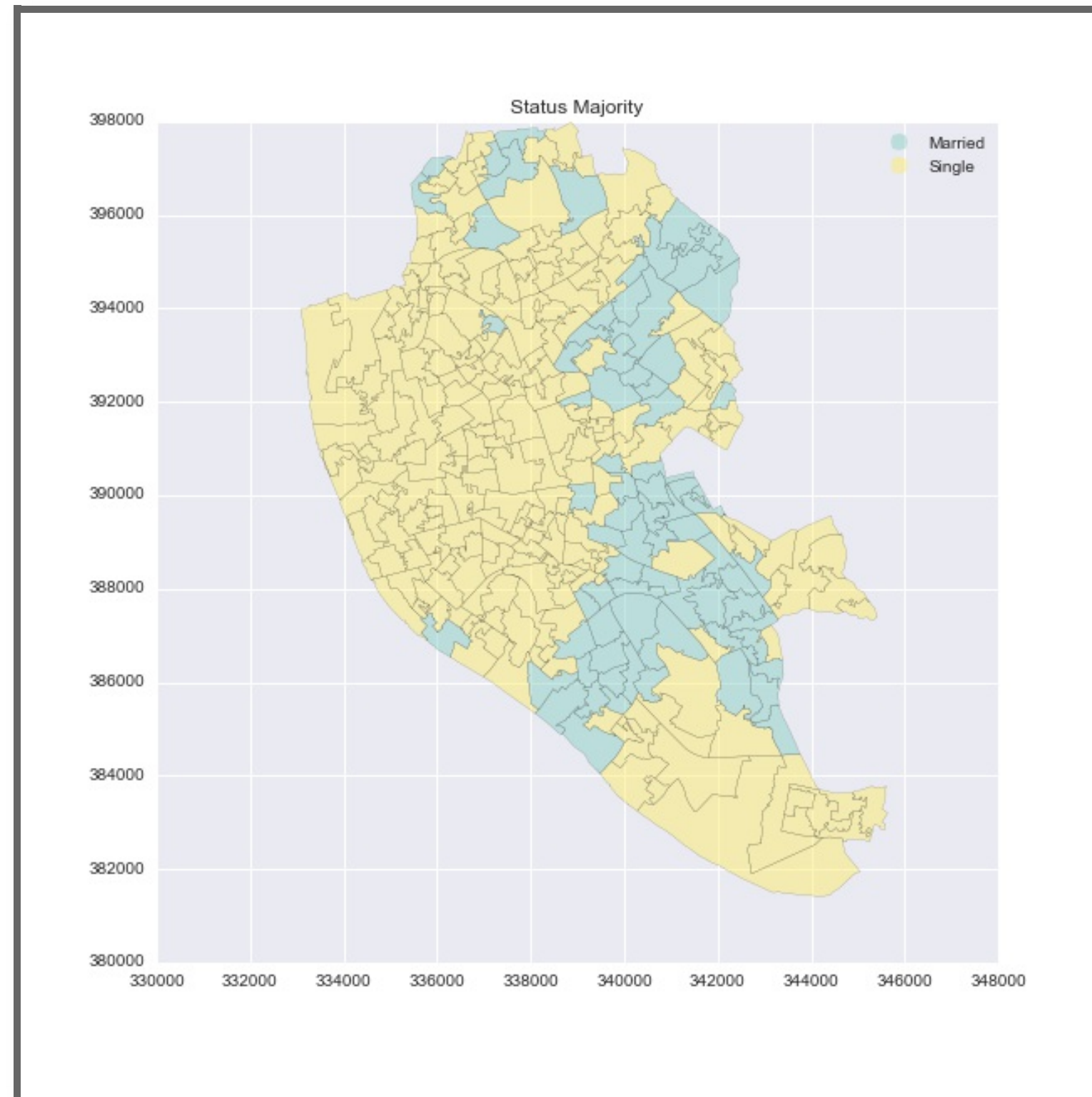
EXPORT



# Unique values

- Categorical data
- No gradient (reflect it with the color scheme!!!)
- Examples: Religion, country of origin...

