

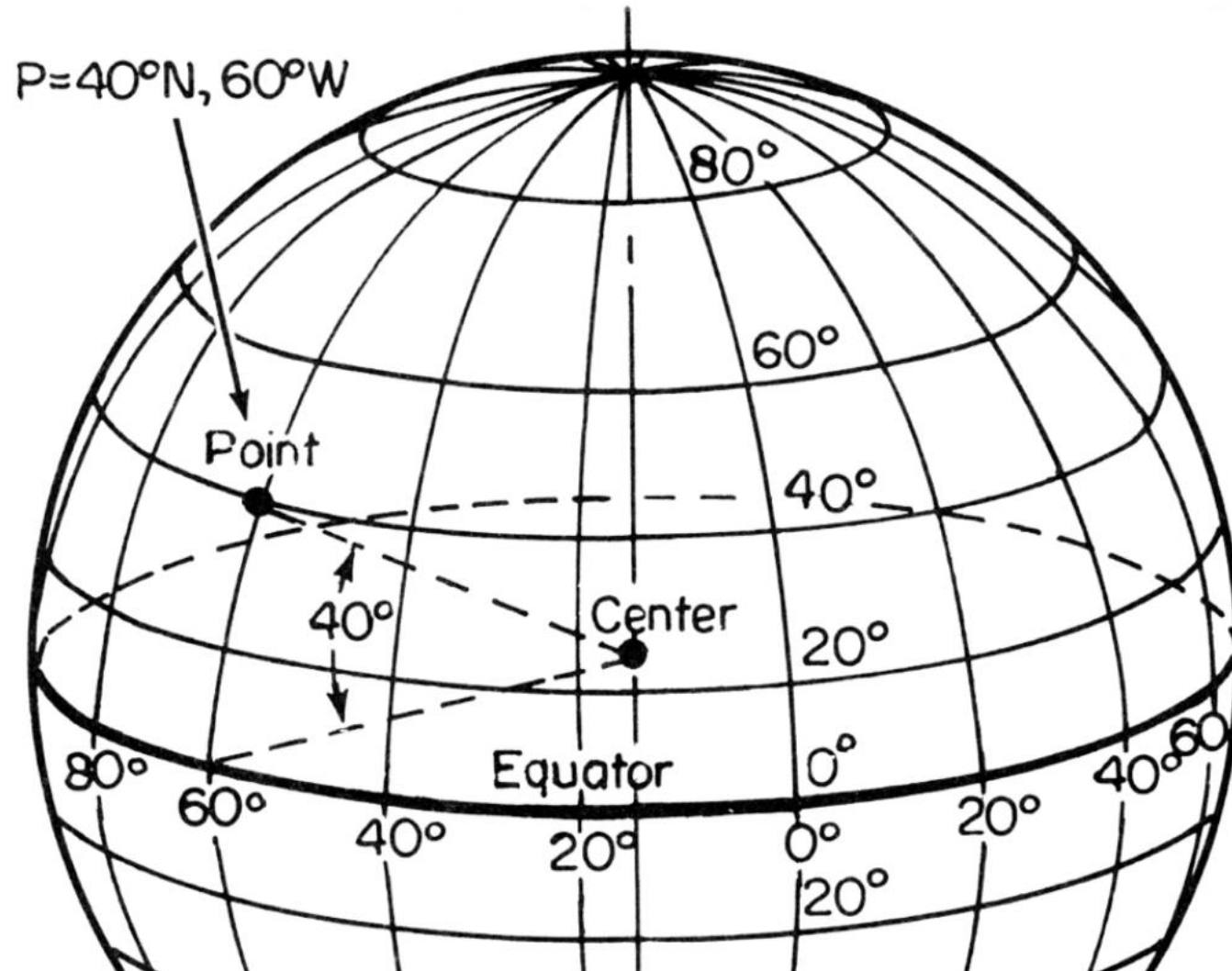
# Geospatial visualization



# Considerations for geospatial visualization

- Projection
- Context
- Distortion
- Interpolation

# Latitude and longitude



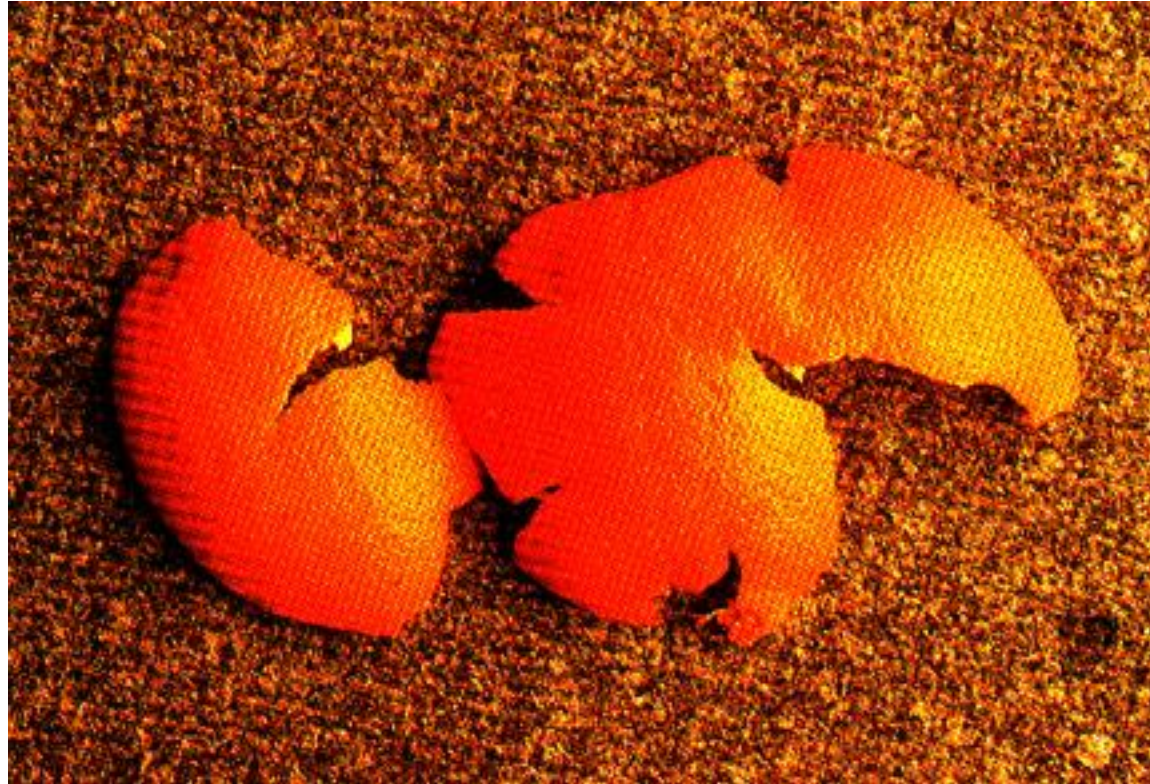
