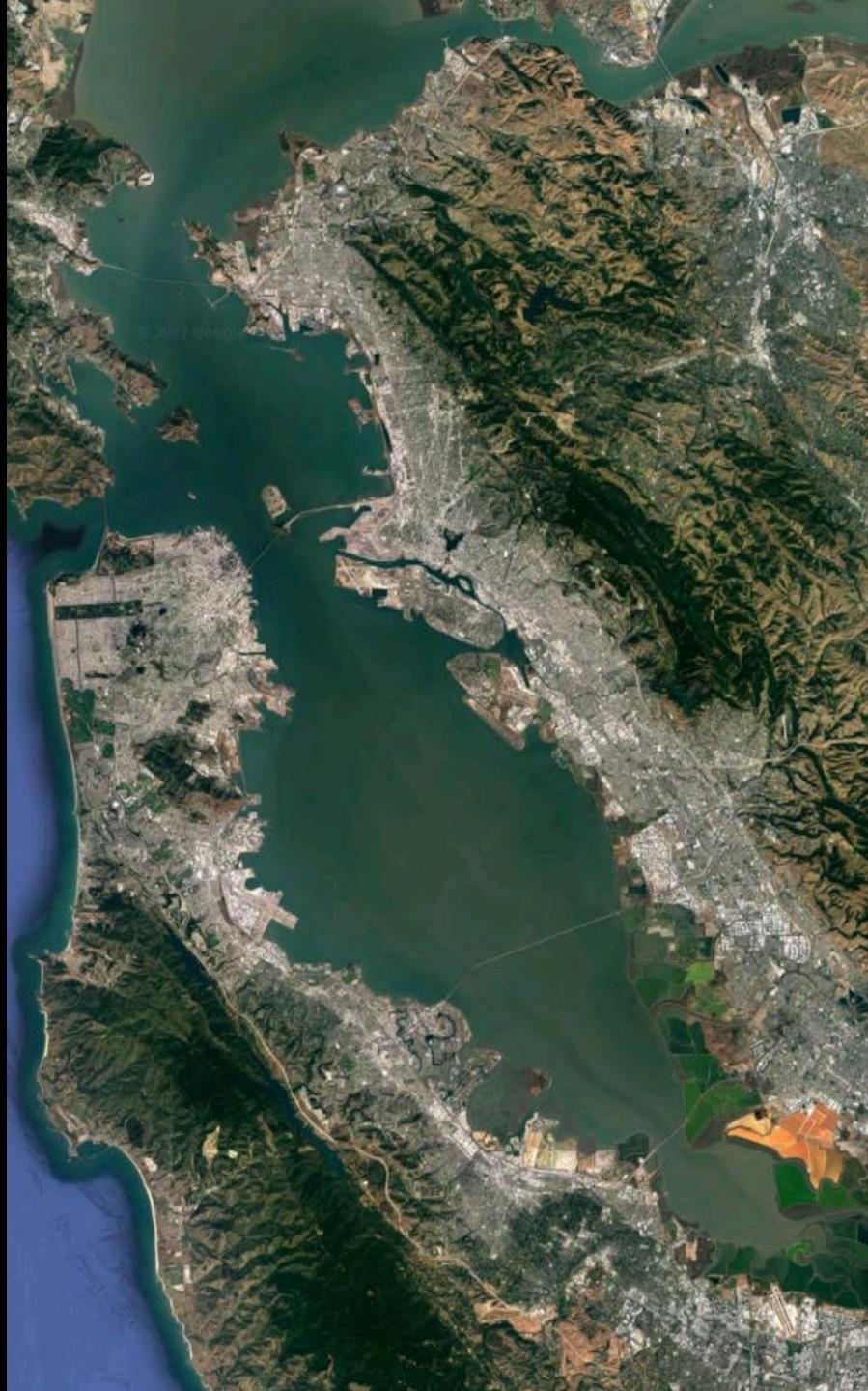


# Spatial Data & GIS Fundamentals

CP101, Summer 2020

Department of City + Regional Planning, UC Berkeley  
Chester Harvey

It's hard to  
study cities  
without maps



# Maps are places to visualize and compare spatial data

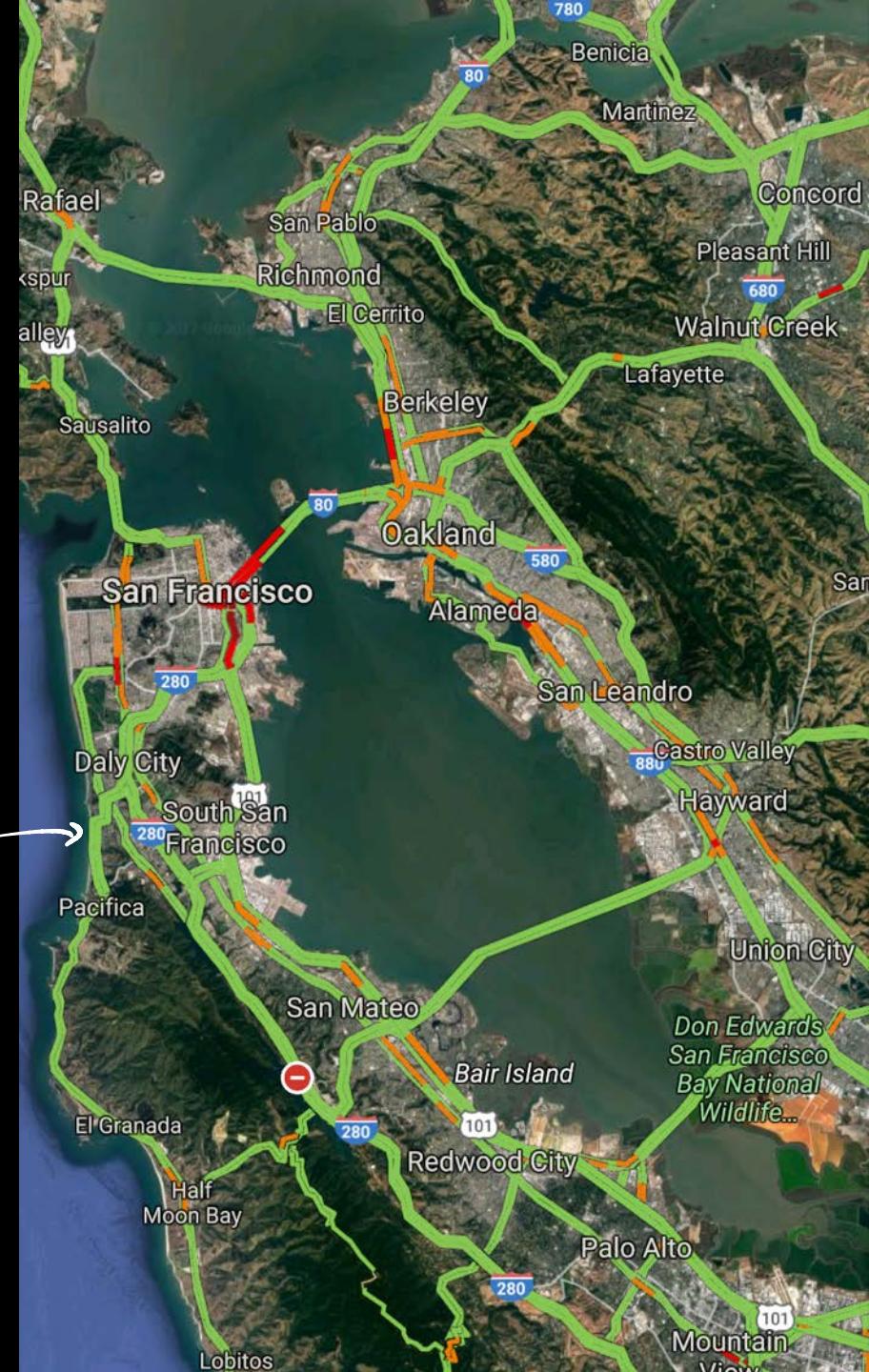
SPACE

MAPS → Visualize BY OVERLAY

↳ UNDERSTAND CHARACTERISTIC

OVERLAY

Google, 2017

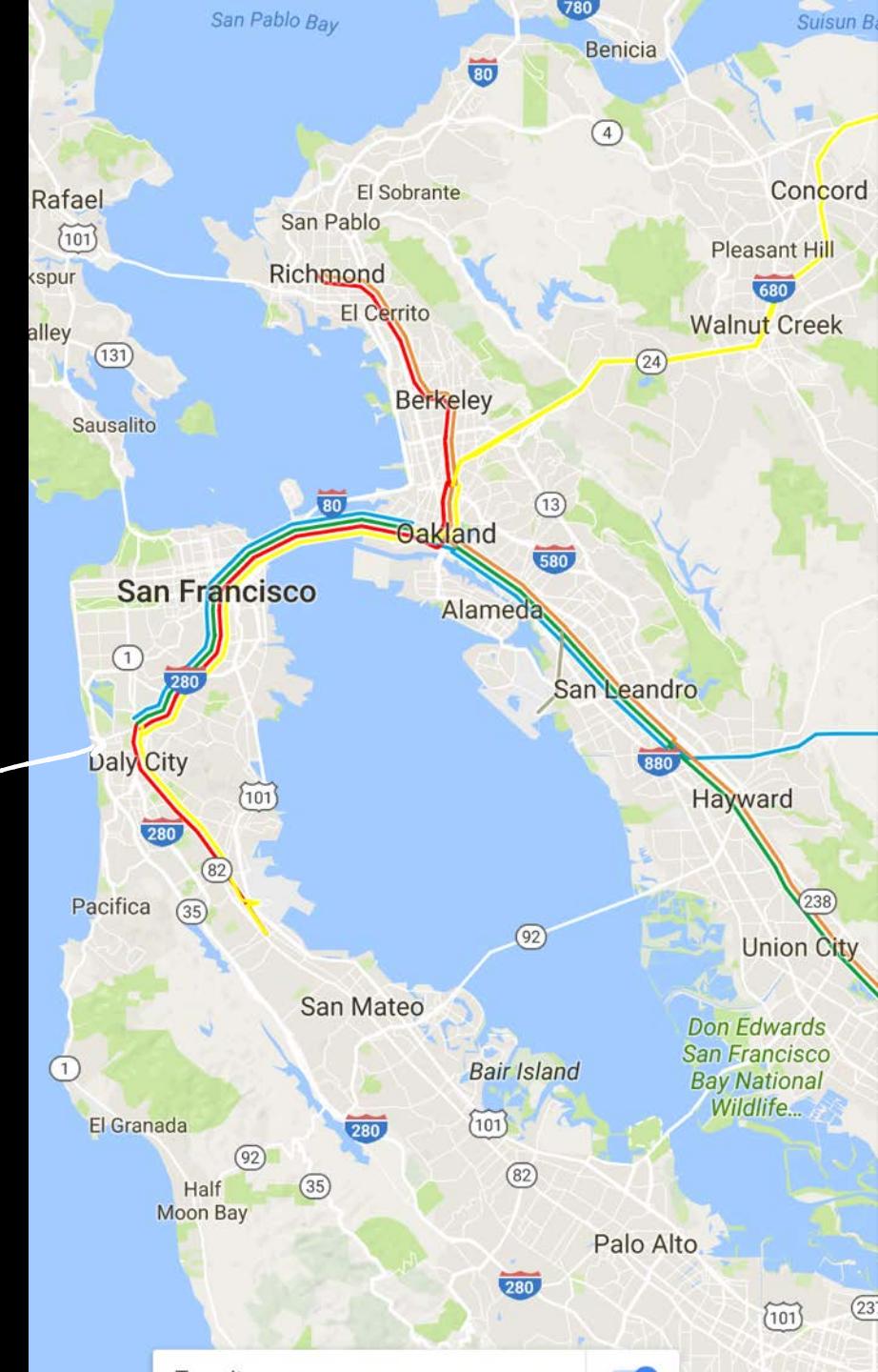


# Maps are places to visualize and “compare spatial data”

DIFFERENT BART LINE

→ ABSTRACTING EACH LINE.

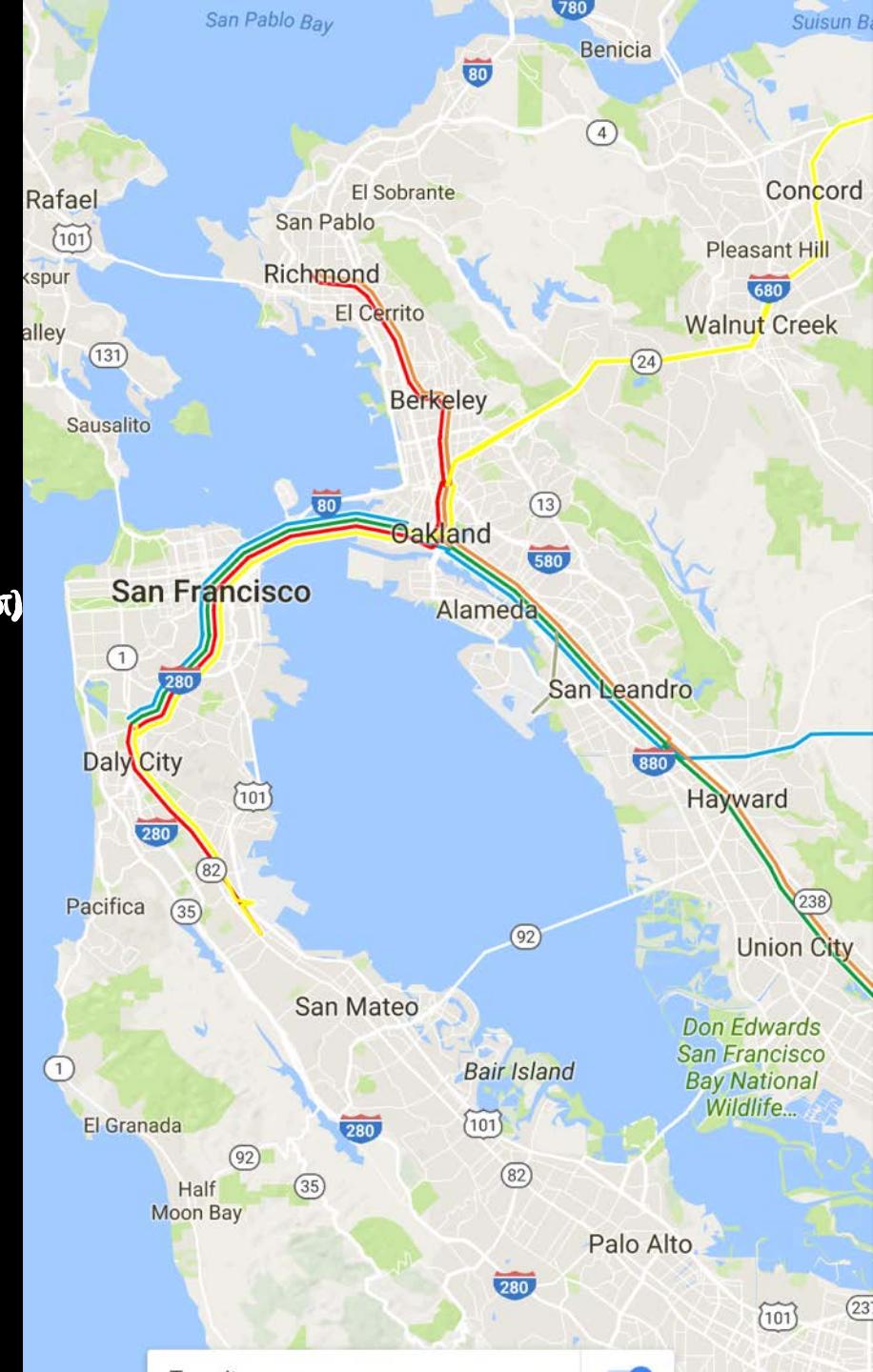
Google, 2017



# What goes into a map?

- Data
    - Structure
    - Generalization
  - Analysis
  - Cartography
    - Projection
    - Graphic Design
- FOCUS ON TOPIC  
(WHAT'S IMPORTANT)*