# Unique values

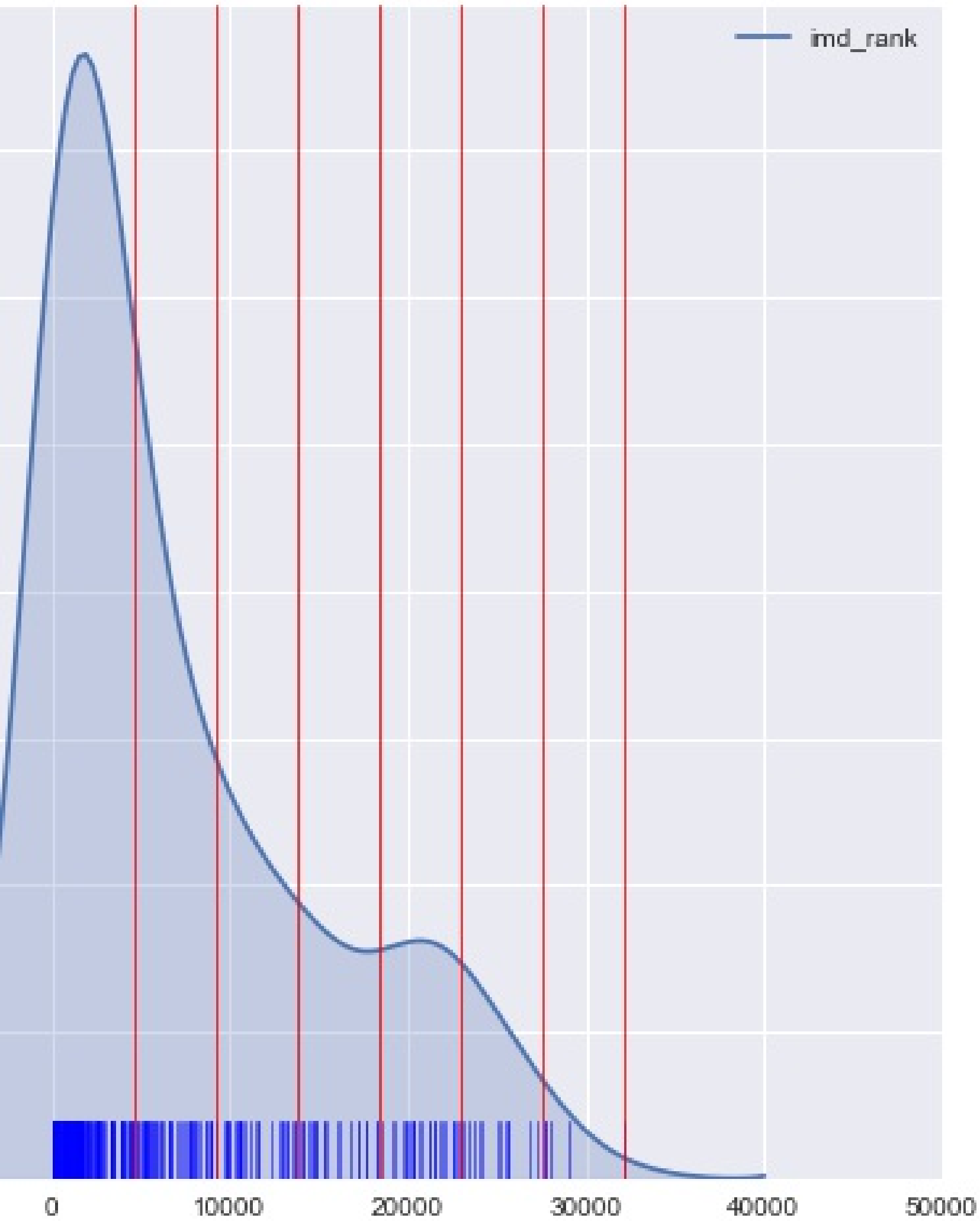


# Equal interval

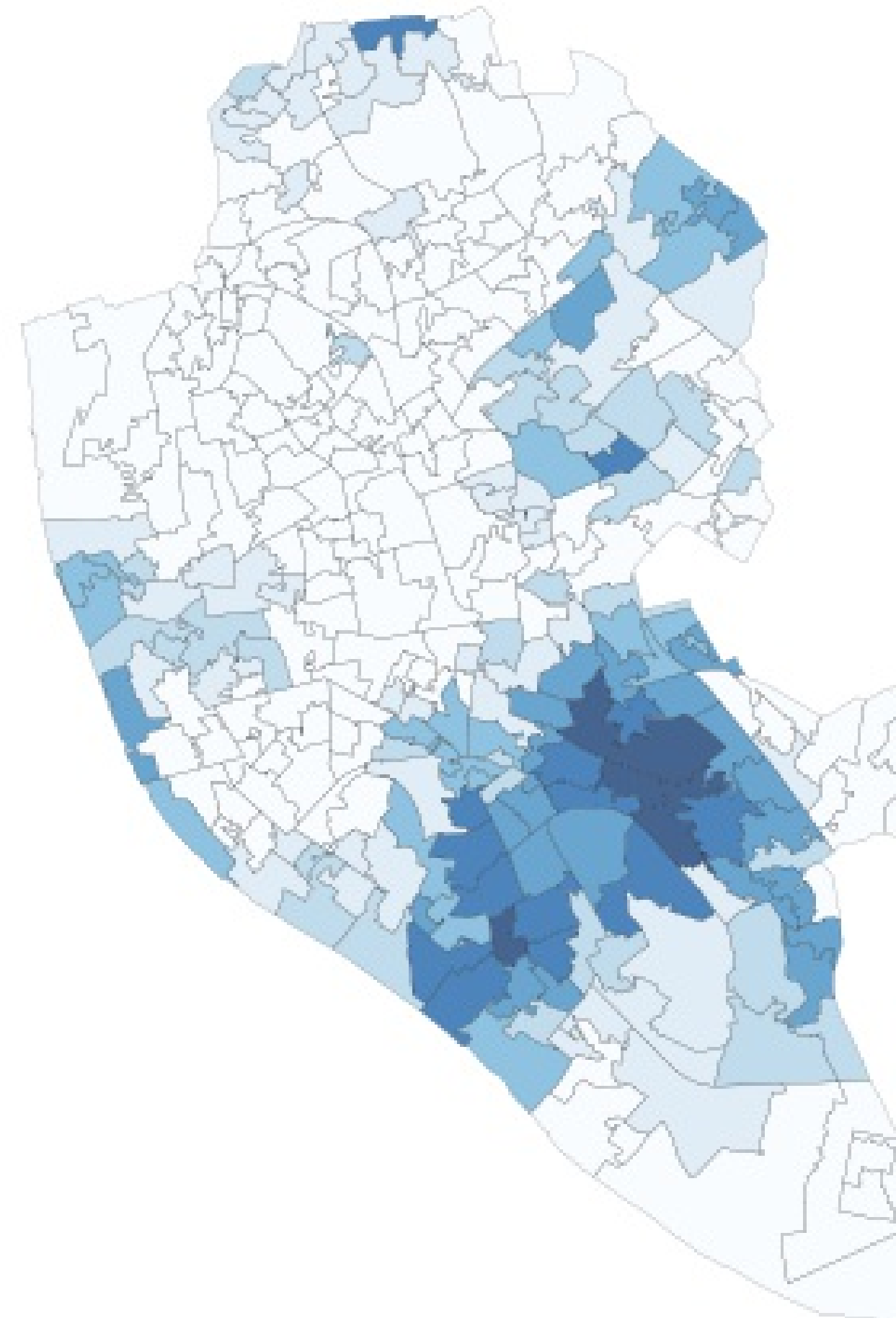
- Take the **value** span of the data to represent and split it equally
- **Splitting** happens based on the **numerical value**
- Gives more weight to outliers if the distribution is skewed

