# Geographic Data Science -Lecture IX

Points

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### Today

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

## The point of points

Points can represent "fixed" entities

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In this case, points are qualitatively similar to

polygons/lines

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

Examples:

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• Cities (in most cases)

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- Buildings

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•

# When points are not polygons

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

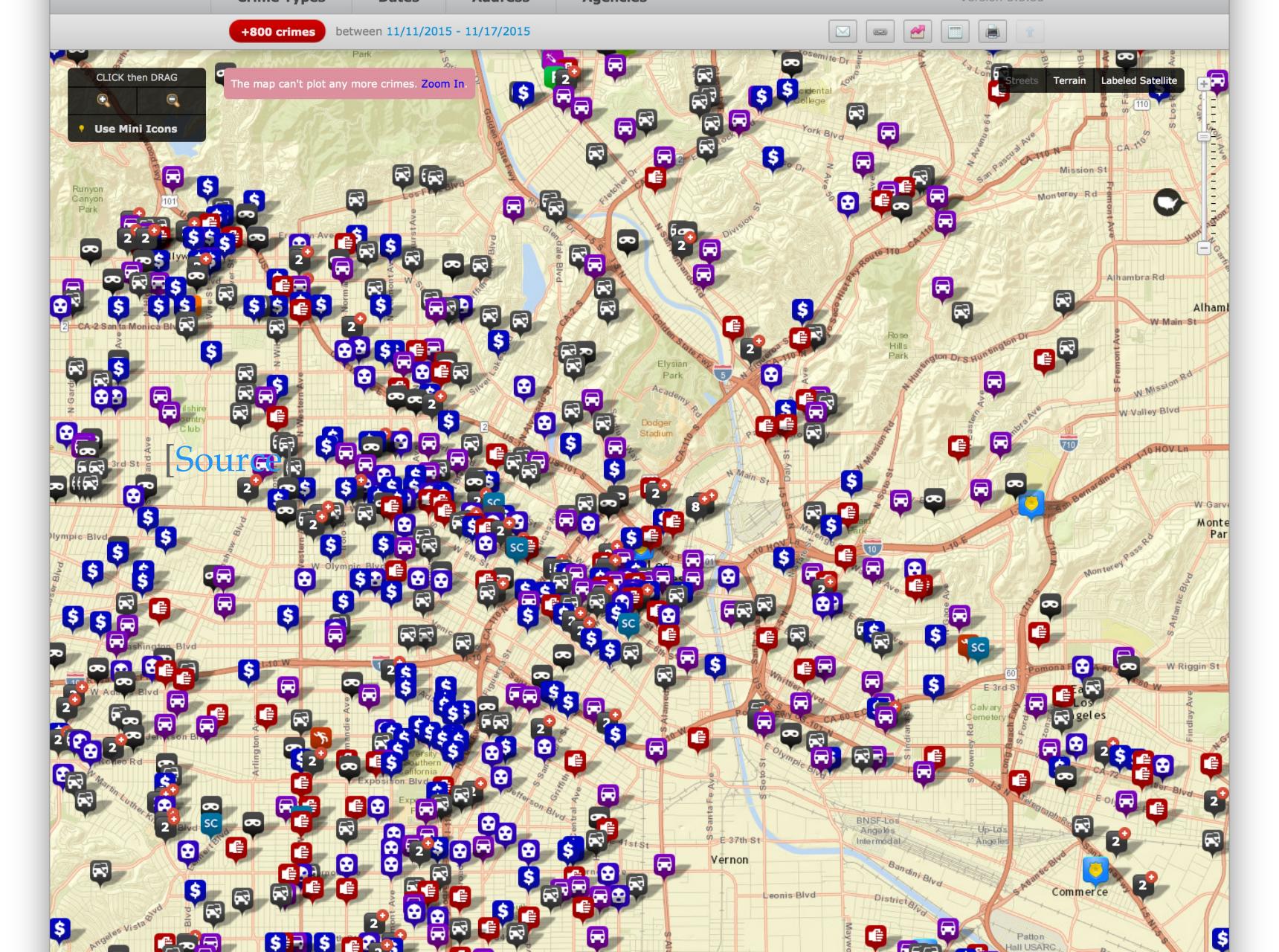
• Rather than exhausting the entire space, points can be events subject to occur anywhere