Google, 2017

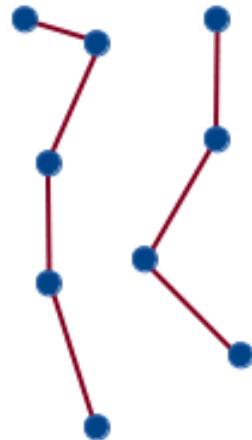


# Spatial Data Fundamentals

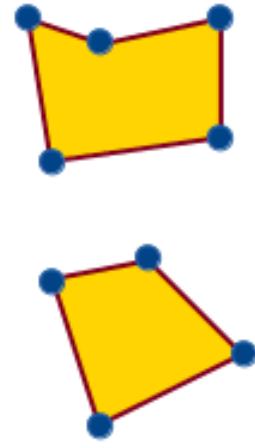
Points



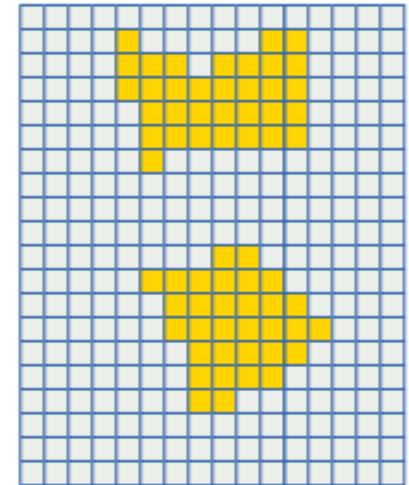
Lines



Polygons



Raster



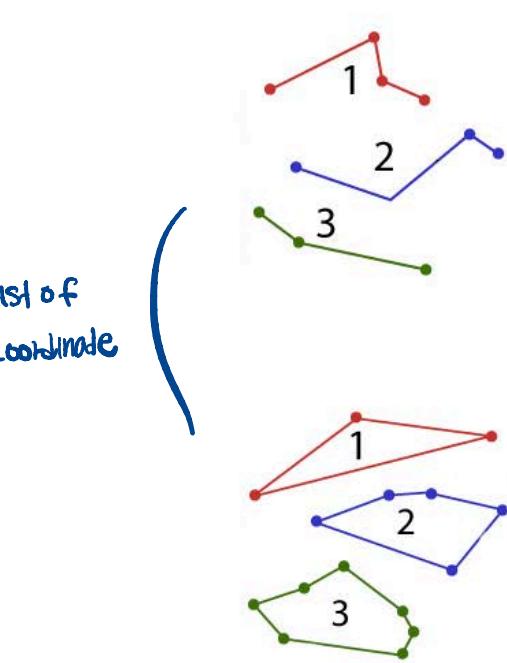
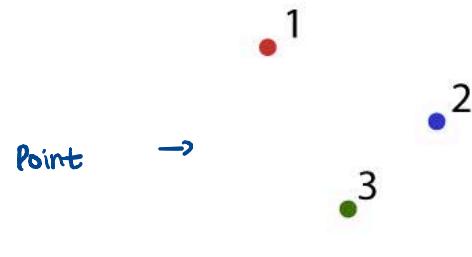
L ]

VECTOR DATA  
(ORDER OF COORDINATES)

<http://michaelminn.net/tutorials/gis-data/>

RASTER DATA  
(Define location on GRID)

# Vector Data



List of  
coordinates

ID	Plot Size	Type	VegClass
1	40	Vegetation	Conifer
2	20	Vegetation	Deciduous
3	40	Vegetation	Conifer

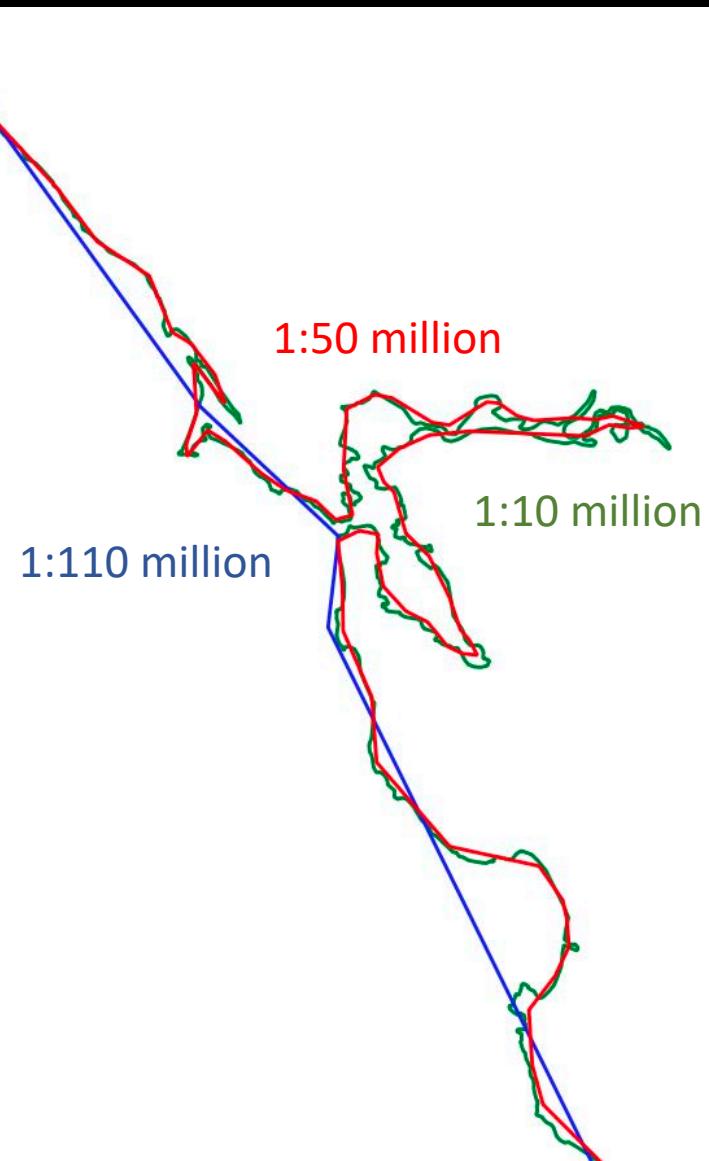
ID	Type	Status	Maintenance
1	Road	Open	Year Round
2	Dirt Trail	Open	Summer
3	Road	Closed	Year Round

ID	Type	Class	Status
1	Herbaceous	Grassland	Protected
2	Herbaceous	Pasture	Open
3	Herbaceous / Woody	Grassland	Protected

