

# Geographic Data Science - Lecture VIII

## Points

Dani Arribas-Bel

# Today

- The *point* of points
- Point patterns
- Visualization of point patterns

The *point* of points

# Points like polygons

Points *can* represent "fixed" entities

# Points like polygons

Points *can* represent "fixed" entities

In this case, points are **qualitatively similar to polygons/lines**

# Points like polygons

Points *can* represent "fixed" entities

In this case, points are **qualitatively similar to polygons/lines**

The **goal** here is, taking location fixed, to model other aspects of the data

# Points like polygons

Examples:

# Points like polygons

Examples:

- Cities (in most cases)

# Points like polygons

Examples:

- Cities (in most cases)
- Buildings

# Points like polygons

Examples:

- Cities (in most cases)
- Buildings
- Polygons represented as their centroid

# Points like polygons

Examples:

- Cities (in most cases)
- Buildings
- Polygons represented as their centroid
- ...

# When points are not polygons

Point data are not only a different geometry than polygons or lines...

# When points are not polygons

Point data are not only a different geometry than polygons or lines...

... Points can also represent a fundamentally different way to approach spatial analysis

Points unlike polygons

# Points unlike polygons

- Rather than exhausting the entire space, points can be **events** subject to **occur anywhere**

# Points unlike polygons

- The **location** of the event is **part** of what we are trying to understand/**model**

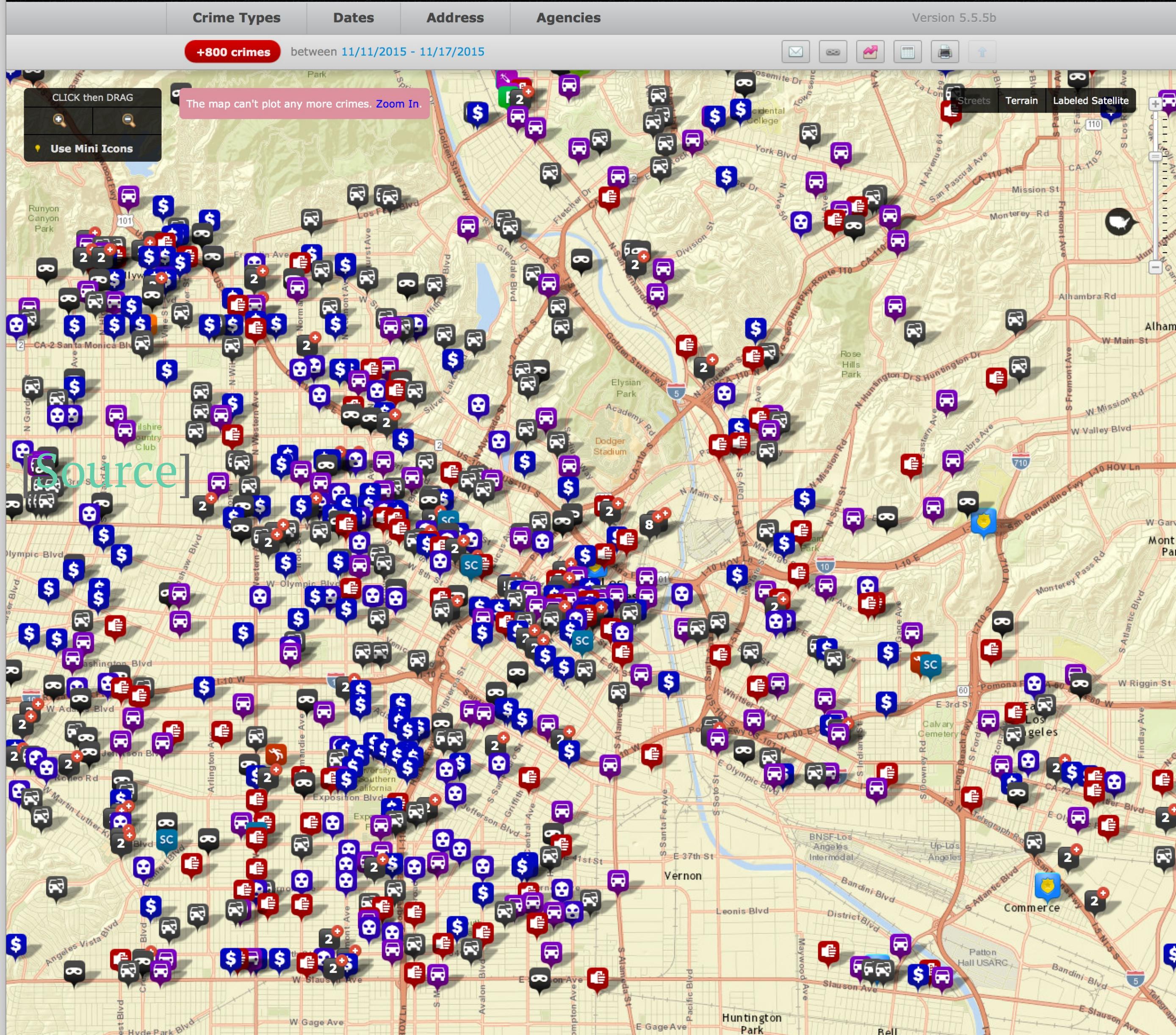
# Points unlike polygons

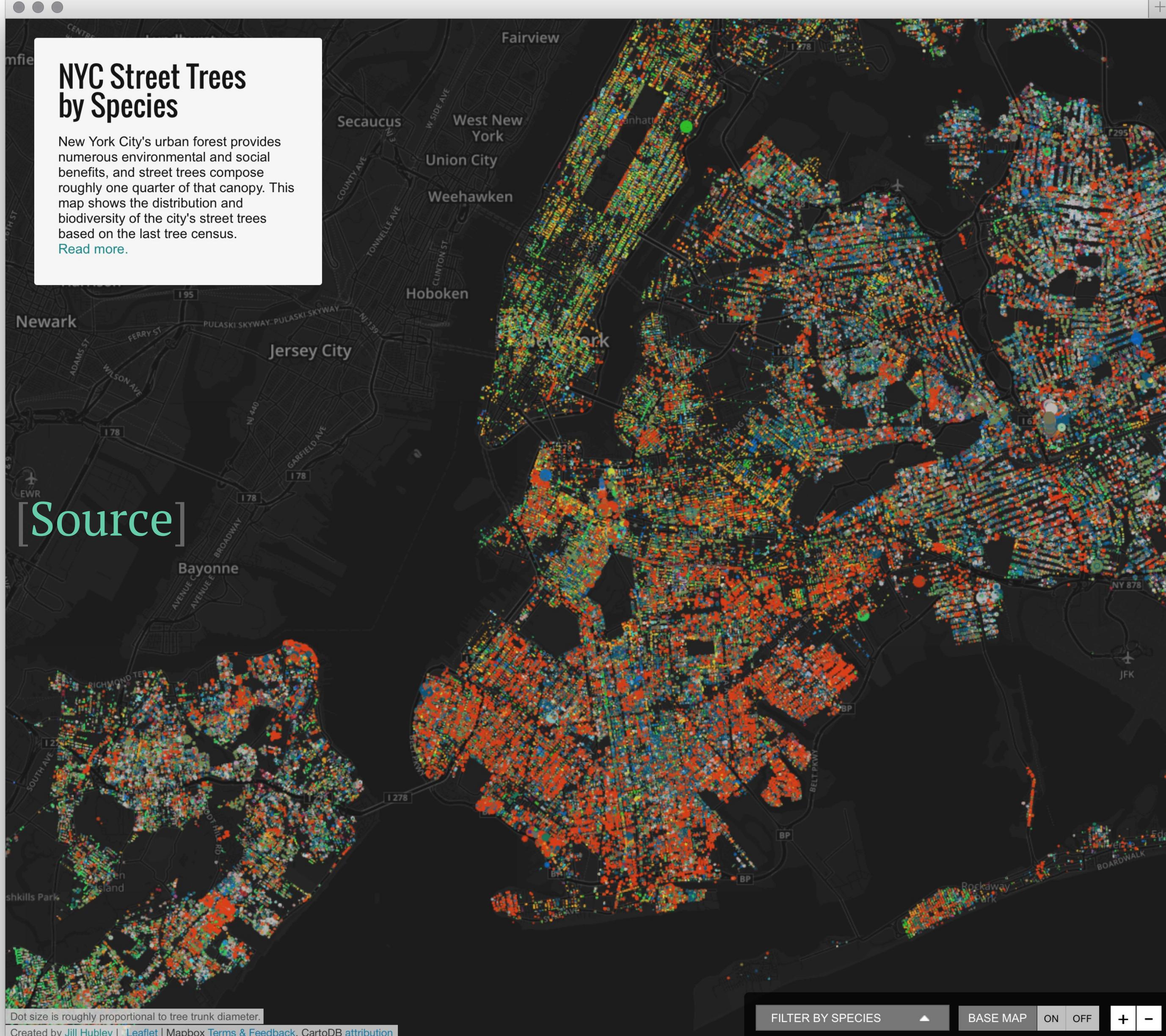
- The interest focuses on **characterizing the pattern** that the points follow **over space**