# equal\_interval

Value distribution



Geographical distribution



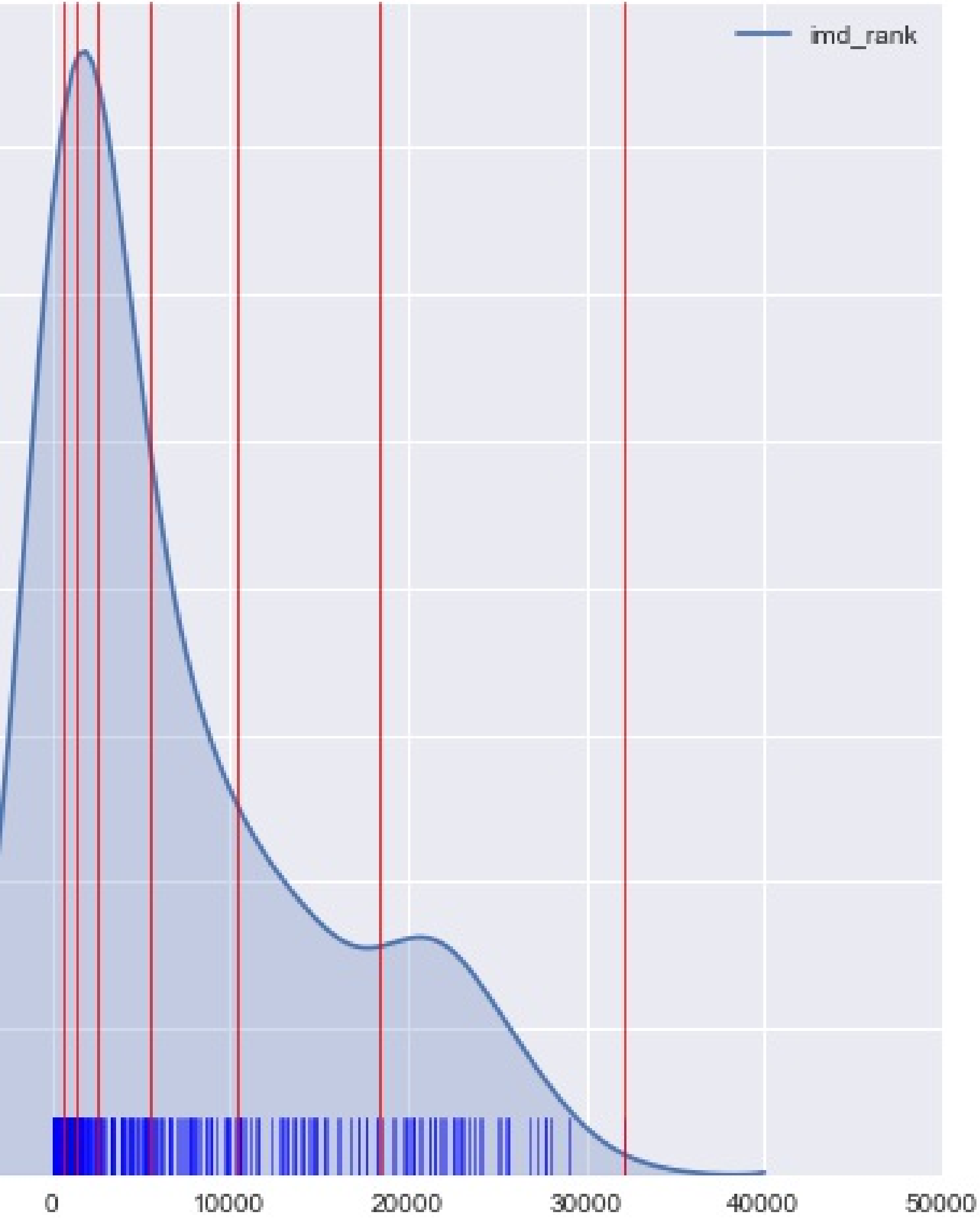
# Quantiles

- Regardless of numerical values, split the distribution keeping the same amount of values in each bin
- **Splitting** based on the **rank** of the value
- If distribution is skewed, it can put very different values in the same bin