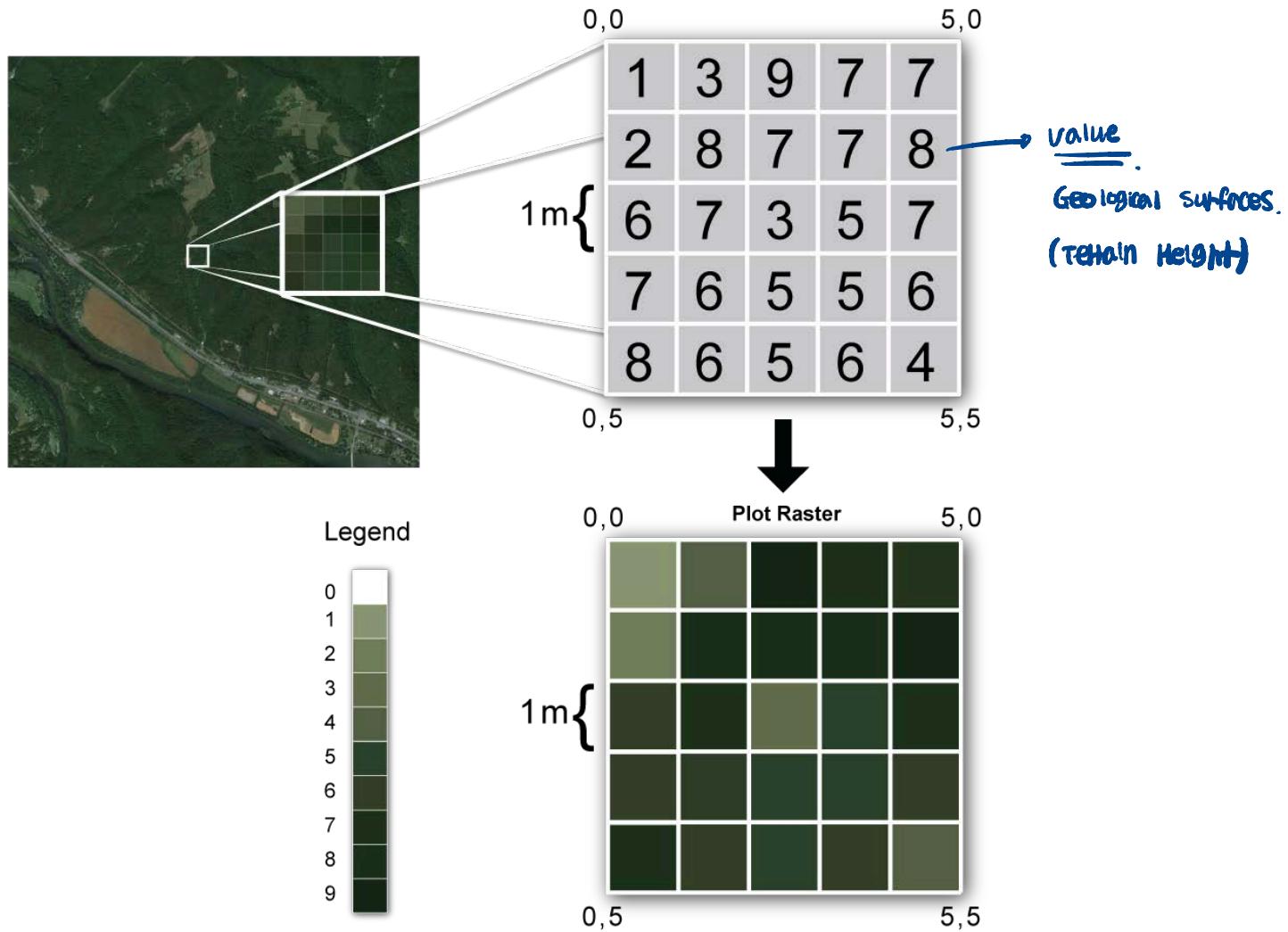
# Vector Data

## Scale/Generalization

Fidelity



# Raster Data



# Raster Data



a. Landsat ETM+

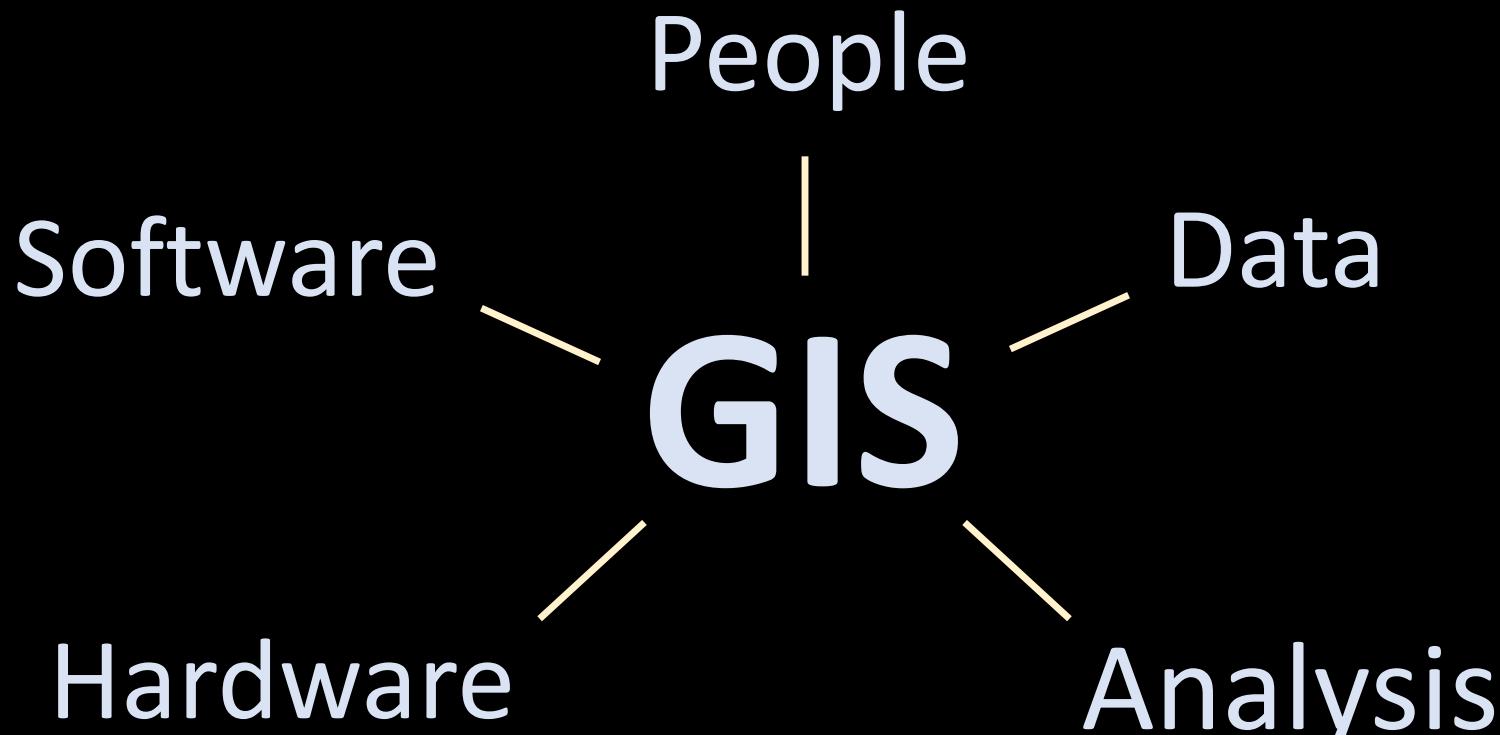


b. ATLAS



c. QuickBird

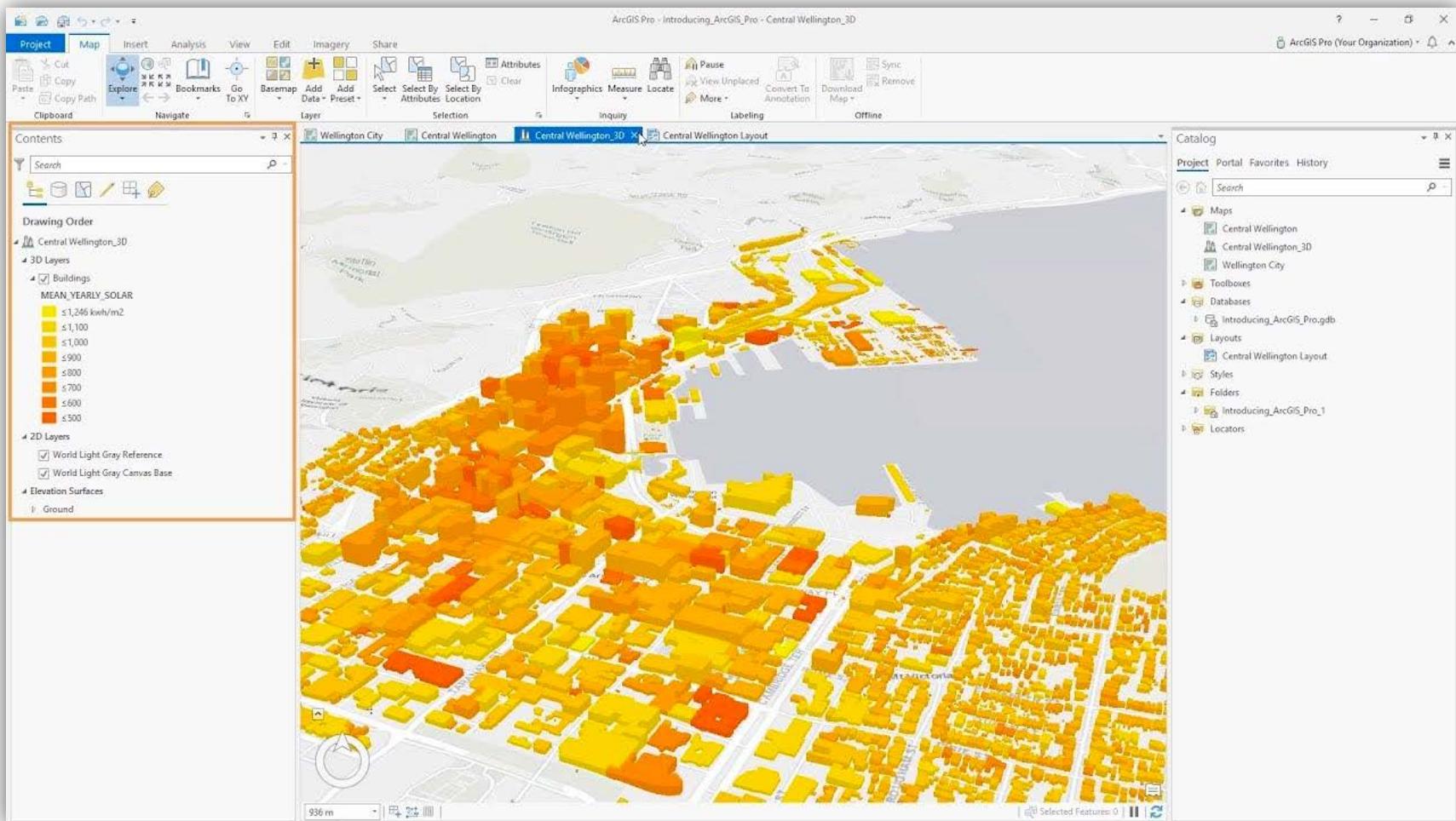
# GIS: Geographic Information System



↳ NOT IMPORTANT THESE PARS.

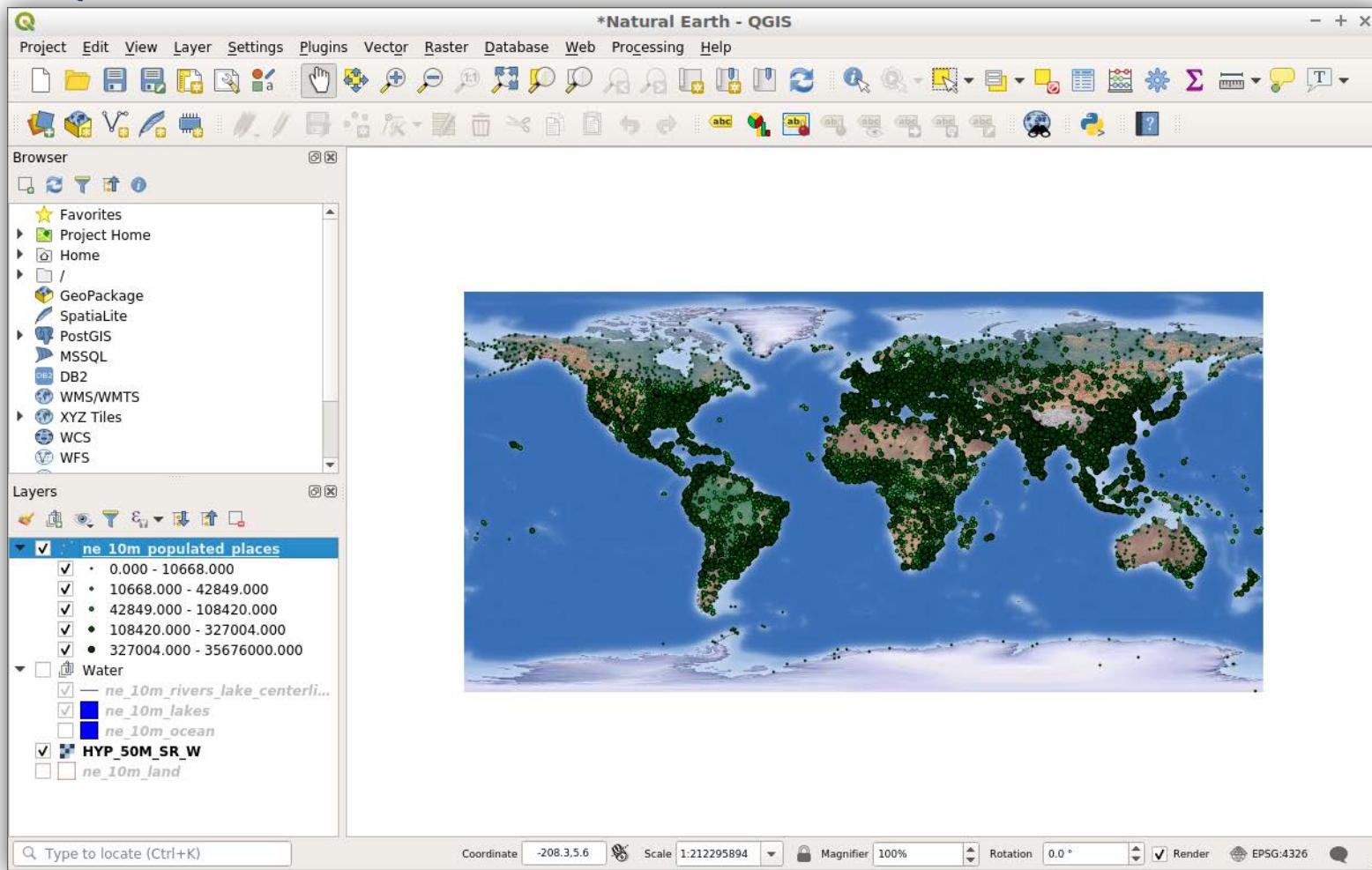
# GIS Software

## ESRI ArcGIS Pro



# GIS Software

## QGIS



# GIS Software

## Python

Jupyter WorkingScript1d Last Checkpoint: 4 hours ago (unsaved changes)

File Edit View Insert Cell Kernel Help

CellToolbar

```
primary.plot(ax=ax, edgecolor='deepskyblue',
secondary.plot(ax=ax, edgecolor='dodgerblue',
residential.plot(ax=ax, edgecolor='steelblue',
rivers.plot(ax=ax, edgecolor='cornflowerblue',
parks.plot(ax=ax, color='lawngreen',edgecolor=
```

```
ax.grid(which='major', alpha=0.5)
plt.legend(loc='center left', title= 'Layers'
plt.show()
```

18 \*\*\*
19
20- ## American Community Survey (ACS)
21
22 The [American Community
Survey](https://www.census.gov/programs-surveys/acs/about.html) differs from
the [decennial US Census
](https://www.census.gov/programs-surveys/decennial-census.html) in that ACS
data are based on an annual sample of approximately 3 million households,
rather than a more complete enumeration of the US population. The ACS provides
estimated values with error margins.
23
24 This plot sorts median income by state.
25
26 '''(r incomeplot, fig.height=7, echo=FALSE)
27 inc\_state %>%
28 filter(NAME != "Puerto Rico") %>%
29 ggplot(aes(x = estimate, y = reorder(NAME, estimate))) +
30 geom\_errorbarh(aes(xmin = estimate - moe, xmax = estimate + moe)) +
31 geom\_point(color = "red") +
32 labs(title = "Household income by state - US",
33 subtitle = "2012-2016 American Community Survey (excl. Puerto Rico)",
34 y = "",
35 x = "ACS estimate (bars represent margin of error)")
36 ...
37
38- ## Estimated household income per county
39
40- '''(r map)
41 mapview(inc\_county, zcol = "estimate", legend = TRUE, color = "#eeeeee", lwd = 0.5)
42 ...
43
44
45- '''(r datatable, echo=FALSE)
46 inc\_county %>%
47 head() %>%
48 select("NAME", "variable", "estimate", "moe") %>%
49 F

R Studio

Project: (None)

Files Plots Packages Help Viewer

Estimated household income per county

mapview(inc\_county, zcol = "estimate", legend = TRUE, color = "#eeeeee", lwd = 0.5)

NAME variable estimate moe

Autauga County, Alabama	medincome	63099	2631
Baldwin County, Alabama	medincome	51365	991

# GIS Software

## Web Applications & Services

CARTO

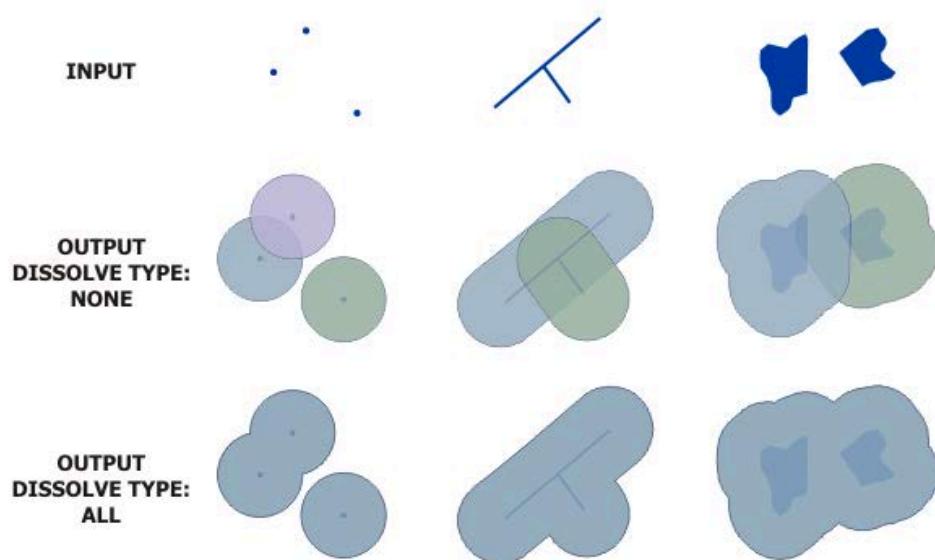


ArcGIS Online



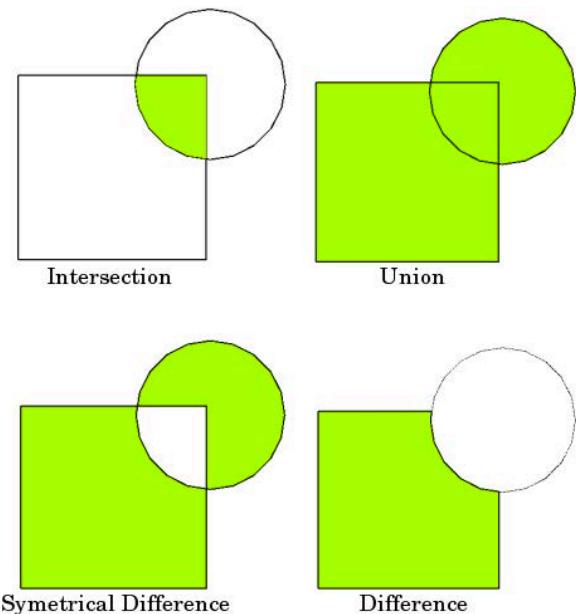
# Analysis

## Buffer



<https://desktop.arcgis.com/en/arcmap/10.3/tools/analysis-toolbox/GUID-267CF0D1-DB92-456F-A8FE-F819981F5467-web.png>

## Overlay (geometric query)



[https://docs.qgis.org/2.8/en/\\_images/overlay\\_operations.png](https://docs.qgis.org/2.8/en/_images/overlay_operations.png)

# Analysis

## Operating Scope

### Raster Algebra

A diagram illustrating Raster Algebra. It shows three input rasters (InRas1, InRas2, InRas3) and their sum (OutRas). The inputs are 4x4 grids with values ranging from 0 to 8. The output is also a 4x4 grid with values ranging from 0 to 8. An equals sign separates the inputs from the output.