# Points unlike polygons

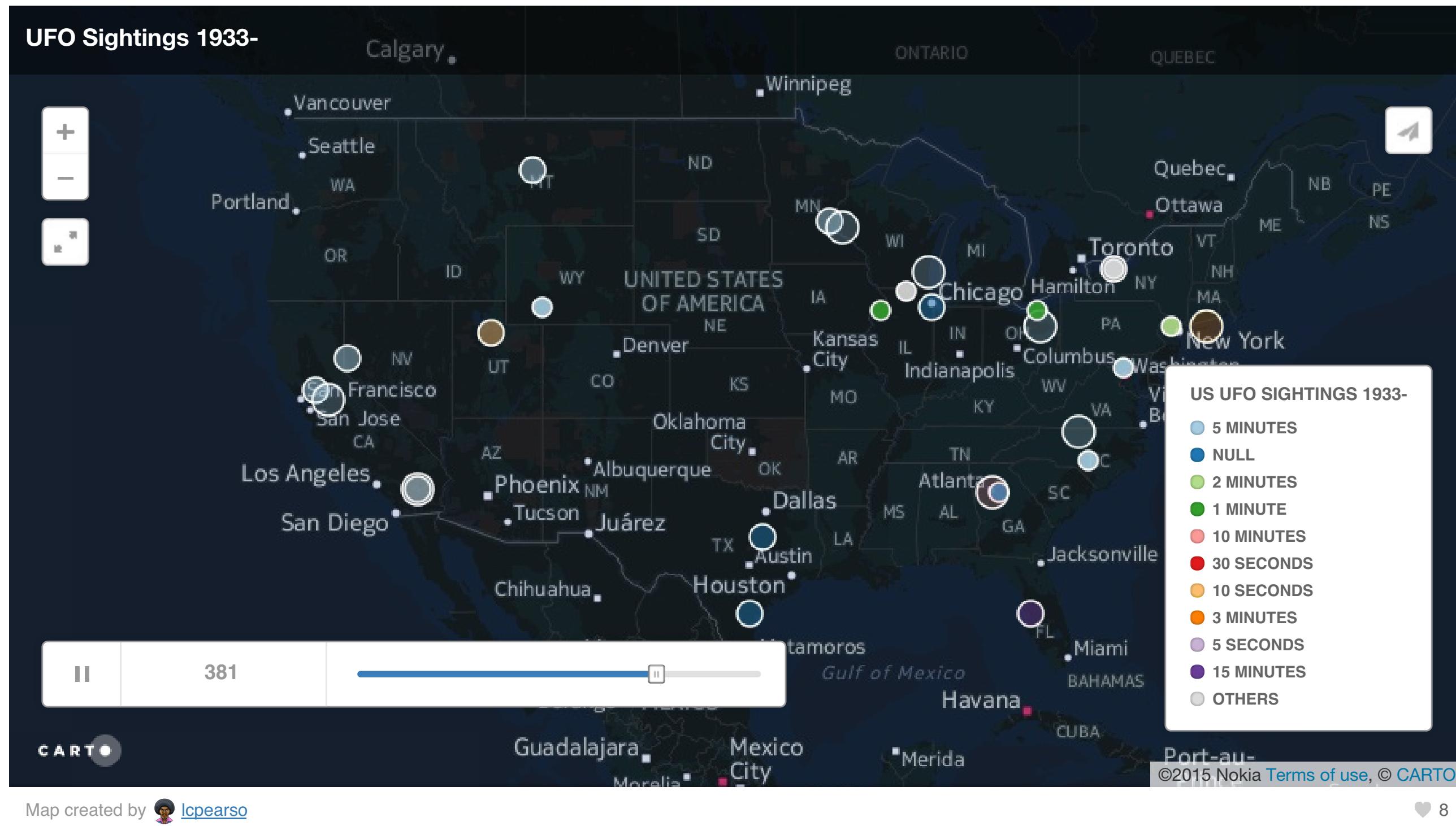
- Rather than exhausting the entire space, points can be **events** subject to **occur anywhere**
- The **location** of the event is **part** of what we are trying to understand/**model**
- The interest focuses on **characterizing** the **pattern** that the points follow **over space**

A few examples...





# UFO Sightings (1933-)



# Geo-tagged tweets



# Point patterns

# Point patterns

Distribution of **points** over a portion of **space**

Assumption is a point can happen anywhere on that space, but only happens in specific locations

# Point patterns

Distribution of **points** over a portion of **space**

Assumption is a point can happen anywhere on that space, but only happens in specific locations

- **Unmarked:** locations only
- **Marked:** values attached to each point

## Point Pattern Analysis

Describe, characterize, and explain point patterns,  
focusing on their **generating process**

- Visual exploration
- Clustering properties
- Statistical modeling of the underlying processes

# Visualization of PPs

# Visualization of PPs

Two routes (today):

- *Aggregate*
- *Smooth*

# Visualization of PPs

Two routes (today):

- *Aggregate*  $\leftrightarrow$  "Histogram"
- *Smooth*

# Visualization of PPs

Two routes (today):

- *Aggregate*  $\leftrightarrow$  "Histogram"
- *Smooth*  $\leftrightarrow$  KDE

# Aggregation

*Points meet polygons*

*Points meet polygons*

Use **polygon** boundaries and **count** points per area

*Points meet polygons*

Use **Polygon** boundaries and **count** points per area

[Insert your skills for **choropleth mapping** here!!!]

*Points meet polygons*

Use **polygon** boundaries and **count** points per area

[Insert your skills for **choropleth mapping** here!!!]