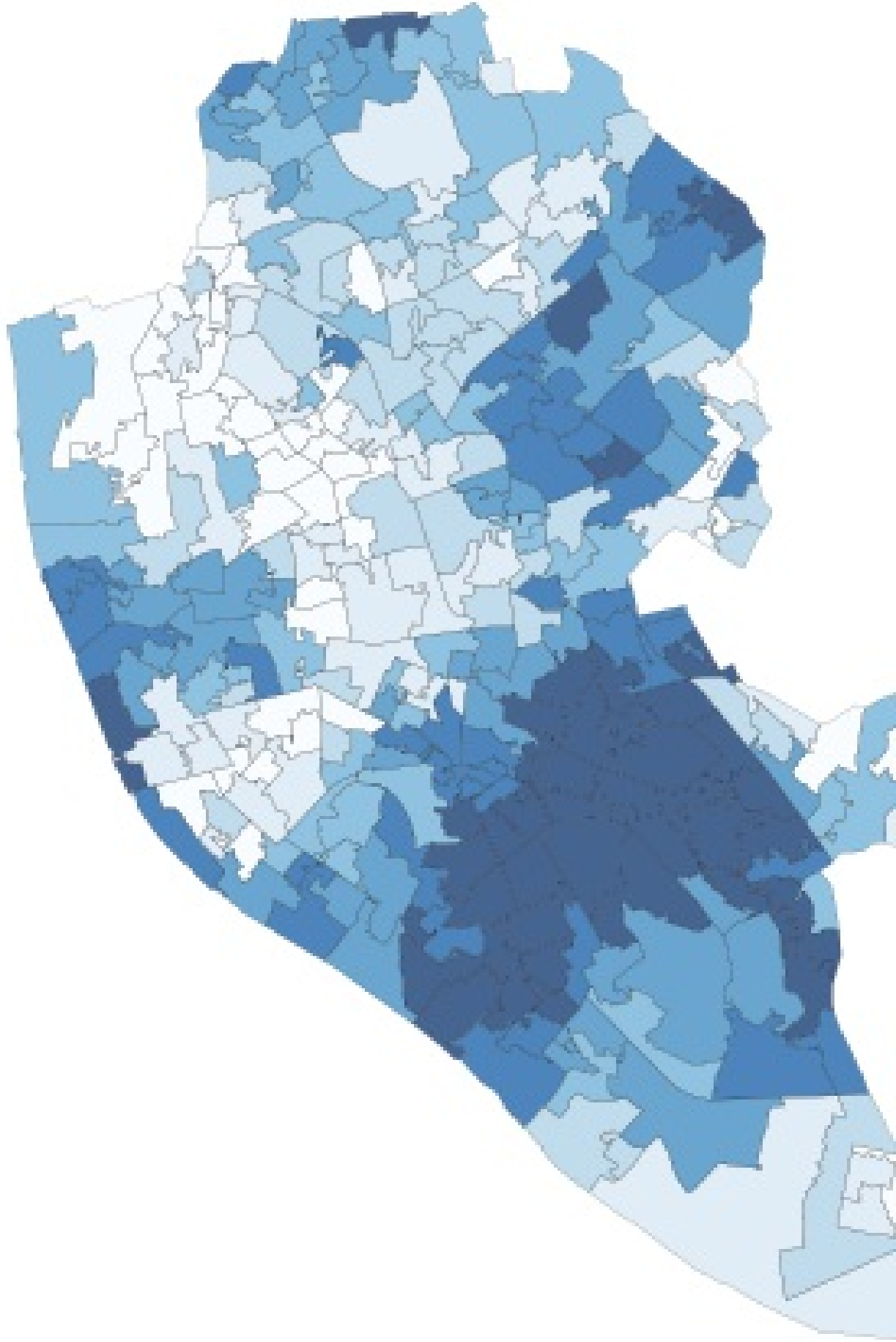


# quantiles

Value distribution



Geographical distribution



# Other

- Fisher-Jenks
- Natural breaks
- Outlier maps: box maps, std. maps...

# Tips

Different classification schemes can produce widely different maps as a result of:

- The distribution of the values
- The inherent simplification that a choropleth implies

Best advice is to **explore** different ones and **combine** choropleths with other graphical devices like histograms or density plots

# Cartograms

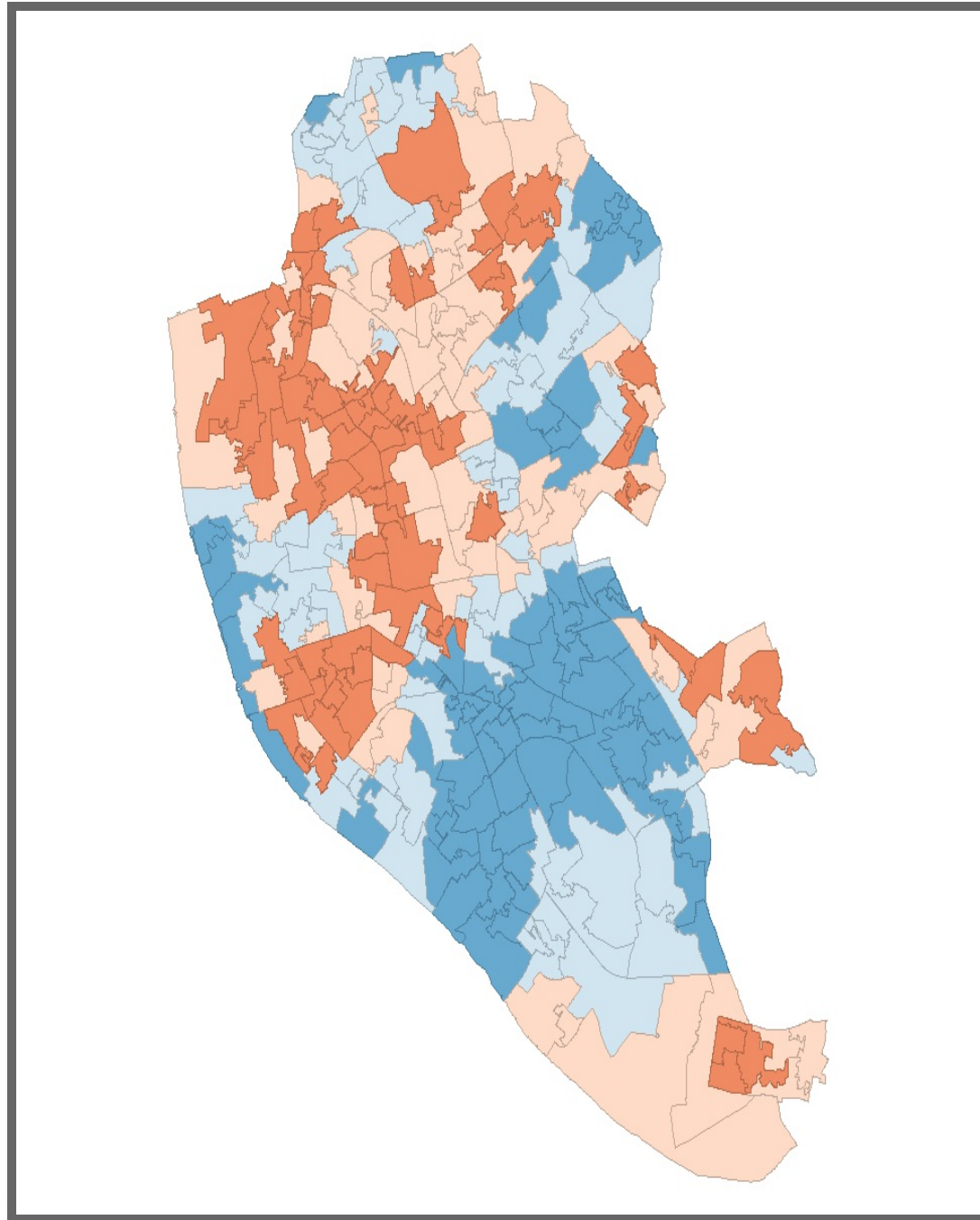
*"Data maps where the variable is encoded, not by a color gradient, but by distorting the shape/size of the geographical objects"*

