

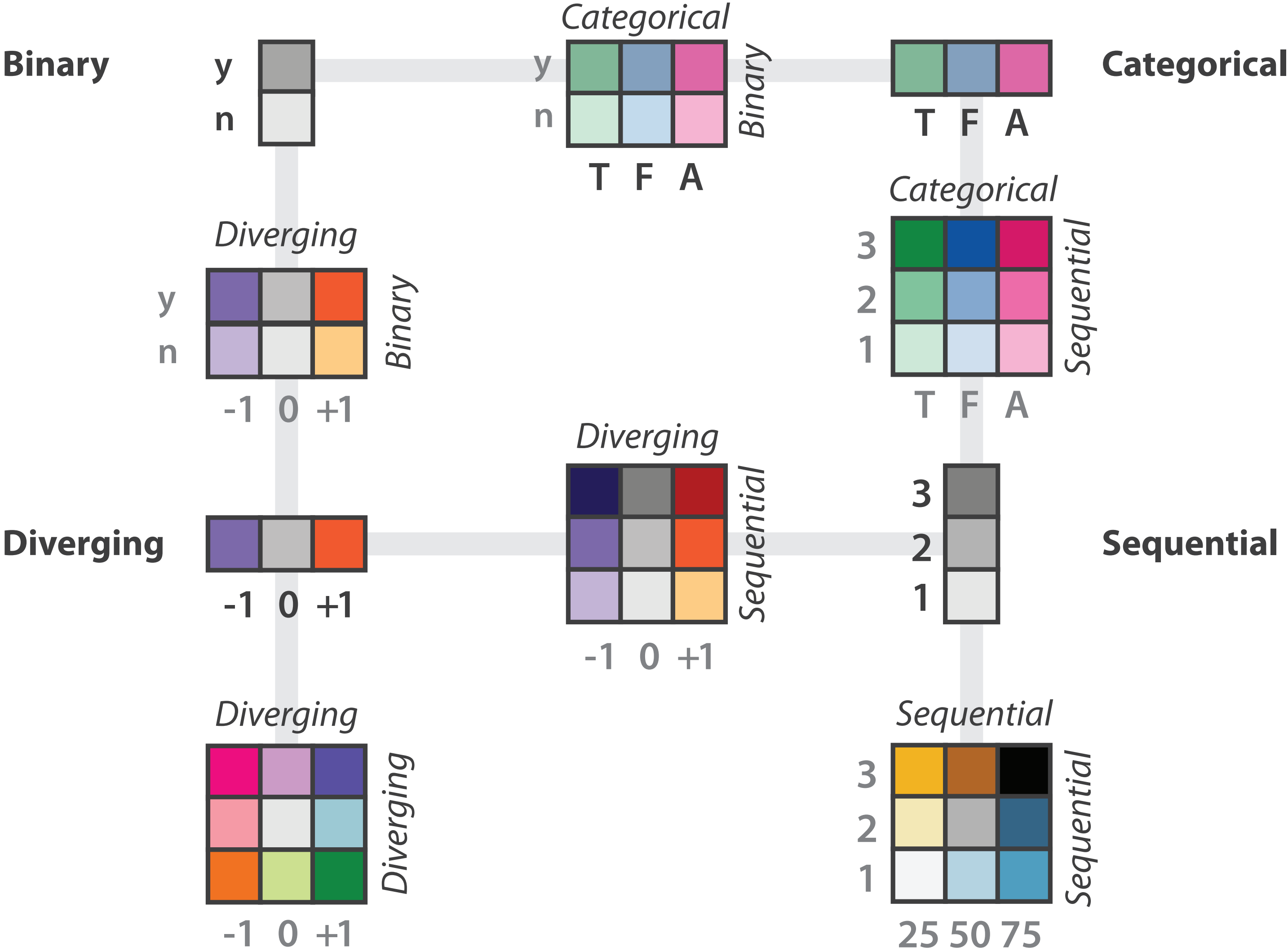
# Data Visualization (CSCI 627/490)

---

Maps & Networks

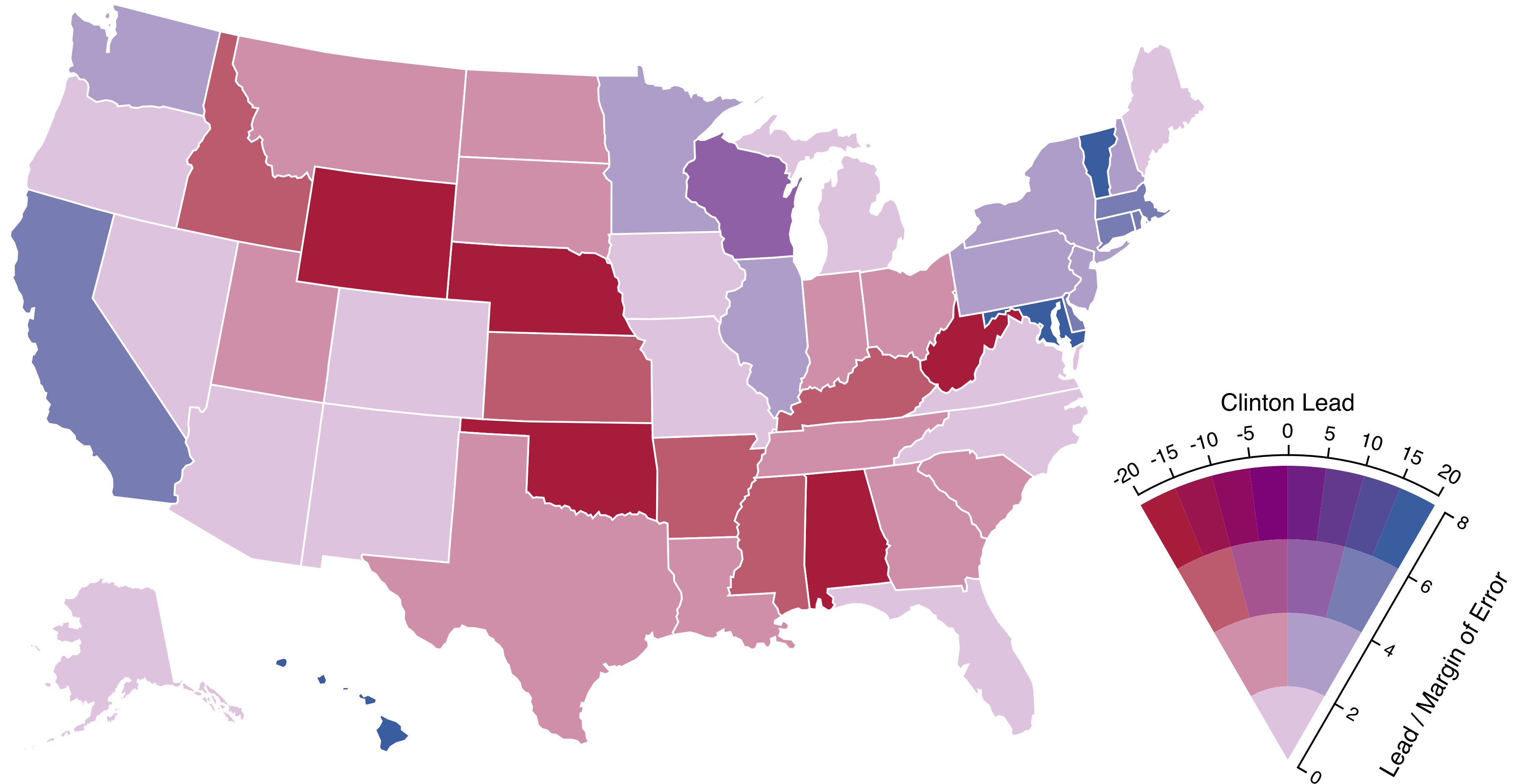
Dr. David Koop

# Bivariate Colormaps



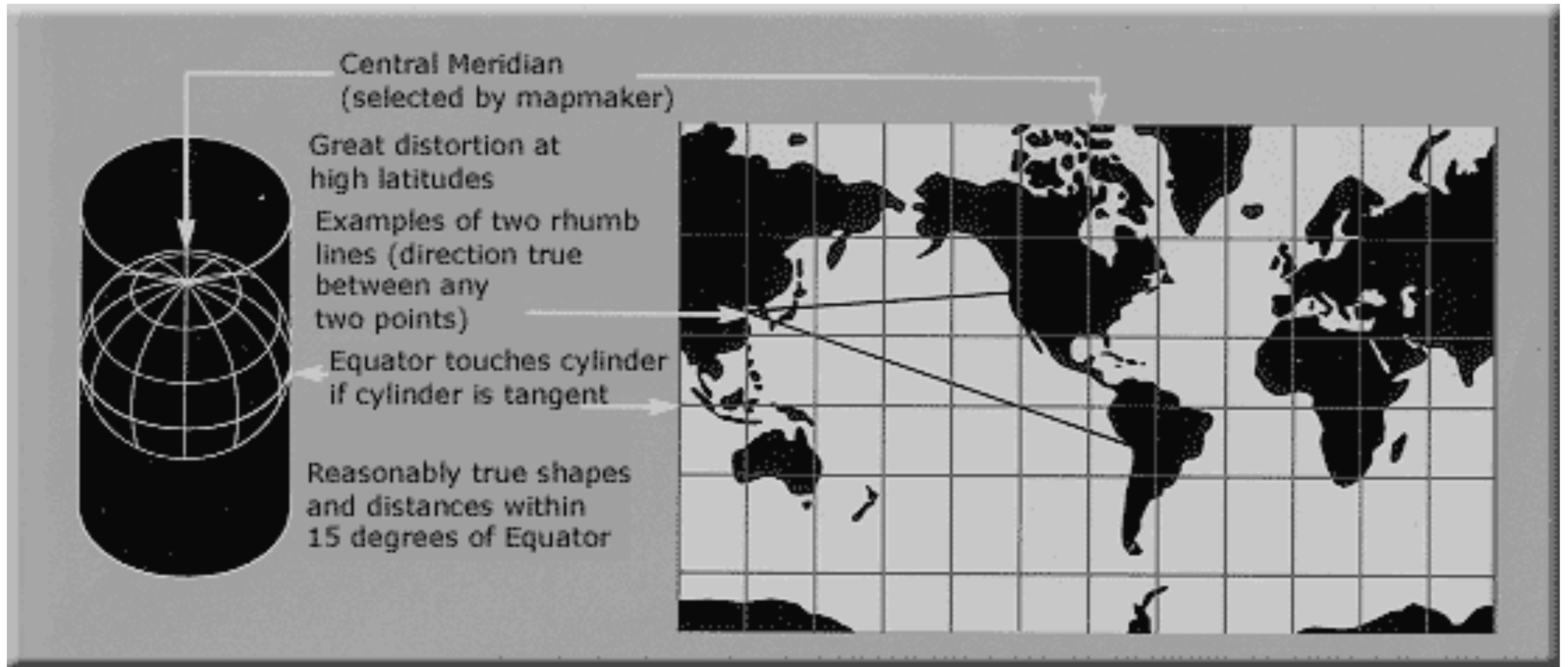
[Munzner (ill. Maguire), 2014]

# Value-Suppressing Uncertainty Palette



[Correll et al., 2018]

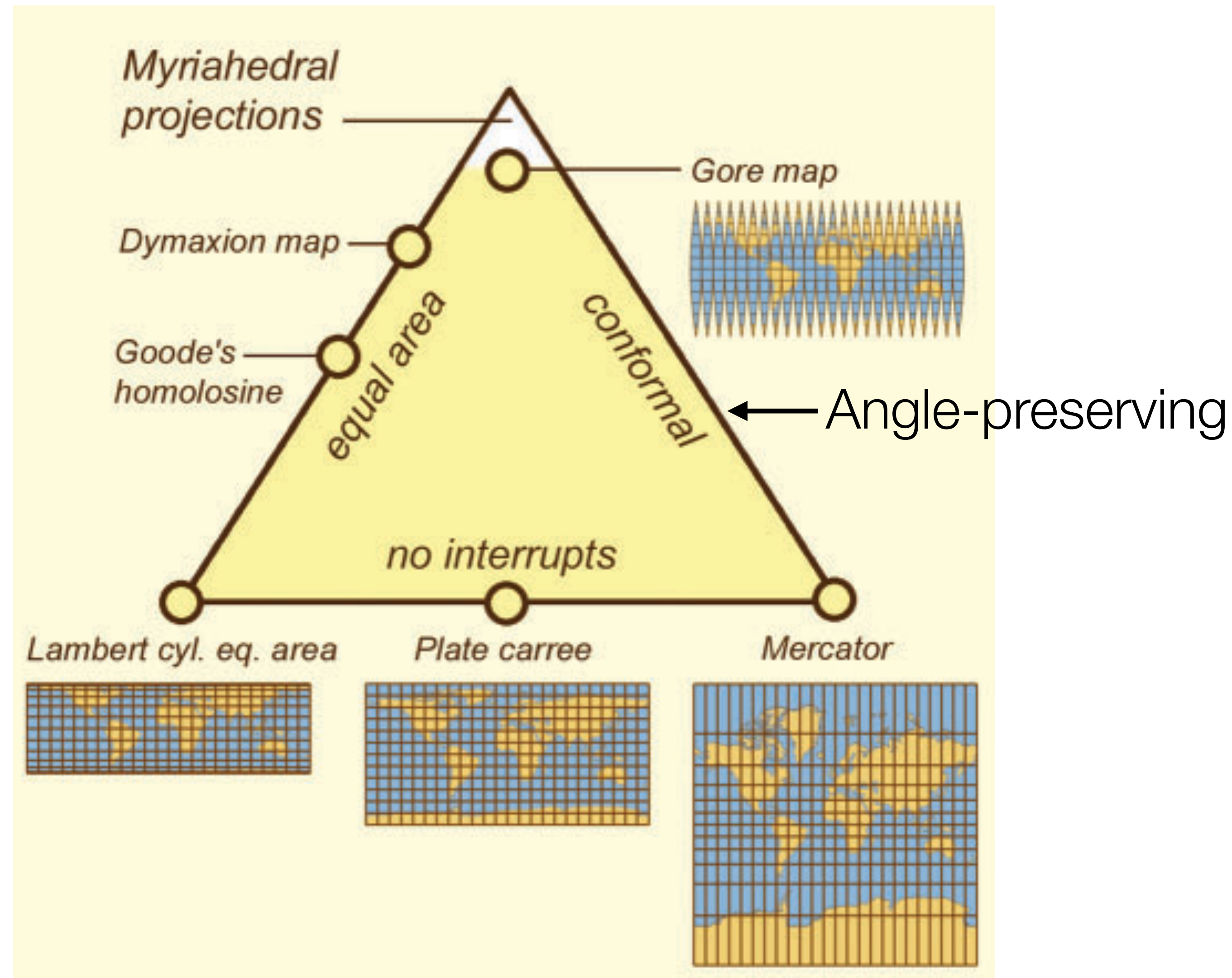
# Geographic Data: 3D to 2D: Projection



[USGS Map Projections]

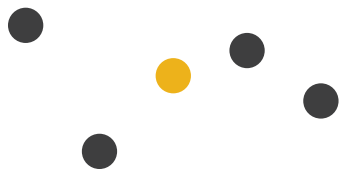
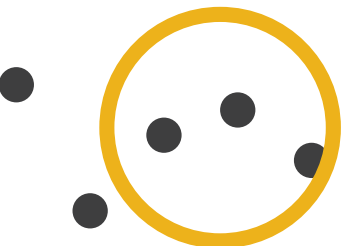
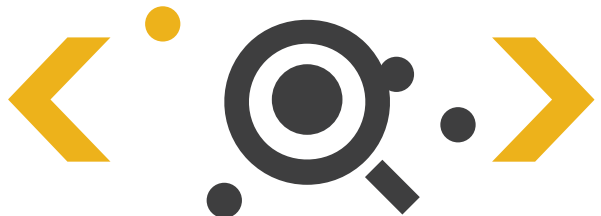



# Projection Classification



[J. van Wijk, 2008]

# Search Tasks

	Target known	Target unknown
Location known	 <i>Lookup</i>	 <i>Browse</i>
Location unknown	 <i>Locate</i>	 <i>Explore</i>

[Munzner (ill. Maguire), 2014]



This map displays the geographic distribution of 1000 points across the New York City metropolitan area. The points are represented by small circles, with a color gradient from white to red. The highest concentration of points is located in the Lower East Side of Manhattan, where they are colored red. From this central cluster, the points radiate outwards, becoming white as they move into the surrounding areas of Queens, Brooklyn, and the New Jersey suburbs. Major transportation corridors, including the New Jersey Turnpike, I-95, I-278, and the Long Island Expressway, are clearly visible. The map also shows the Hudson River, the East River, and various local airports like Newark Liberty and John F. Kennedy International.



# Adding Continuous Data to a Map: Isolines



[USGS via Wikipedia]



# Assignment 4

---

- To be announced soon
- Colormaps, geospatial vis, networks



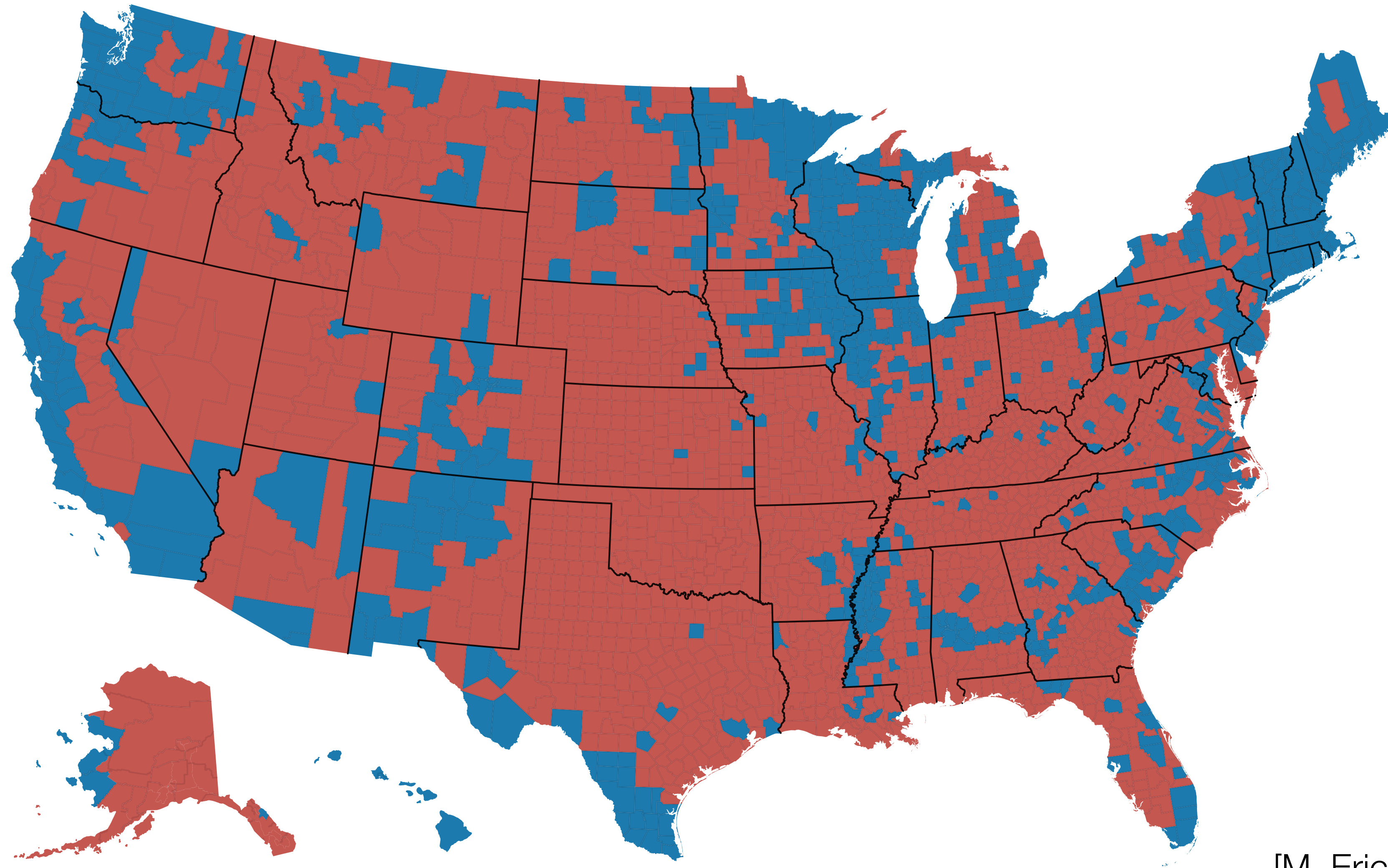
# Project

---

- Next Step: Design
  - Given dataset (what) and tasks (why), work on the how
  - Don't do this the other way around: do not start with "I want to make a streamgraph" and then decide what tasks could work with that
  - This includes interactive design

# Choropleth (Two Hues)

---



[M. Ericson, New York Times]