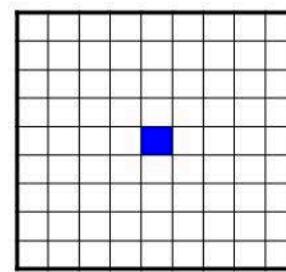
1	1	0	0
1	2	2	2
4	0	0	2
4	0	1	1

0	1	1	0
3	3	1	2
0	0	2	2
3	2	1	0

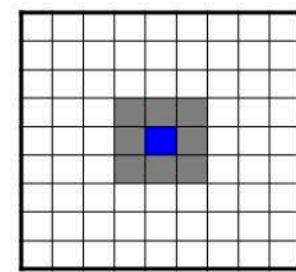
1	0	0	0
2	0	3	3
0	0	3	2
1	1	0	0

=

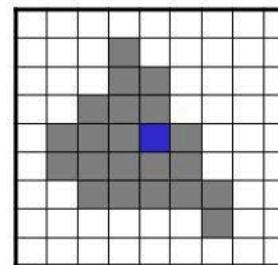
3	1	0	0
4	6	7	7
0	3	6	6
8	3	1	1



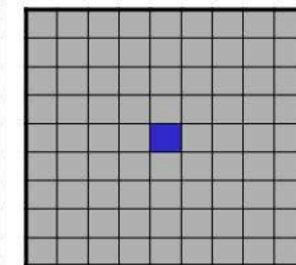
Local



Focal



Zonal

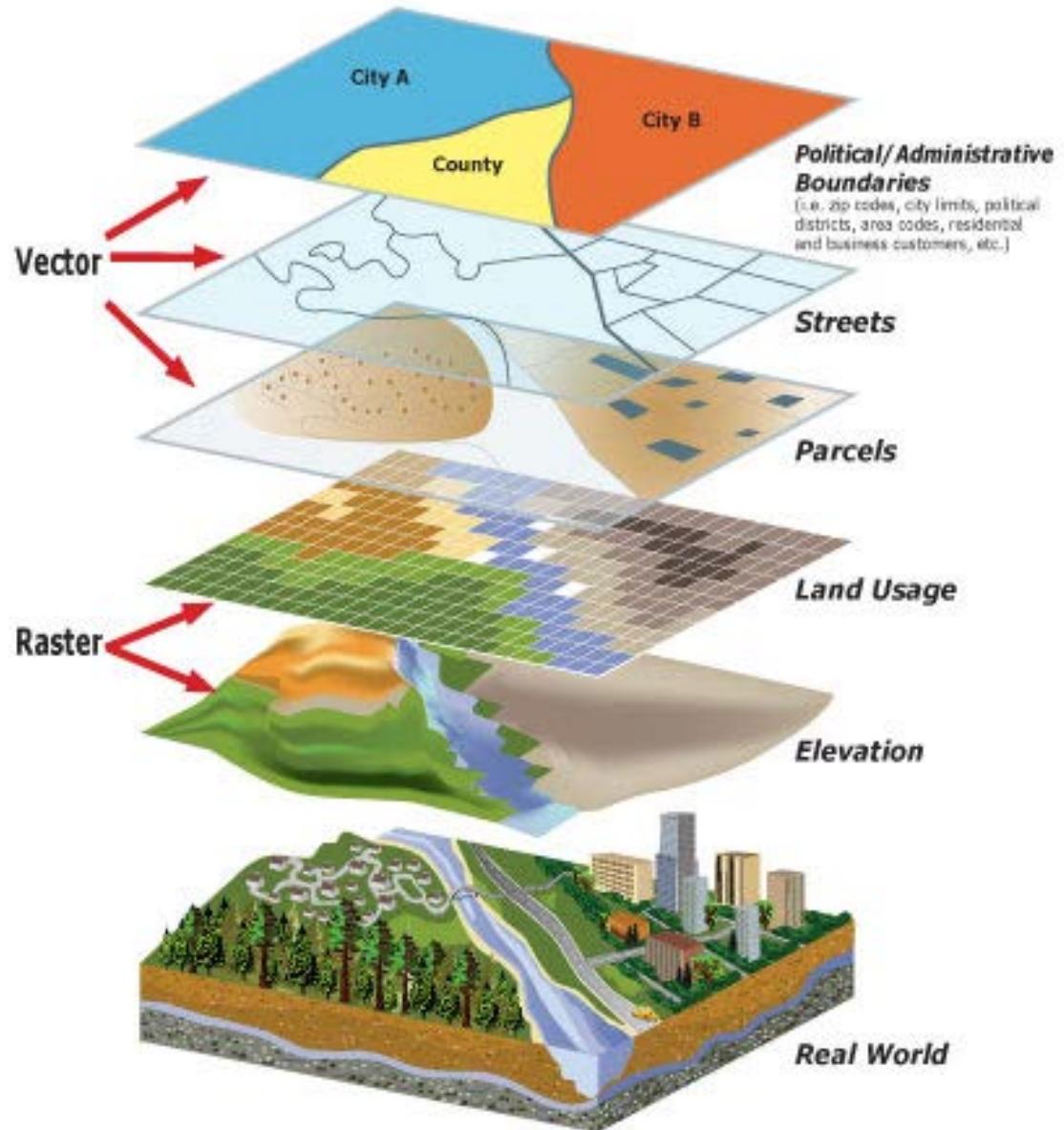


Global

<https://desktop.arcgis.com/en/arcmap/10.3/tools/spatial-analyst-toolbox/GUID-D8BD897C-B7A2-46A5-93DC-121409FF4C0D-web.png>

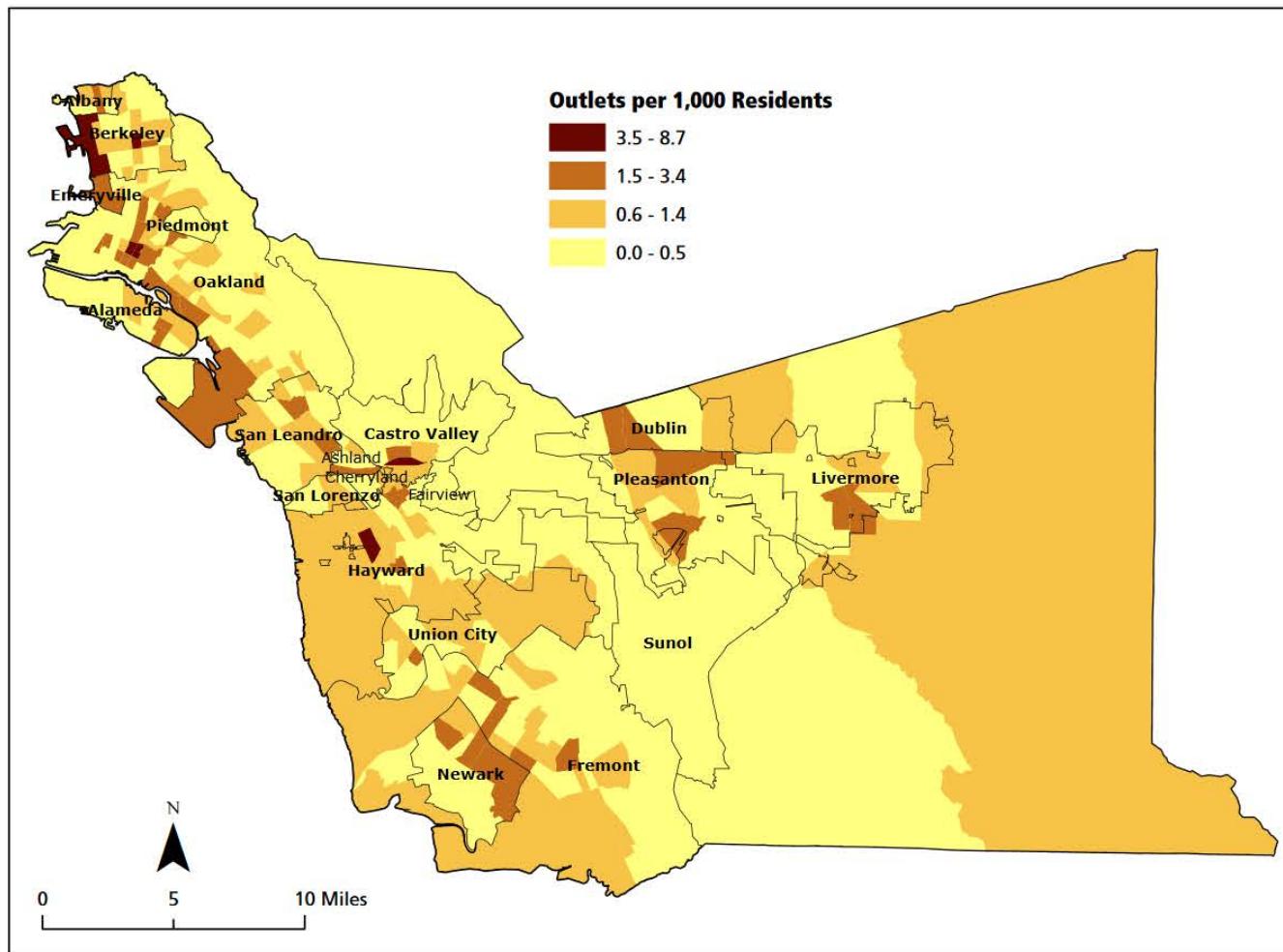
<https://slideplayer.com/slide/5822638/19/images/4/Map+Algebra+Operations.jpg>

# What sorts of questions can we ask with these data and tools?



# How do you think this analysis was done?

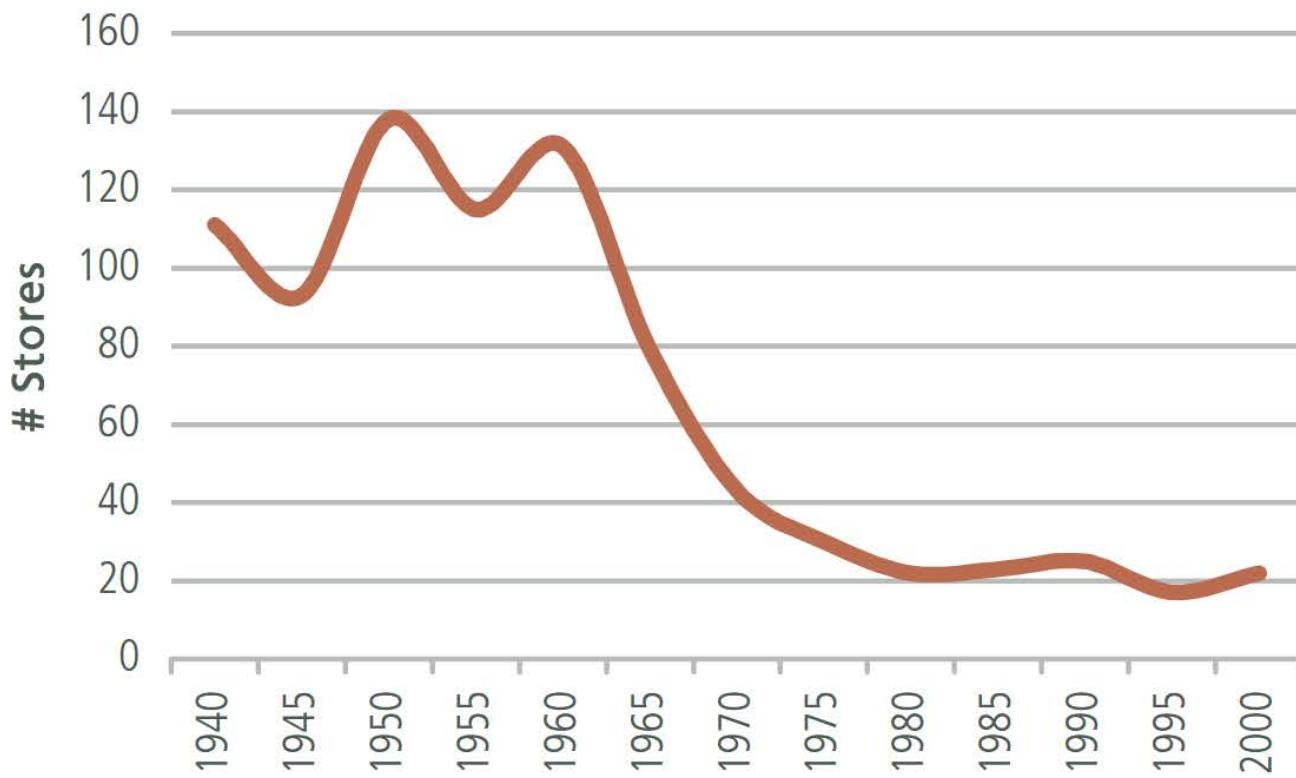
Map 9: Fast Food and Convenience Store Density, Alameda County



Source: California Center for Public Health Advocacy, with data from ESRI/InfoUSA 2005.

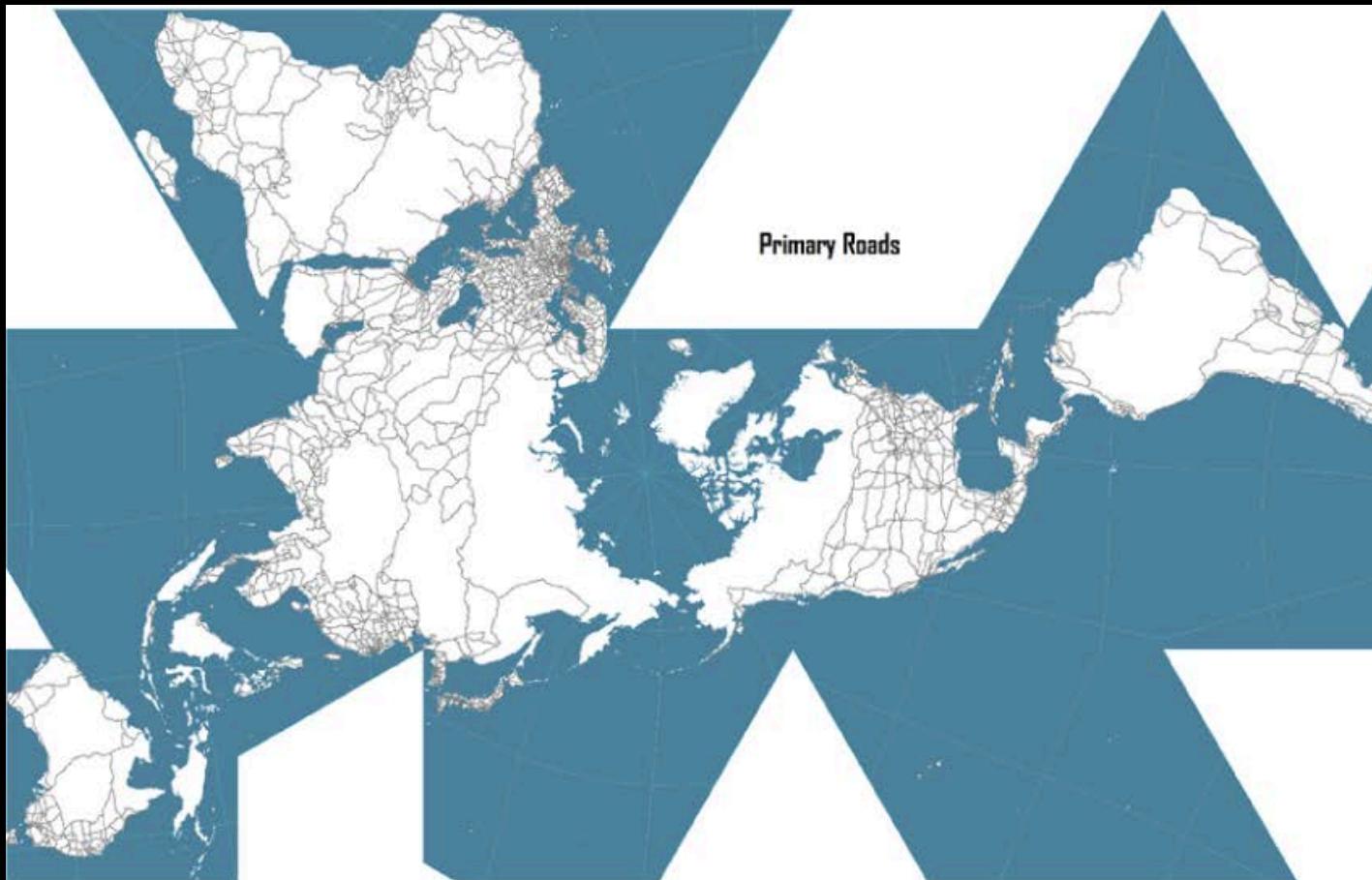
# Spatial analyses can have non-spatial outputs

Figure 47: Number of Food Stores, West Oakland



Source: Fuller A, 2006.