# Cartograms

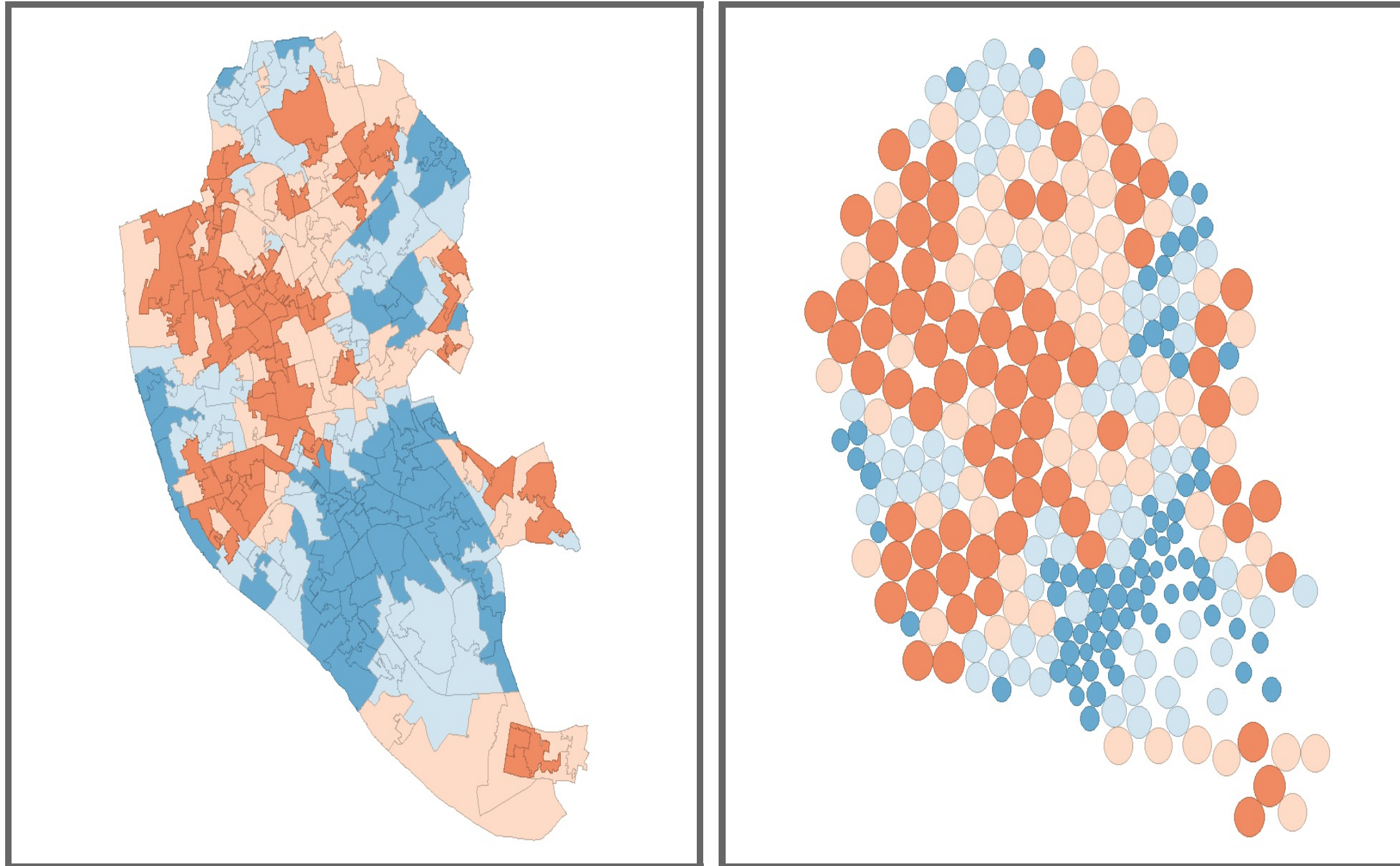
*"Data maps where the variable is encoded, not by a color gradient, but by distorting the shape/size of the geographical objects"*

- Useful in cases where the natural size / shape induces to wrong interpretation, or obscures the intended representation.
- If not done carefully, it can distort the message in unintended ways