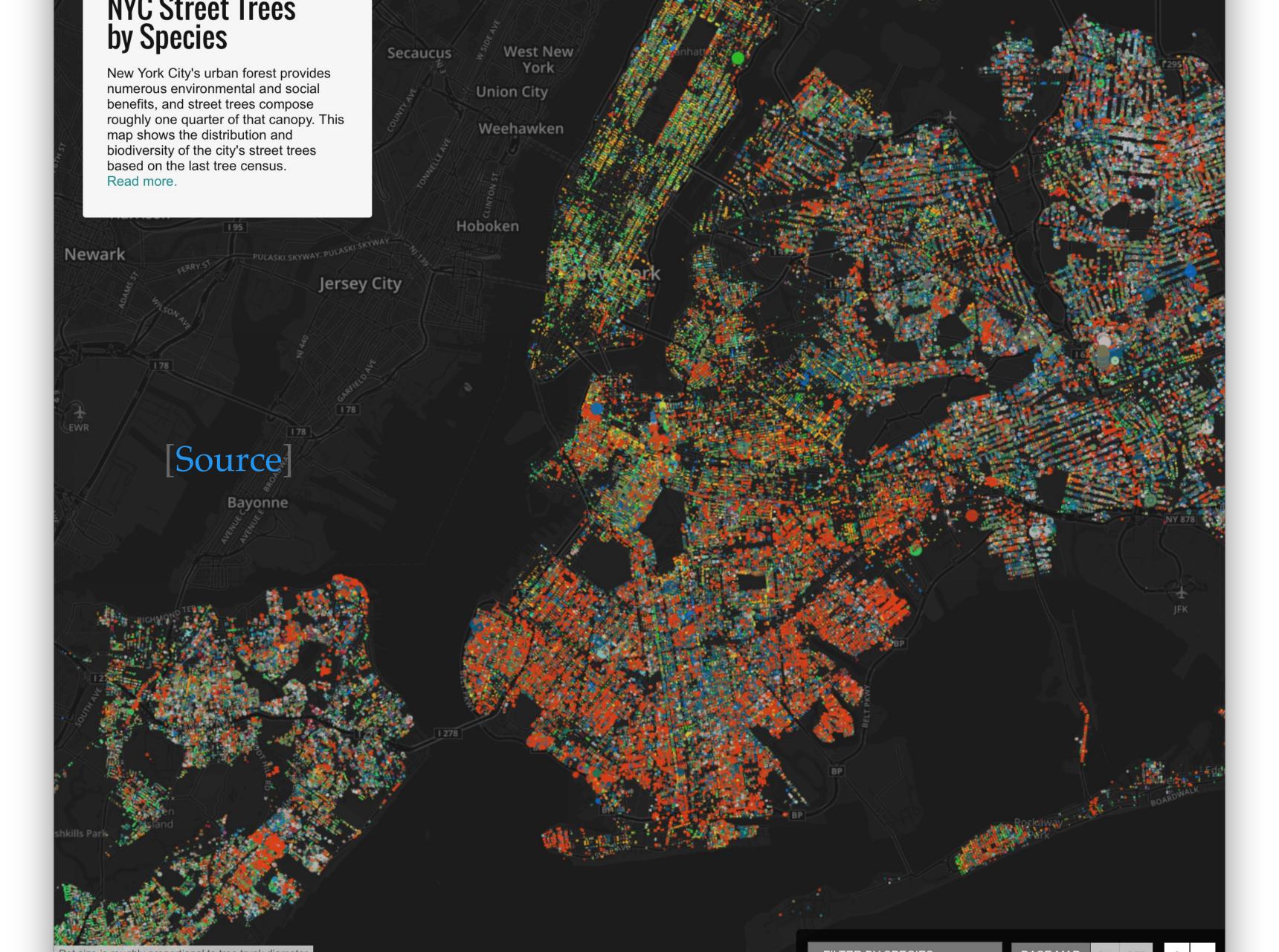
• The **location** of the event is **part** of what we are trying to understand/**model** 