Cartography = Representation  
of Spatial Data

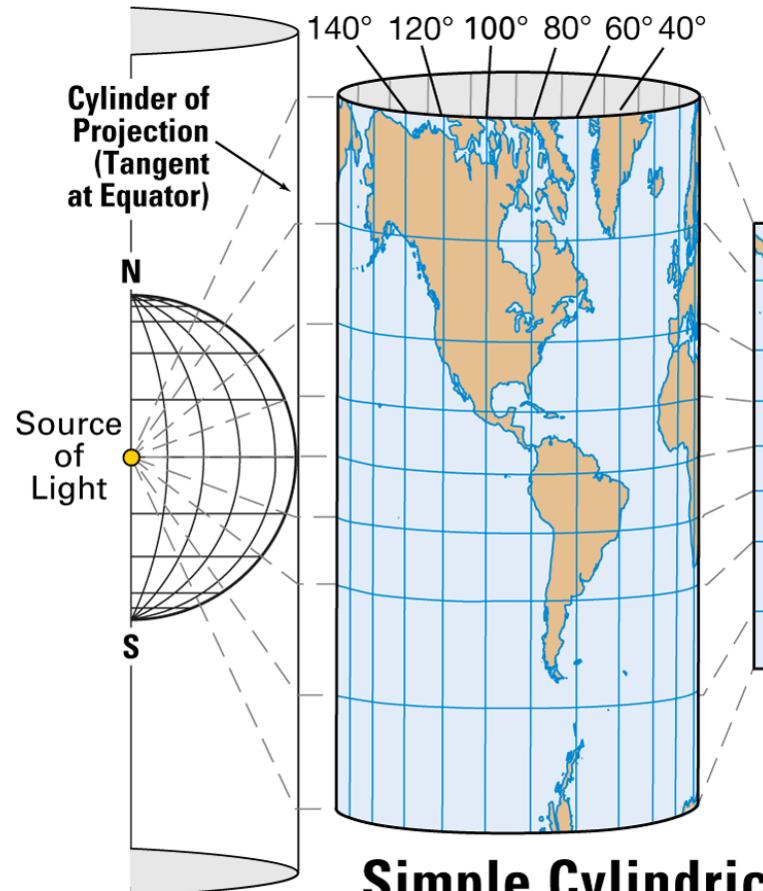


Dymaxion Projection, by Buckminster Fuller

# Projection

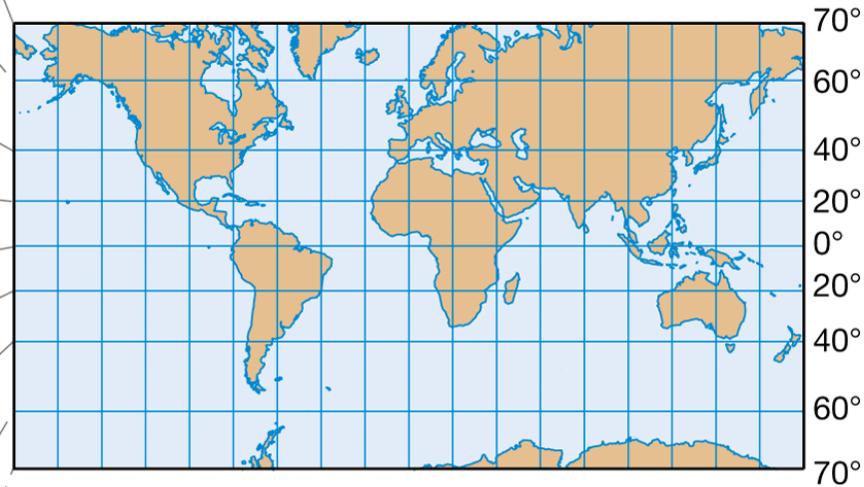


# Projection

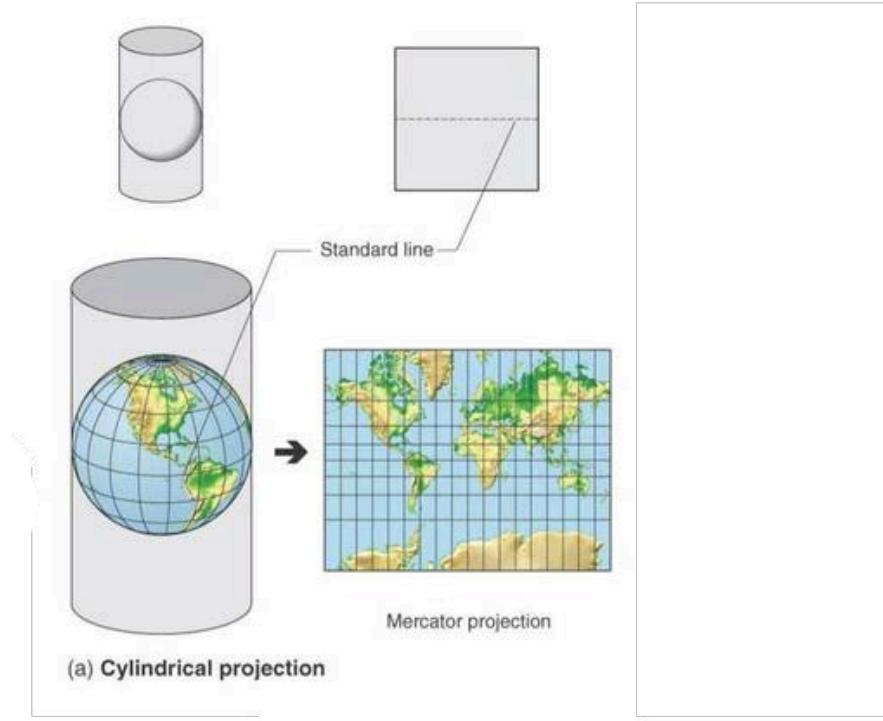


**Simple Cylindrical  
Projection**

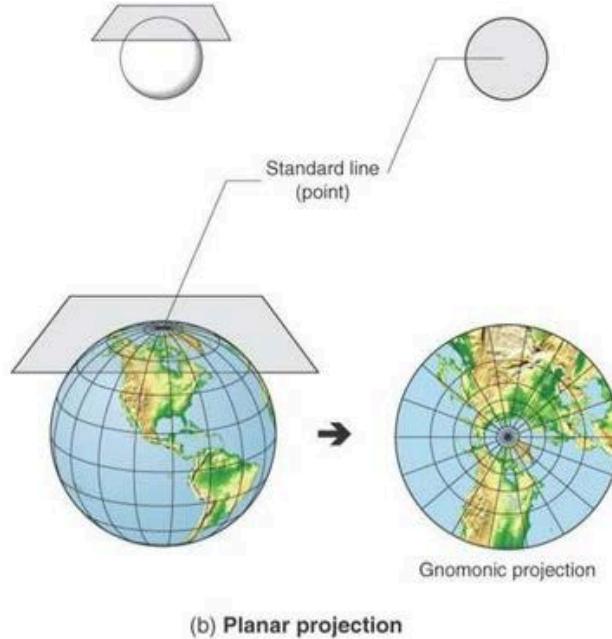
## Mercator Projection



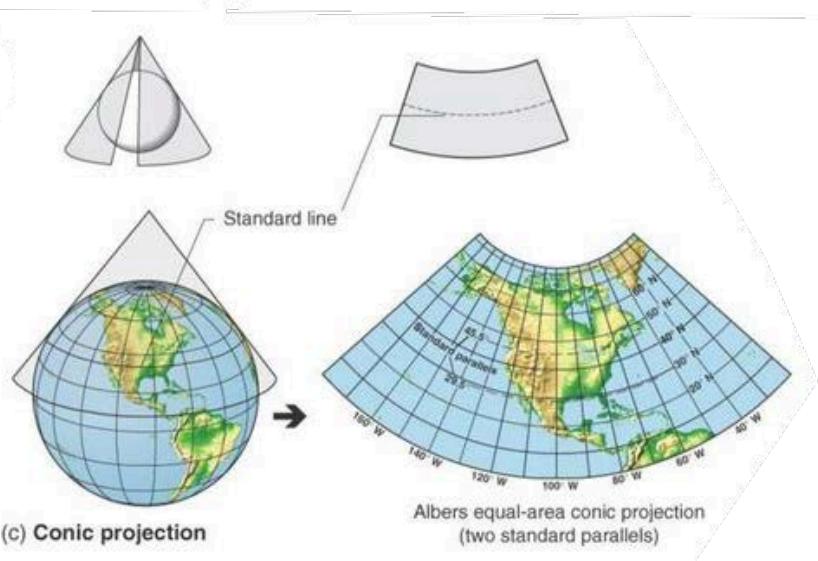
© 2010 EB, Inc.



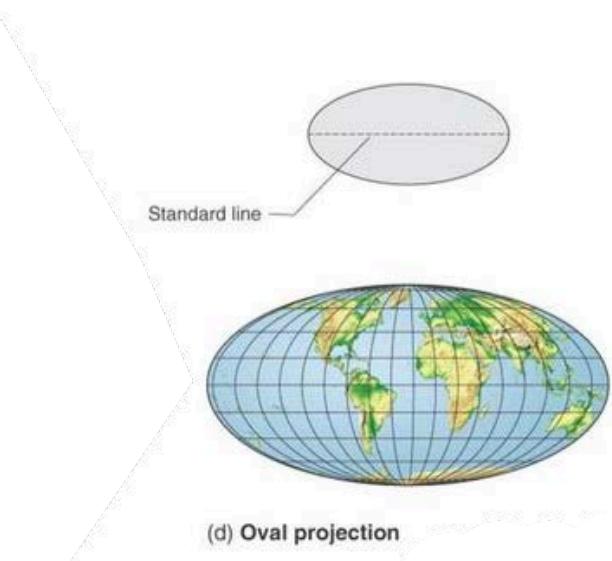
(a) Cylindrical projection



(b) Planar projection



(c) Conic projection

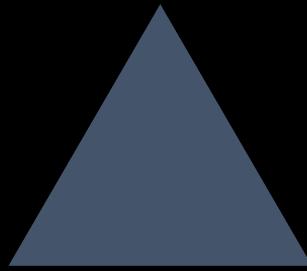


(d) Oval projection

# Projection Compromise

(choose 1 or 2)