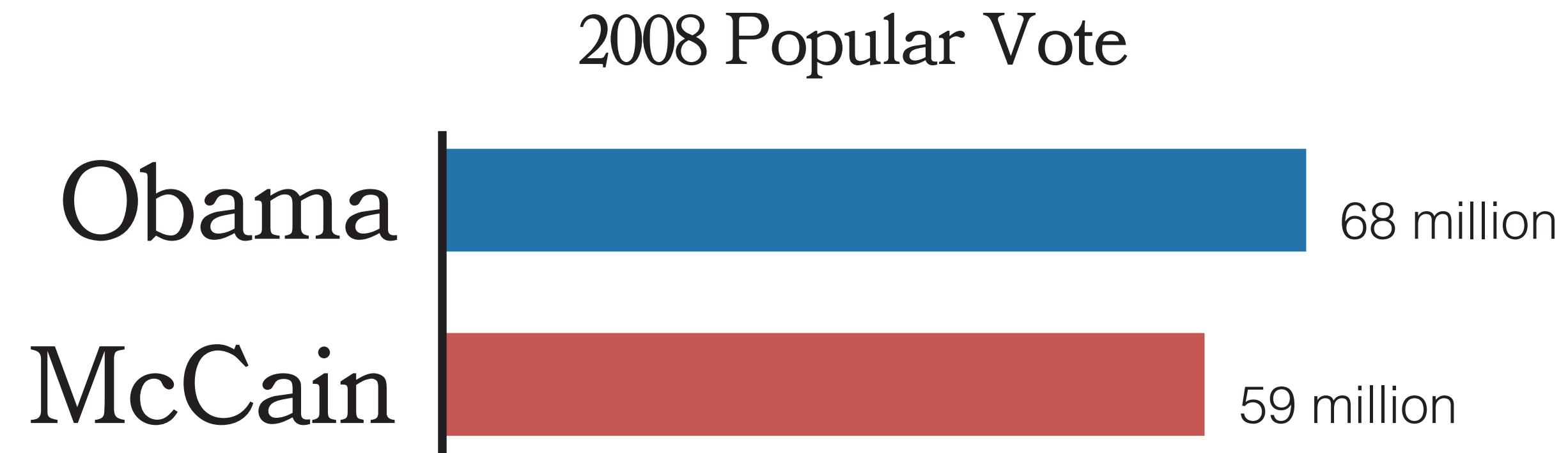
# Choropleth Map

---

- Data: geographic geometry data & one quantitative attribute per region
- Tasks: trends, patterns, comparisons
- How: area marks from given geometry, color hue/saturation/luminance
- Scalability: thousands of regions
  
- Design choices:
  - Colormap
  - Region boundaries (level of summarization)

# Problem?

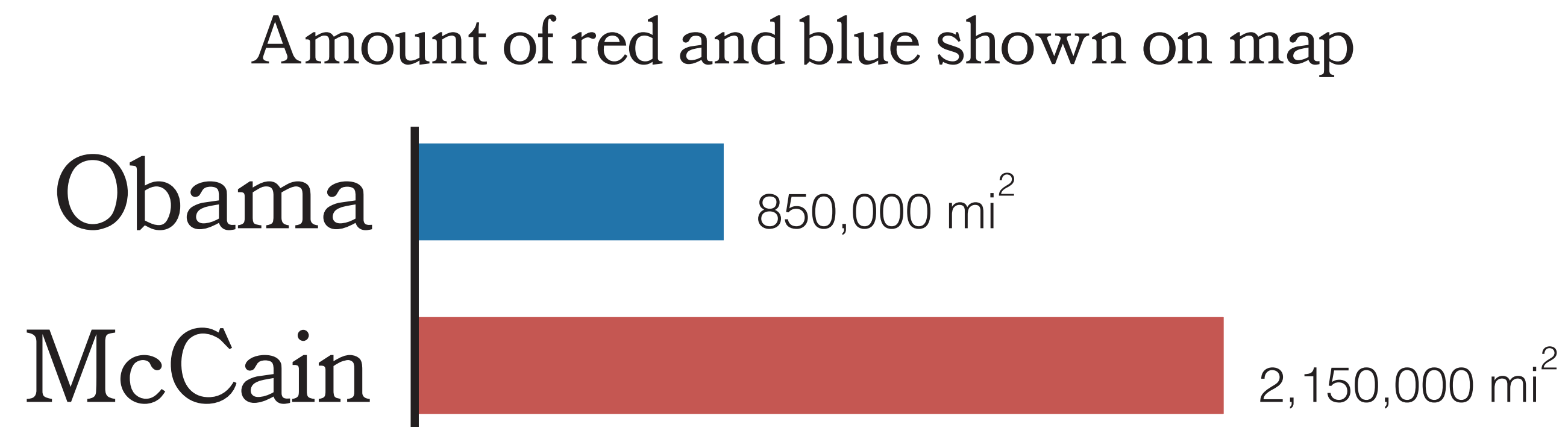
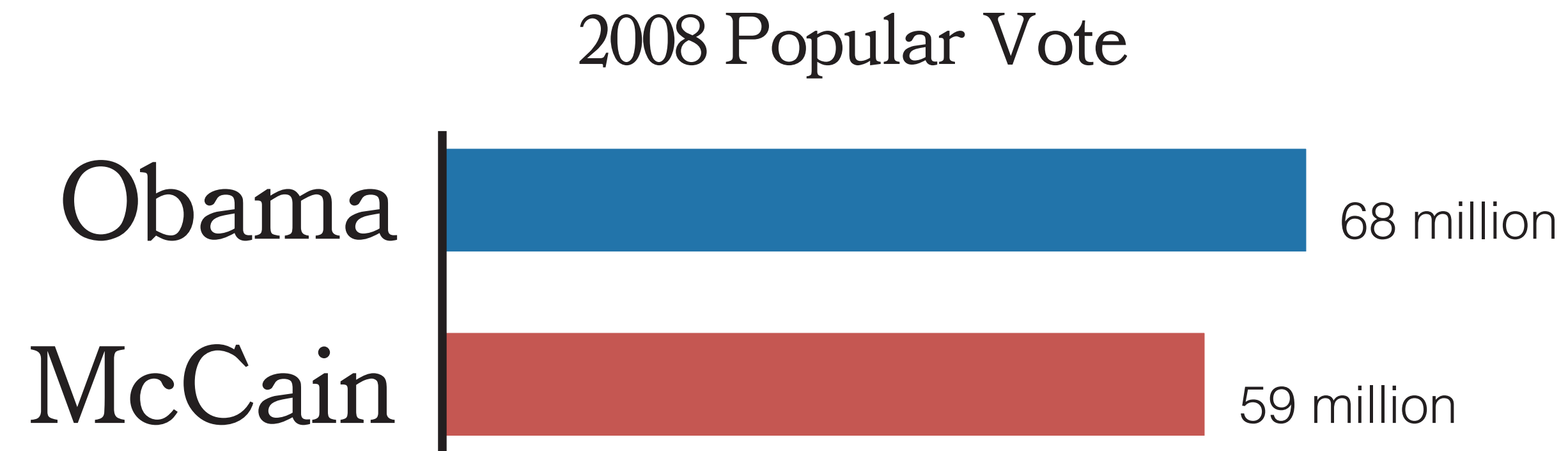
---



[M. Ericson, New York Times]

# Problem?

---

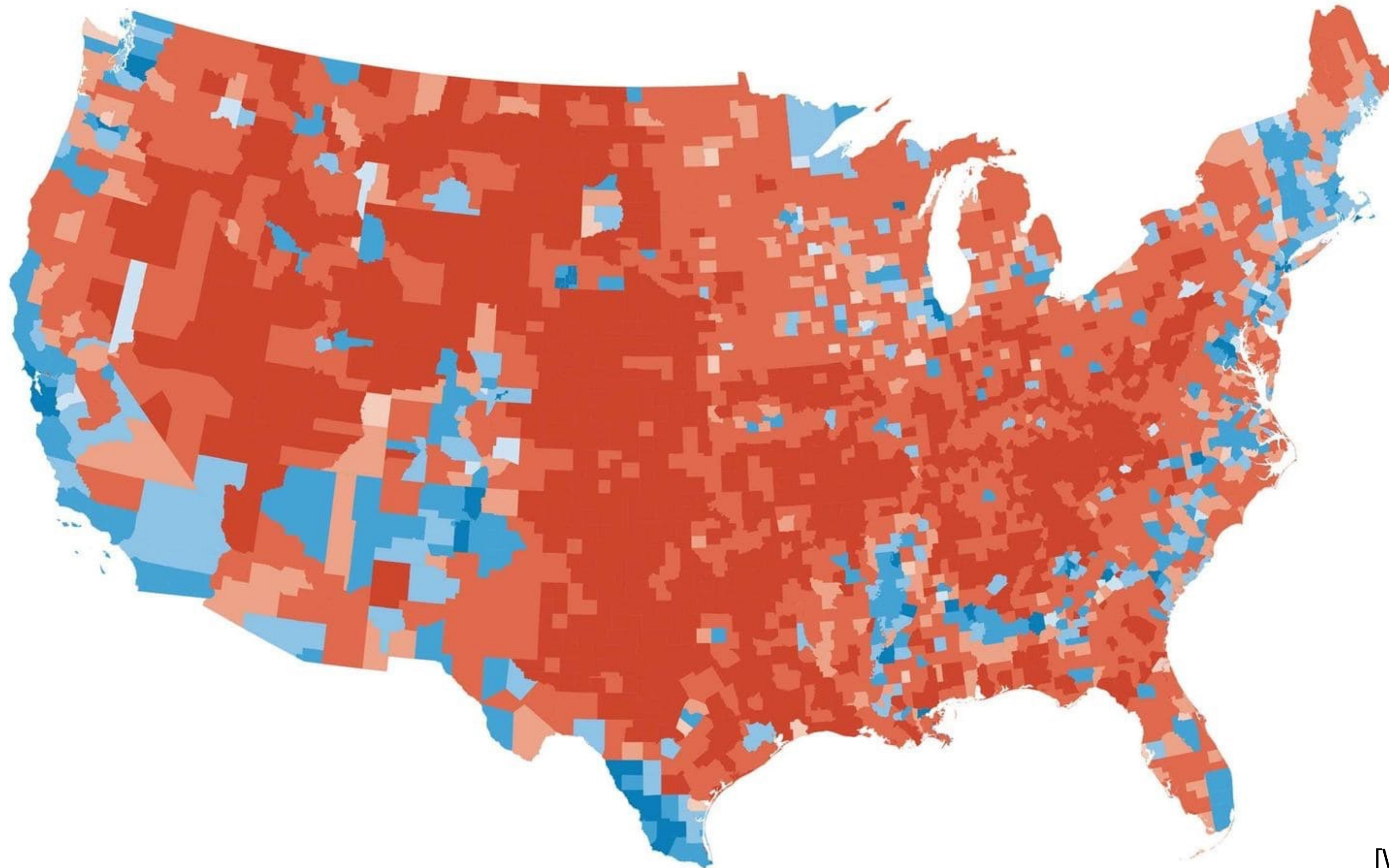


[M. Ericson, New York Times]



# Adding Saturation

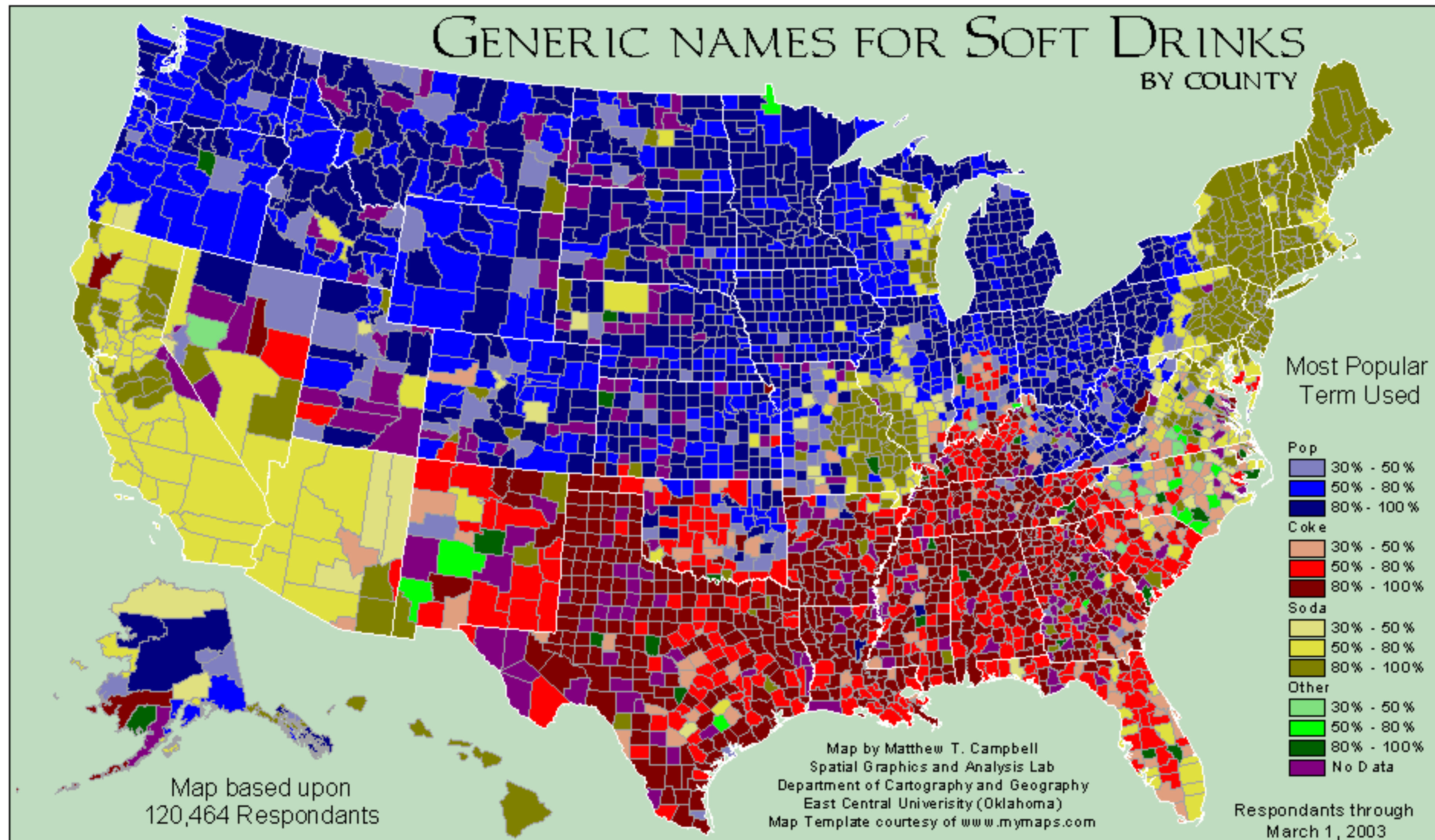
Clinton +50-100 +15-50 +2.1-15 +0-2.1 Trump +0-2.1 +2.1-15 +15-50 +50-100



[Washington Post, 2018]



# Area Marks and Color Hue & Saturation

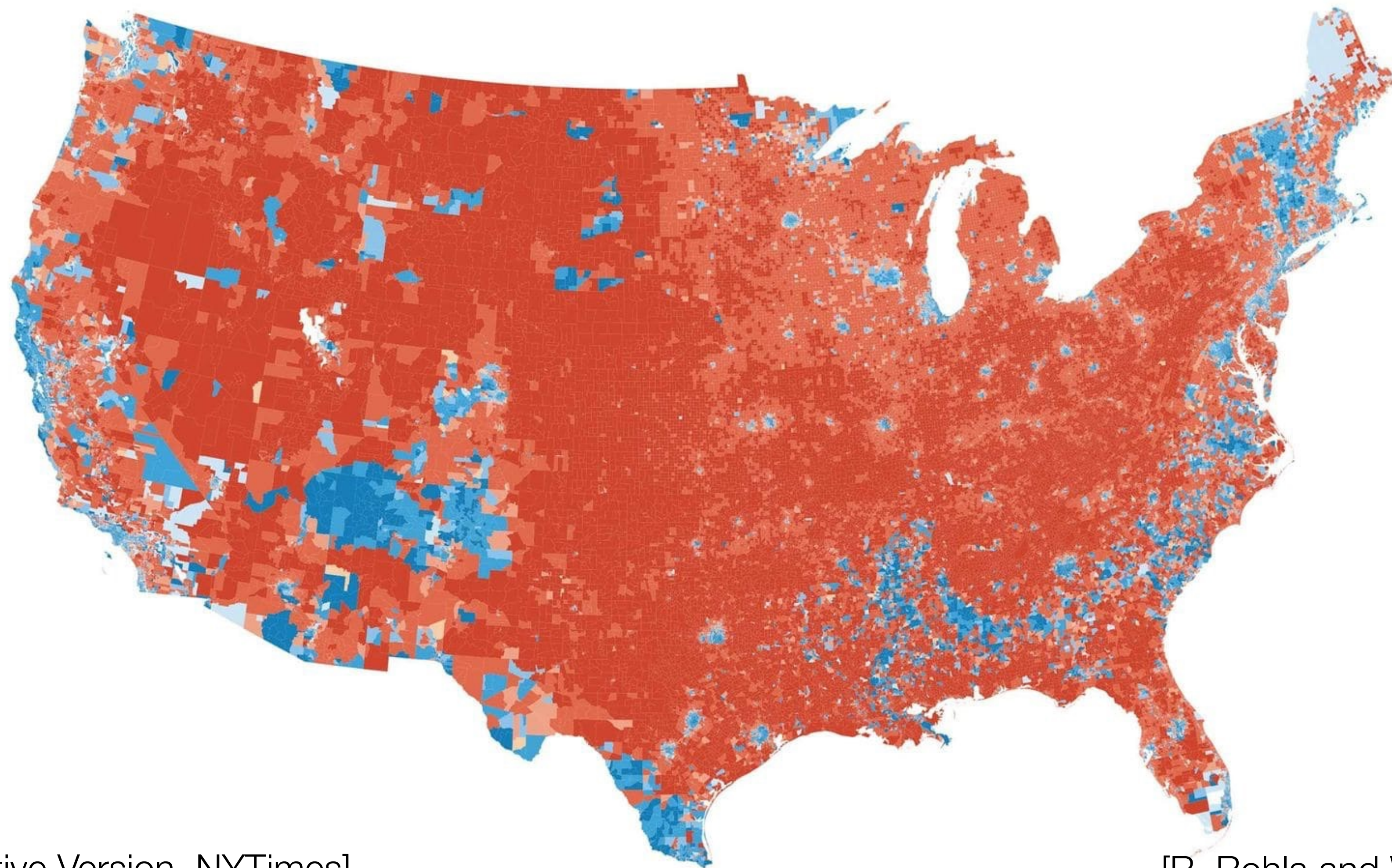


[[popvssoda.com](http://popvssoda.com)]



# Aggregation: 2016 Election by Precinct

Clinton +50-100 +15-50 +2.1-15 +0-2.1 Trump +0-2.1 +2.1-15 +15-50 +50-100



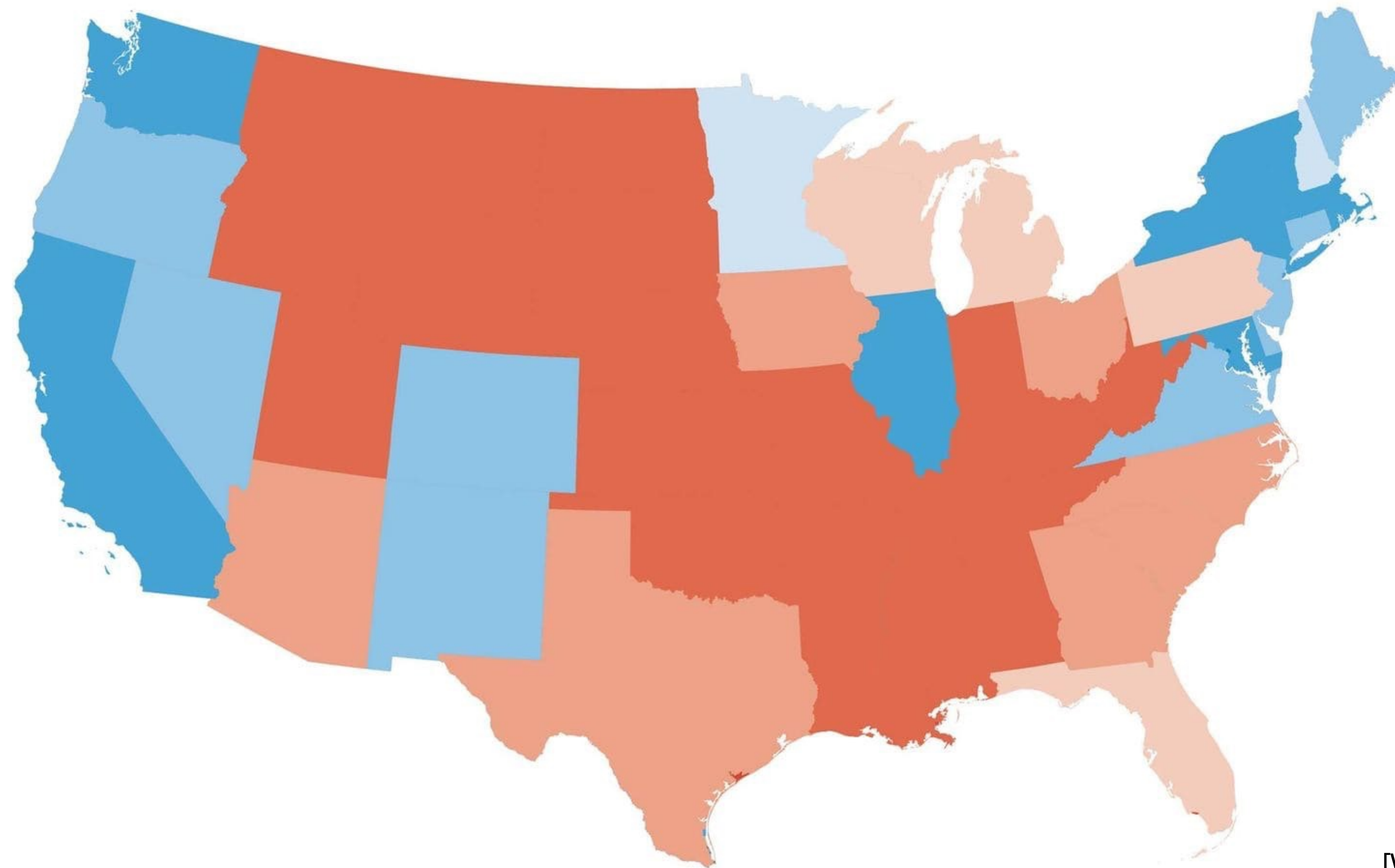
[[Interactive Version](#), NYTimes]