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

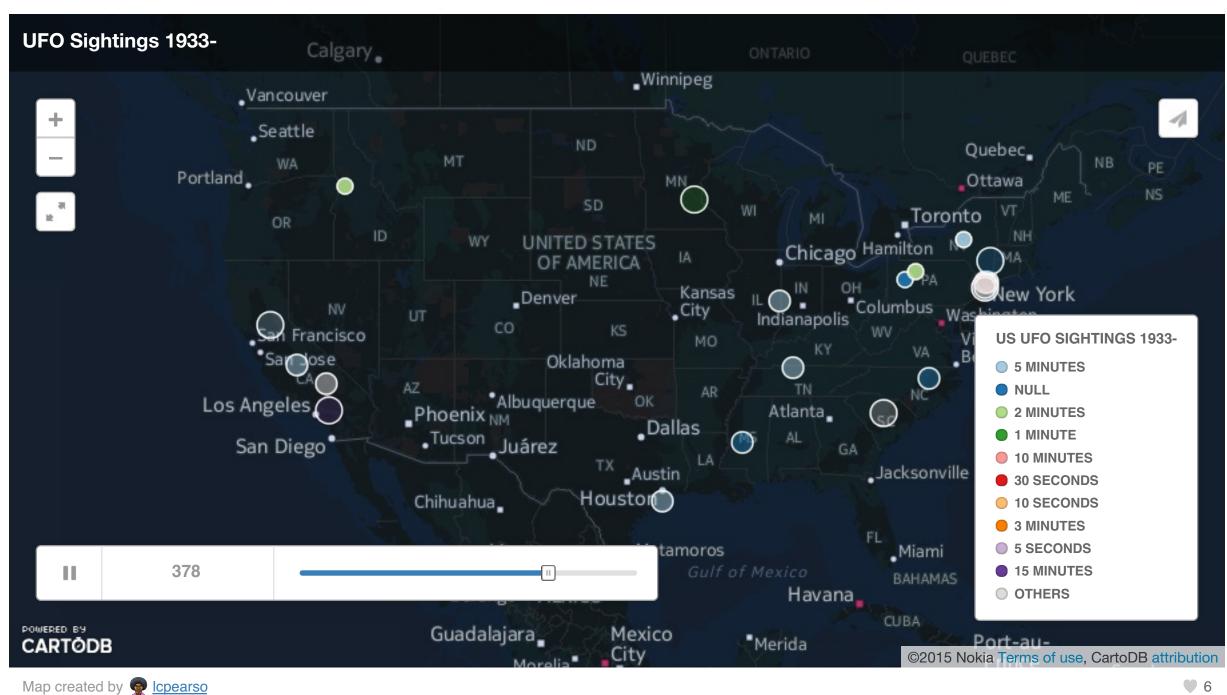
- Rather than exhausting the entire space, points can be events subject to occur anywhere
- The **location** of the event is **part** of what we are trying to understand/**model**
- The interest focuses on **characterizing** the **pattern** that the points follow **over space**

A few examples...





## UFO Sightings (1933-)



## Geo-tagged tweets



## Point patterns

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Distribution of points over a portion of space

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

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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

Two routes (today):

- Aggregate
- Smooth

Two routes (today):

- *Aggregate* ↔ "Histogram"
- Smooth

Two routes (today):

- *Aggregate* ↔ "Histogram"
- $Smooth \Leftrightarrow KDE$

## Aggregation



# Points meet polygons Use polygon boundaries and count points per area

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

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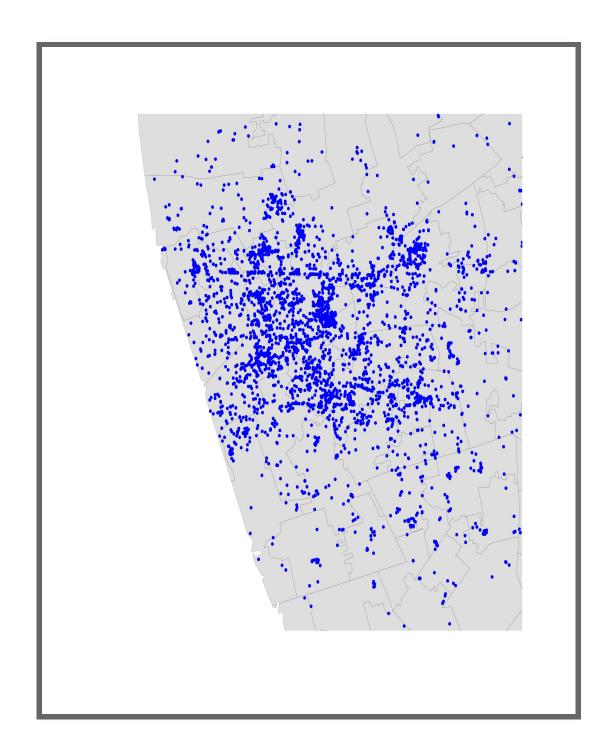
But,