# Orthographic projection

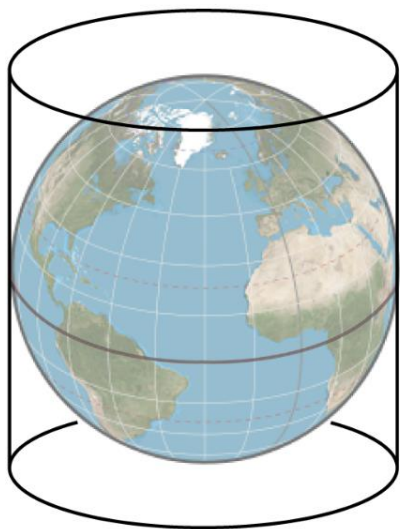


$$x = R \cos \varphi \sin(\lambda - \lambda_0)$$

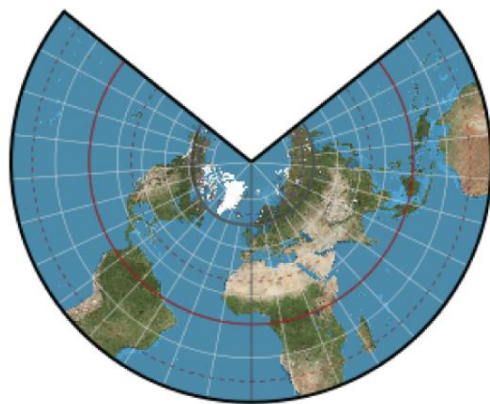
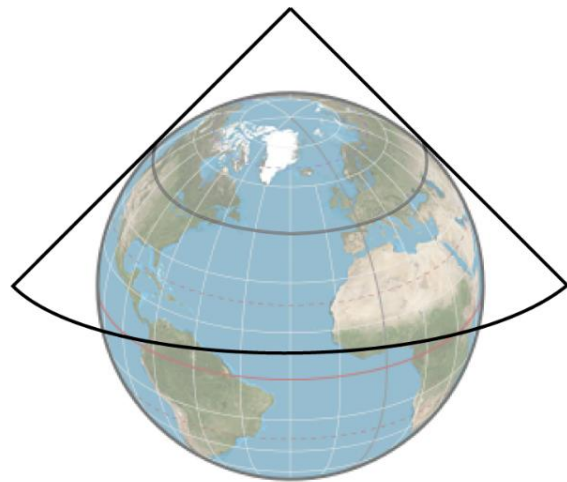