[R. Rohla and [Washington Post](#), 2018]



# Aggregation: 2016 Election by State

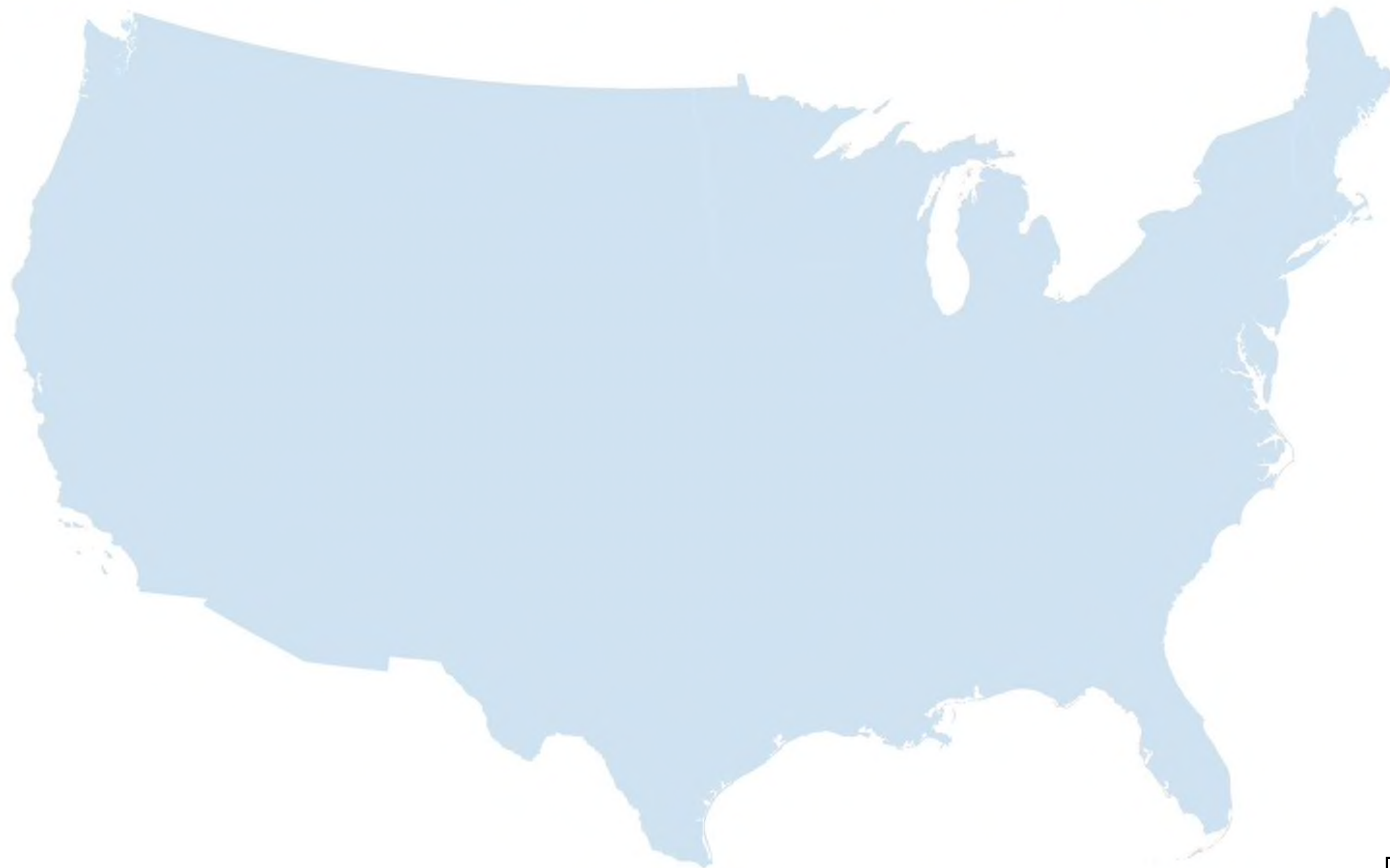
Clinton +50-100 +15-50 +2.1-15 +0-2.1 Trump +0-2.1 +2.1-15 +15-50 +50-100



[Washington Post, 2018]

# Aggregation: 2016 Election by Country

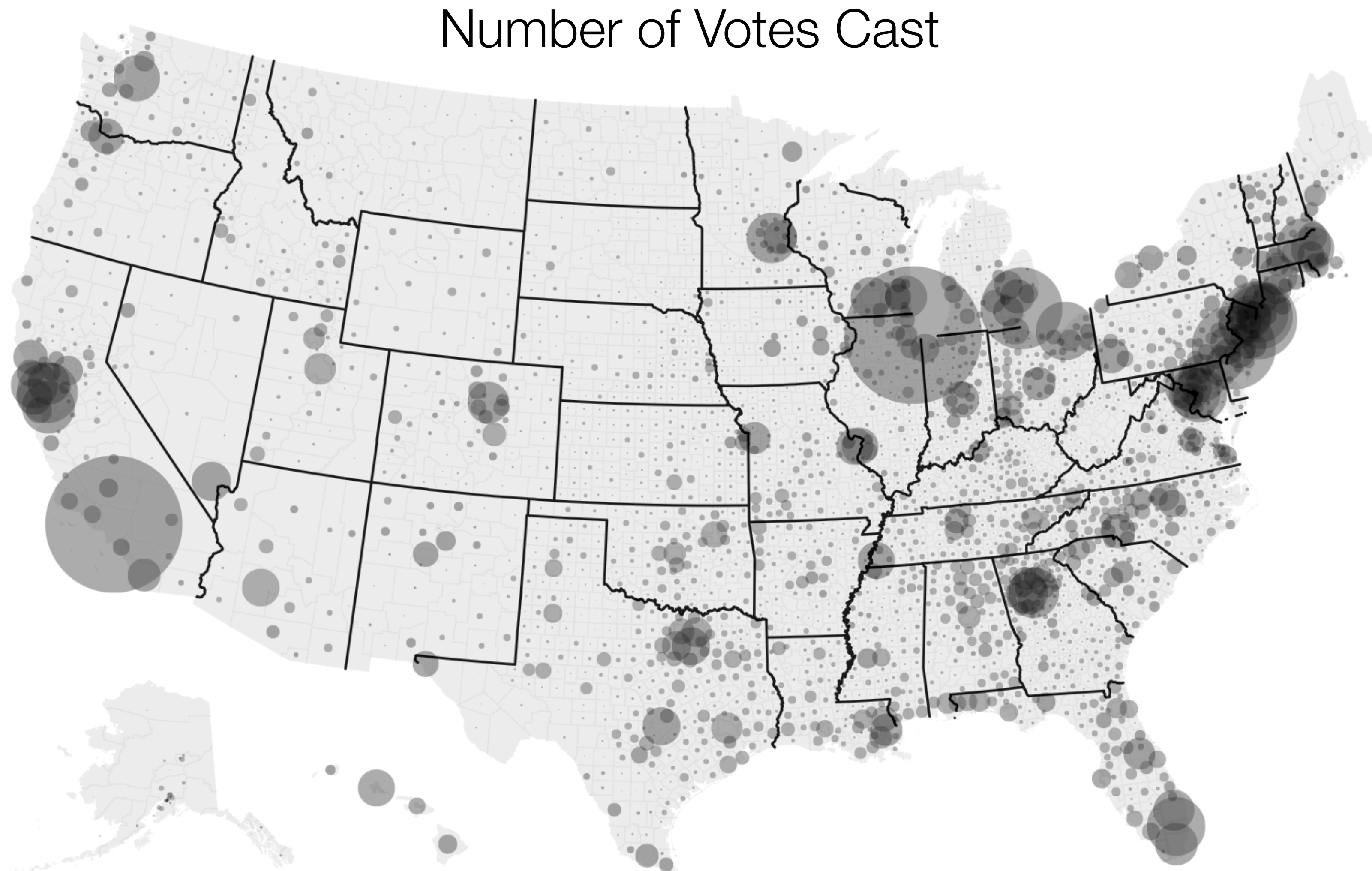
■ Clinton +50-100 ■ +15-50 ■ +2.1-15 ■ +0-2.1 ■ Trump +0-2.1 ■ +2.1-15 ■ +15-50 ■ +50-100



[Washington Post, 2018]

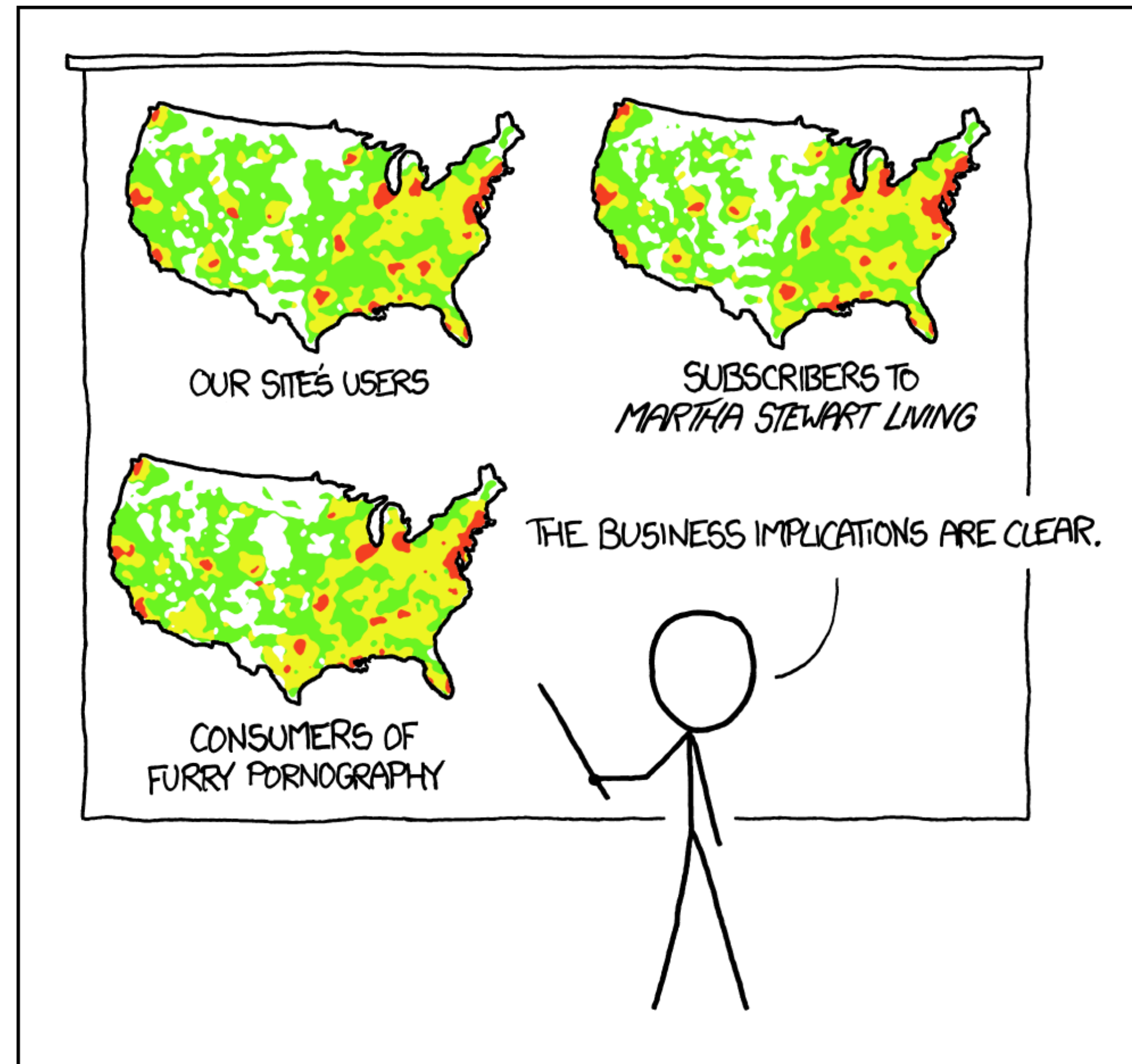


# Maps: What trends do you see?



[Desaturated by D. Koop, M. Ericson, New York Times]

# Don't Just Create Population Maps!

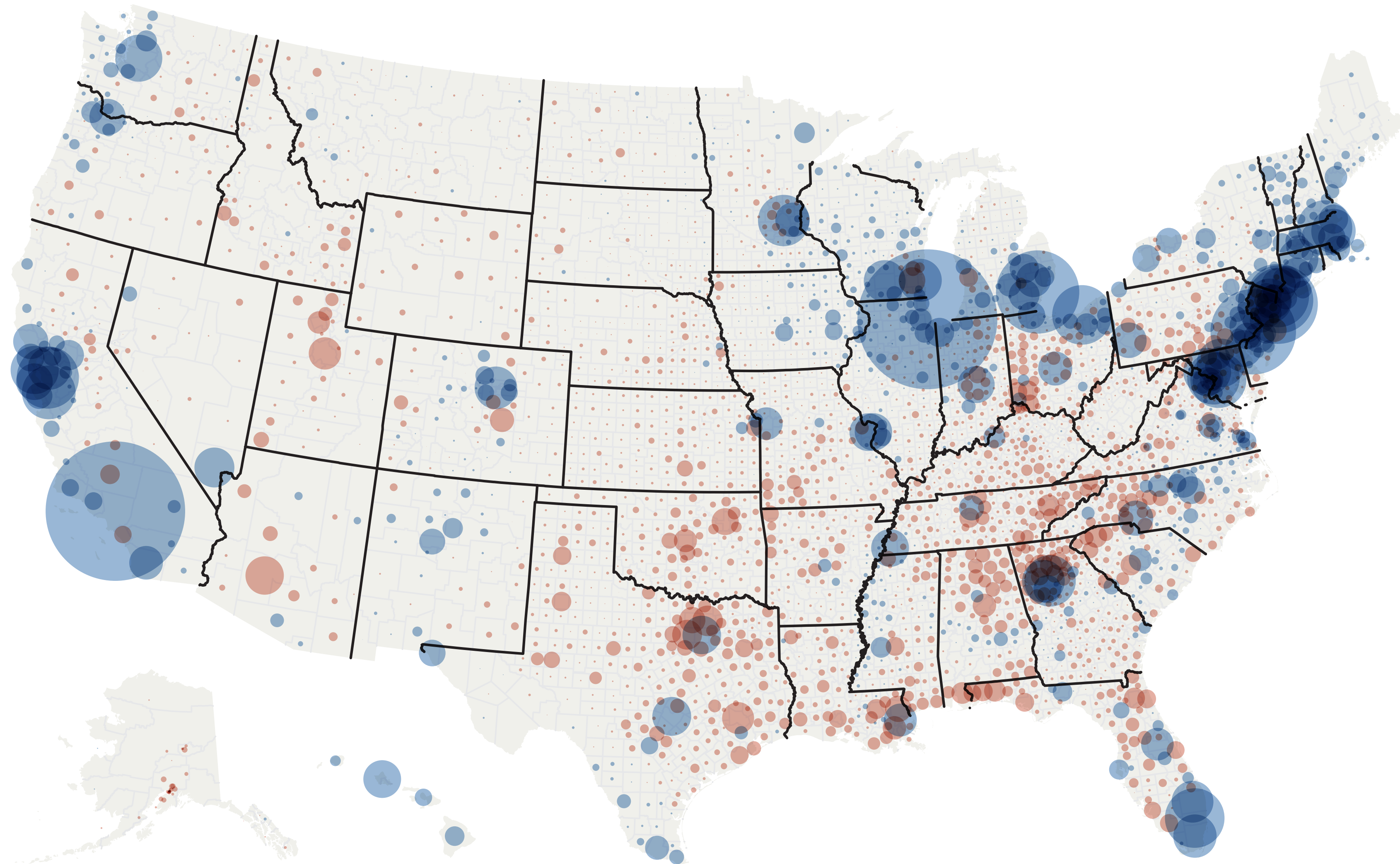


PET PEEVE #208:  
GEOGRAPHIC PROFILE MAPS WHICH ARE  
BASICALLY JUST POPULATION MAPS

[xkcd]



# Size Encoding

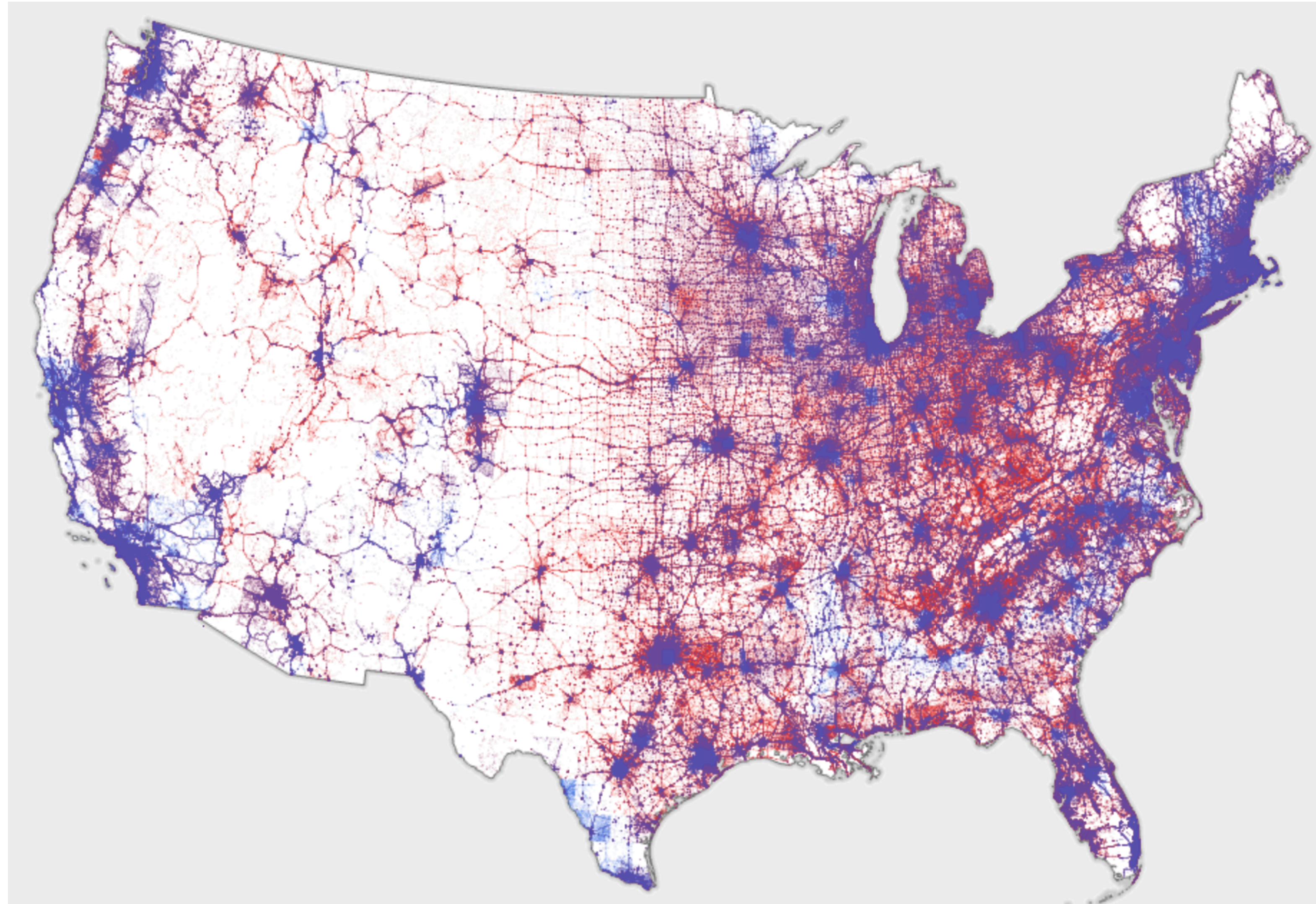


[M. Ericson, New York Times]