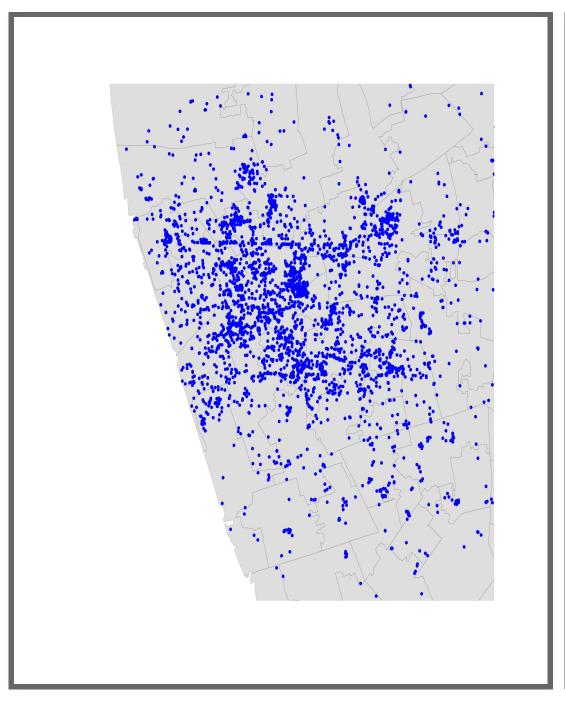
#### Points meet polygons

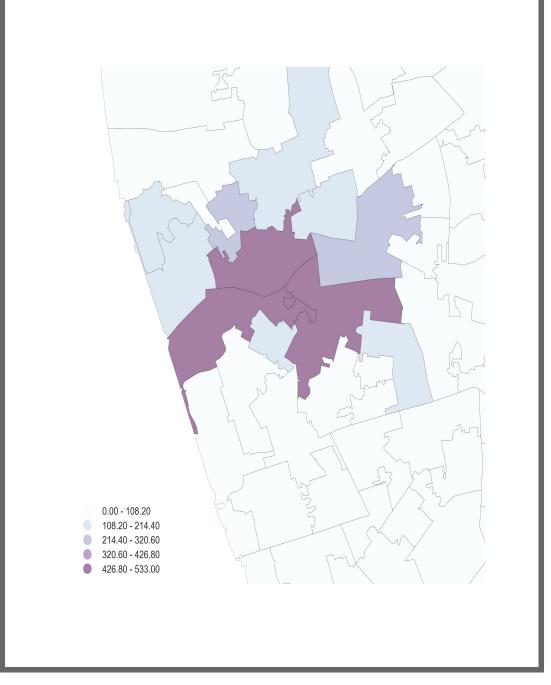
Use polygon boundaries and count points per area

[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...