Area



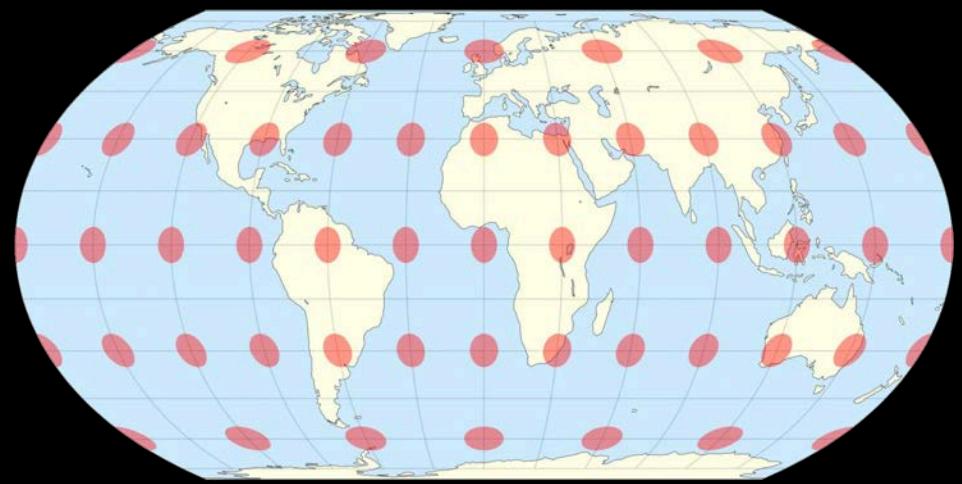
Shape

Distance

# Mercator

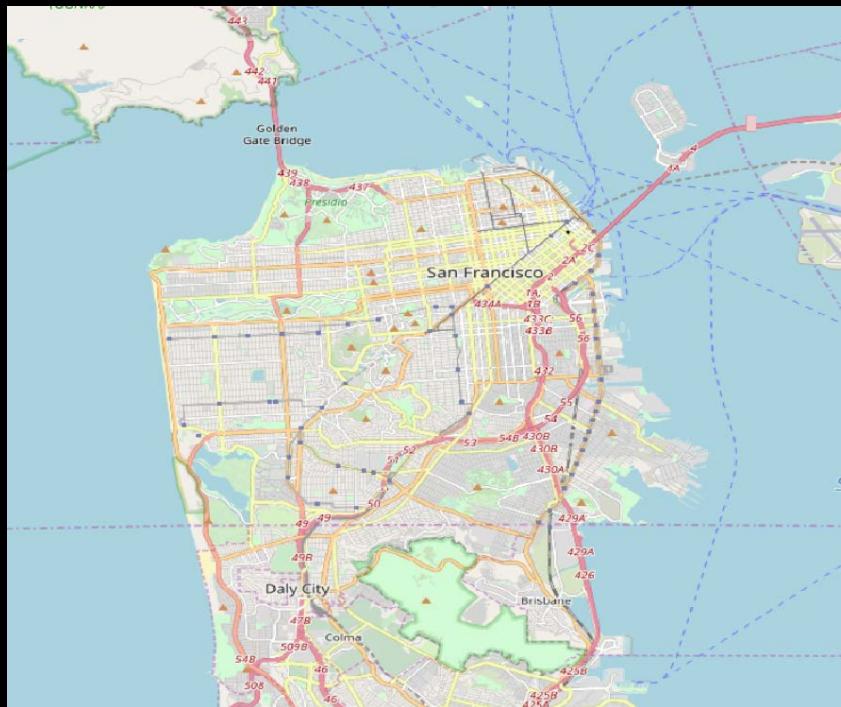


# Equal Earth

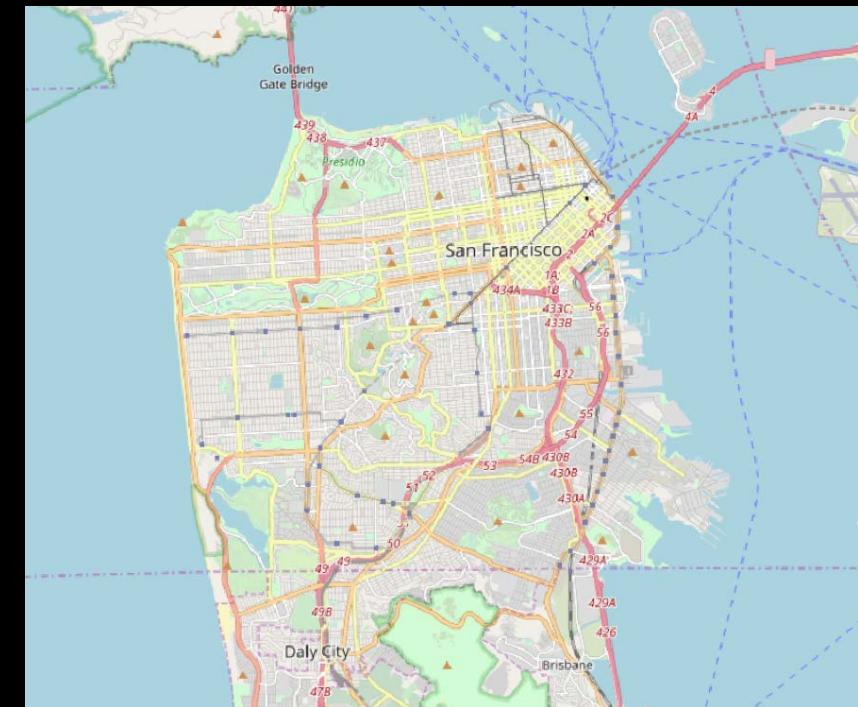


Other Projections

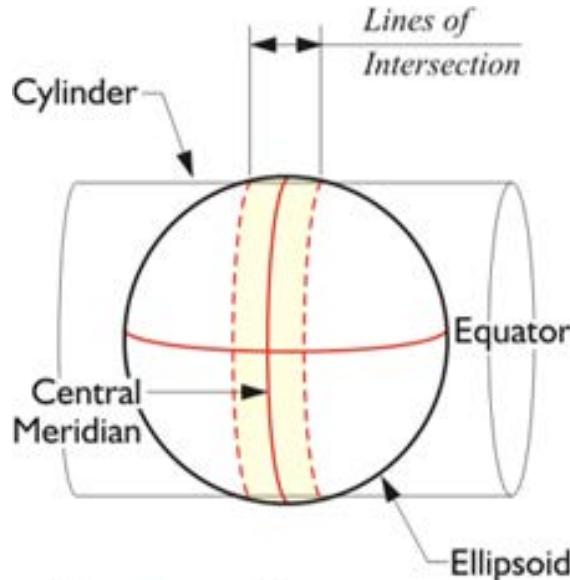
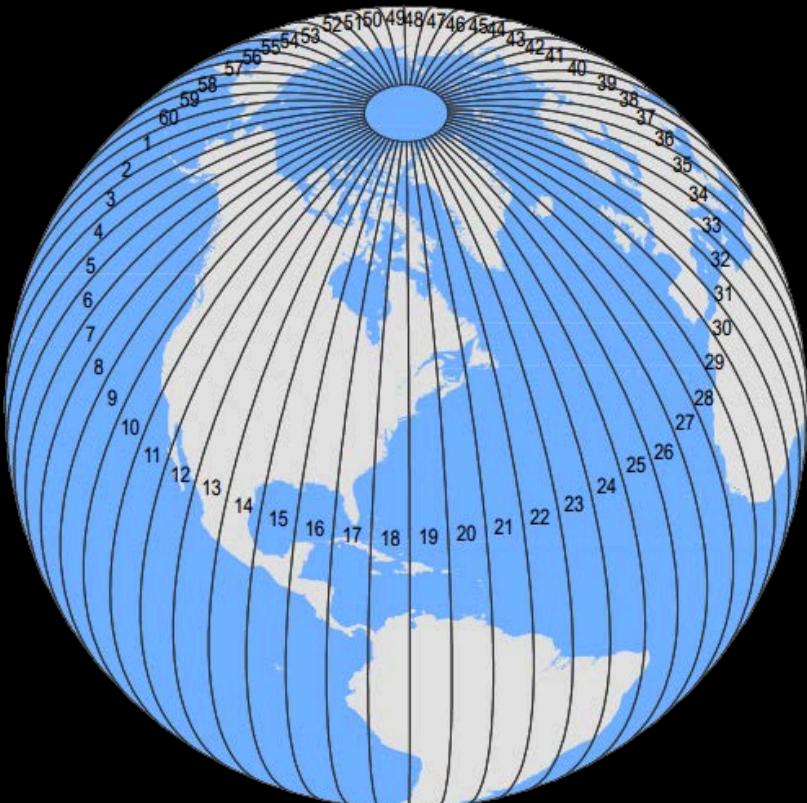
**Latitude, Longitude  
(WGS84)**



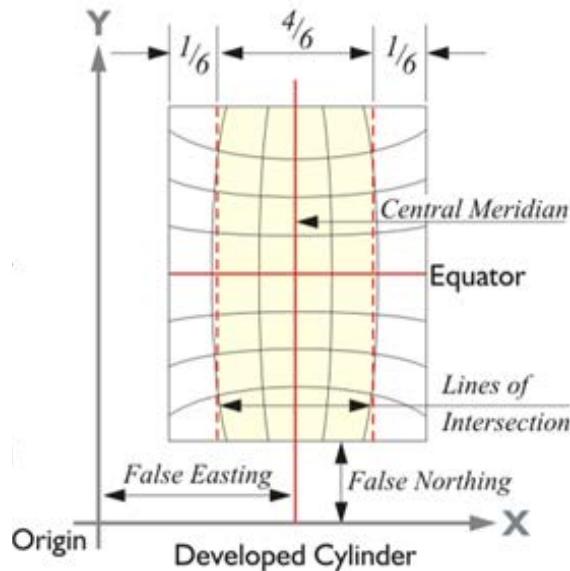
**Meters  
(UTM10N)**



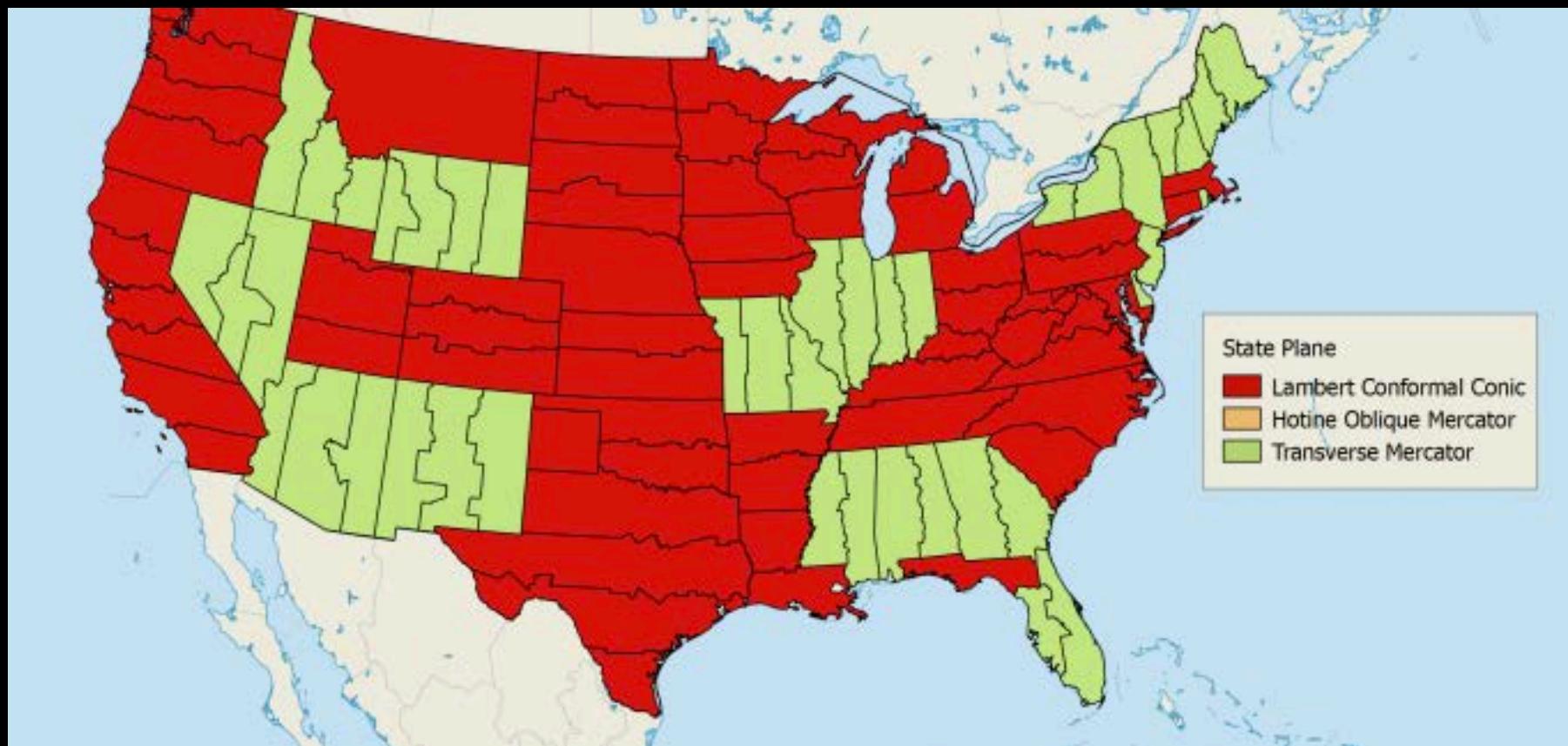
# UTM Zones



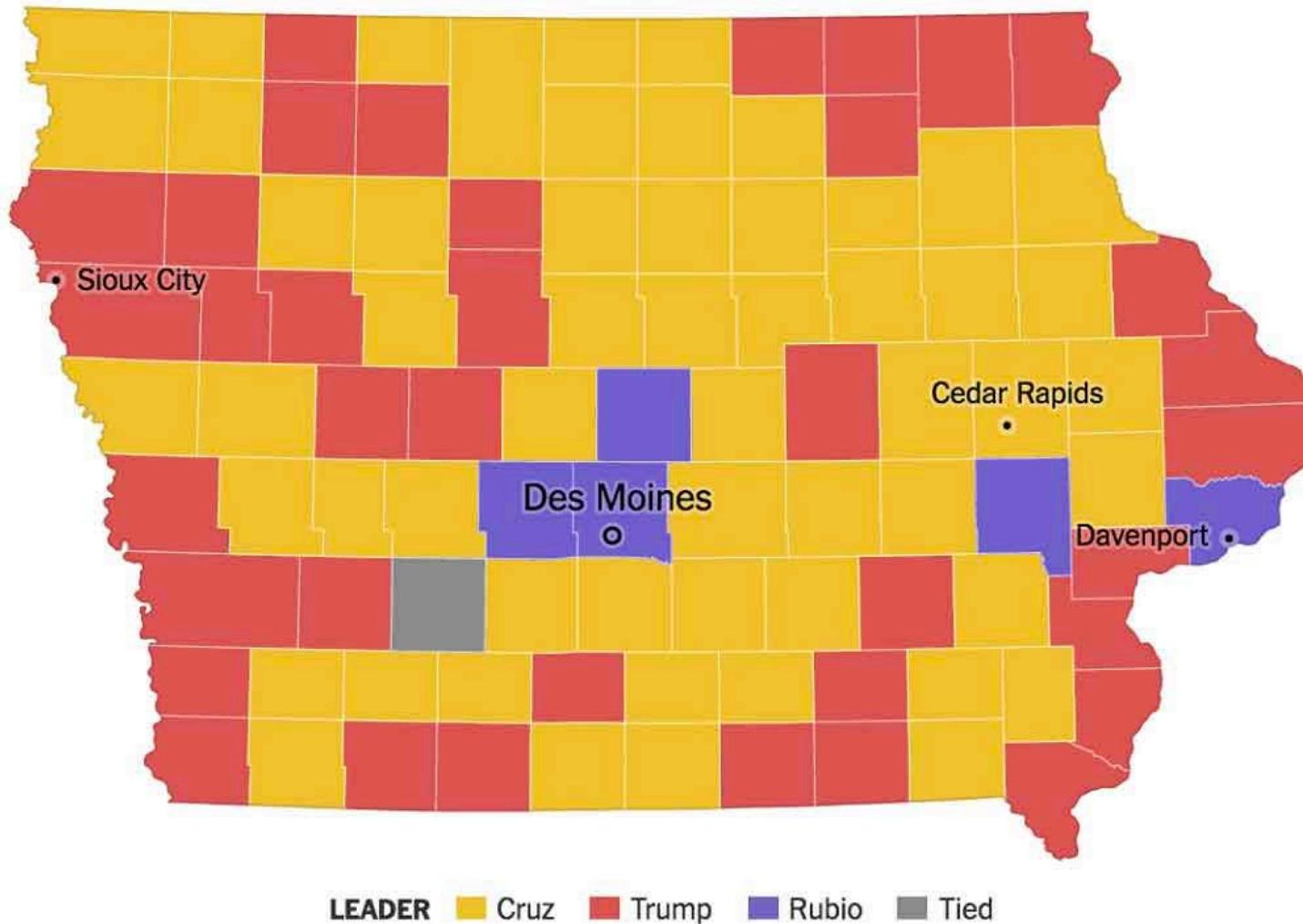
**Transverse Mercator  
Projection**



# State Plane Coordinate Systems

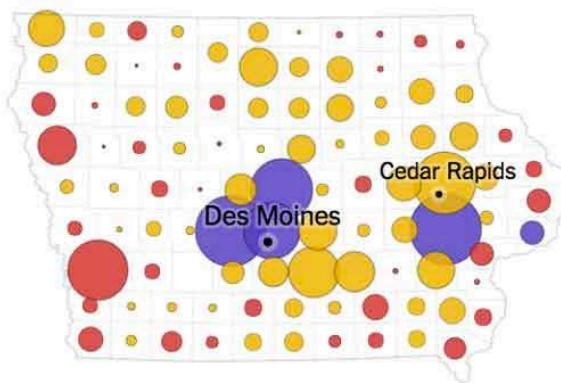


# Symbology Matters



# How Republicans Voted

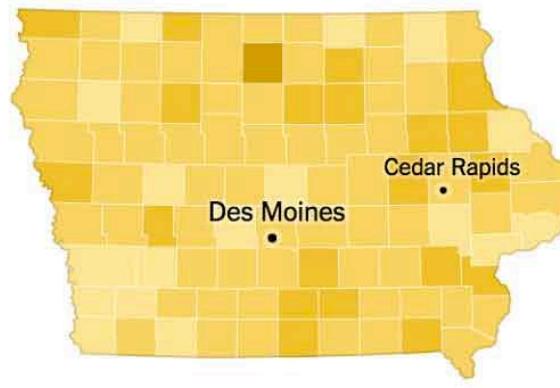
Size of lead



LEADER    ■ Cruz    ■ Trump    ■ Rubio

Circle size is proportional to the size of a candidate's lead.