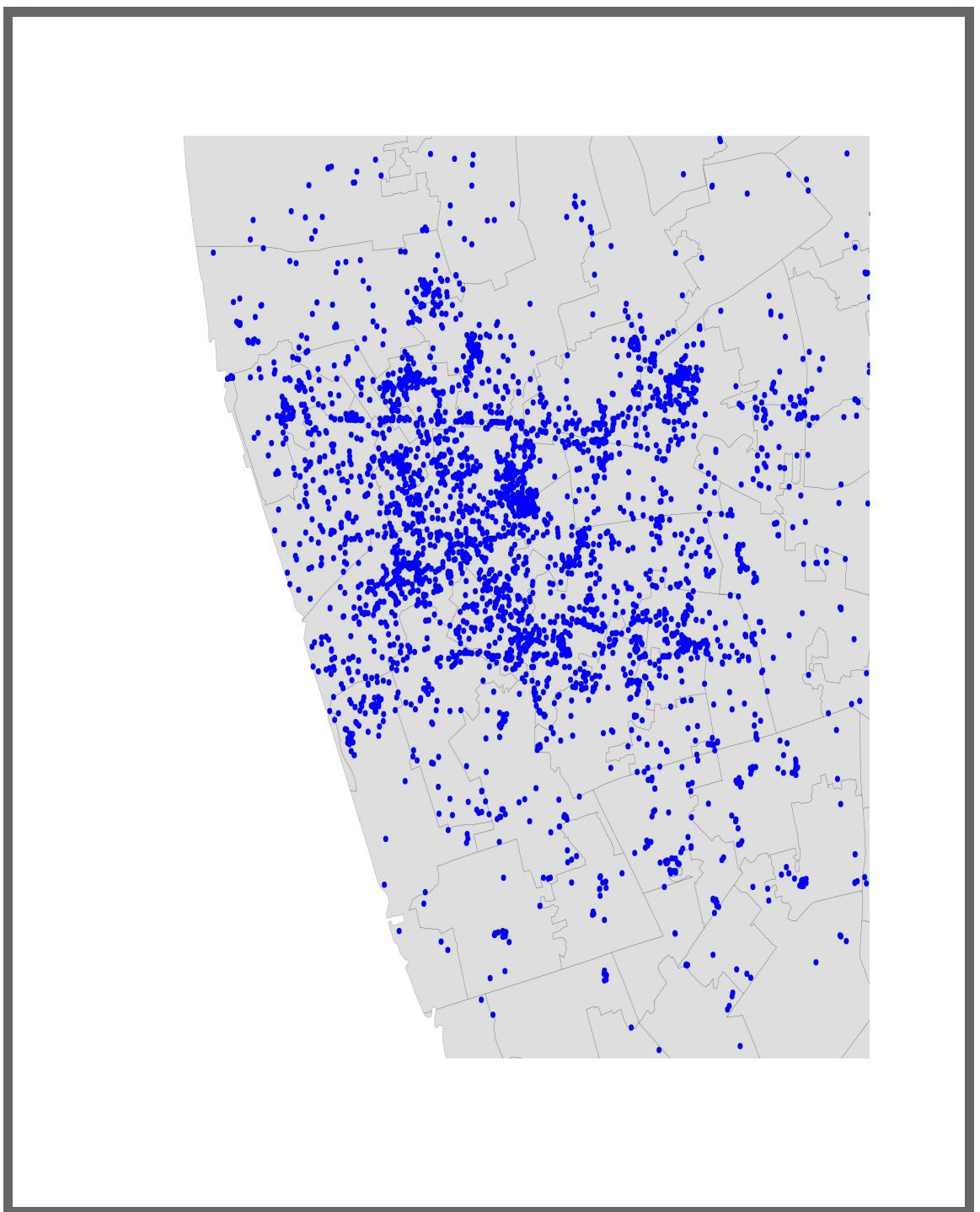
But,

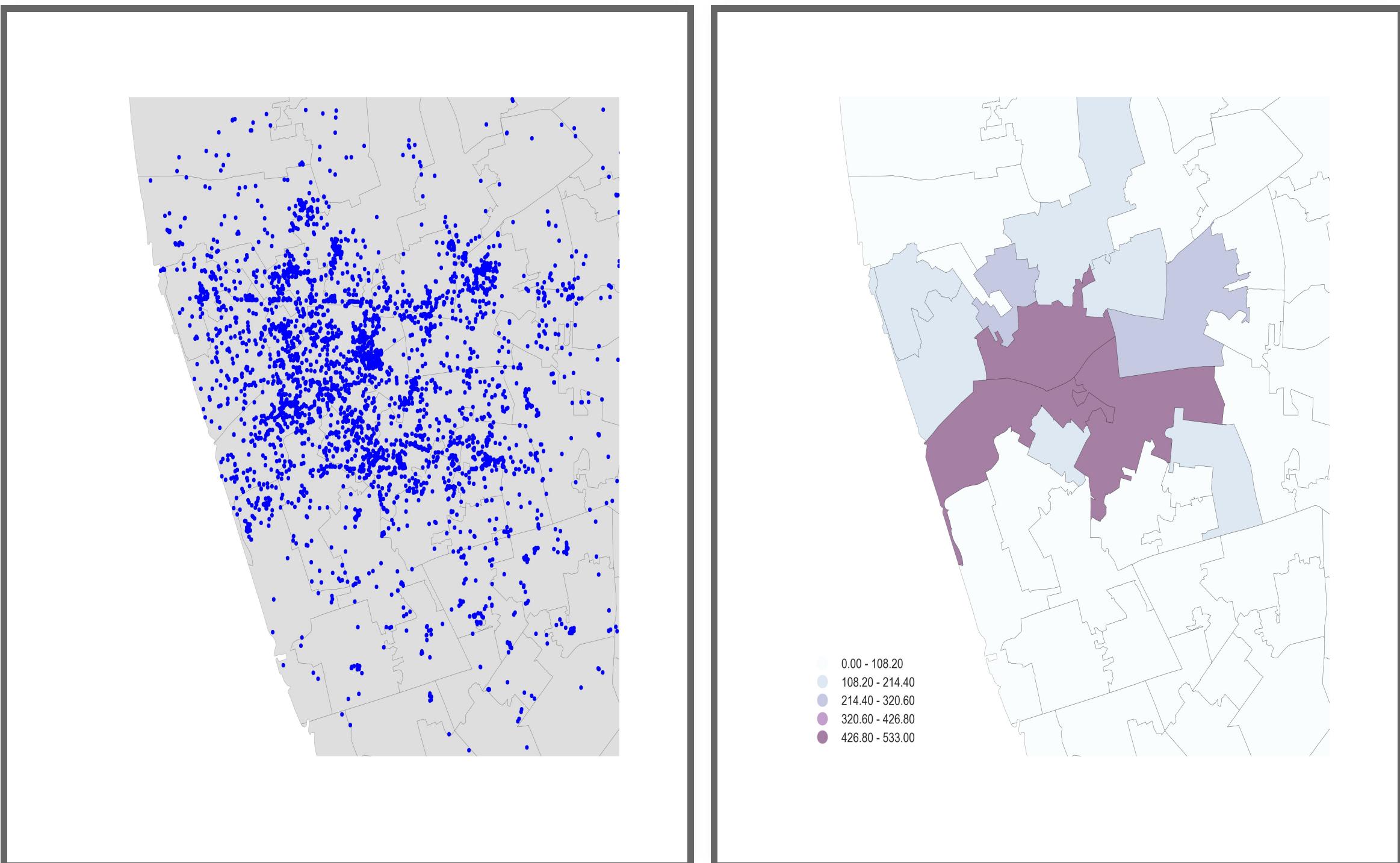
## *Points meet polygons*

Use **polygon** boundaries and **count** points per area

[Insert your skills for **choropleth mapping** here!!!]

But, the polygons need to "make sense" (their delineation needs to relate to the point generating process)





# Hex-binning

If no polygon boundary seems like a good candidate  
for aggregation...

# Hex-binning

If no polygon boundary seems like a good candidate for aggregation...