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

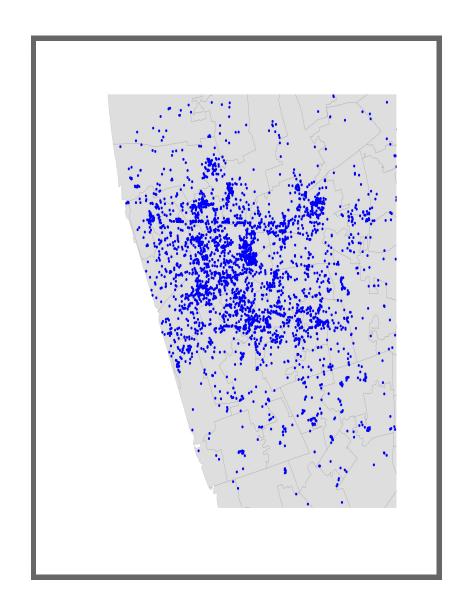
### Hex-binning

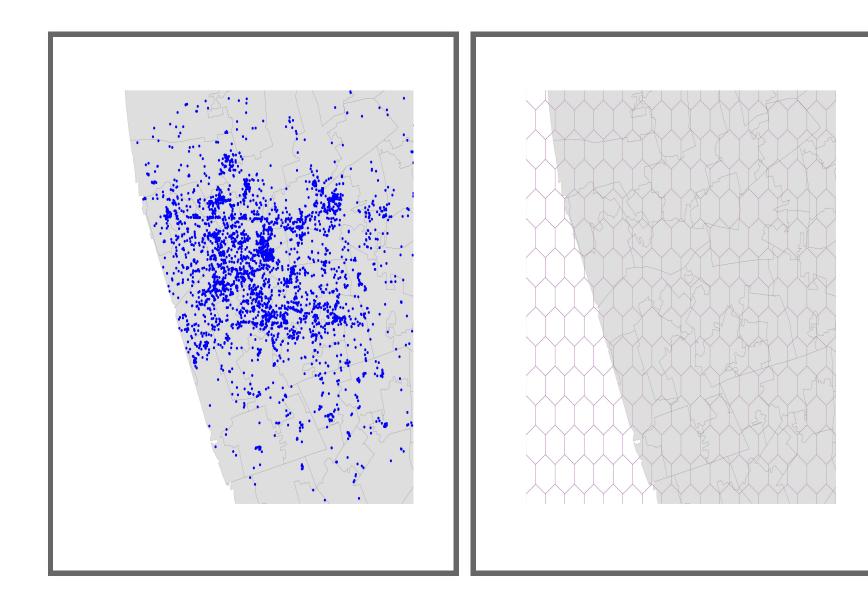
If **no** polygon boundary seems like a **good candidate** for aggregation...

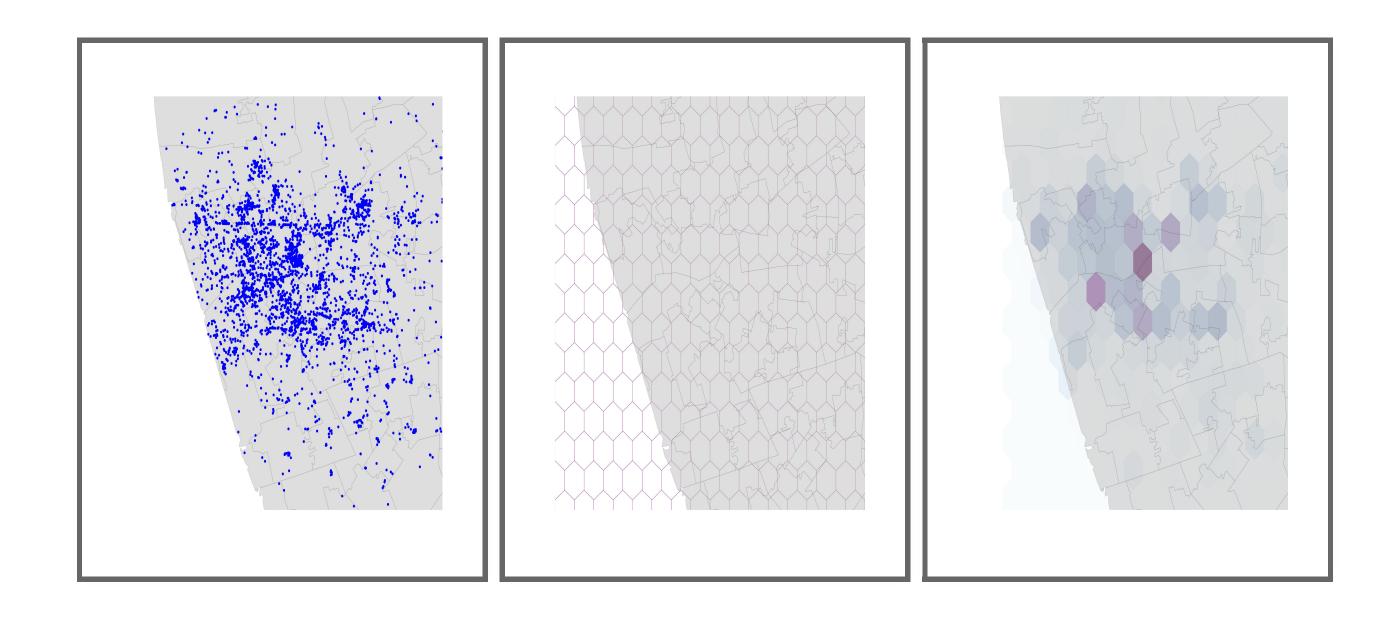
...draw a hexagonal (or squared) tesselation!!!