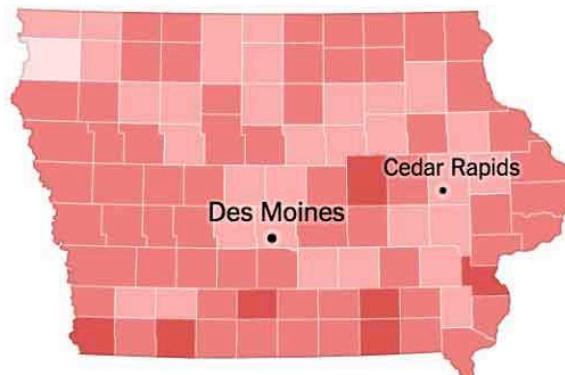
Cruz



CRUZ'S VOTE SHARE

15 25 35 45%

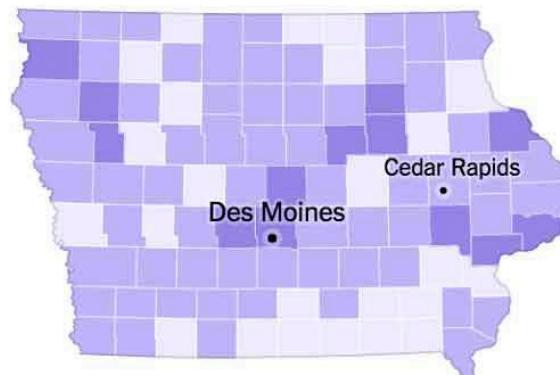
Trump



TRUMP'S VOTE SHARE

15 25 35 45%

Rubio

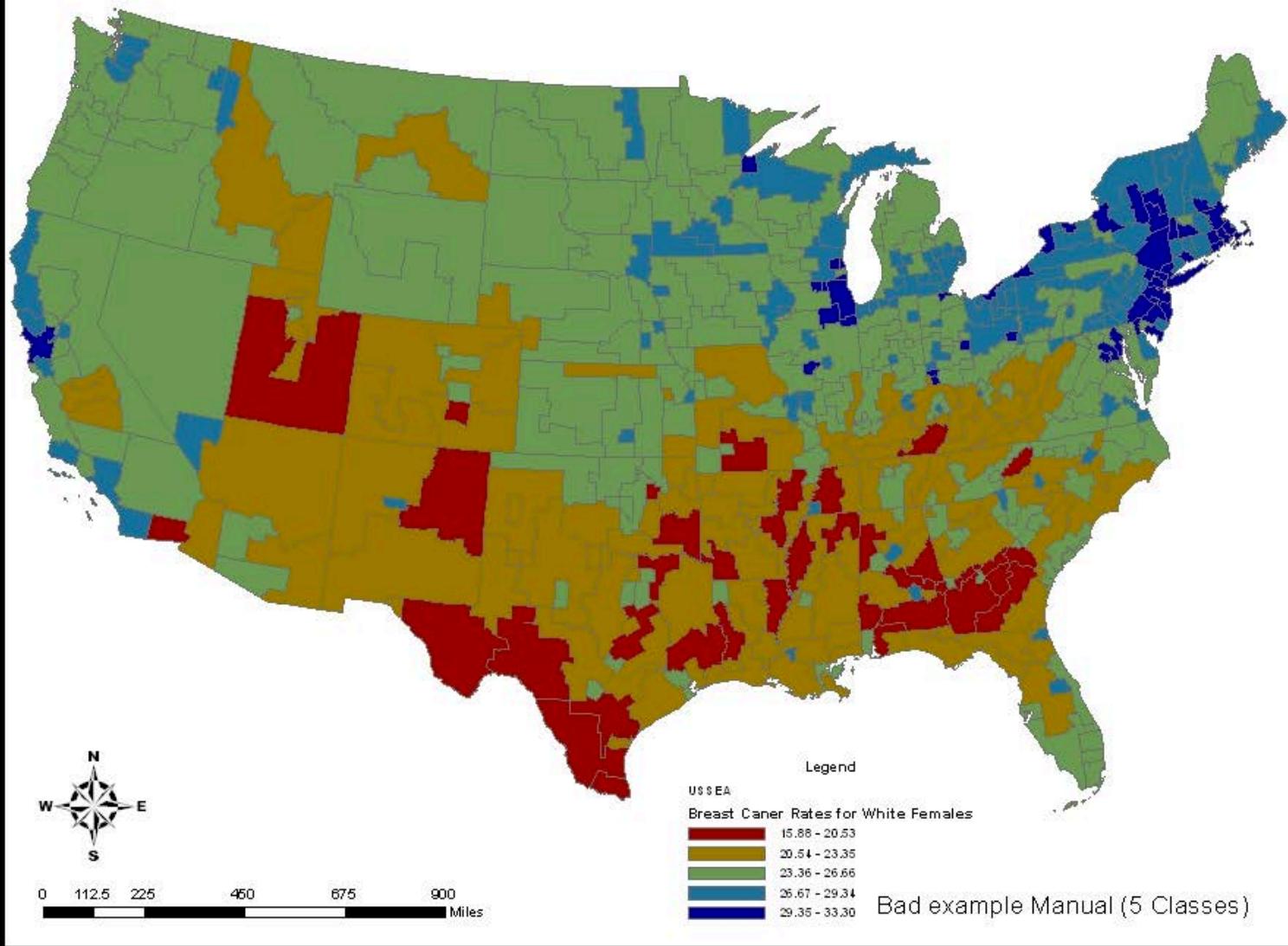


RUBIO'S VOTE SHARE

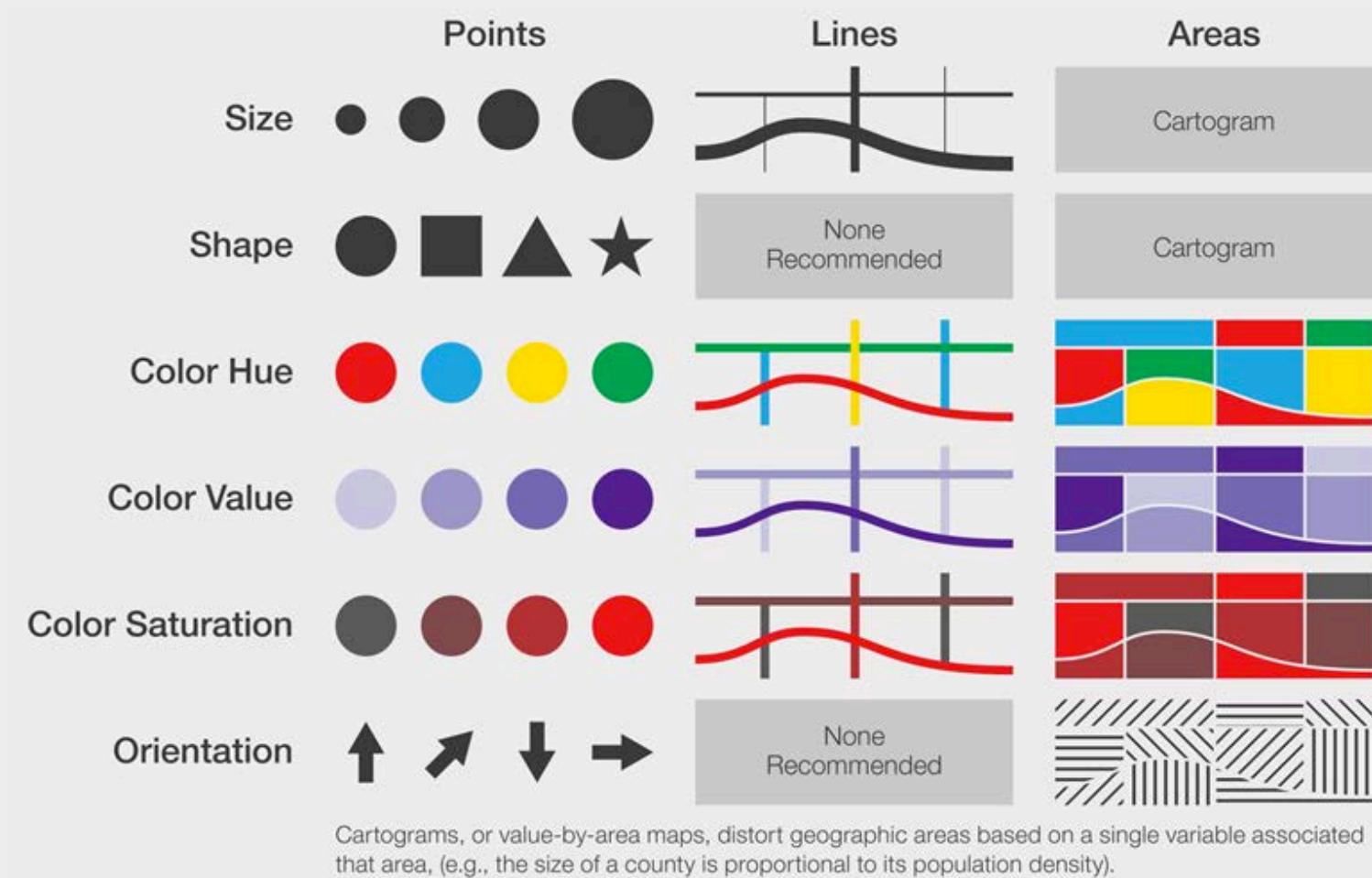
15 25 35 45%

# What's Bad About This Map?

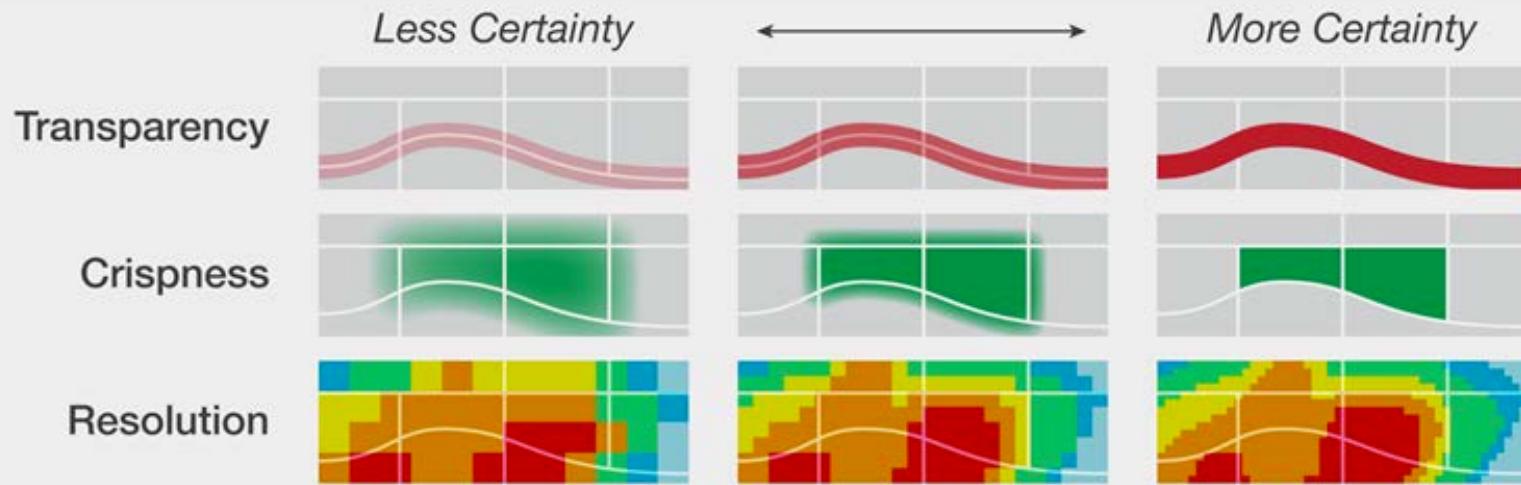
Cancer Rates in the United States



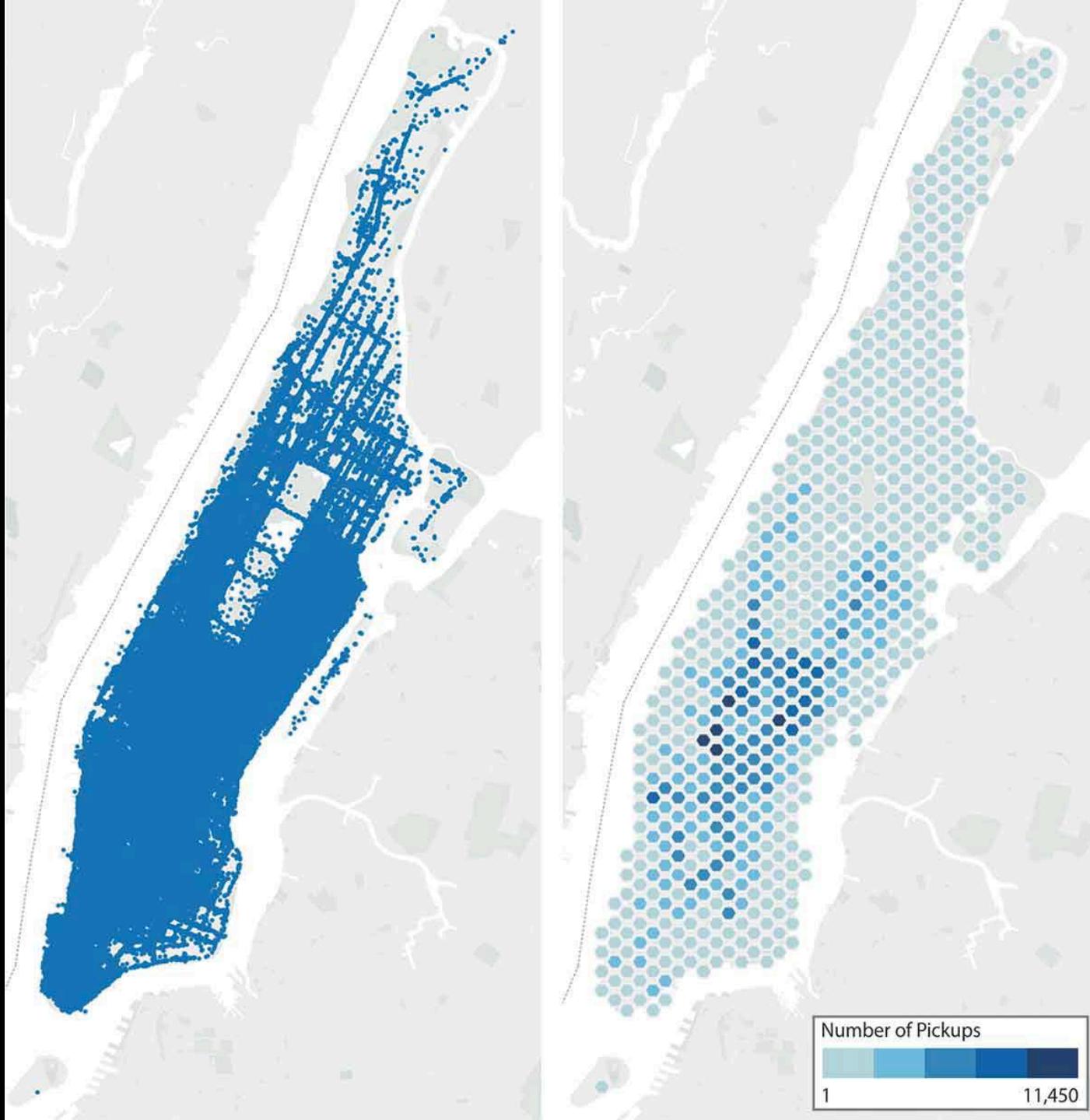
# Visual Variables



# Visual Variables



# Binning



Battersby, S. E., Strebe, D. D. & Finn, M. P. (2017). Shapes on a plane: Evaluating the impact of projection distortion on spatial binning. *Cartography and Geographic Information Science*, 44(5), 410-421.