# Cartograms

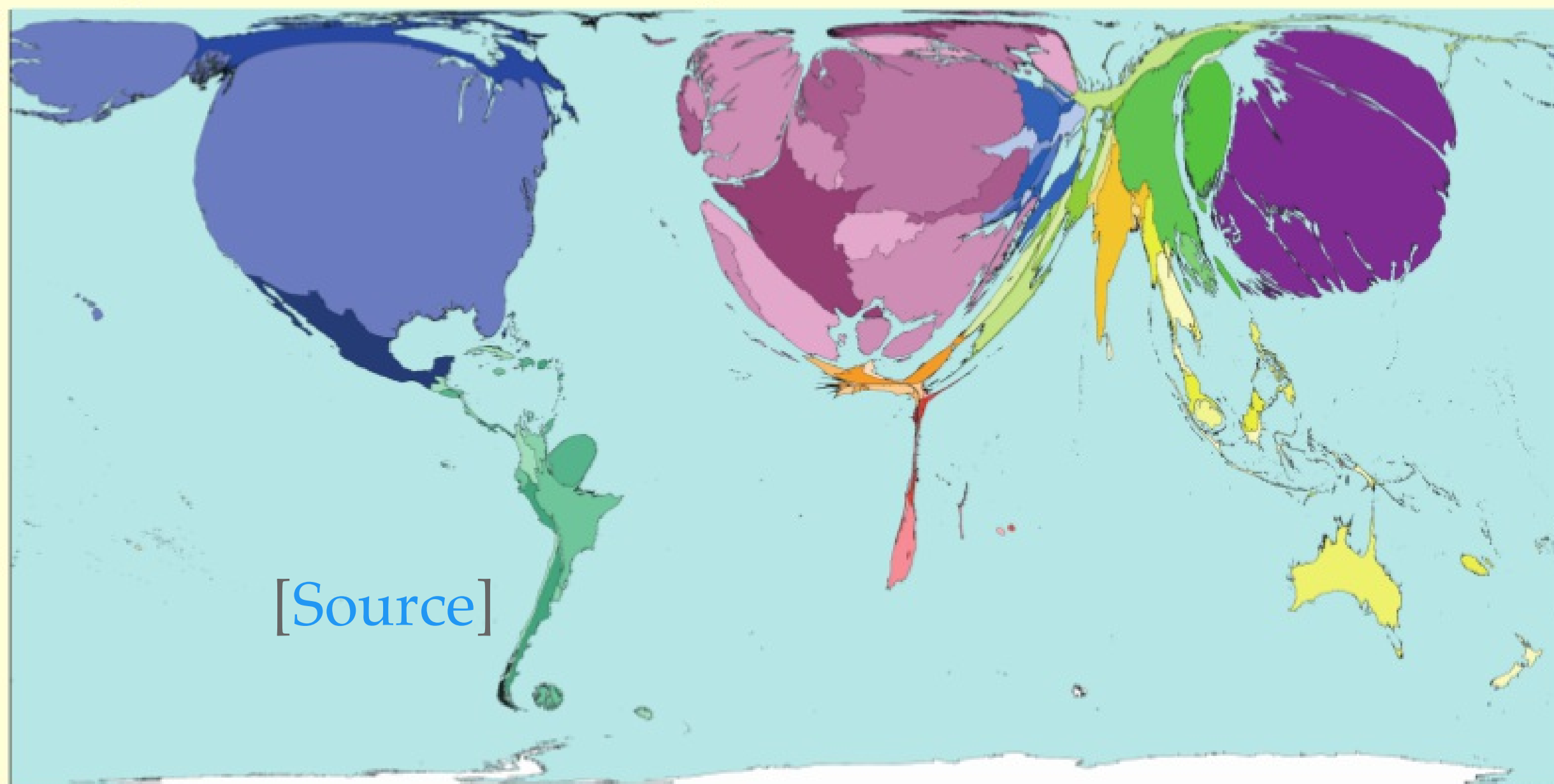


# Cartograms





# Capital Consumption



[Source]

Shown here is the cost of maintaining fixed assets and replacing them when necessary. Fixed assets include buildings, vehicles, communication and transport infrastructures. Deterioration occurs due to use, time, and becoming outdated. Fixed capital consumption is the cost of preventing this deterioration.

Large, technologically advanced infrastructures are likely to cost more to maintain than those that are small and simple. The United States, Japan, Germany and France have the highest absolute costs of fixed capital consumption. Costs per person are high in Norway, Switzerland and Denmark - roughly one thousand times higher than in Nepal, Ethiopia and Burundi.

Territory size shows the proportion of all fixed capital consumption that occurred there.



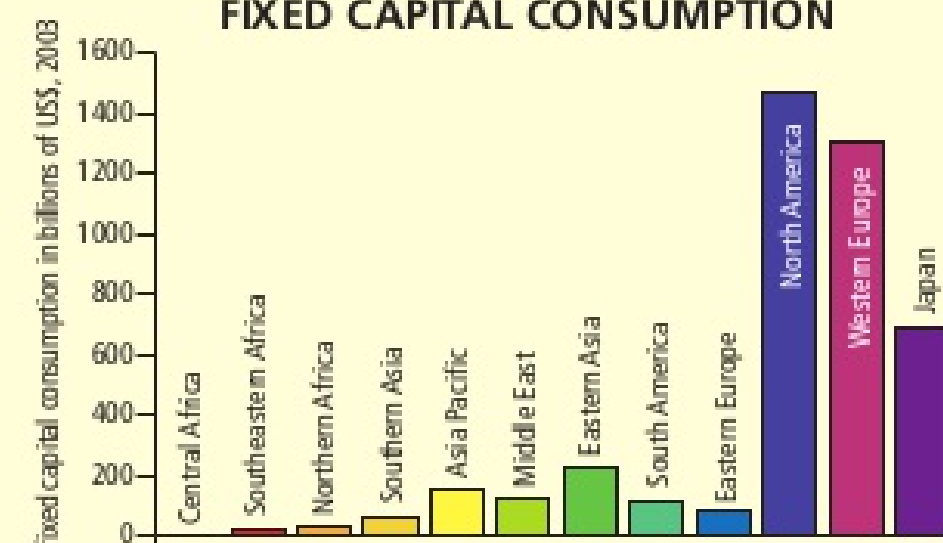
Land area

## HIGHEST AND LOWEST FIXED CAPITAL CONSUMPTION

Rank	Territory	Value	Rank	Territory	Value
1	Norway	7.0	191	Eritrea	0.0126
2	Switzerland	6.2	192	Sierra Leone	0.0112
3	Japan	5.4	193	Malawi	0.0101
4	Denmark	5.1	194	Guinea-Bissau	0.0099
5	United States	4.5	195	Liberia	0.0091
6	Finland	4.4	196	Haiti	0.0072
7	Sweden	4.0	197	Democratic Republic Congo	0.0070
8	Netherlands	4.0	198	Nepal	0.0058
9	Austria	3.9	199	Ethiopia	0.0055
10	Germany	3.8	200	Burundi	0.0054

Fixed Capital Consumption in thousands of US\$ per person per year, 2003

## FIXED CAPITAL CONSUMPTION



Technical notes  
 • Data are from the World Bank's 2005 World Development Indicators.  
 • See website for further information.