Hexagons...

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







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

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+

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

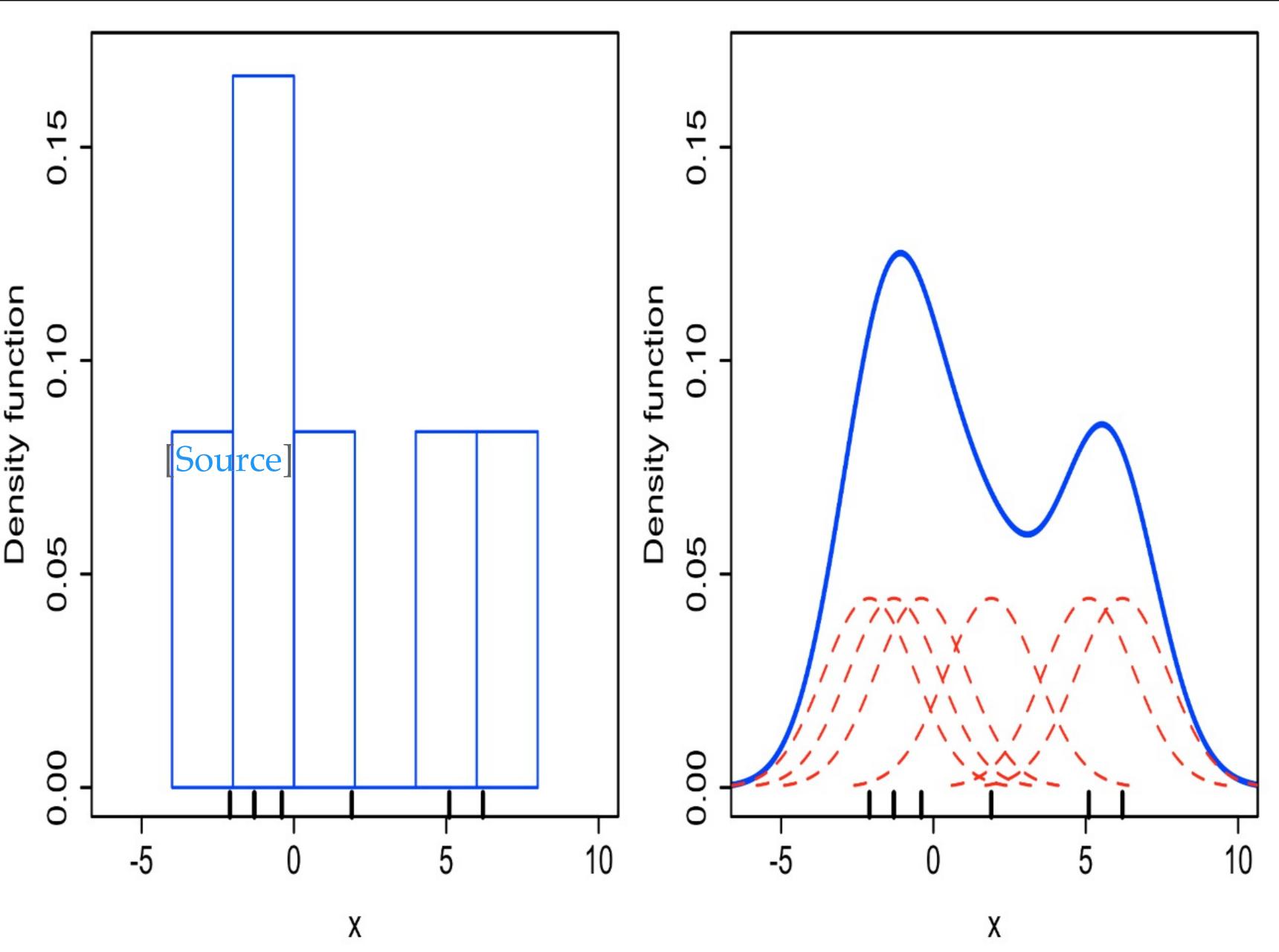