# Dasymetric Dot Density

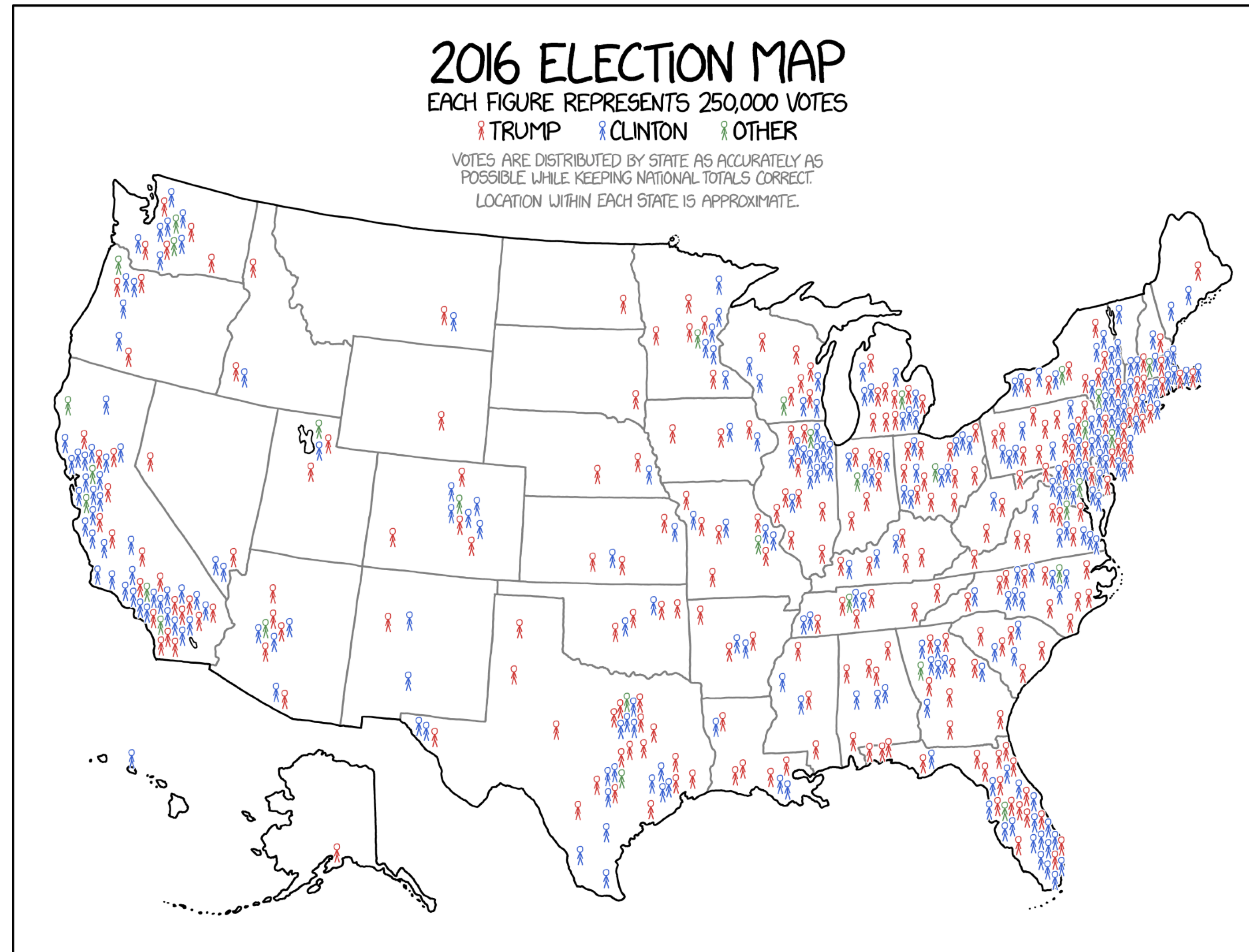
---



[K. Field]



# Glyphs: xkcd's Map



[xkcd]



# Cartograms

## US Presidential Election 2016

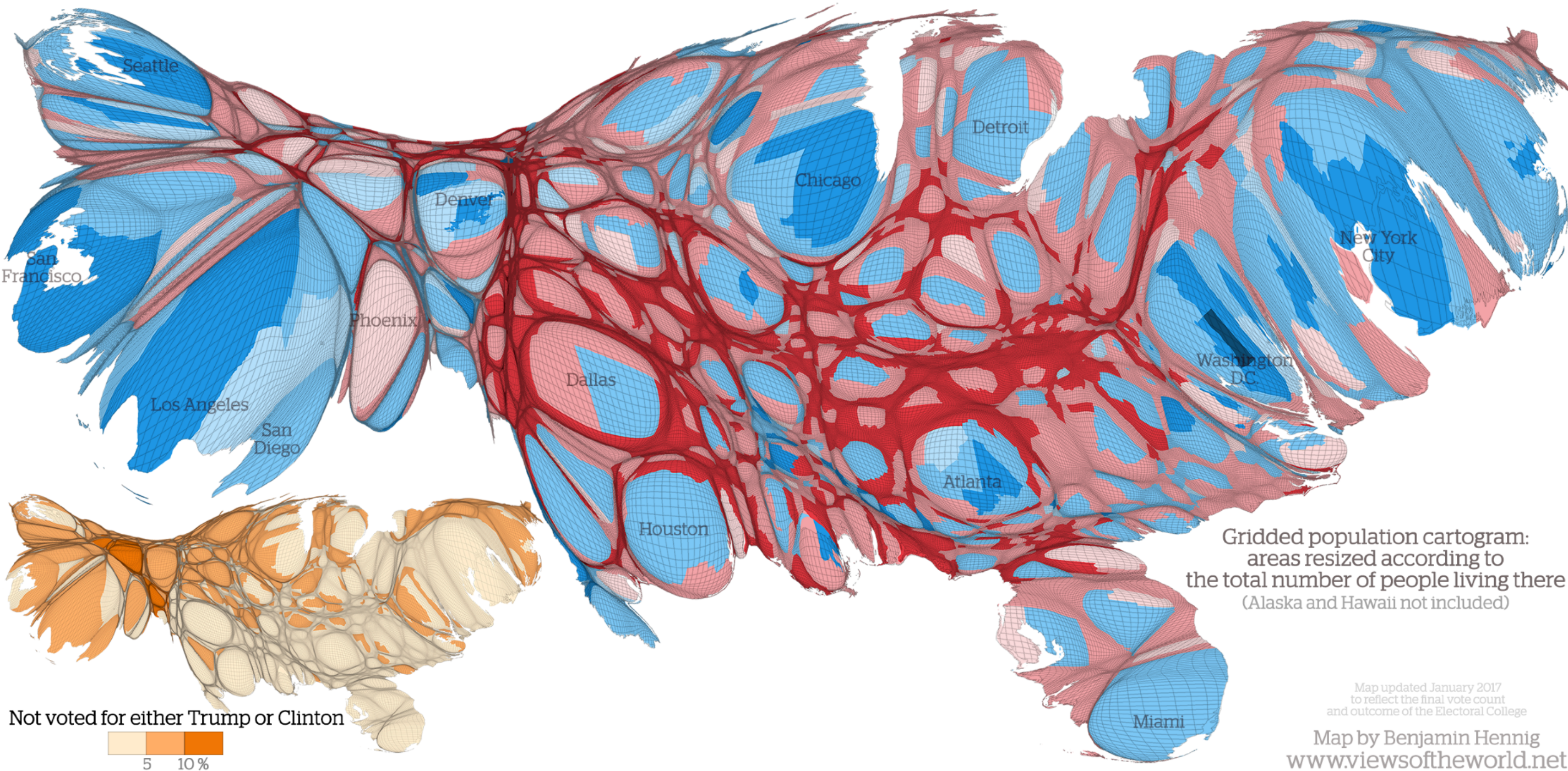
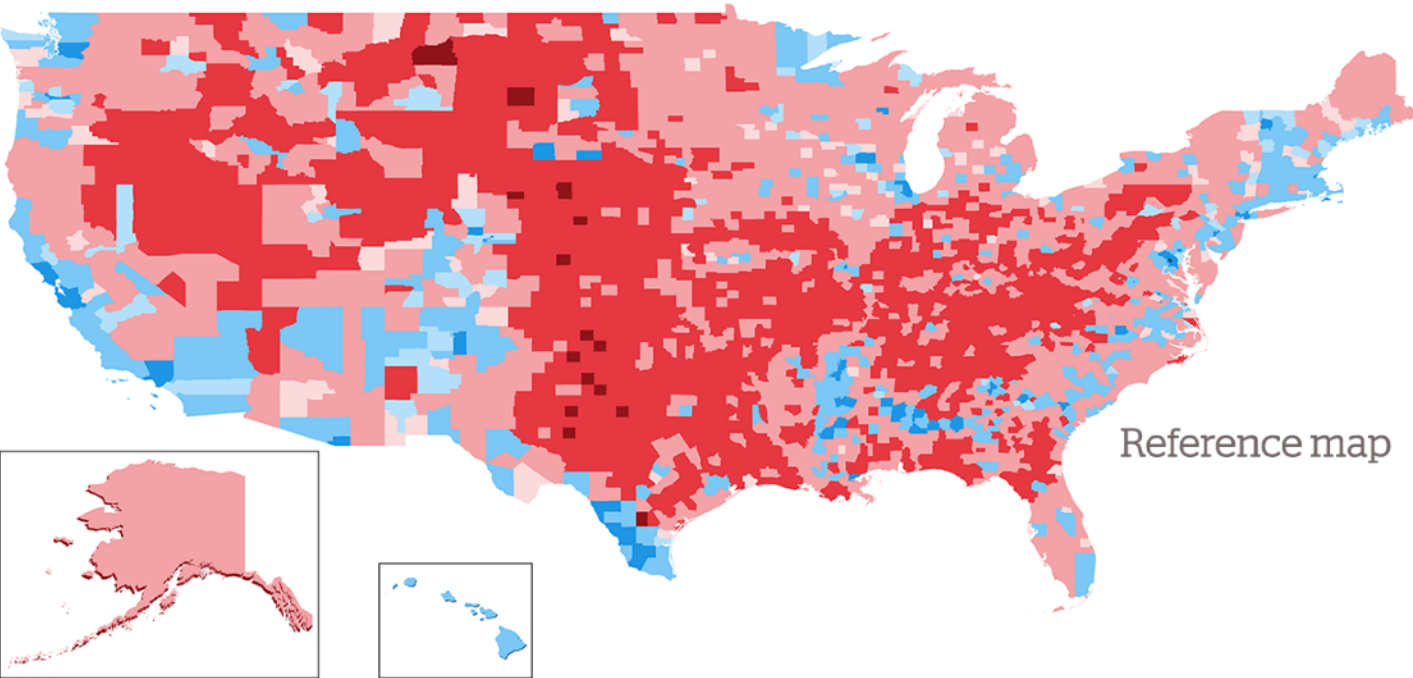
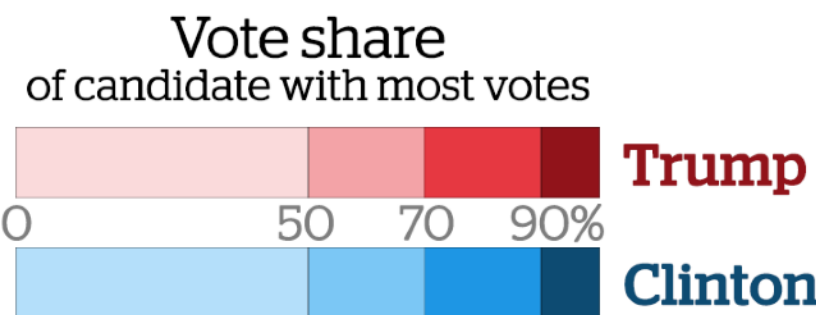
Results mapped at county level showing the candidate with the largest vote share in each area

**Overall result:**

**Trump**  
62,979,636 votes (46.1%)  
306 electoral votes  
(received 304 in the Electoral College)

**Clinton**  
65,844,610 votes (48.2%)  
232 electoral votes  
(received 227 in the Electoral College)

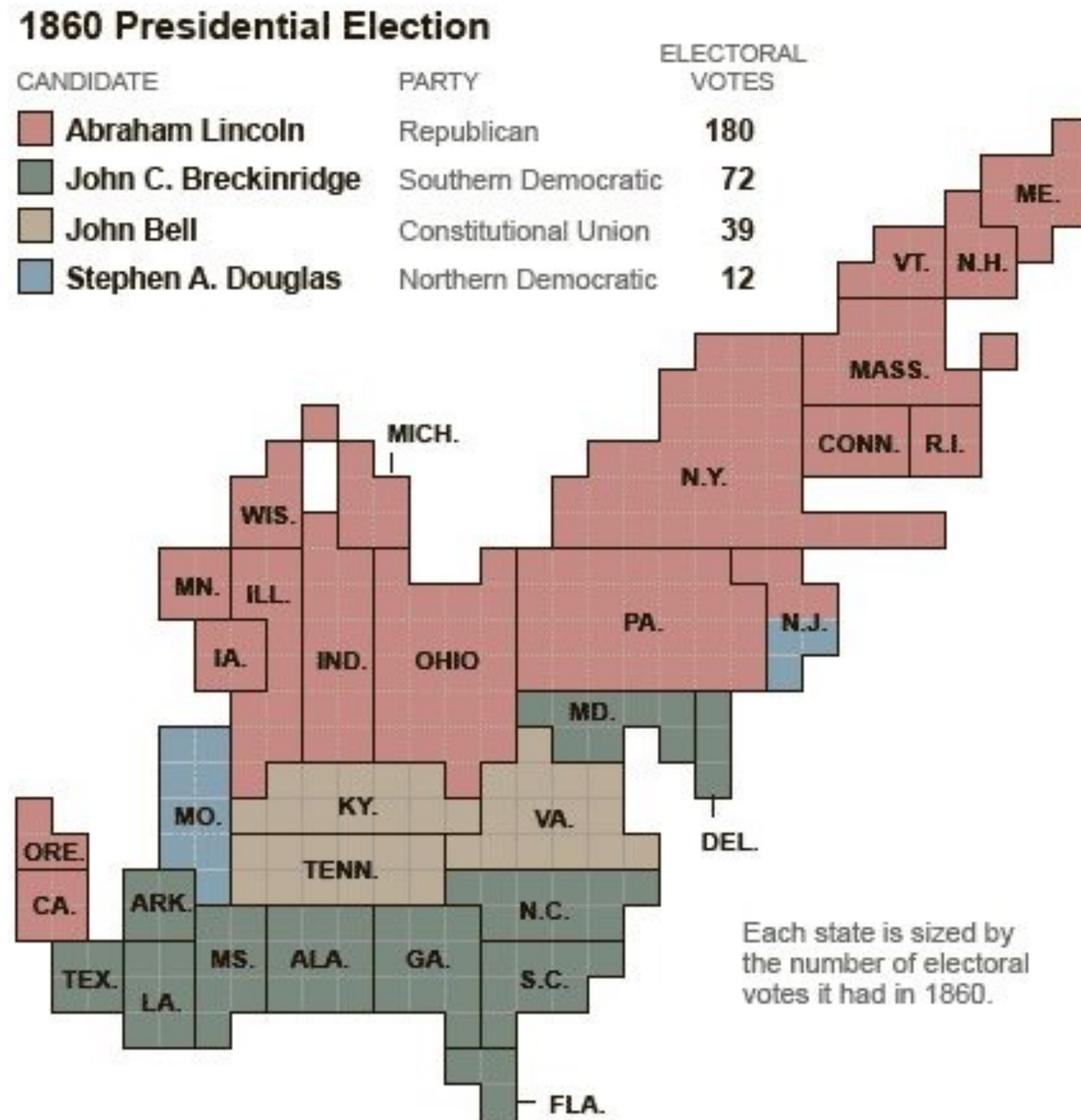
**Other candidates**  
7,804,213 votes (5.7%)



[B. Hennig]



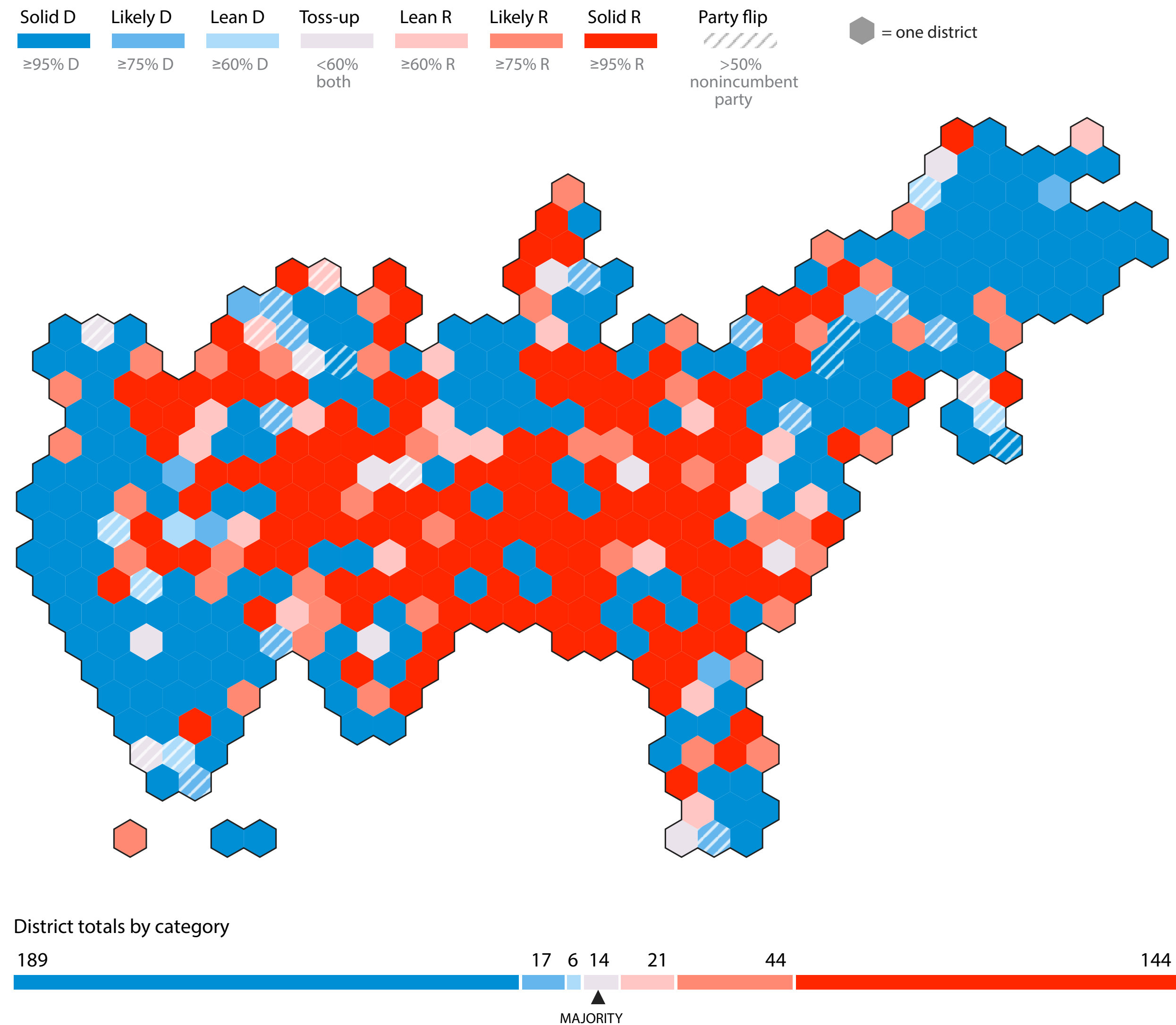
# Cartograms



- Data: geographic geometry data & **two** quantitative attributes (one part-of-whole)
- Derived data: new geometry derived from the part-of-whole attribute
- Tasks: trends, comparisons, part-of-whole
- How: area marks from derived geometry, color hue/saturation/luminance
- Scalability: thousands of regions
- Design choices:
  - Colormap
  - Geometric deformation

[New York Times]