...draw a **hexagonal** (or squared) **tesselation!!!**

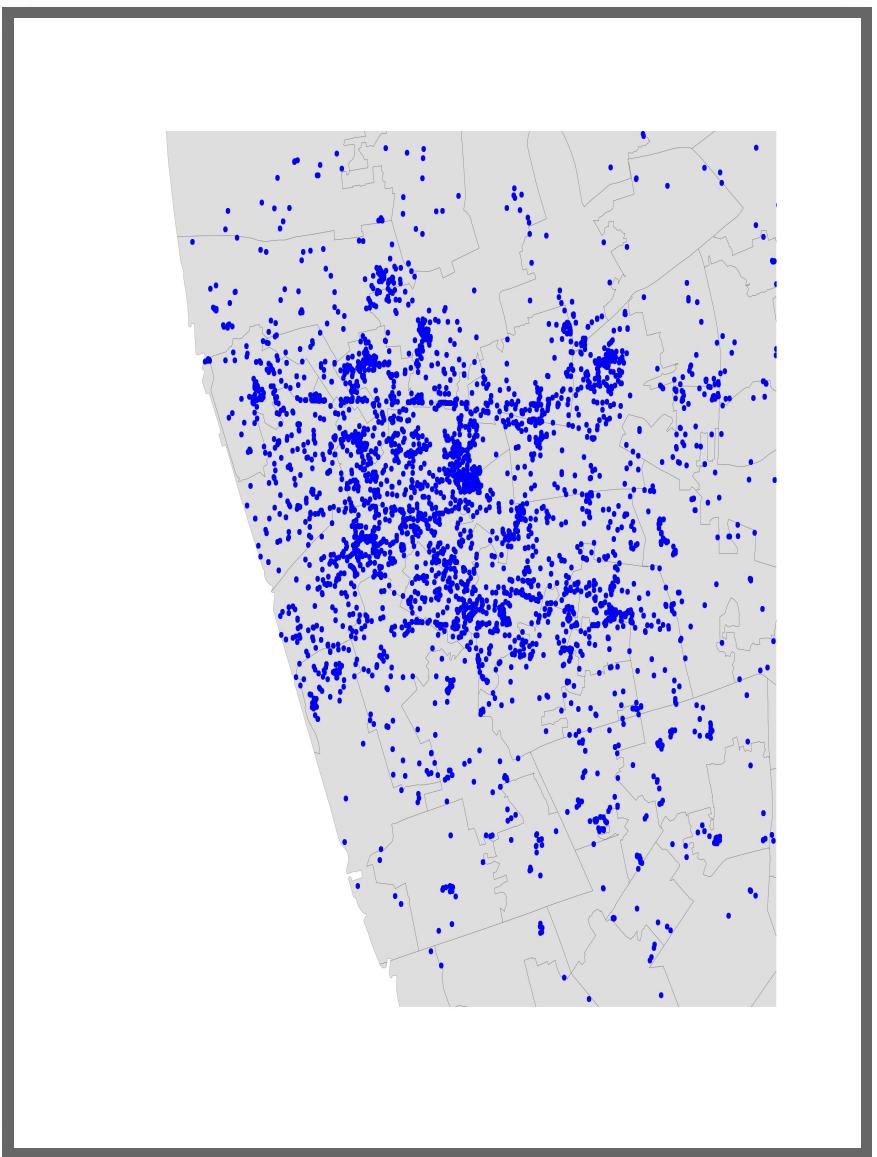
# Hex-binning

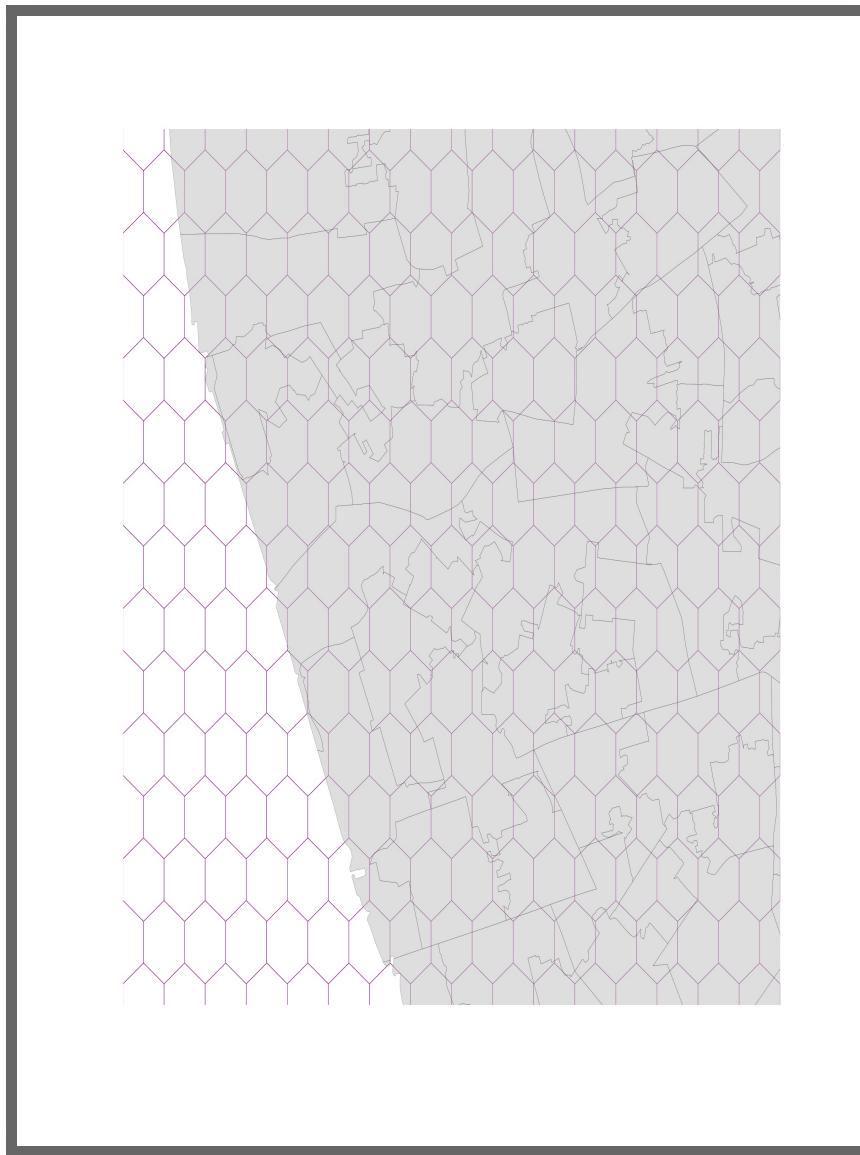
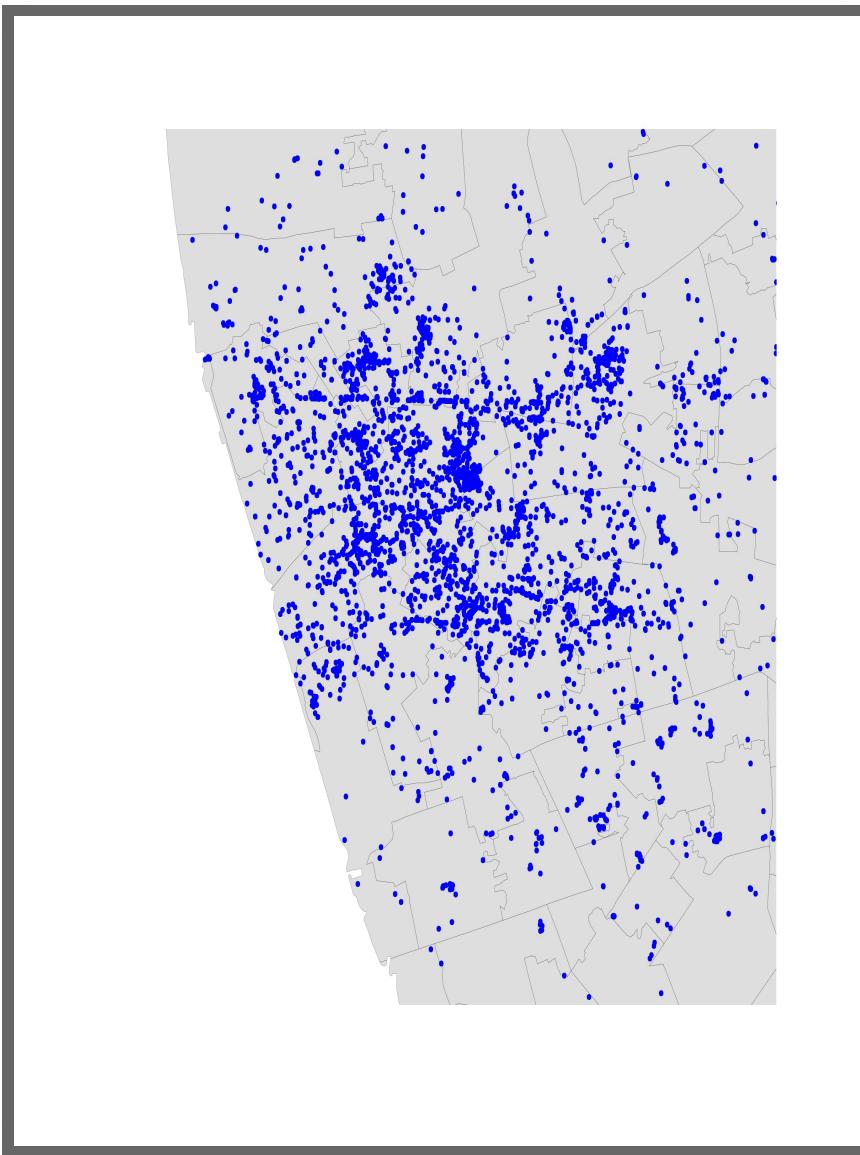
If no polygon boundary seems like a good candidate for aggregation...