### Kernel Density Estimation

Estimate the (continuous) observed distribution of a variable

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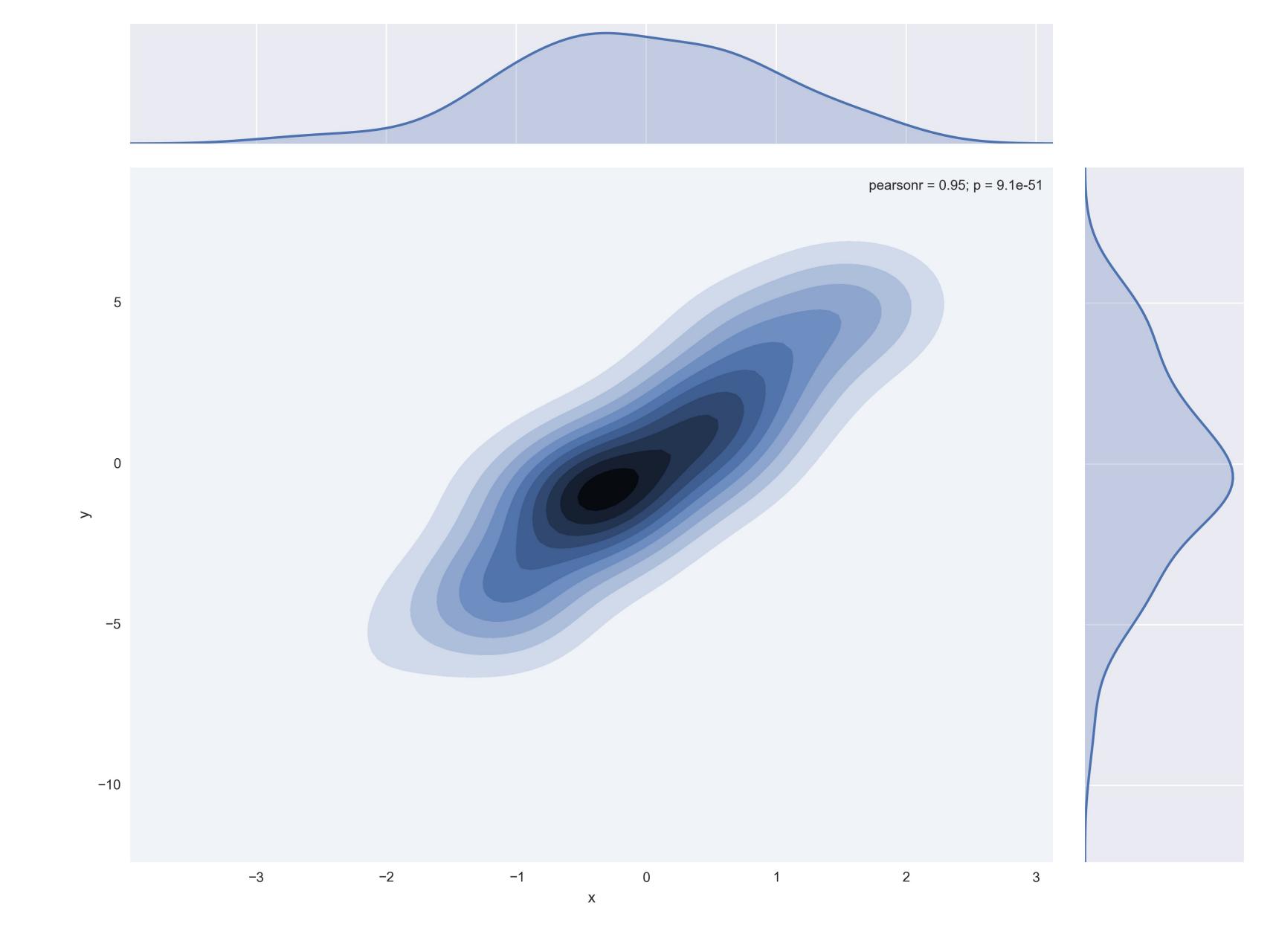
- Probability of finding an observation at a given point
- "Continuous histogram"
- Solves (much of) the MAUP problem, but not the underlying population issue