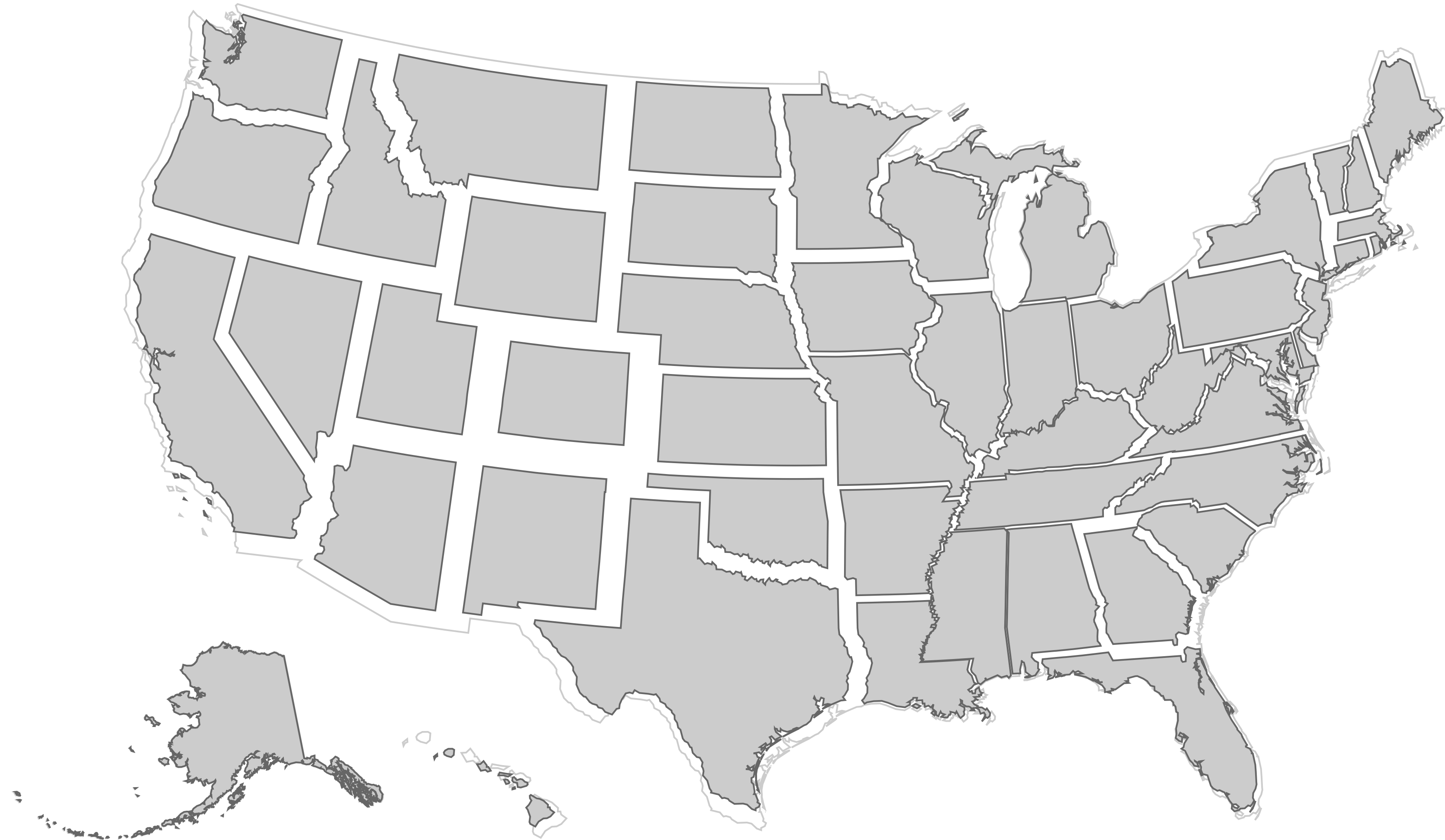
# Hexagonal Cartogram



[FiveThirtyEight, 2018]

# Non-Contiguous Cartogram

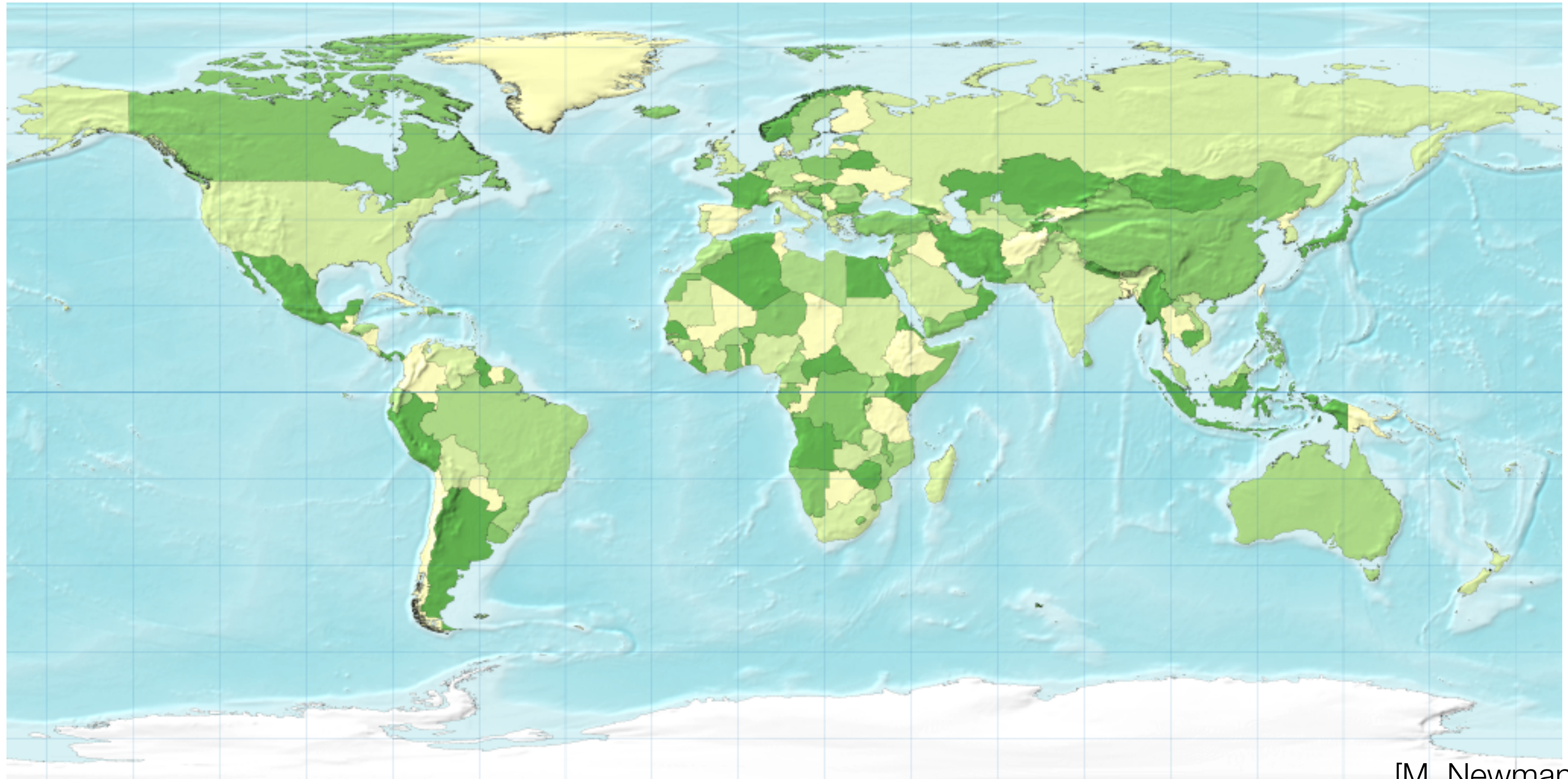
---



[M. Bostock, 2012]



# World Cartograms



[M. Newman, 2009]



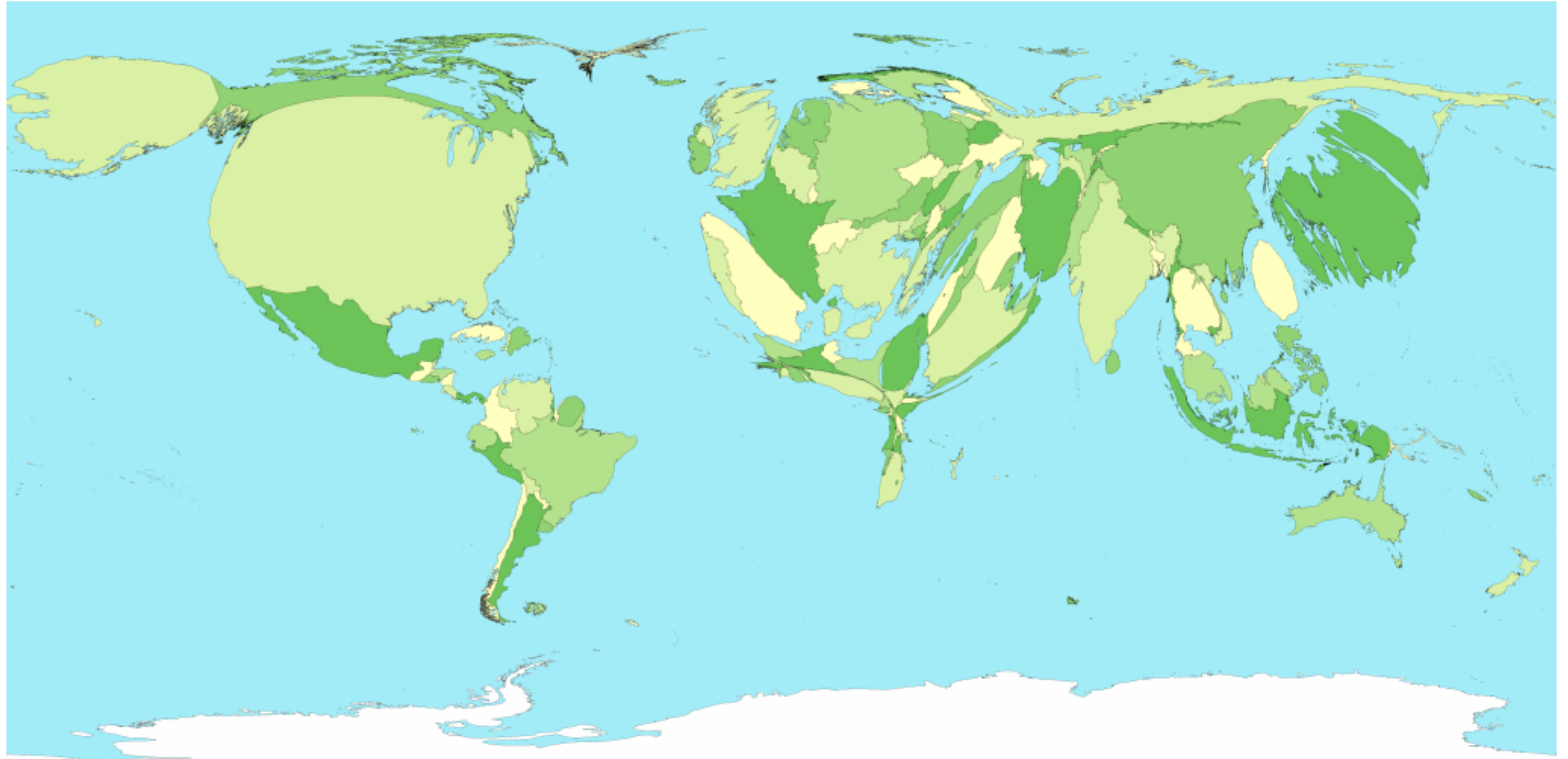
# World Population



[M. Newman, 2009]



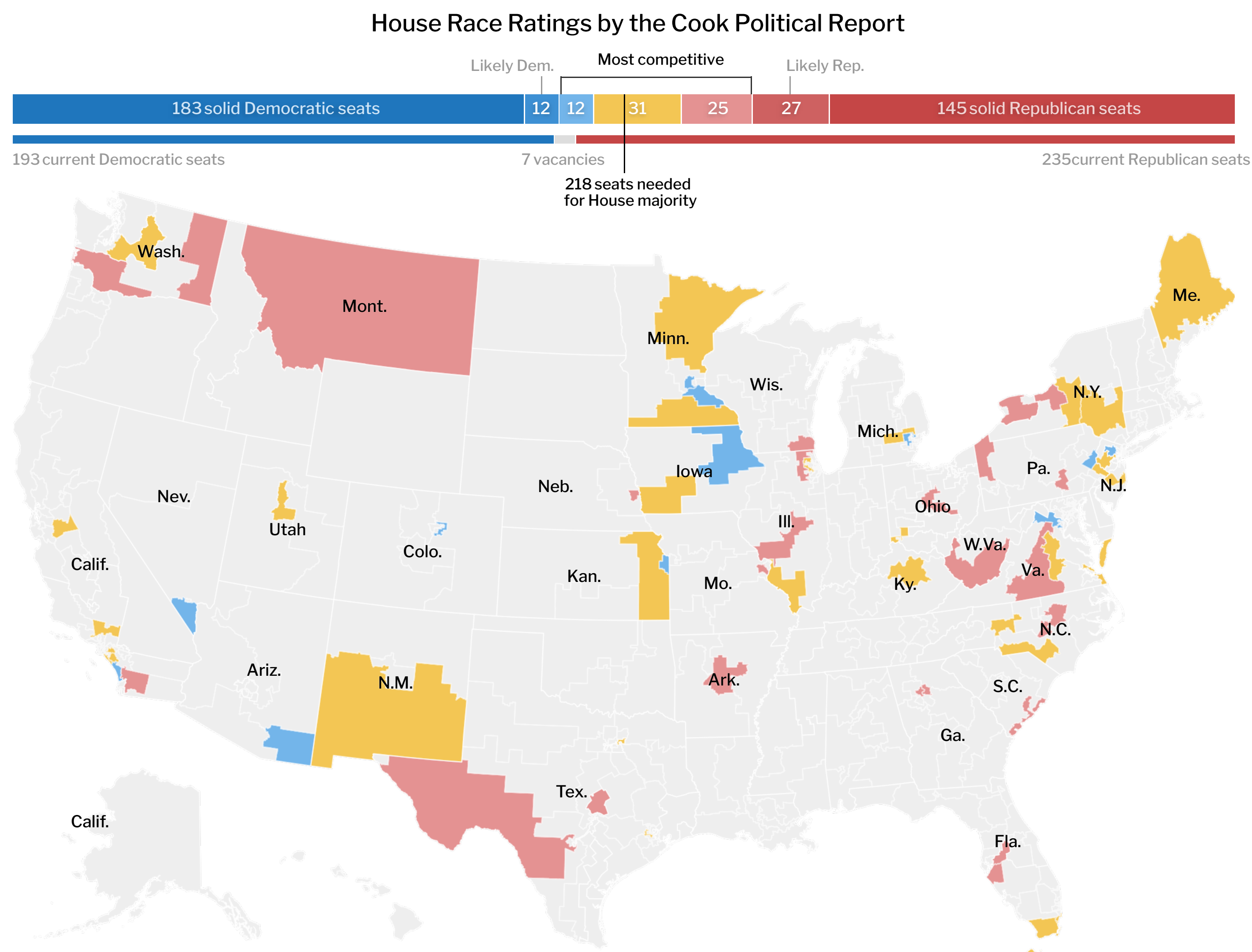
# World Energy Consumption



[M. Newman, 2009]



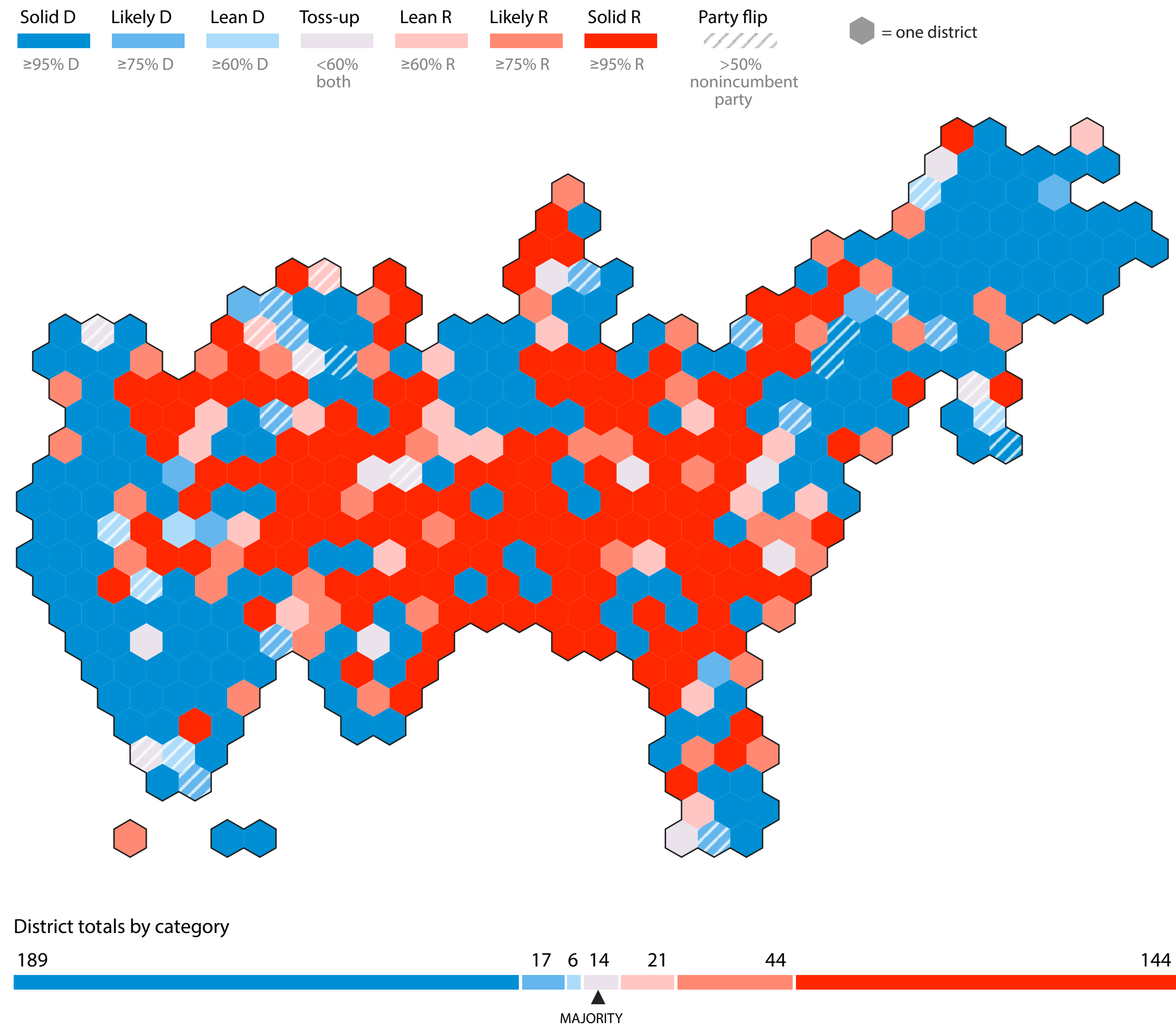
# House Races: Map?



[New York Times, 2018]



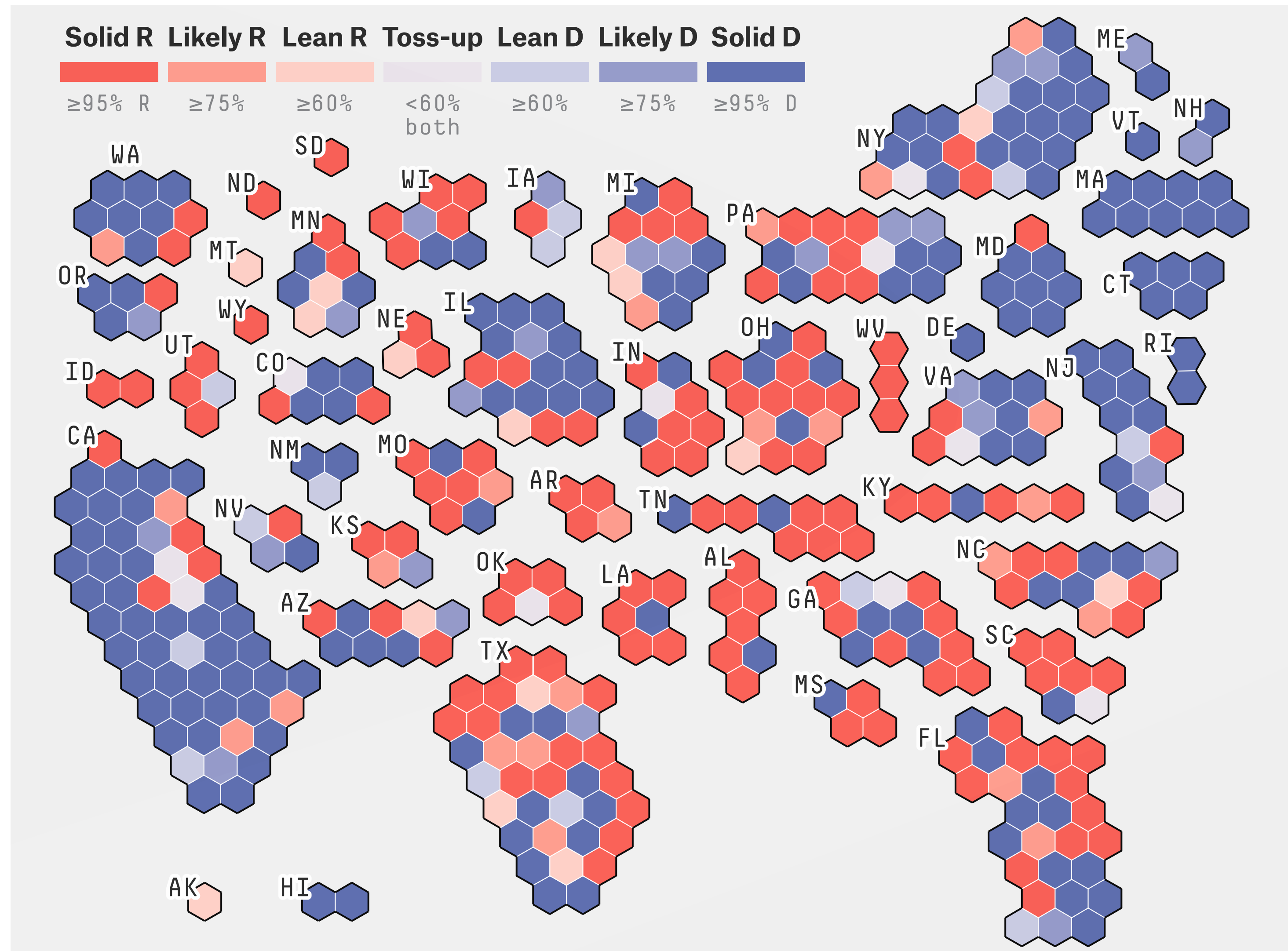
# House Races: Cartogram?