$$y = R(\cos \varphi_0 \sin \varphi - \sin \varphi_0 \cos \varphi \cos(\lambda - \lambda_0))$$

How do we display in 2-d?

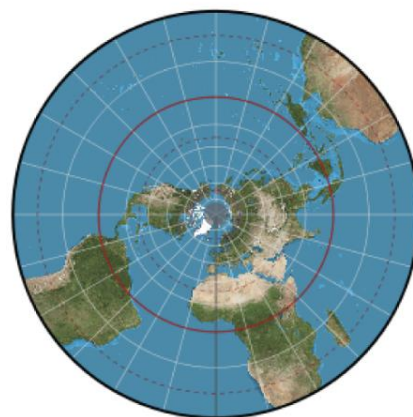
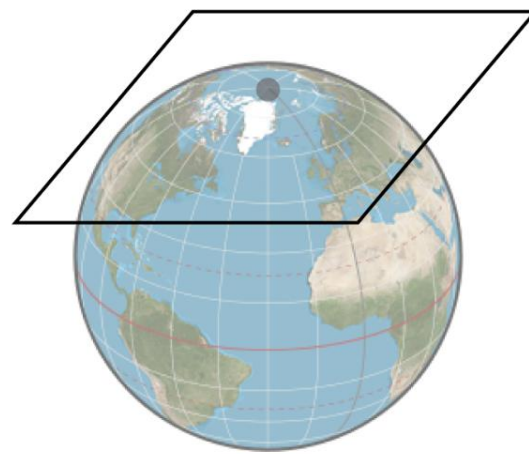