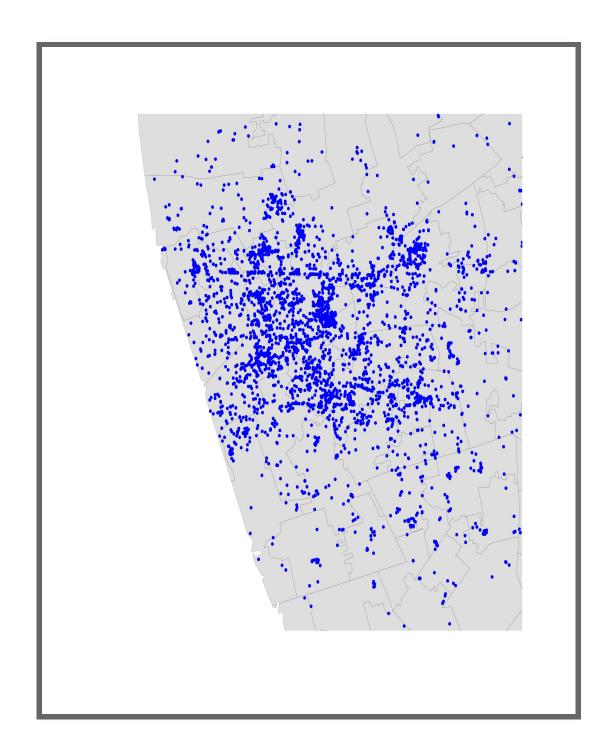


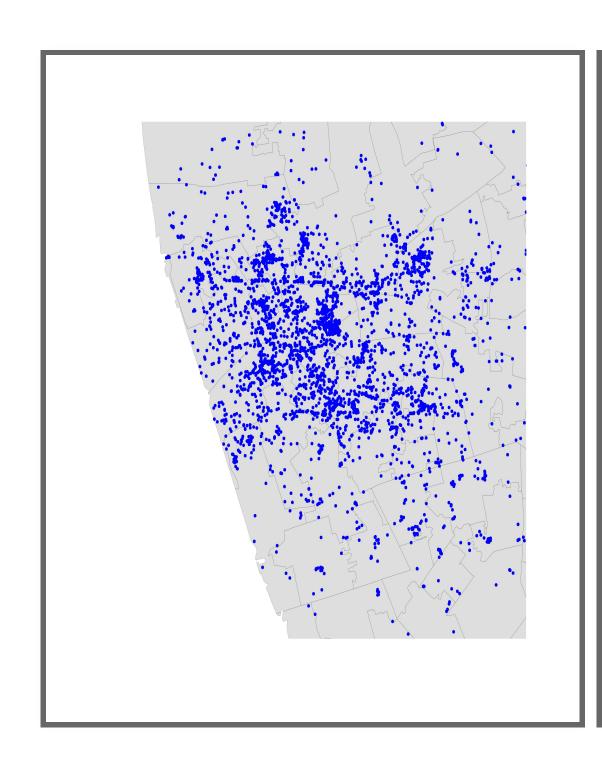
# Bivariate (spatial) KDE

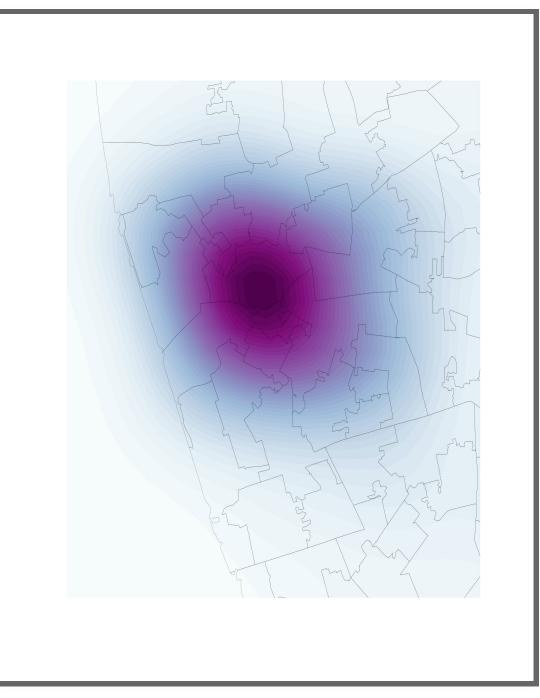
Probability of finding observations at a given point in space

- 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!)



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