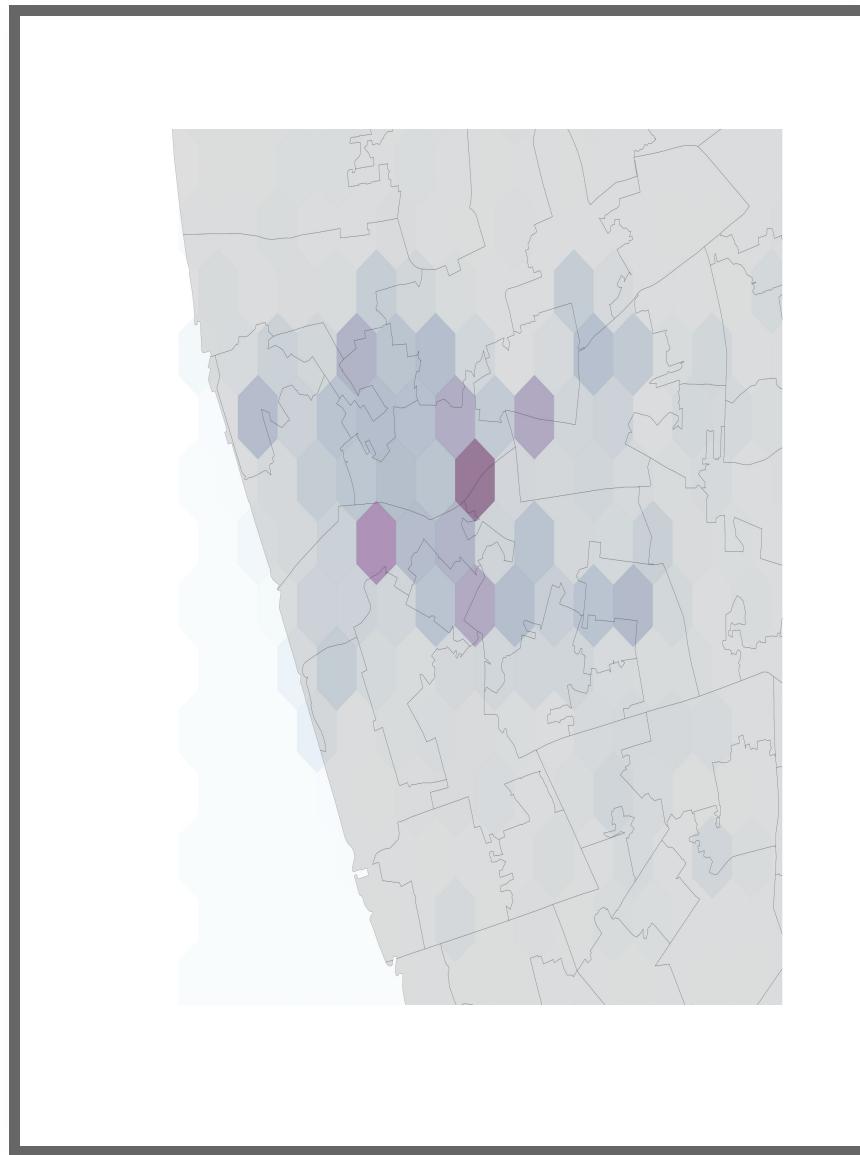
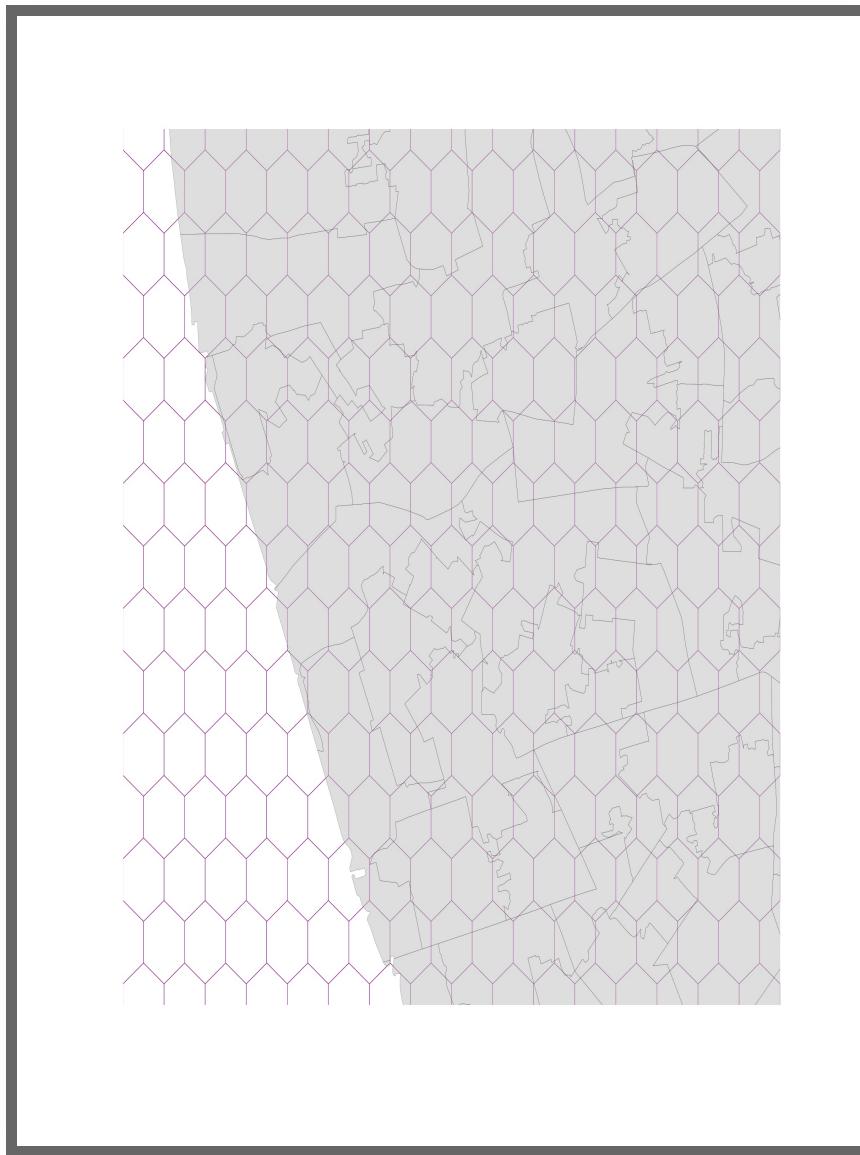
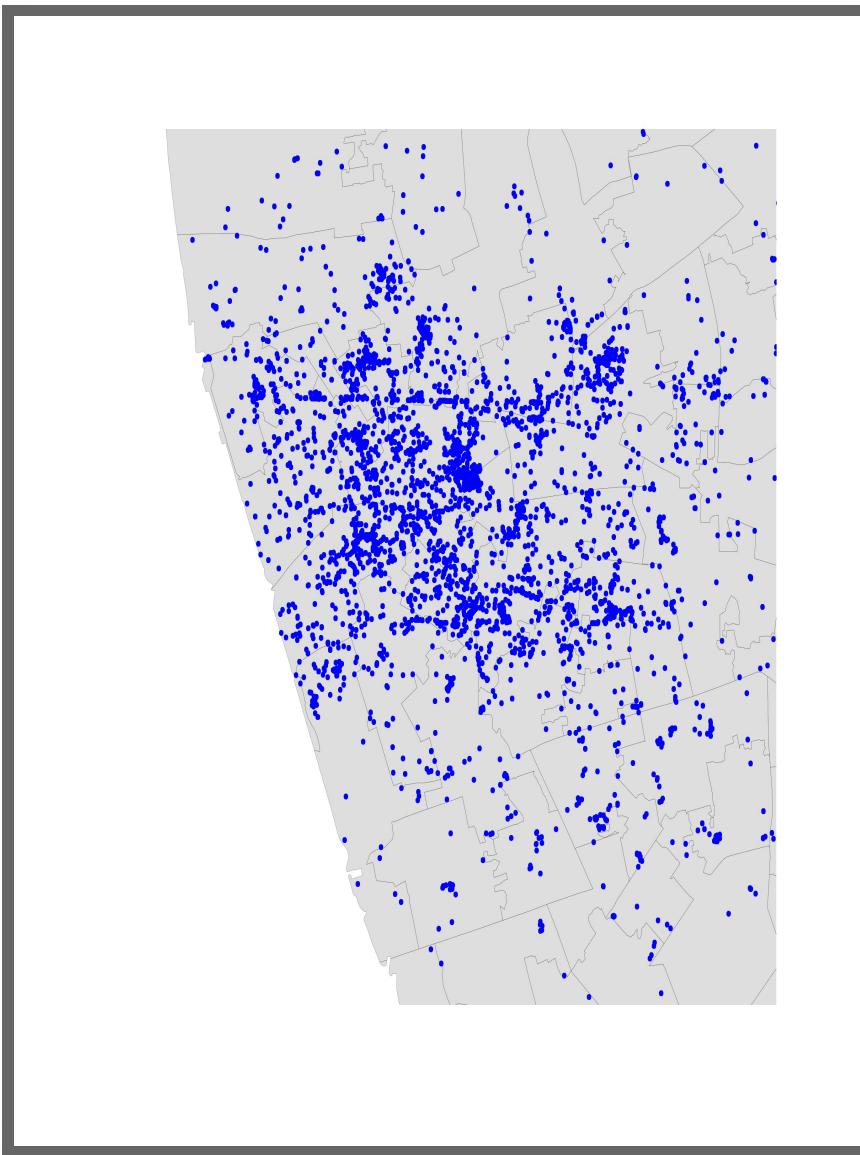
...draw a **hexagonal** (or squared) **tesselation!!!**