[FiveThirtyEight, 2018]



# House Races: Non-Contiguous "Cartogram"



[FiveThirtyEight, 2020]



# Maps Aren't Always Best: Close House Races

## 12 Lean Democratic

- AZ-02 Open (McSally)
- CA-49 Open (Issa)
- CO-06 Coffman
- IA-01 Blum
- KS-03 Yoder
- MI-11 Open (Trott)
- MN-02 Lewis
- MN-03 Paulsen
- NV-03 Open (Rosen)
- NJ-11 Open (Frelinghuysen)
- PA-07 Vacant (formerly Dent)
- VA-10 Comstock

## 31 Tossups

- CA-10 Denham
- CA-25 Knight
- CA-39 Open (Royce)
- CA-45 Walters
- CA-48 Rohrabacher
- FL-26 Curbelo
- FL-27 Open (Ros-Lehtinen)
- IL-06 Roskam
- IL-12 Bost
- IA-03 Young
- KS-02 Open (Jenkins)
- KY-06 Barr

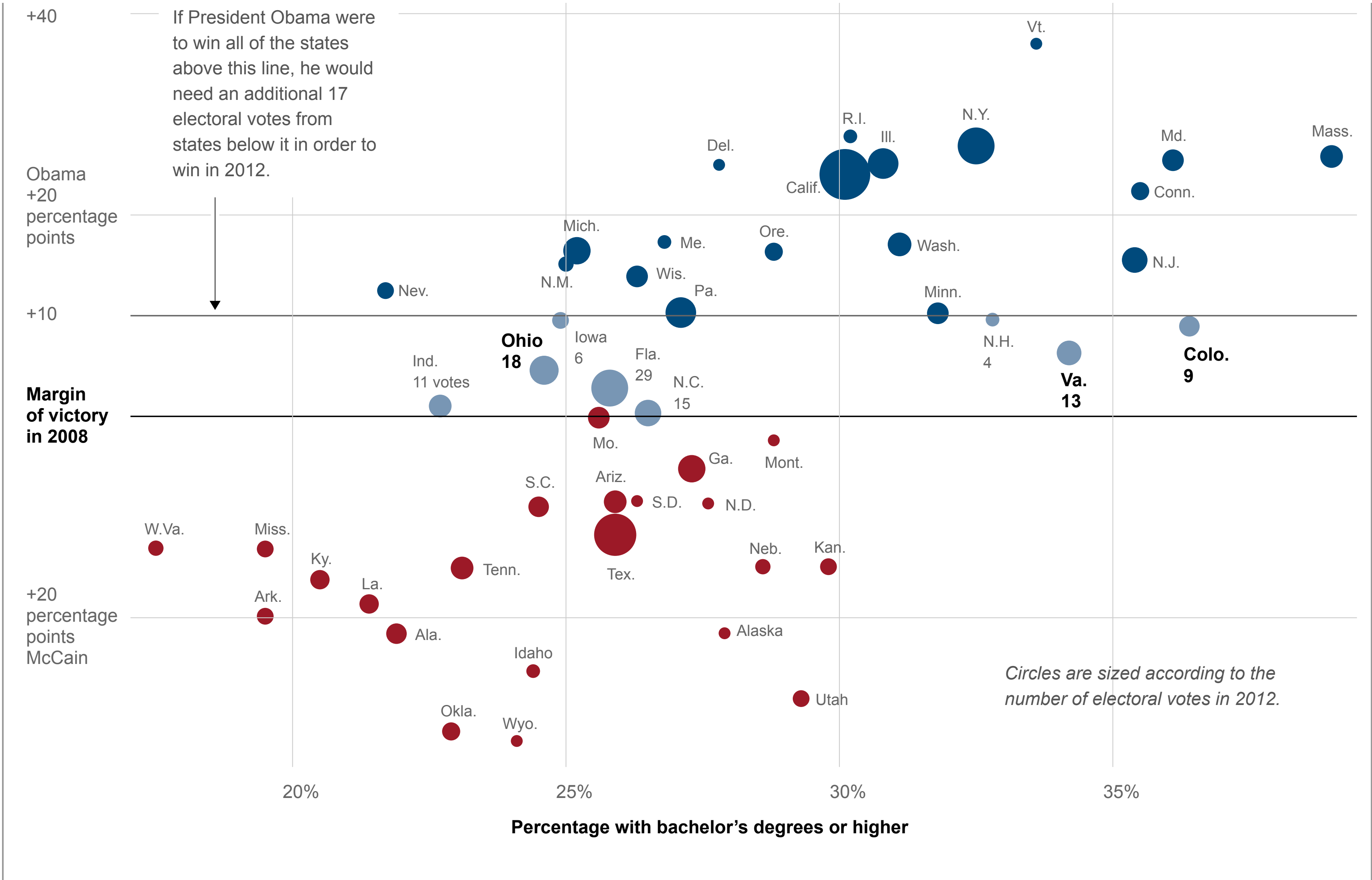
## 25 Lean Republican

- AR-02 Hill
- CA-50 Hunter
- FL-15 Open (Ross)
- FL-16 Buchanan
- GA-06 Handel
- GA-07 Woodall
- IL-13 Davis
- IL-14 Hultgren
- MO-02 Wagner
- MT-AL Gianforte
- NE-02 Bacon
- NY-24 Katko

[New York Times, 2018]



# Maps Aren't Always Best: Obama Targets



[NYTimes]



# Networks

# Networks

---

- Why not graphs?
  - Bar graph
  - Graphing functions in mathematics
- Network: nodes and edges connecting the nodes
- Formally,  $G = (V, E)$  is a set of nodes  $V$  and a set of edges  $E$  where each edge connects two nodes.
- Nodes == items, edges connect items
- **Both** nodes and edges may have **attributes**

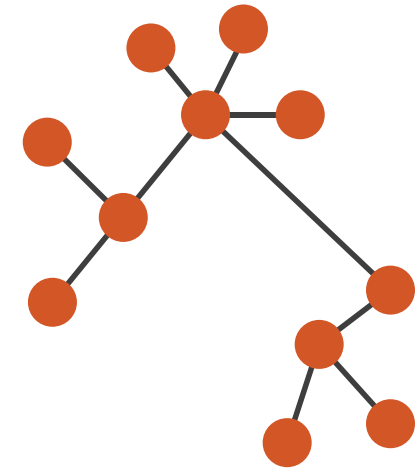


# Arrange Networks and Trees

## → Node–Link Diagrams Connection Marks

✓ NETWORKS

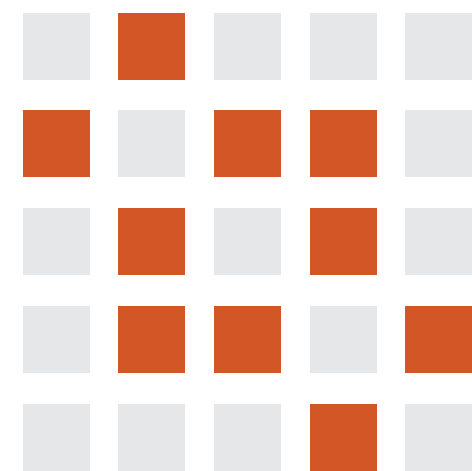
✓ TREES



## → Adjacency Matrix Derived Table

✓ NETWORKS

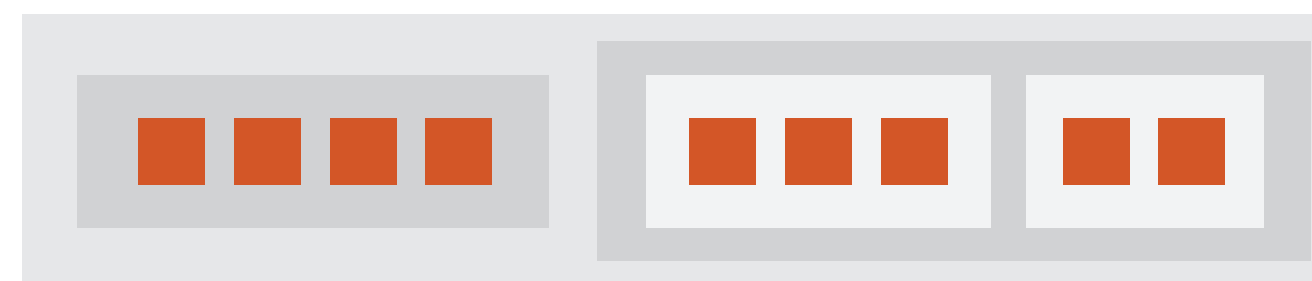
✓ TREES



## → Enclosure Containment Marks

✗ NETWORKS

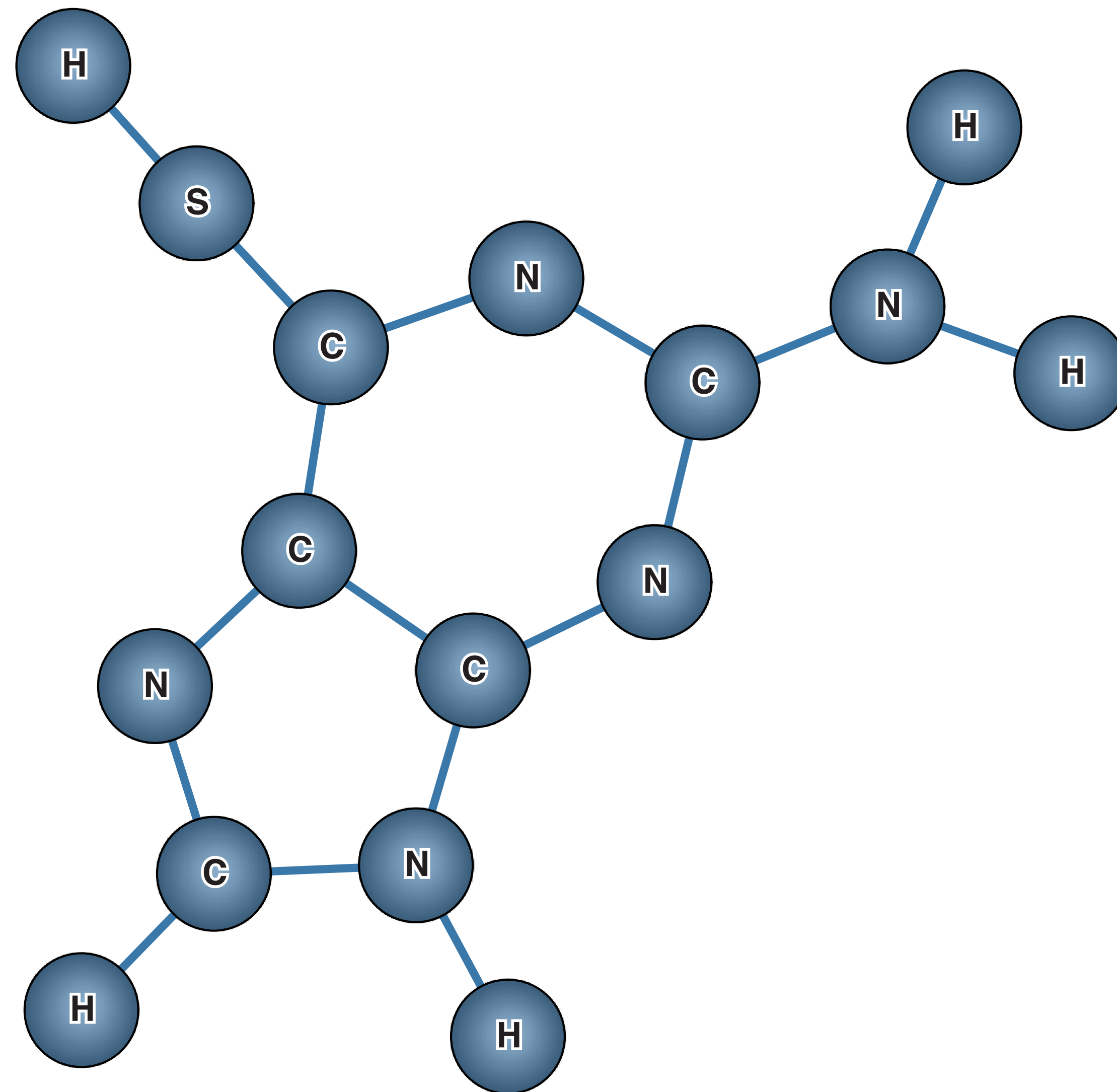
✓ TREES



[Munzner (ill. Maguire), 2014]

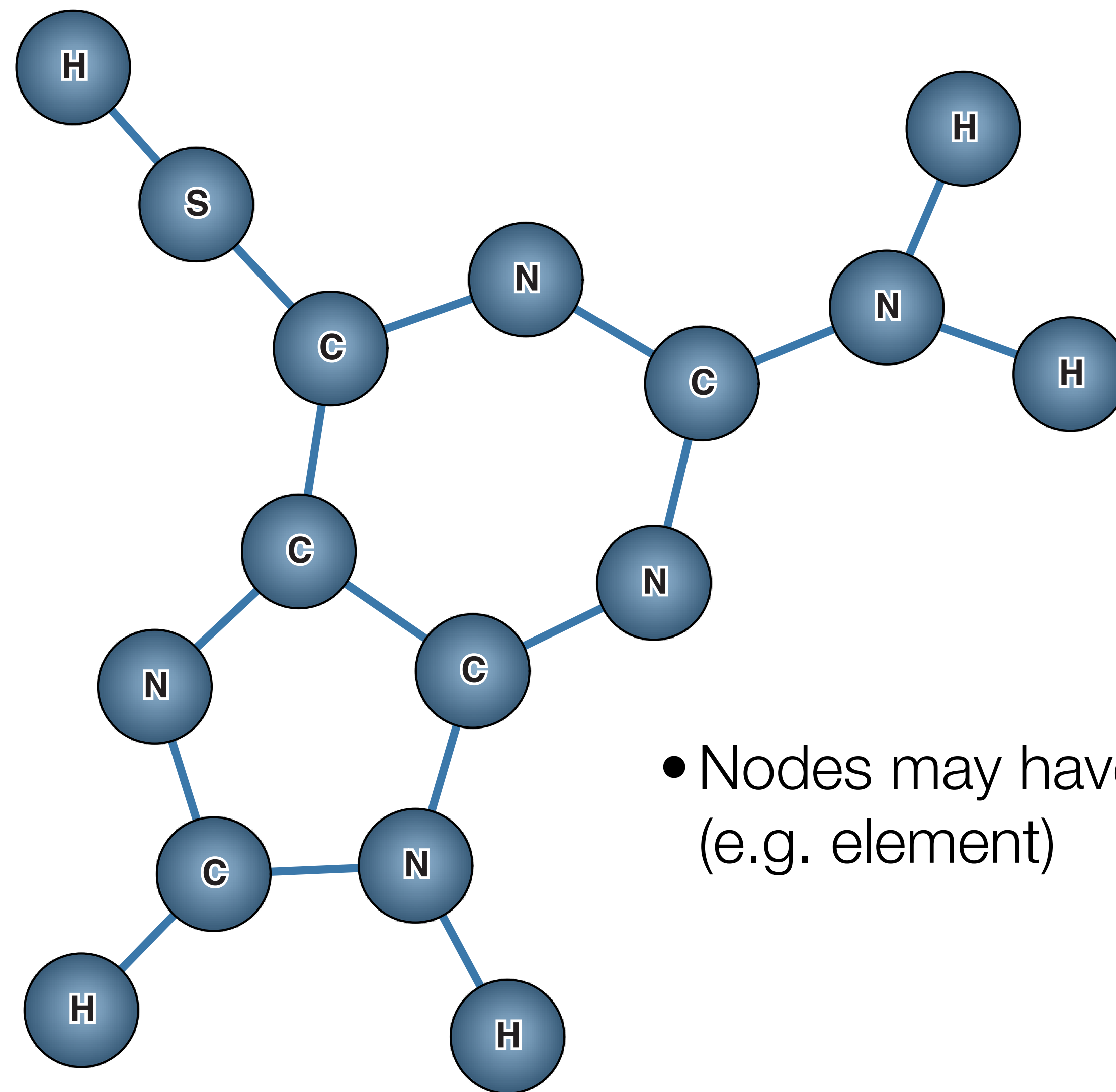
# Molecule Graph

---





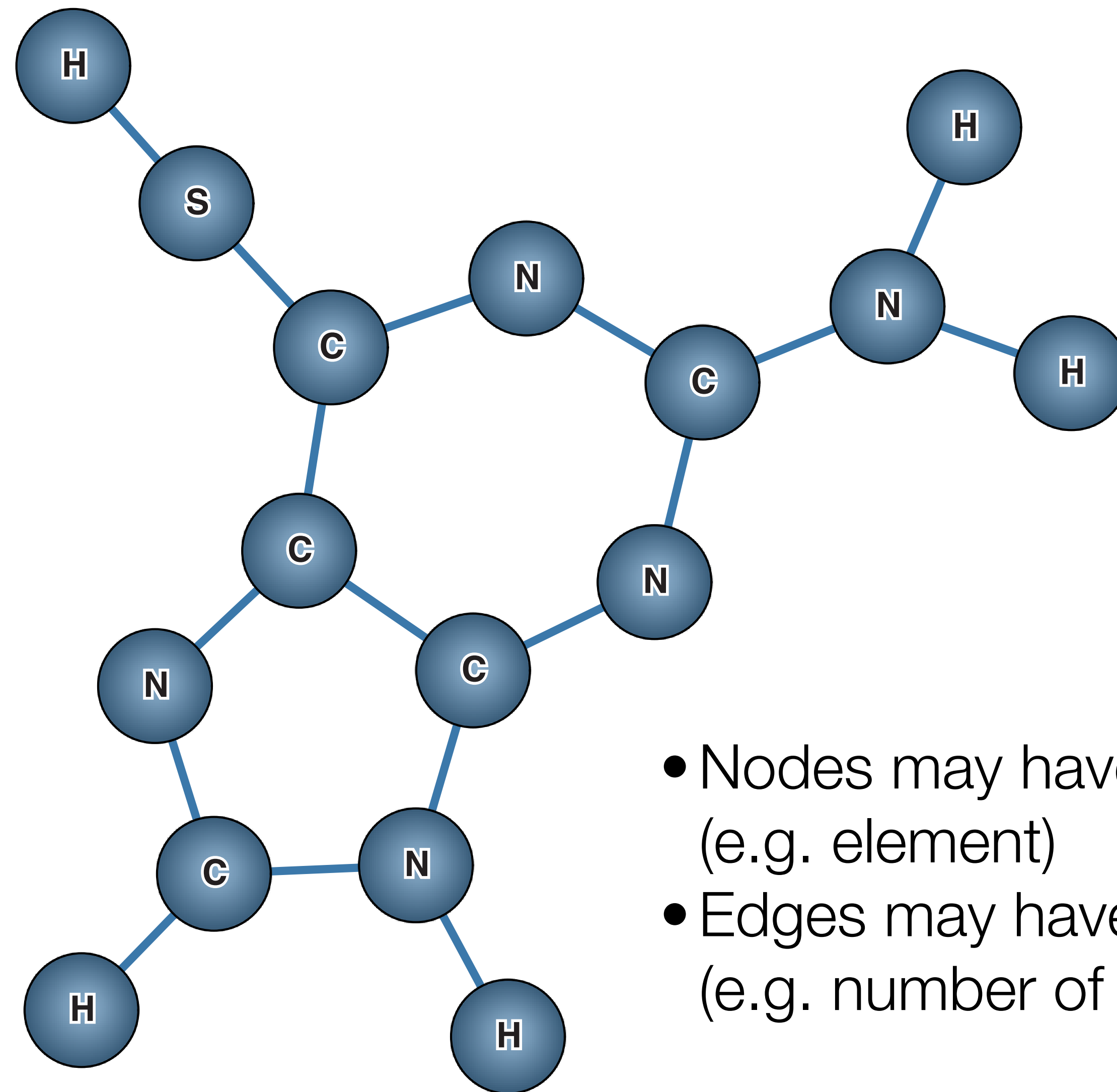
# Molecule Graph



- Nodes may have attributes (e.g. element)