*“... it will be more expensive in the long run if we let existing infrastructure become so run down, that it must eventually be replaced at great costs.”*  
 Education and Training Unit (South Africa), 2007

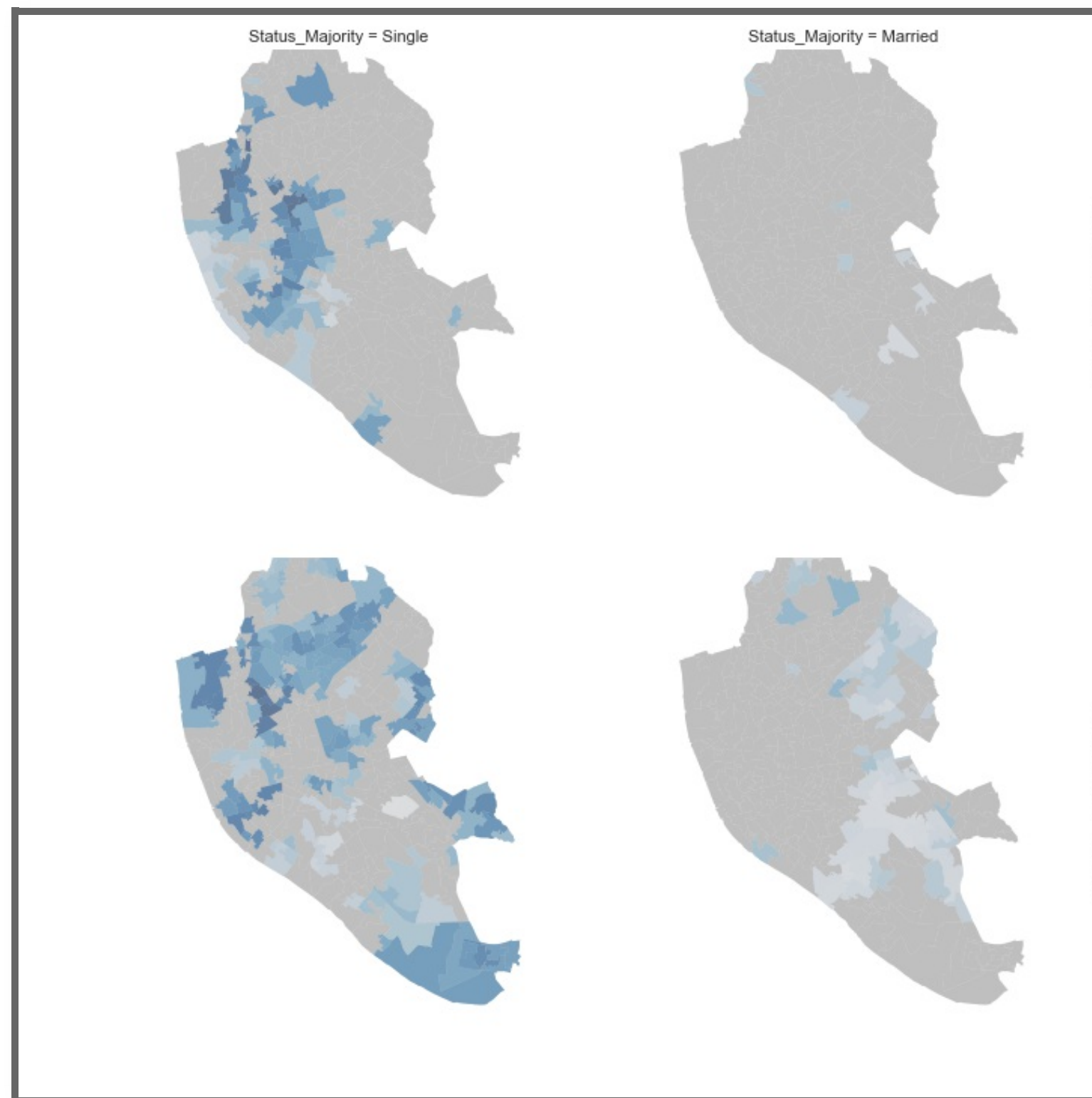


# Conditional maps

Split a dataset in *buckets* by *conditioning* on additional variables, then create a map for each *bucket*

- If no association, maps should look the same
- But, if the conditioning variables are somewhat related to the outcome we are mapping, the spatial distribution can vary substantially
- **Exploration of multivariate relationships**

# Conditional maps



# Space-Time mapping

# Space-Time mapping

- Bringing time into a spatial 2D context is "tricky" (it's really 3D!)
- Traditionally
- More recently