Hexagons...

- Are **regular**
- **Exhaust** the space (Unlike circles)
- Have **many sides** (minimize boundary problems)







But...

But...

(Arbitrary) aggregation may induce MAUP (see  
Lecture 4)

But...

(Arbitrary) aggregation may induce MAUP (see  
Lecture 4)

+

But...

(Arbitrary) aggregation may induce MAUP (see  
Lecture 4)

+

Points usually represent events that affect to only  
**part** of the population and hence are best considered  
as **rates** (see Lecture 4)

# Kernel Density Estimation

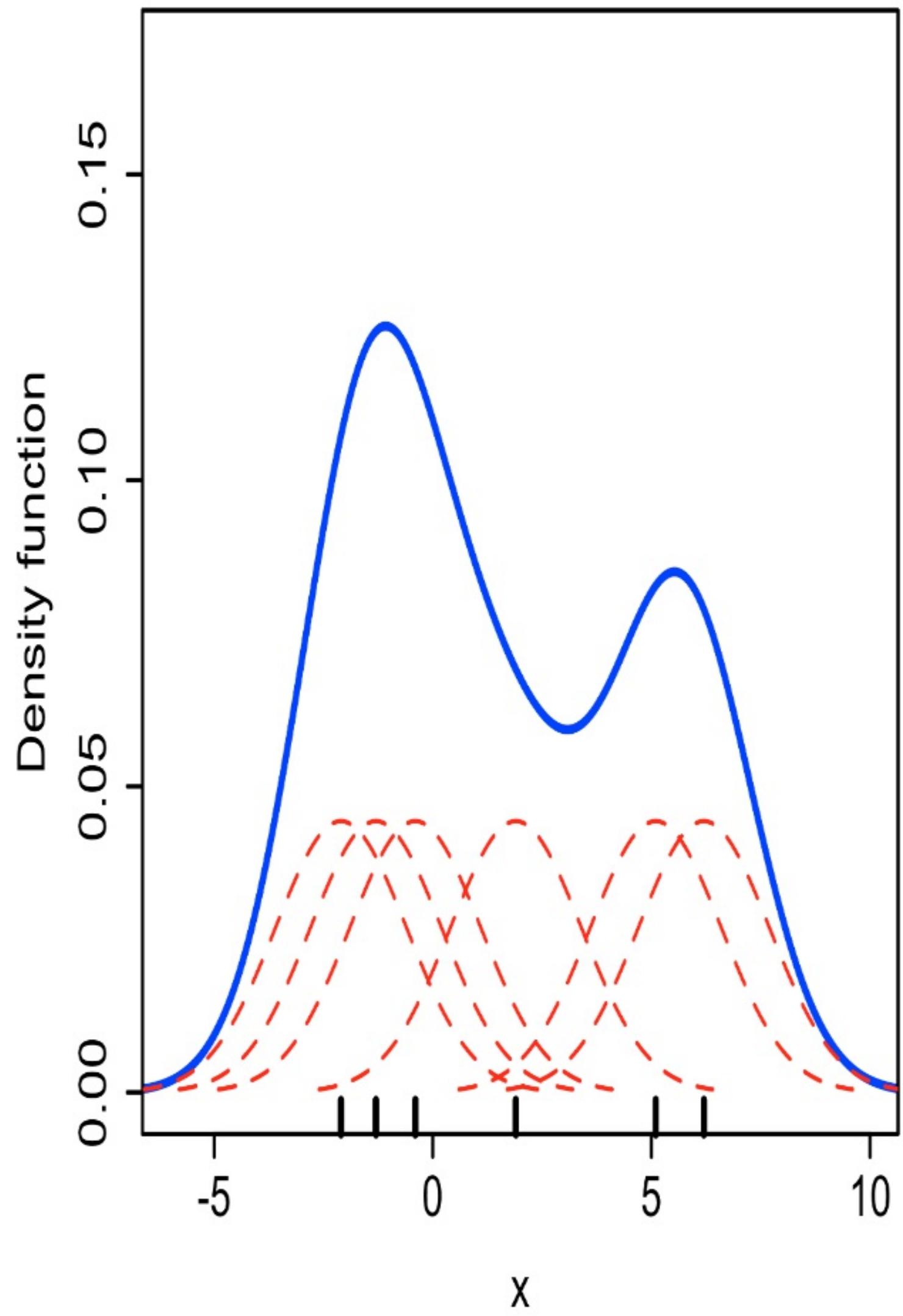
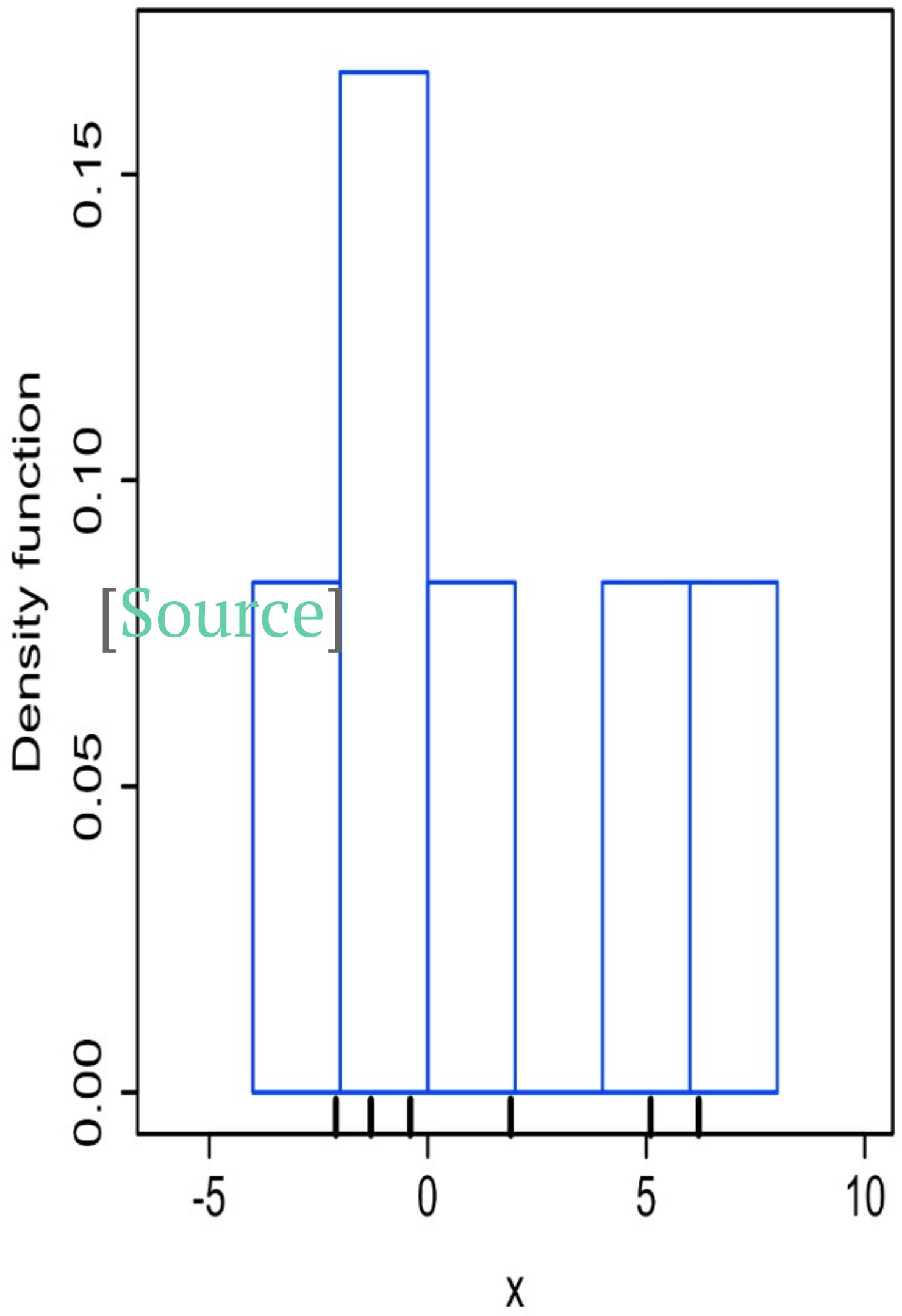
# Kernel Density Estimation