# Molecule Graph

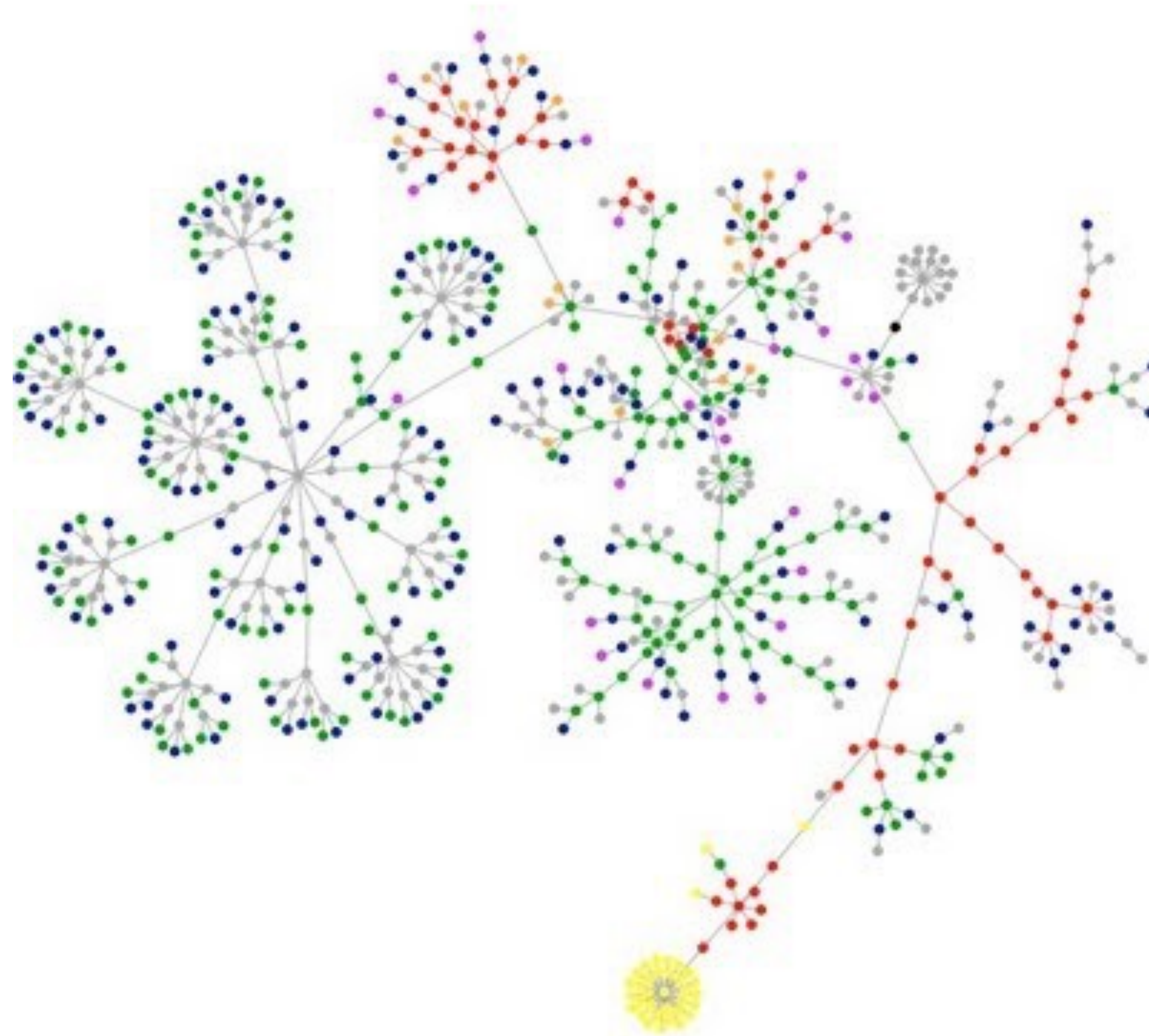


- Nodes may have attributes (e.g. element)
- Edges may have attributes (e.g. number of bonds)



# Web Sites as Graphs (amazon.com)

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[M. Salathe, 2006]



# Social Networks





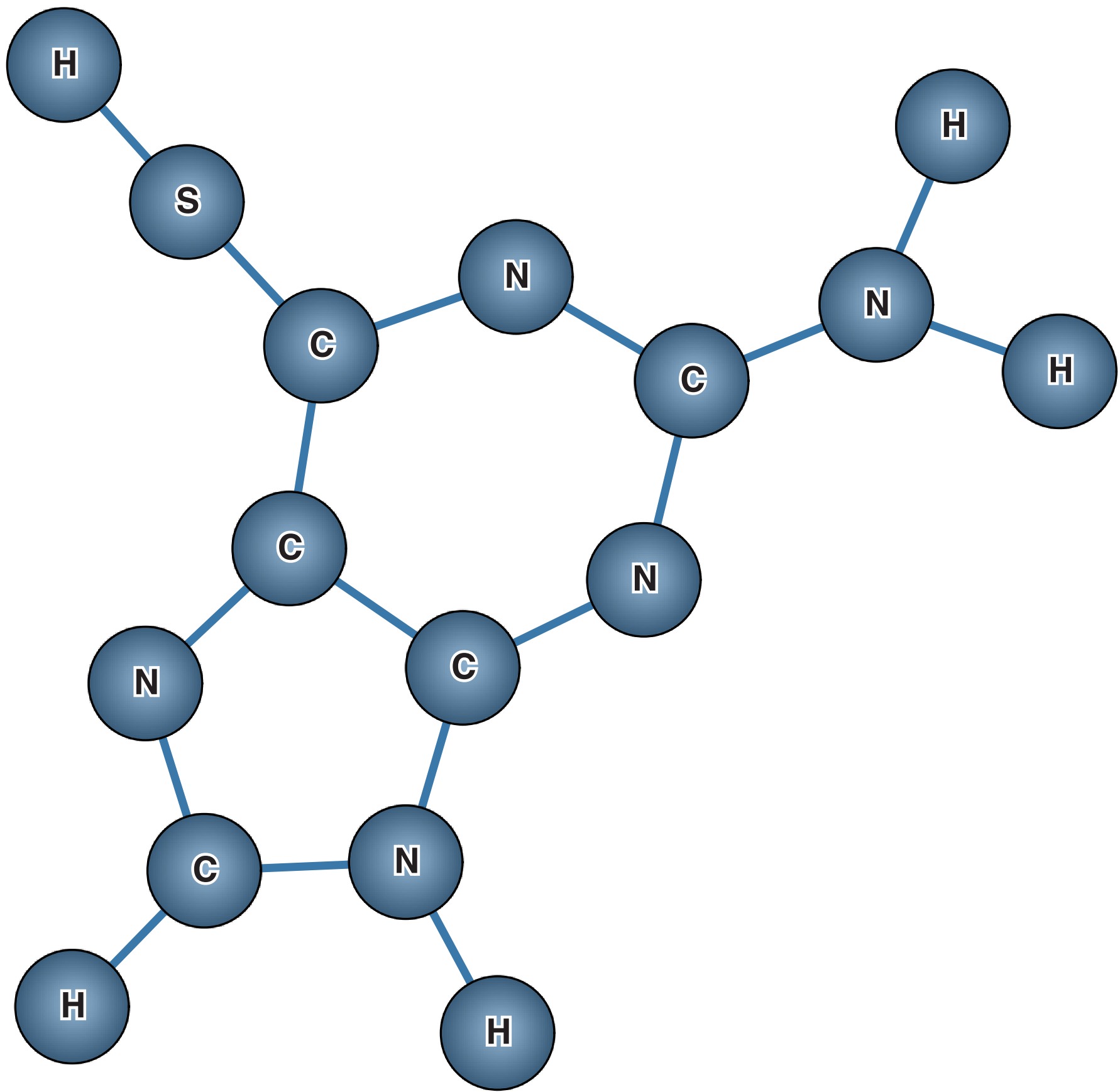
# Networks as Data

Nodes

ID	Atom	Electrons	Protons
0	N	7	7
1	C	6	6
2	S	16	16
3	C	6	6
4	N	7	7

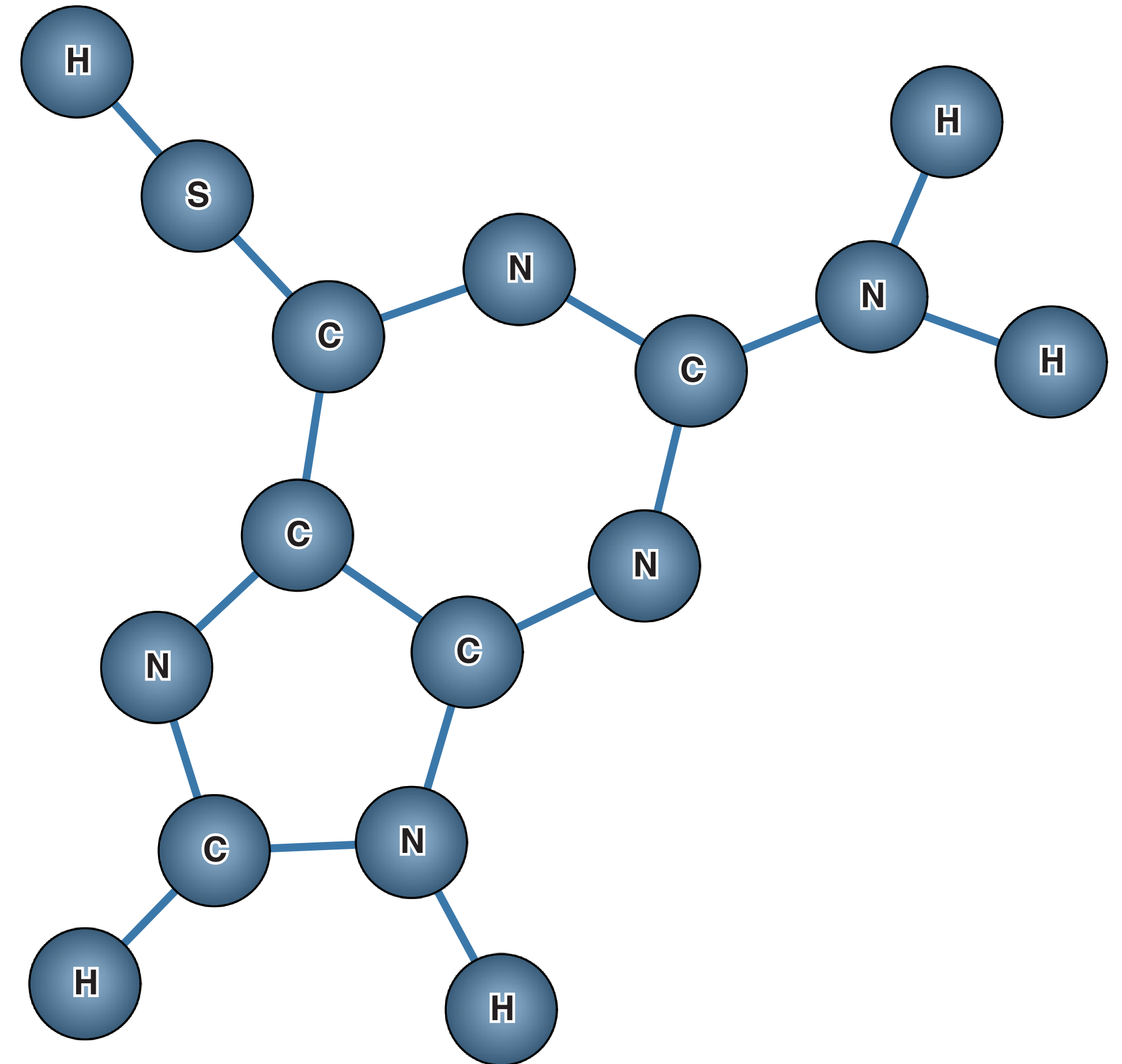
Edges

ID1	ID2	Bonds
0	1	1
1	2	1
1	3	2
3	4	1



# Node-Link Diagrams

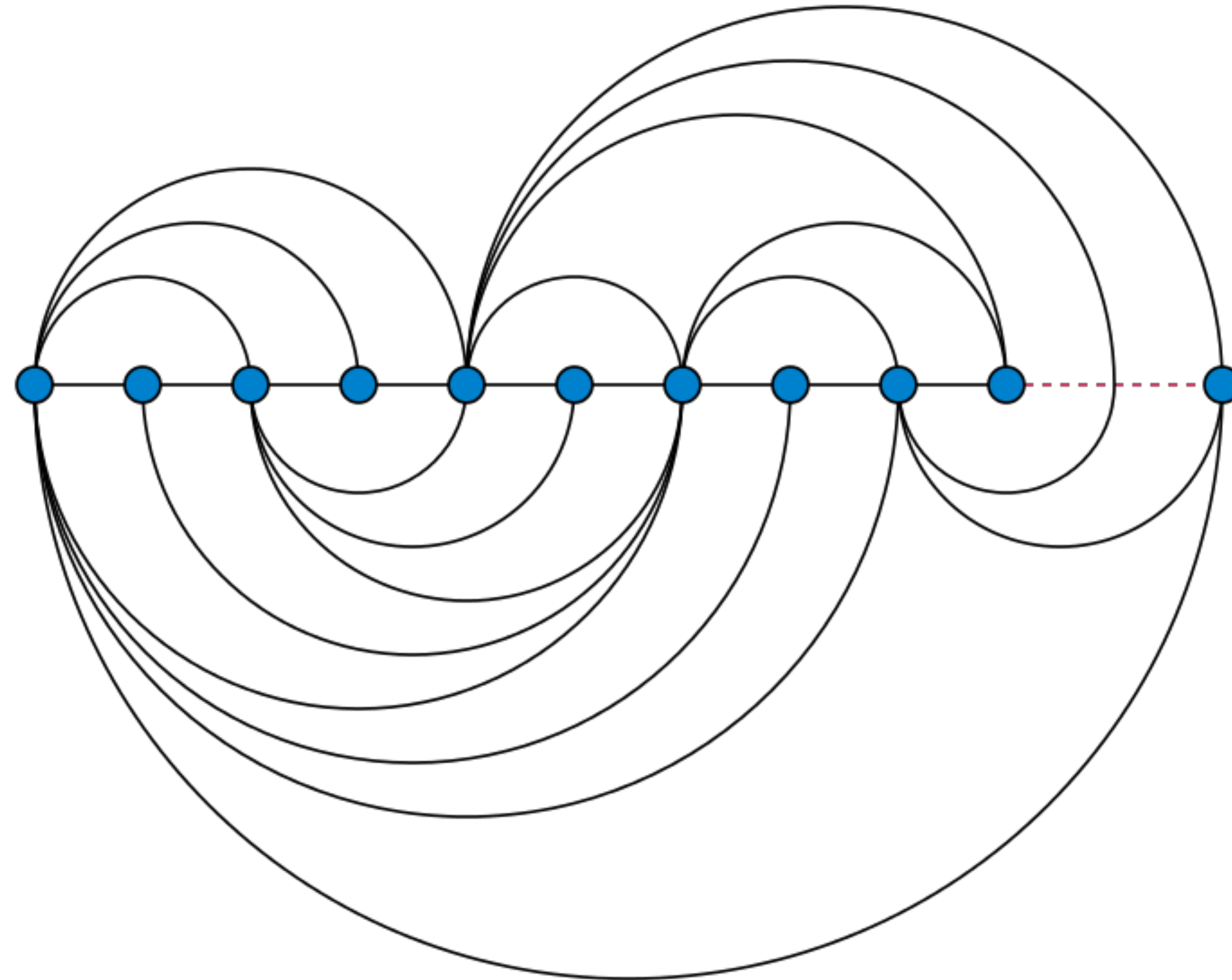
- Data: nodes and edges
- Task: understand connectivity, paths, structure (topology)
- Encoding: nodes as point marks, connections as line marks
- Scalability: hundreds
- ...but high **density** of links can be problematic!
- Issue with the encoding?





# Arc Diagram

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[D. Eppstein, 2013]

# Network Layout

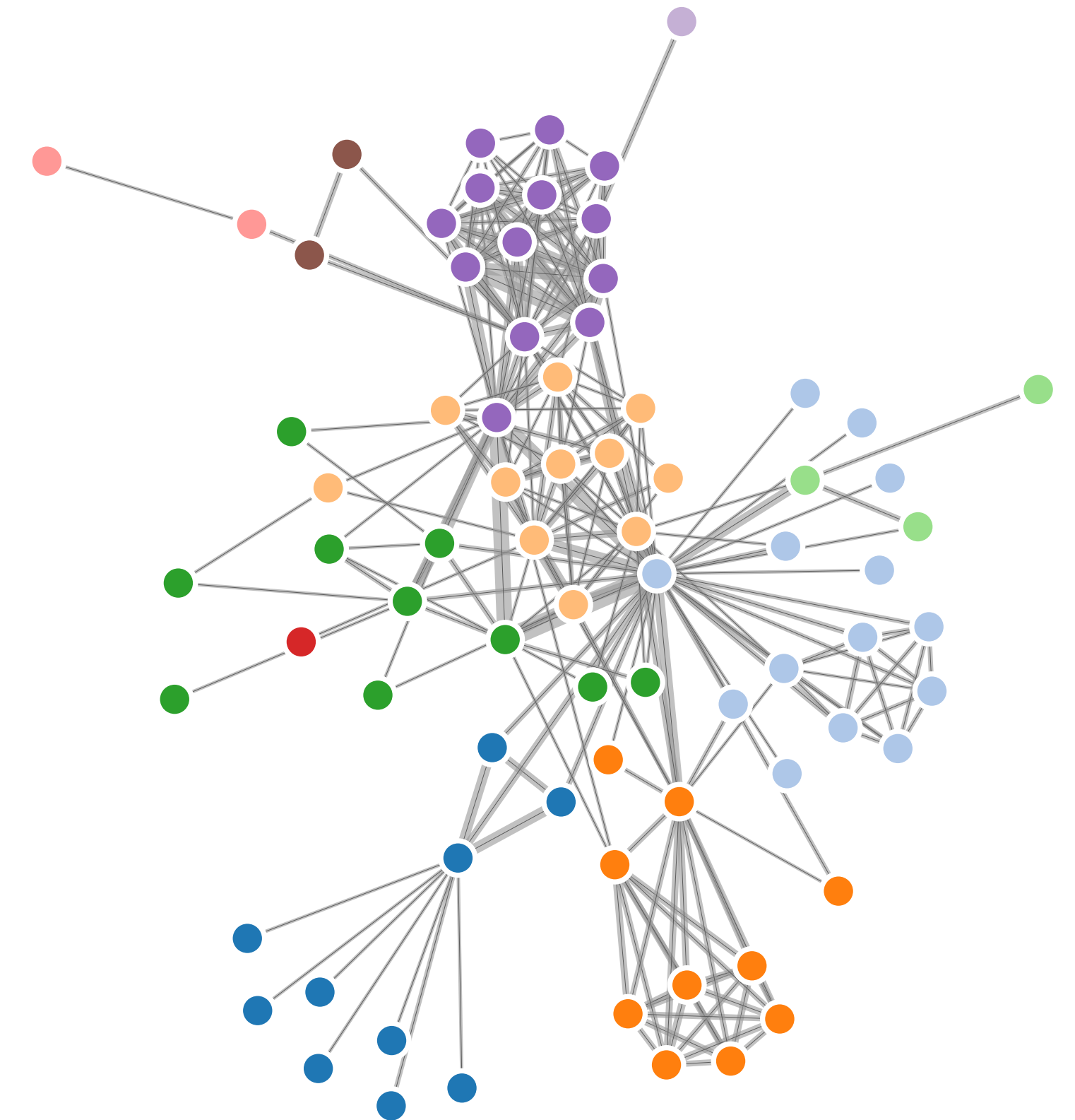
---

- Need to use spatial position when designing network visualizations
- Otherwise, nodes can **occlude** each other, links hard to distinguish
- How?
  - With bar charts, we could order using an attribute...
  - With networks, we want to be able to see connectivity and topology (not in the data usually)
- Possible metrics:
  - Edge crossings
  - Node overlaps
  - Total area



# Force-Directed Layout

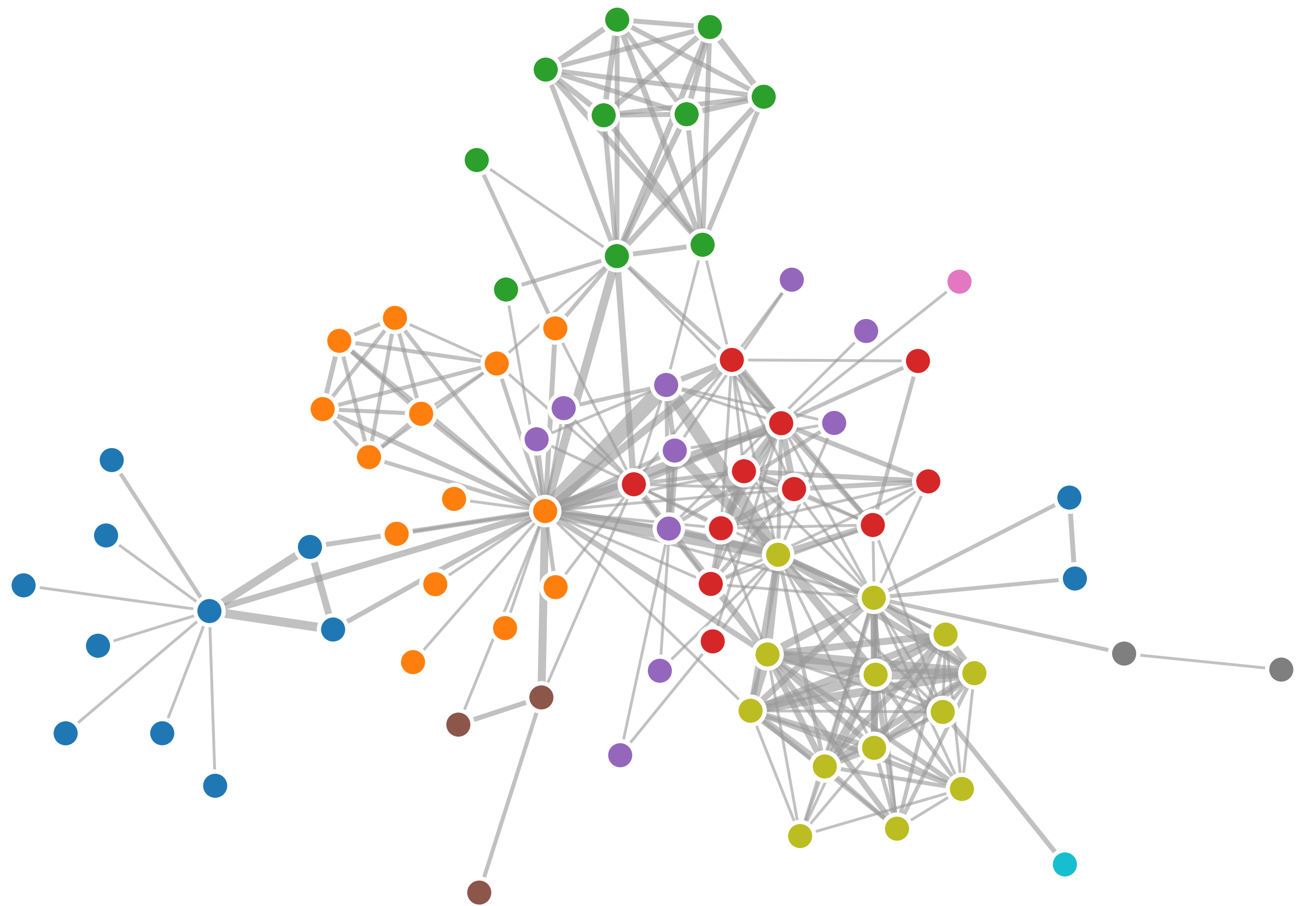
- Nodes push away from each other but edges are springs that pull them together
- Weakness: nondeterminism, algorithm may produce different results each time it runs