Cylindrical

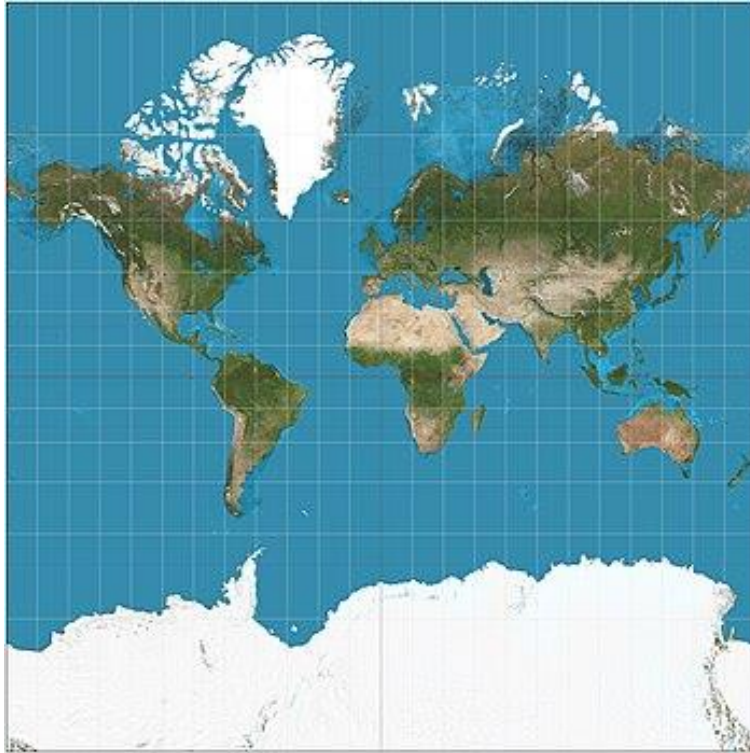
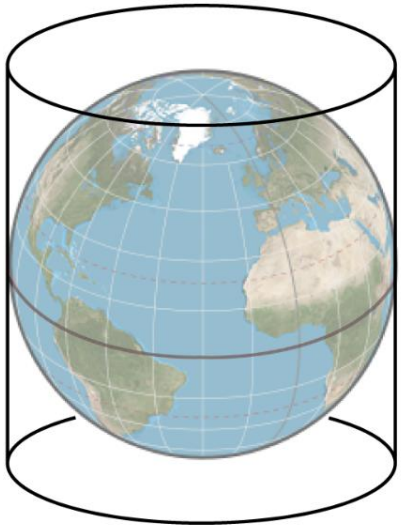


Conical



Azimuthal

# Mercator projection

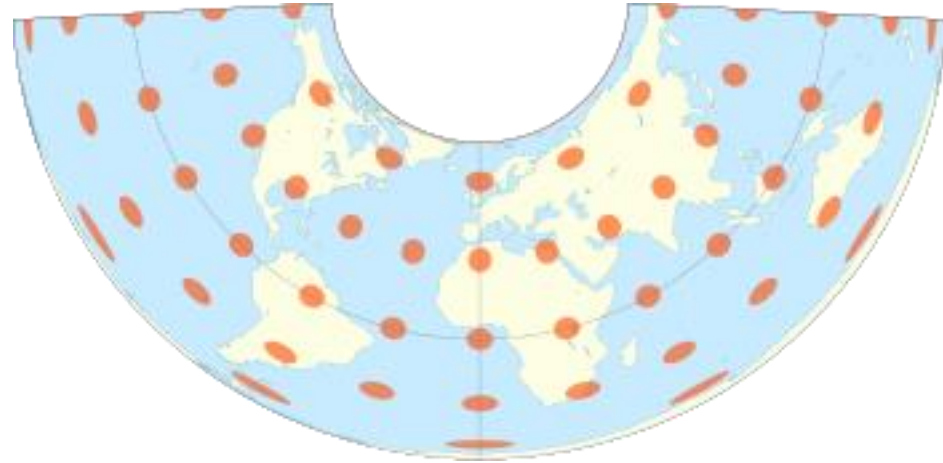
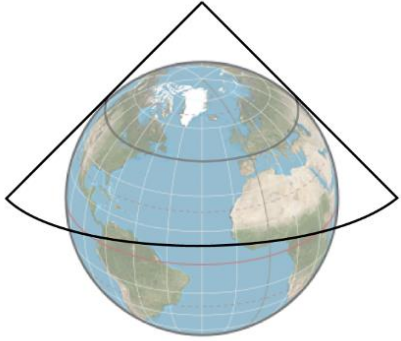


$$x = R(\lambda - \lambda_0), \quad y = R \ln \left[ \tan \left( \frac{\pi}{4} + \frac{\varphi}{2} \right) \right]$$