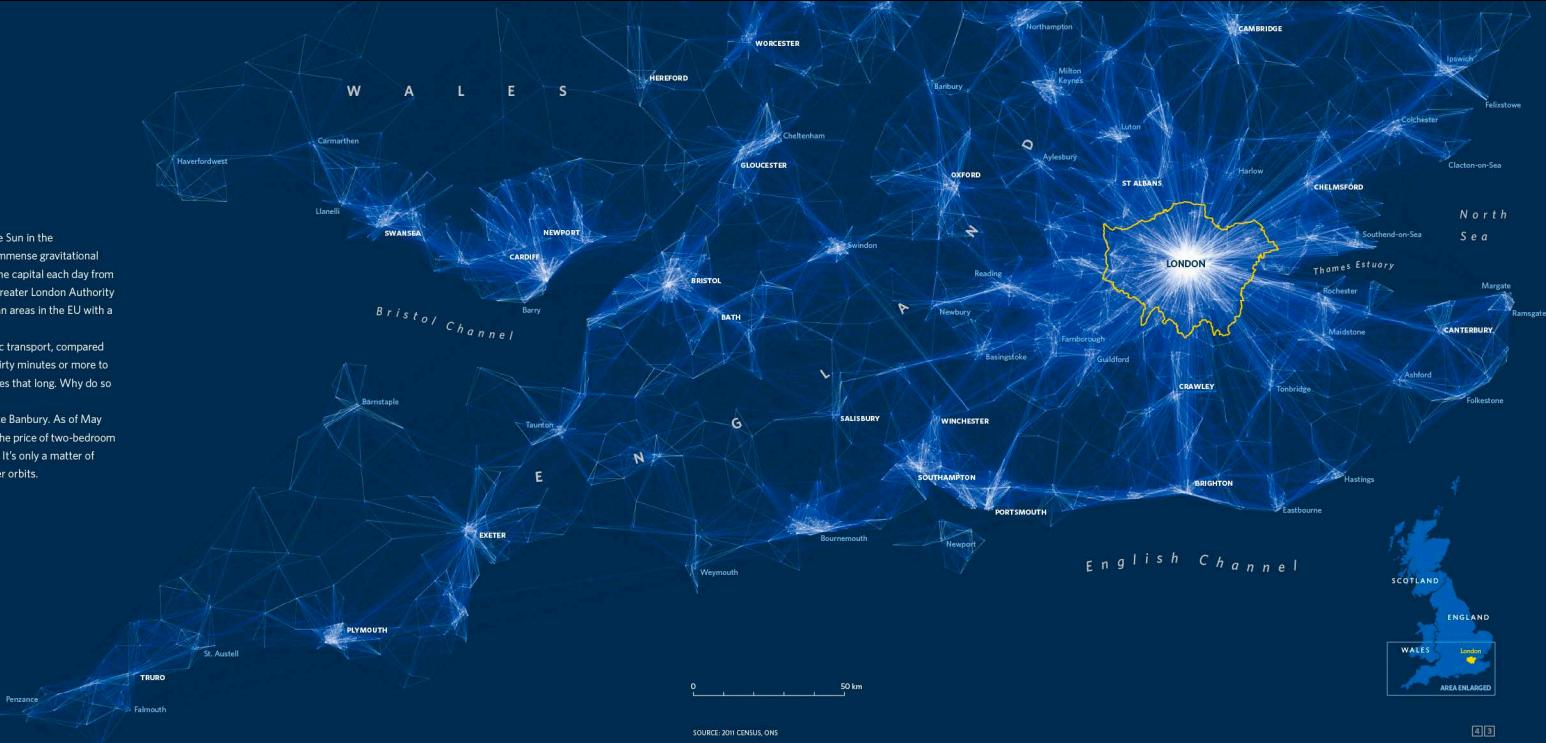
## From Home to Work

High-paying jobs draw workers from far, far away

In this depiction of daily commutes, London shines like the Sun in the constellation of Southern England. Like all stars, it has an immense gravitational pull. Whether by car, train or tube, thousands travel into the capital each day from all directions. Including this 'commuter belt' beyond the Greater London Authority boundary makes the capital one of the largest metropolitan areas in the EU with a population of more than 13 million.

Half of London's workforce make their journey by public transport, compared with only 9% in the rest of the country. Still, most need thirty minutes or more to get to work. Elsewhere in the UK, only 20% have commutes that long. Why do so many live so far away?

For one, London salaries go further in satellite towns like Banbury. As of May 2014, a five-bedroom converted barn there was going for the price of two-bedroom flats along the Underground's Central Line (see pp. 66-7). It's only a matter of time before faster trains propel commuters into even wider orbits.



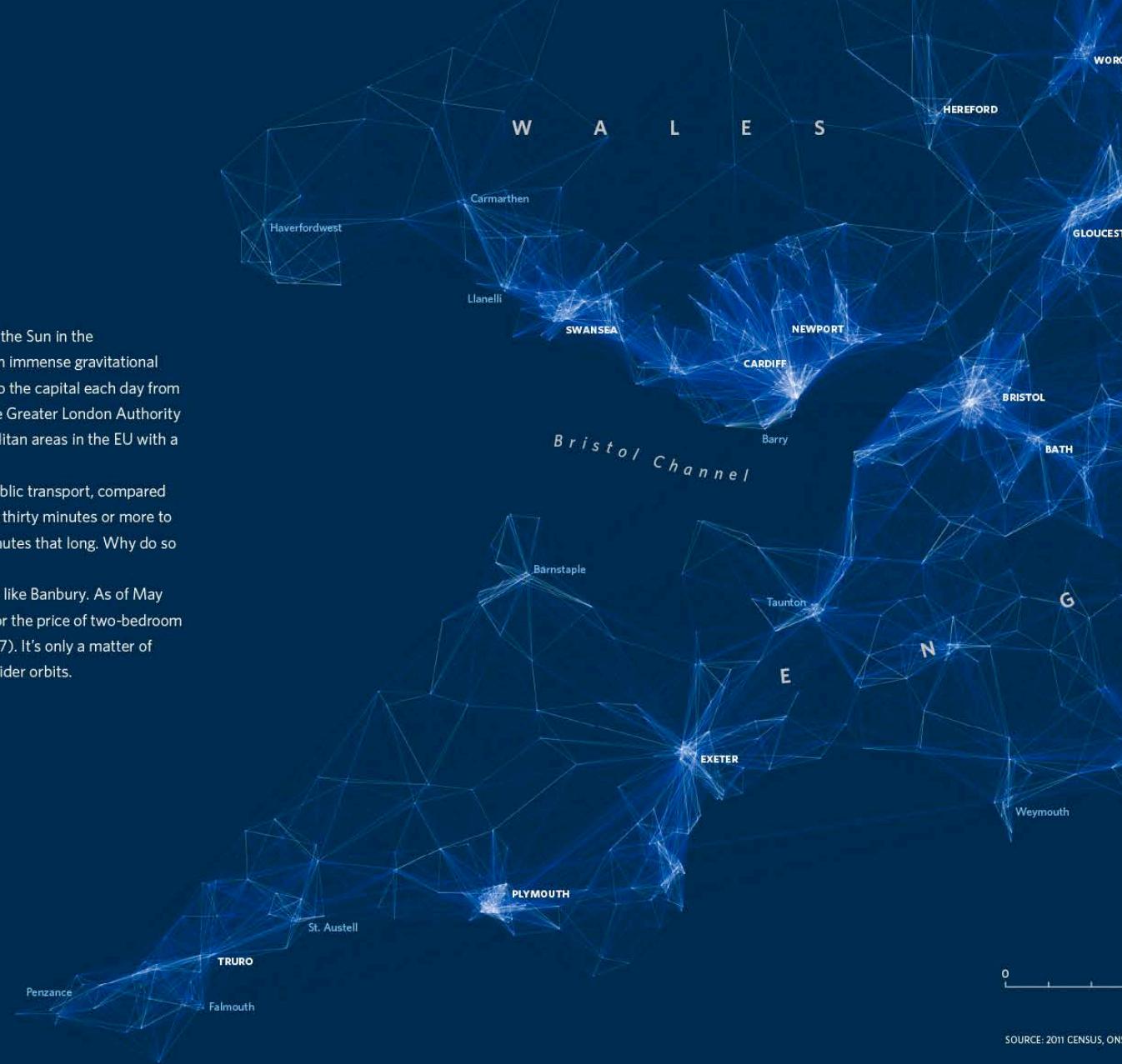
# From Home to Work

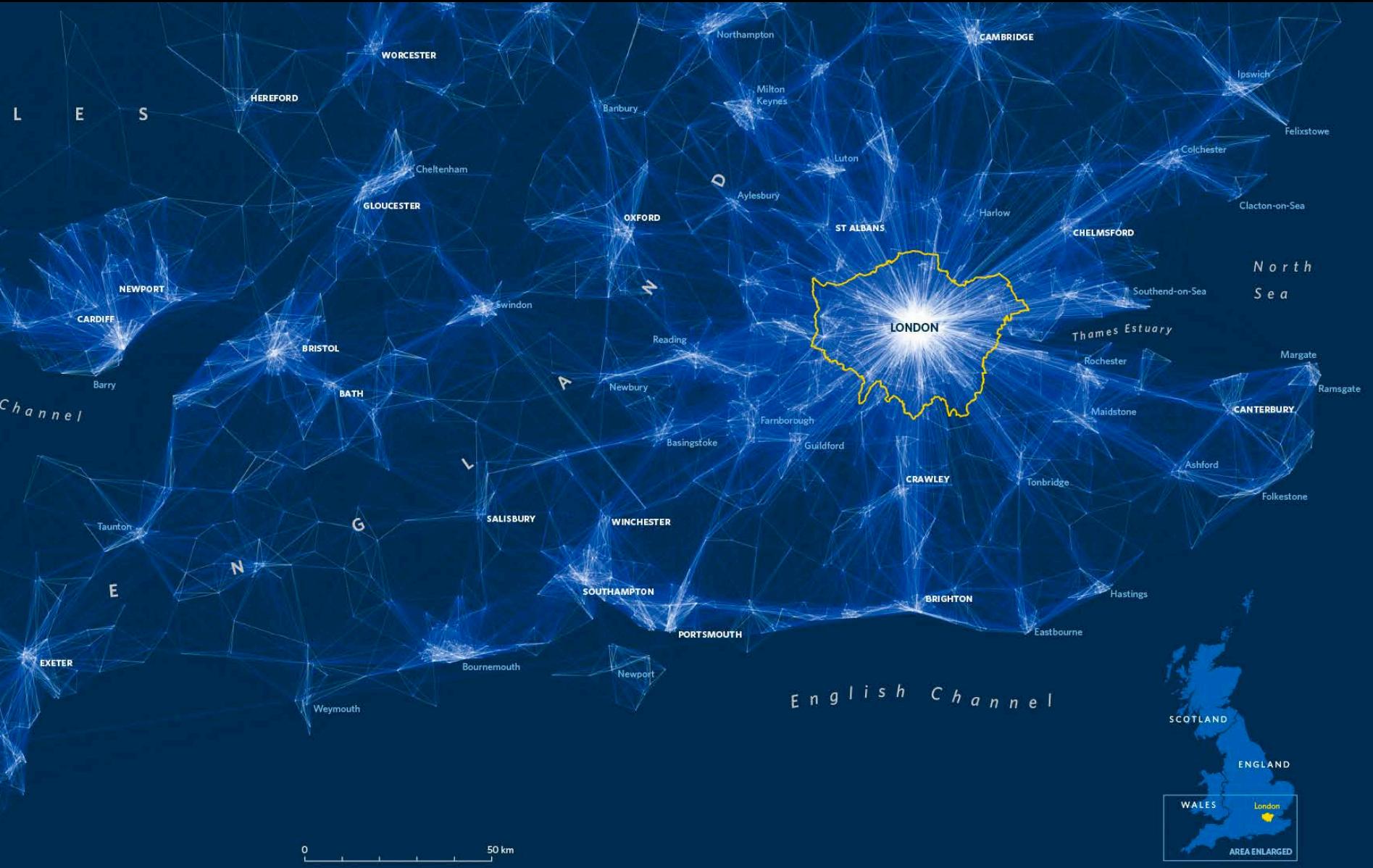
*High-paying jobs draw workers from far, far away*

In this depiction of daily commutes, London shines like the Sun in the constellation of Southern England. Like all stars, it has an immense gravitational pull. Whether by car, train or tube, thousands travel into the capital each day from all directions. Including this 'commuter belt' beyond the Greater London Authority boundary makes the capital one of the largest metropolitan areas in the EU with a population of more than 13 million.

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SOURCE: 2011 CENSUS, ONS

4 | 3