[M. Bostock, 2017]

# Constraint-Based Optimization (CoLa)

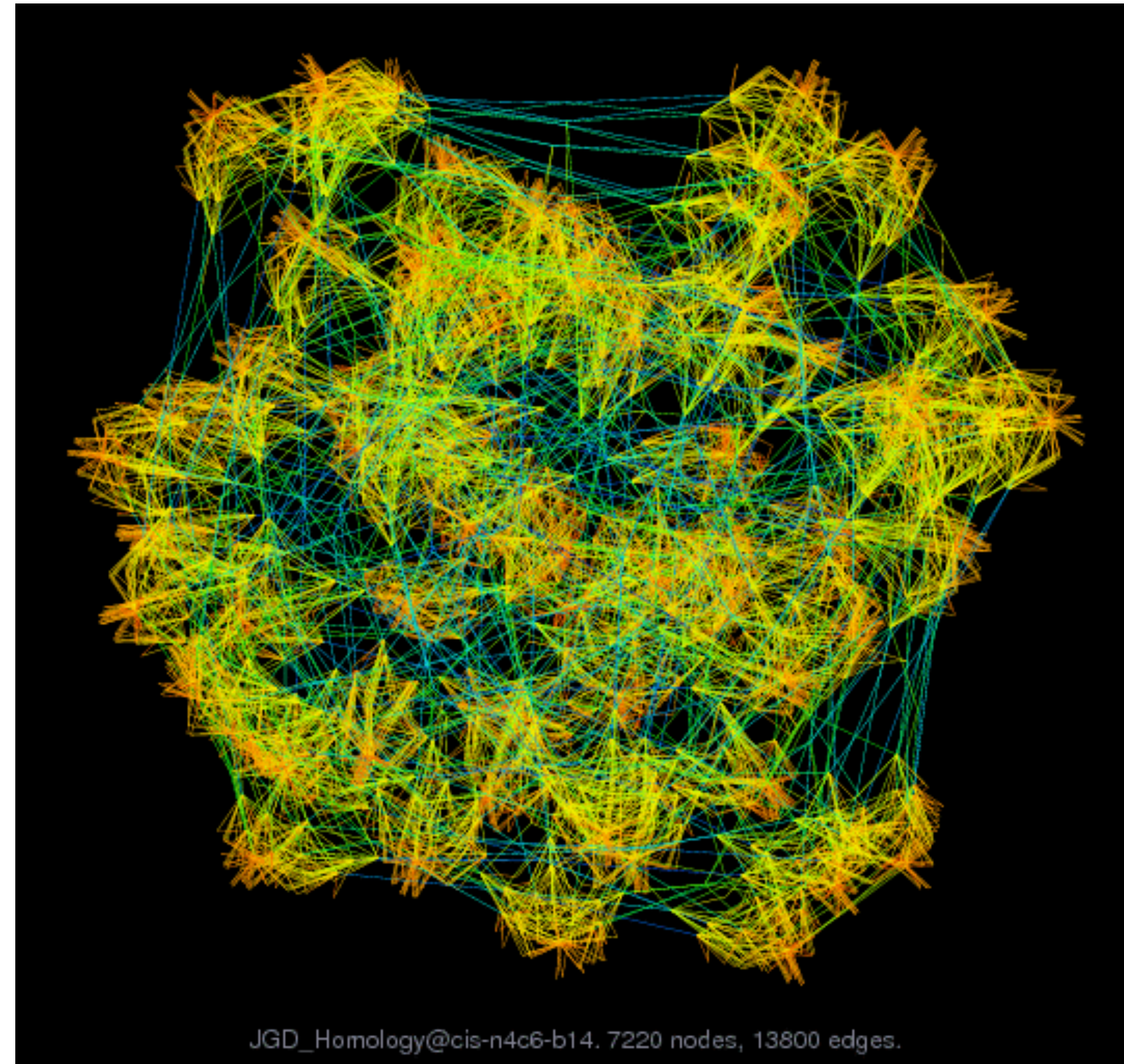
- Higher quality layout
- More **stable** in interactive applications (no "jitter")
- Allows user specified constraints such as alignments and grouping
- Can avoid overlapping nodes
- Provides flow layout for directed graphs
- May be **less scalable** to very large graphs
- Can route edges around nodes



[T. Dwyer et al. (WebCoLa); M. Bostock (Example), 2018]



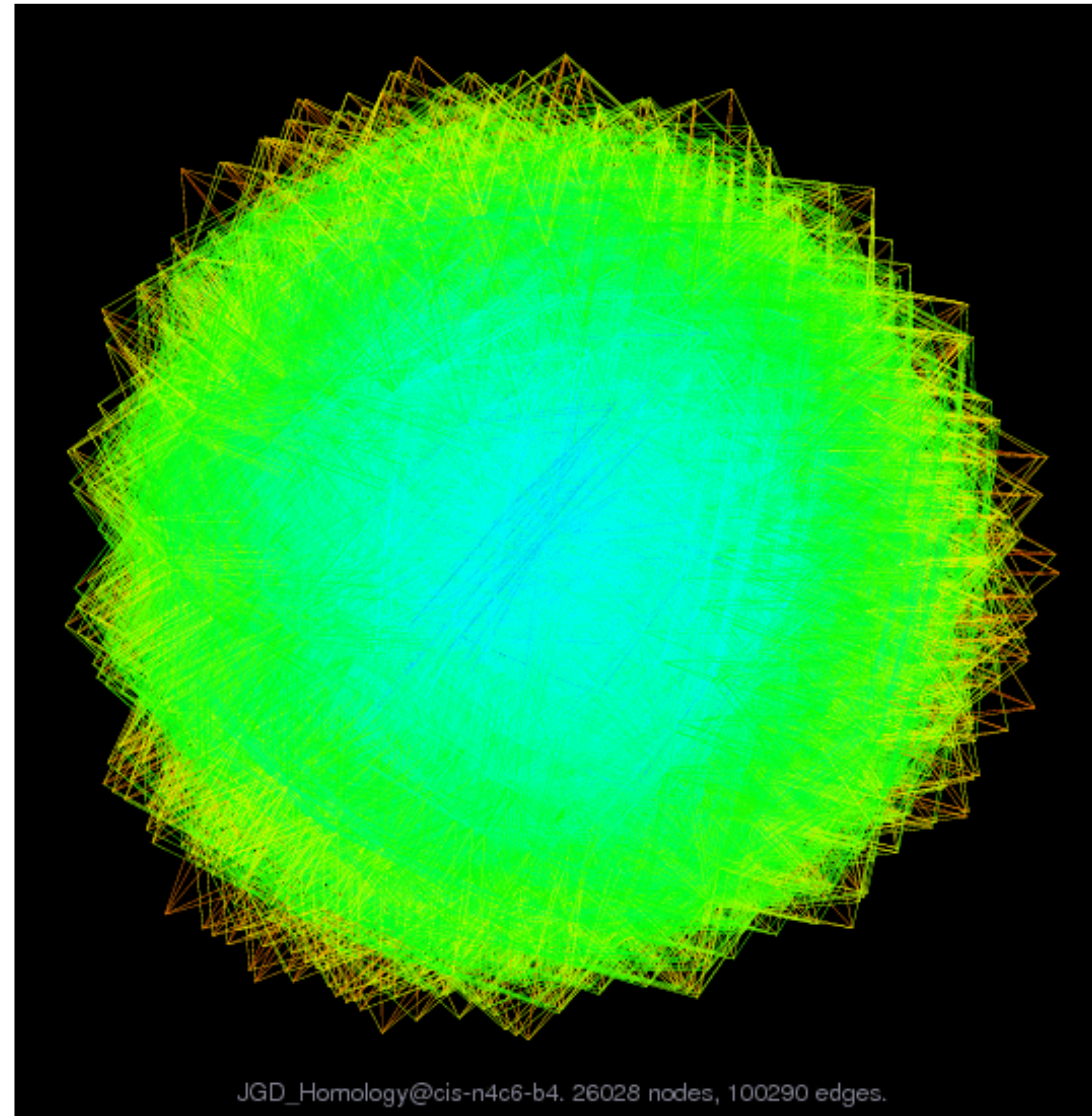
# sfdp



[Hu, 2005]



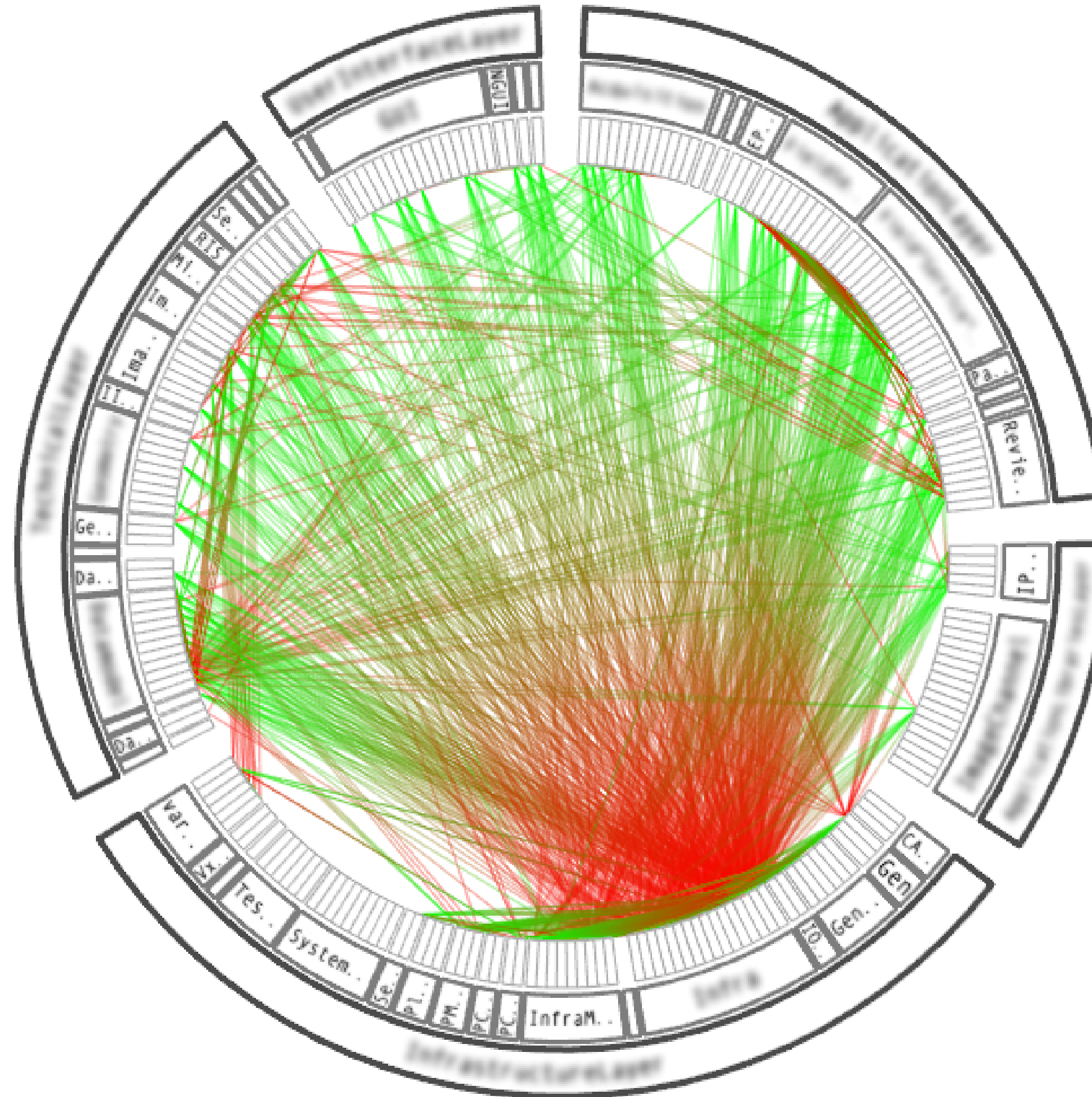
# “Hairball”



[Hu, 2014]

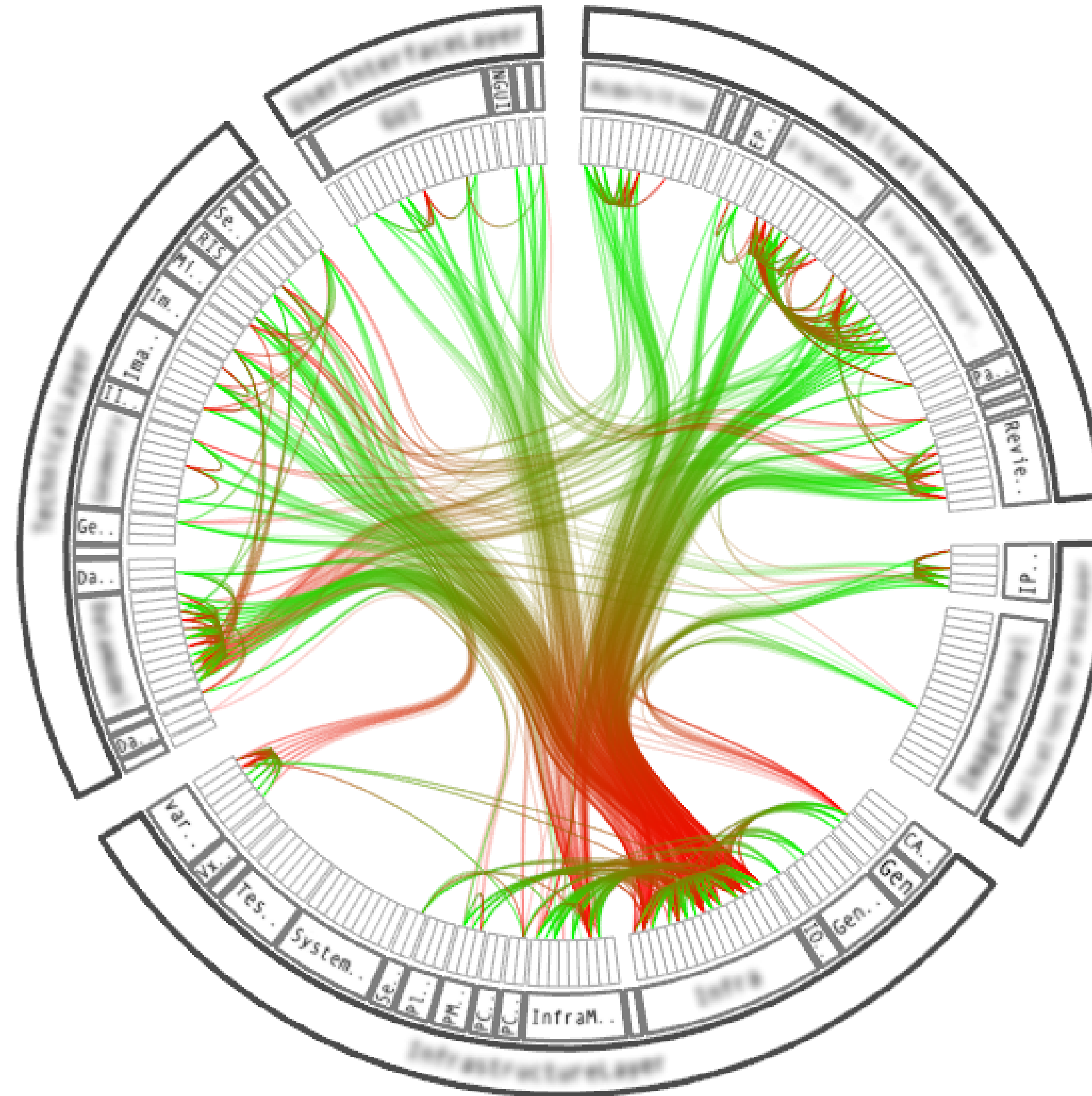


# Hierarchical Edge Bundling



[Holten, 2006]

# Hierarchical Edge Bundling



[Holten, 2006]



# Hierarchical Edge Bundling

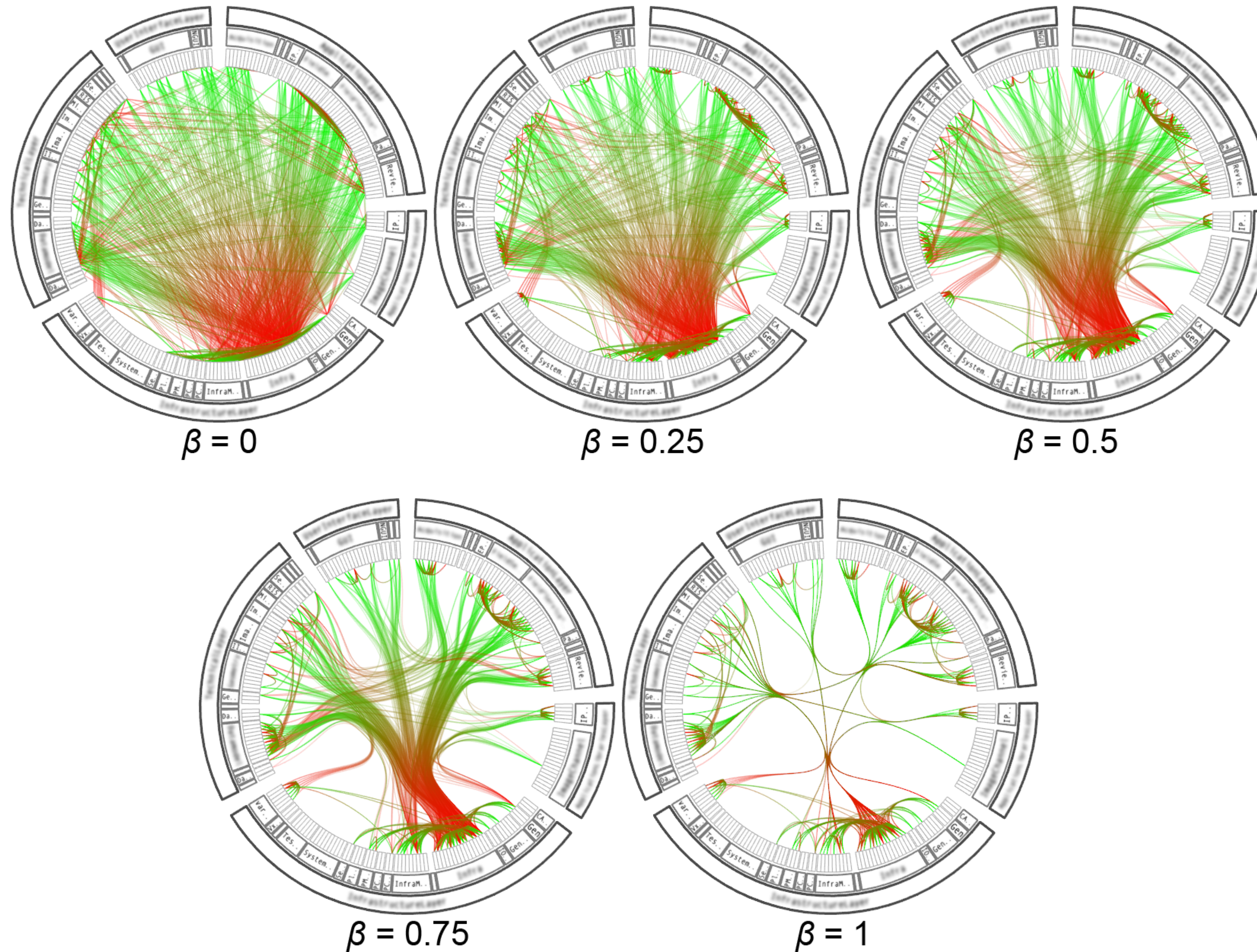
---

- Flexible and generic method
- Reduces visual clutter when dealing with large numbers of adjacency edges
- Provides an intuitive and continuous way to control the strength of bundling.
  - Low bundling strength mainly provides low-level, node-to-node connectivity information
  - High bundling strength provides high-level information as well by implicit visualization of adjacency edges between parent nodes that are the result of explicit adjacency edges between their respective child nodes

[Holten, 2006]



# Bundling Strength



[Holten, 2006]