*Estimate the (continuous) observed distribution of a variable*

# Kernel Density Estimation

*Estimate the (continuous) observed distribution of a variable*

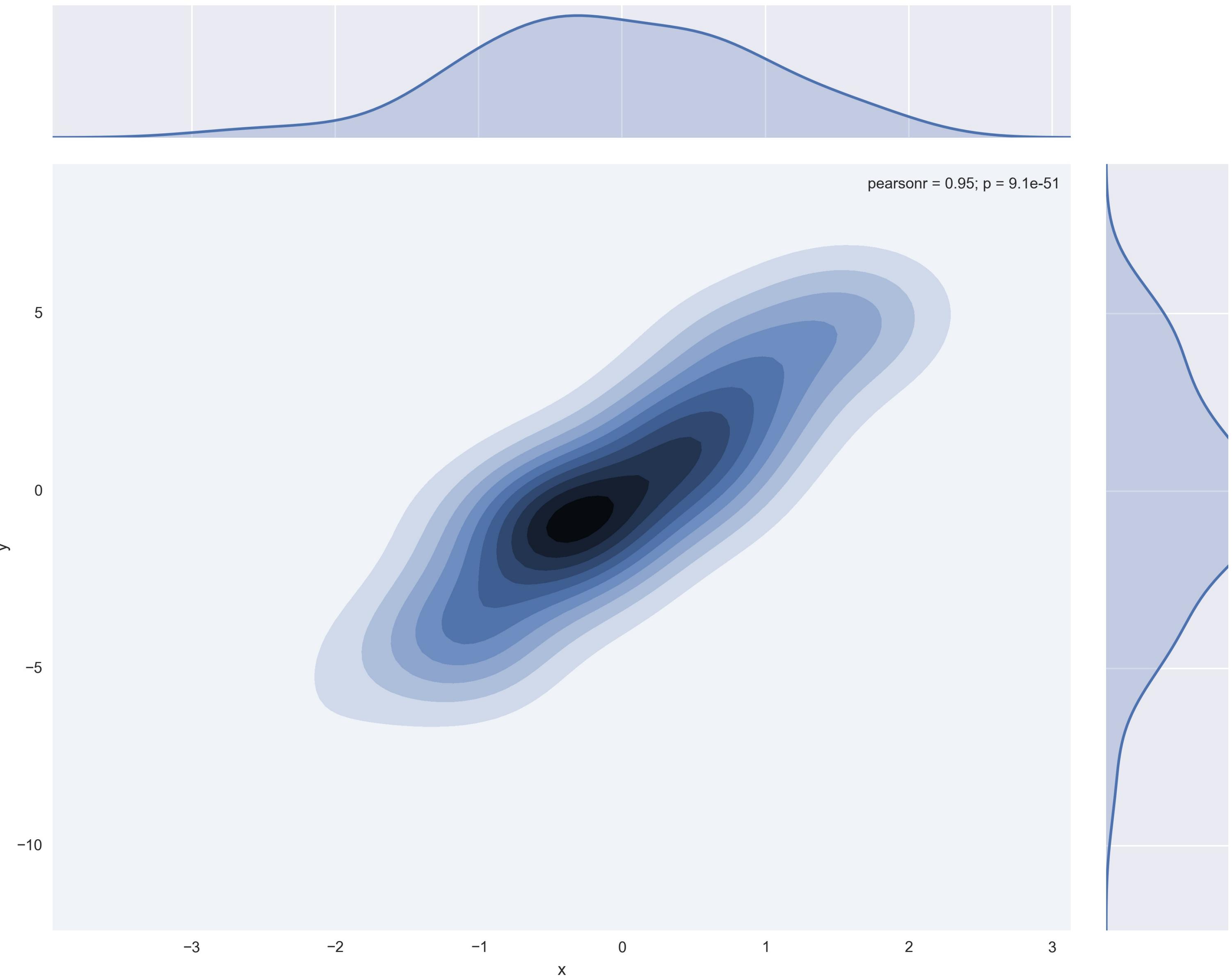
- Probability of finding an observation at a given point
- "Continuous histogram"
- Solves (much of) the MAUP problem, but not the underlying population issue