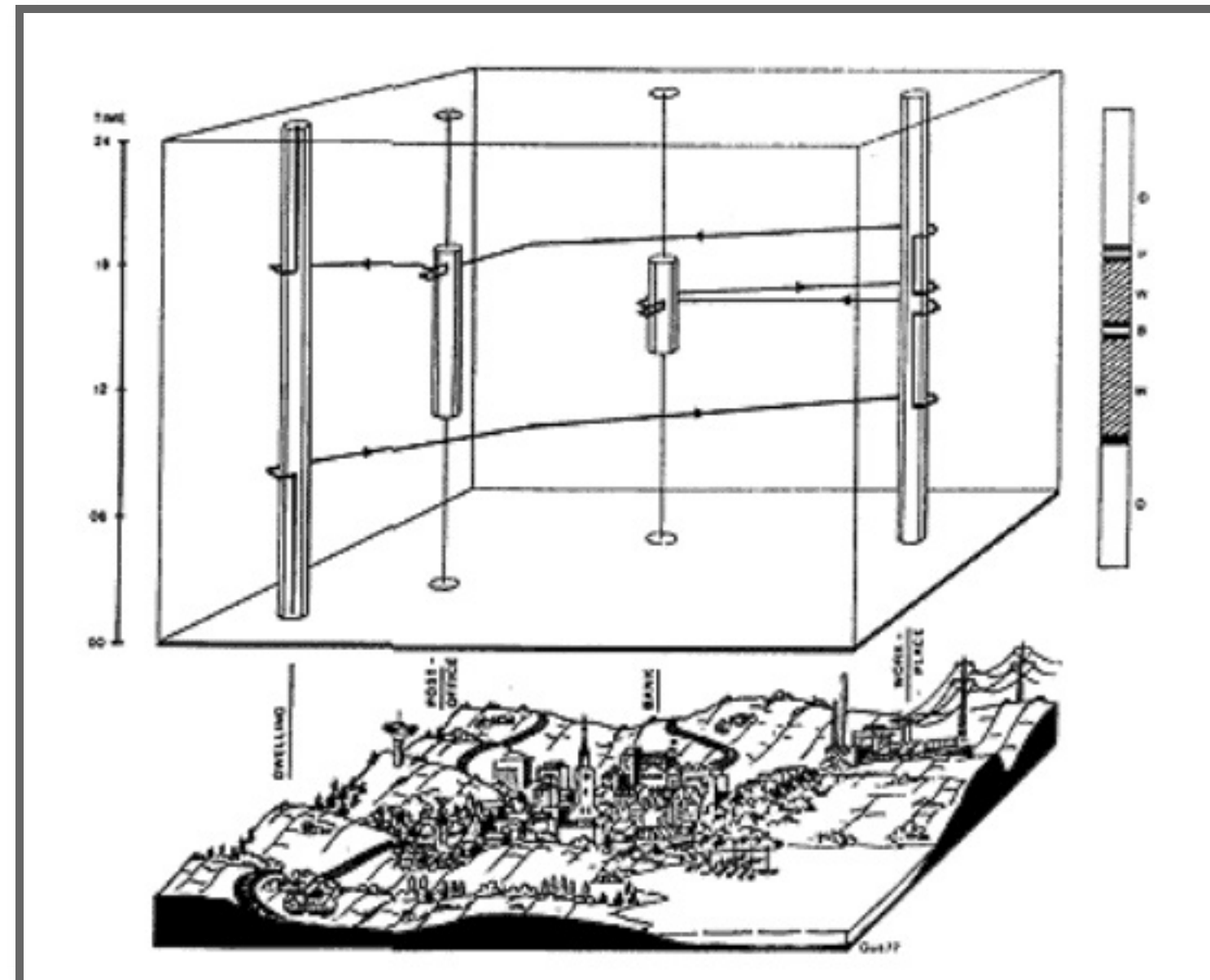
# Space-Time mapping

- Bringing time into a spatial 2D context is "tricky" (it's really 3D!)
- Traditionally --> sequence of time periods, 3D plots
- More recently

# Space-Time mapping

- Bringing time into a spatial 2D context is "tricky" (it's really 3D!)
- Traditionally --> sequence of time periods, 3D plots
- More recently --> animation and interactivity

[Source]

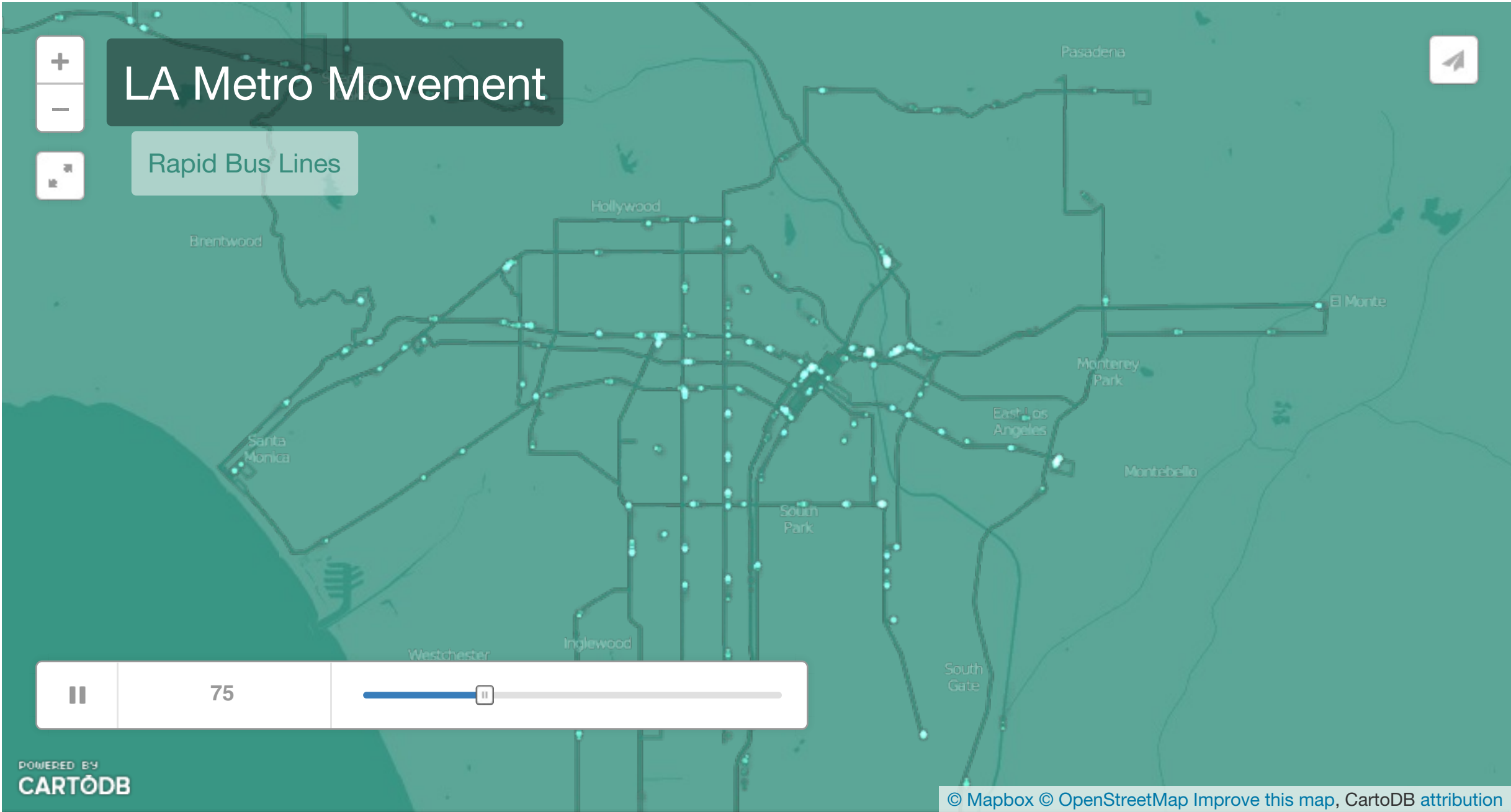


[Source]





[Source]



Map created by [★ d9a](#)

♥ 4



Geographic Data Science'15 - Lecture 4 by [Dani Arribas-Bel](#) is licensed under a [Creative Commons Attribution-NonCommercial-ShareAlike 4.0 International License](#).