# Mercator projection – Preserves direction



# Albers Projection



$$x = \rho \sin \theta$$

$$y = \rho_0 - \rho \cos \theta$$

where

$$n = \frac{1}{2} (\sin \varphi_1 + \sin \varphi_2)$$

$$\theta = n (\lambda - \lambda_0)$$

$$C = \cos^2 \varphi_1 + 2n \sin \varphi_1$$

$$\rho = \frac{R}{n} \sqrt{C - 2n \sin \varphi}$$

$$\rho_0 = \frac{R}{n} \sqrt{C - 2n \sin \varphi_0}$$

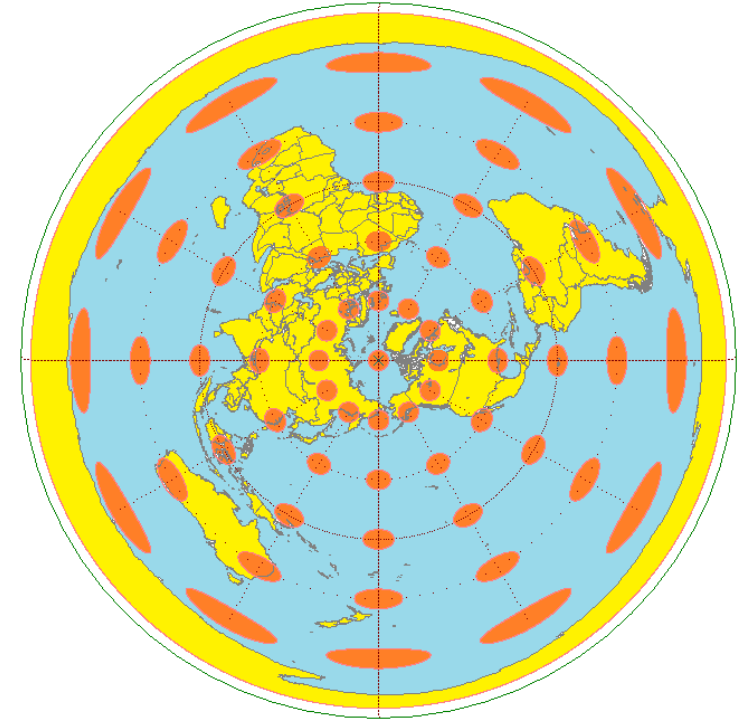
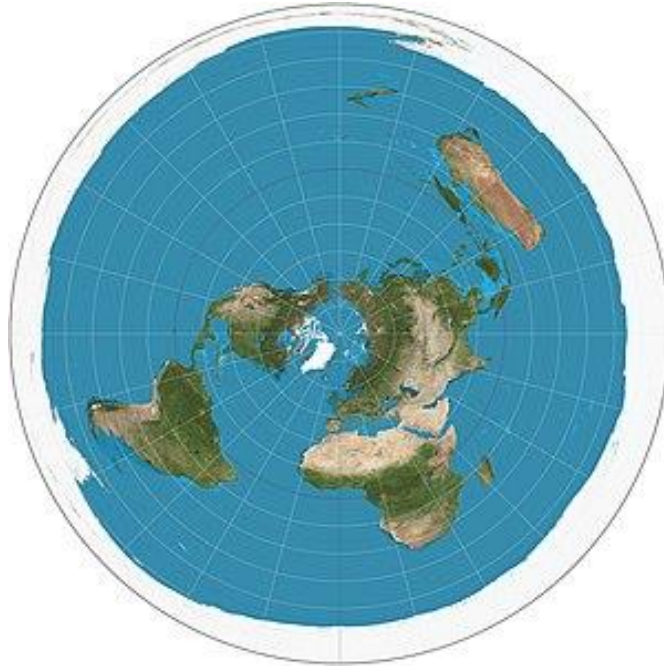
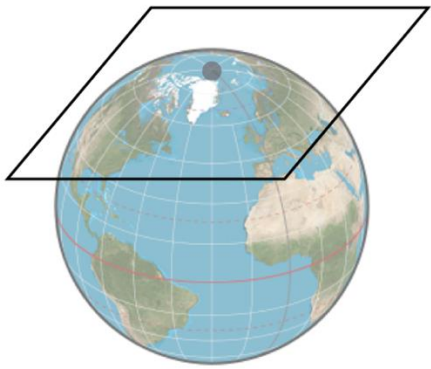
# Albers Projection – preserves area