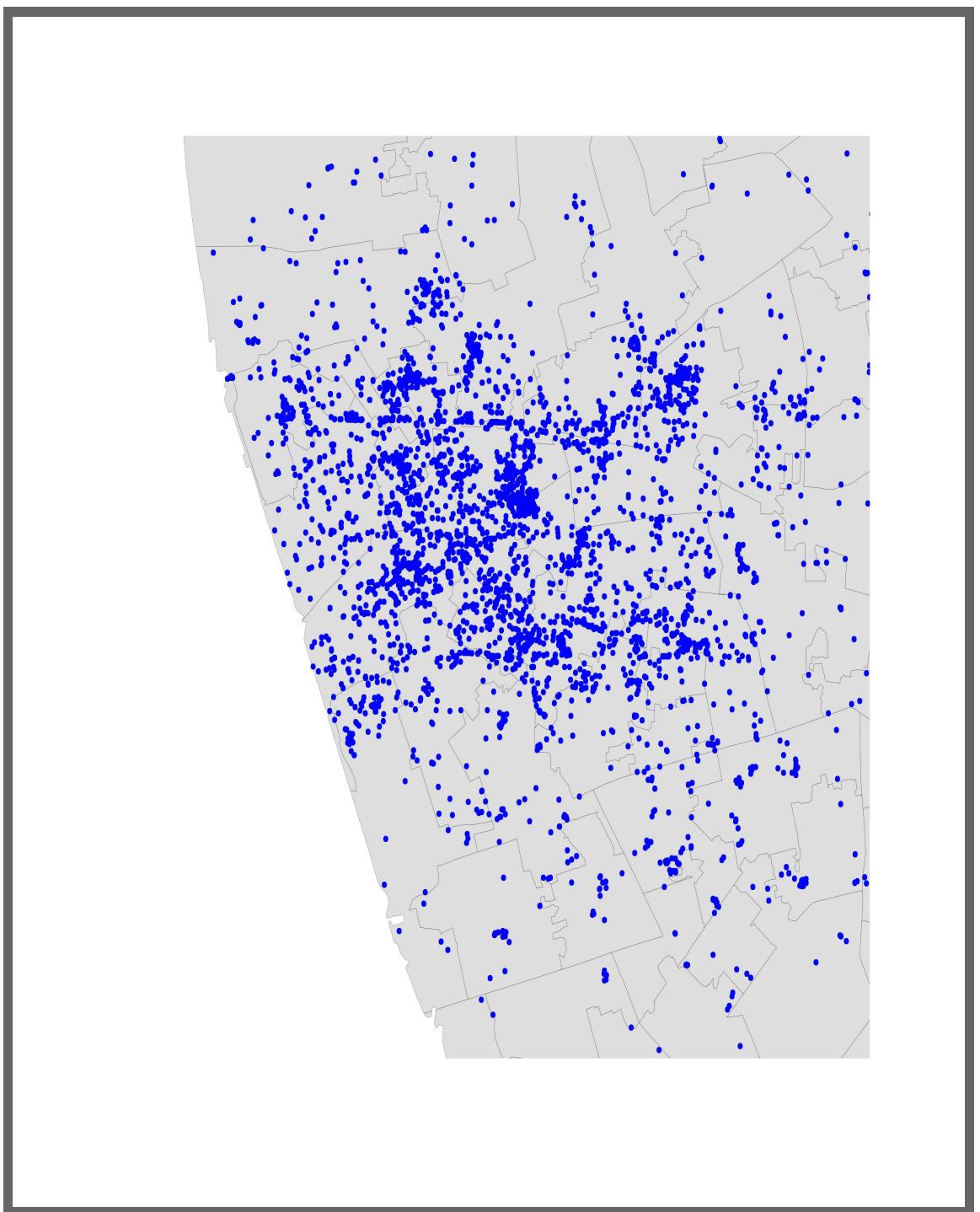


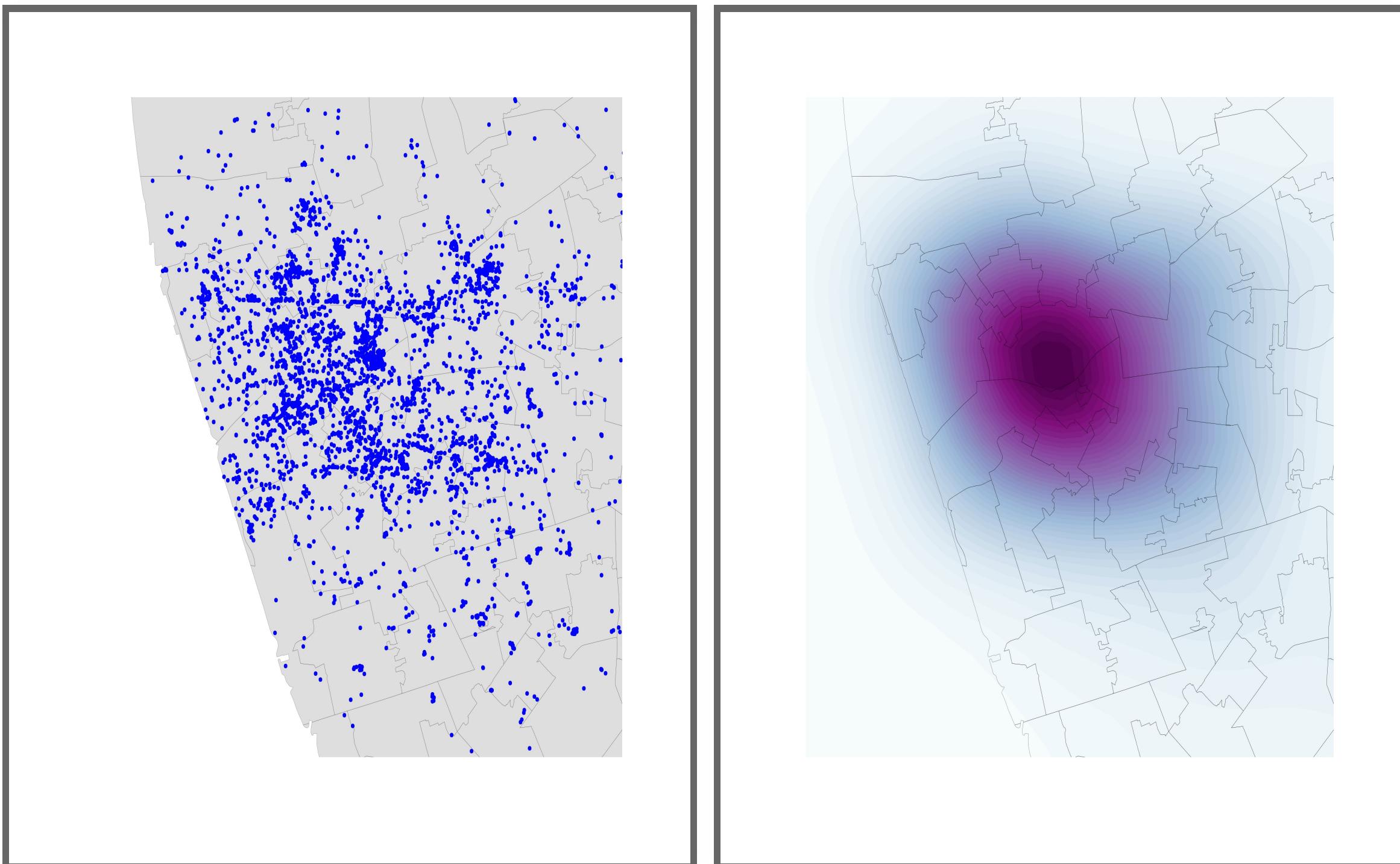
# Bivariate (spatial) KDE

*Probability of finding observations at a given point in space*

- **Bivariate** version: distribution of **pairs of values**
- In **space**: values are coordinates (XY), locations
- Continuous "version" of a choropleth







# Recapitulation

- Points can be understood as a **fixed** or **random** process over space
- If seen as a random, *where* points are located is part of the interest in the (**point pattern**) analysis
- **Visualization** of point patterns can be done through **aggregation** or **smoothing** (but issues relating to the MAUP and underlying populations need to be kept in mind!)



Geographic Data Science'16 - Lecture 8 by [Dani Arribas-Bel](#) is licensed under a [Creative Commons Attribution-NonCommercial-ShareAlike 4.0 International License](#).