# Albers USA



# Azimuthal equidistant

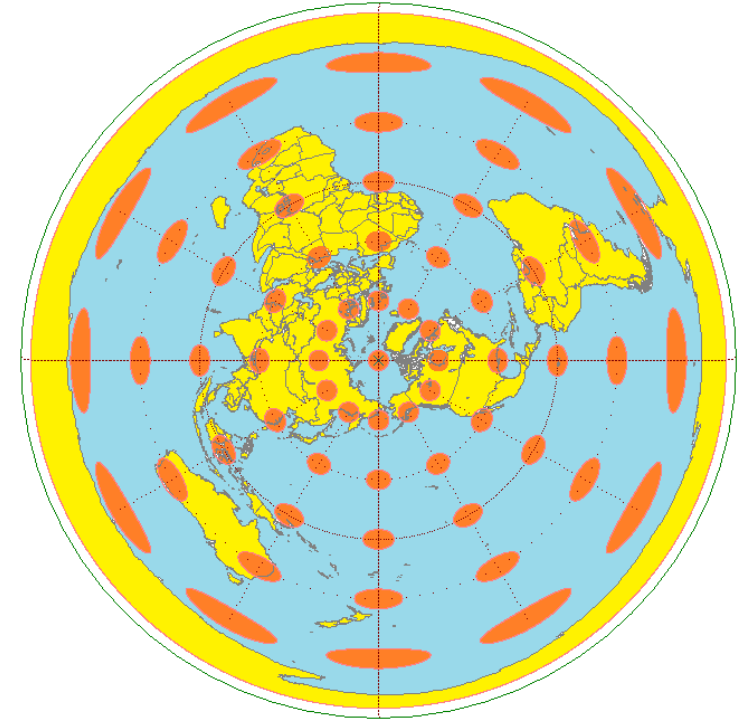
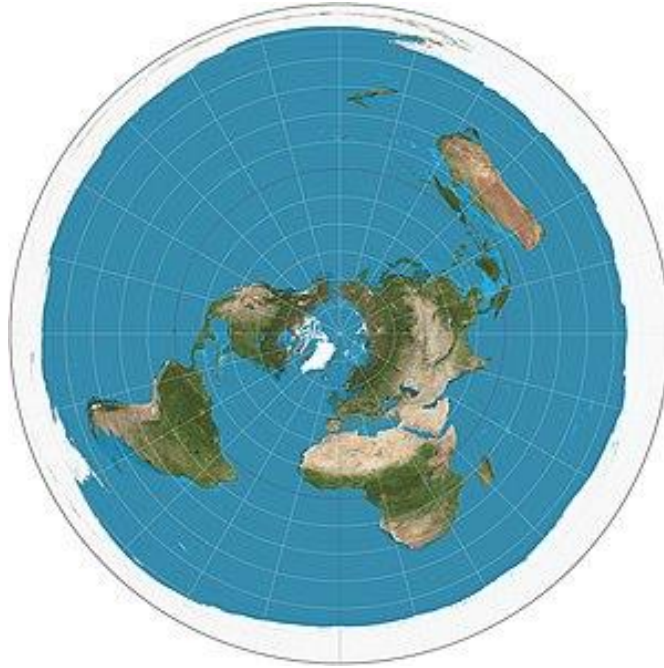
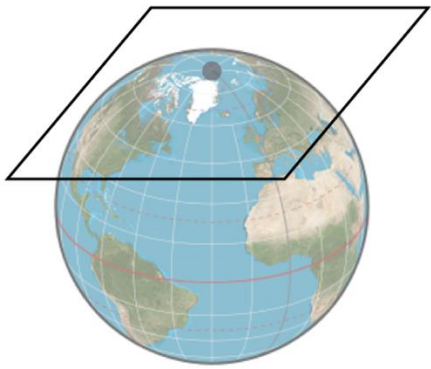


$$x = \rho \sin \theta, \quad y = -\rho \cos \theta$$

$$\cos \frac{\rho}{R} = \sin \varphi_0 \sin \varphi + \cos \varphi_0 \cos \varphi \cos(\lambda - \lambda_0)$$

$$\tan \theta = \frac{\cos \varphi \sin(\lambda - \lambda_0)}{\cos \varphi_0 \sin \varphi - \sin \varphi_0 \cos \varphi \cos(\lambda - \lambda_0)}$$

# Azimuthal equidistant



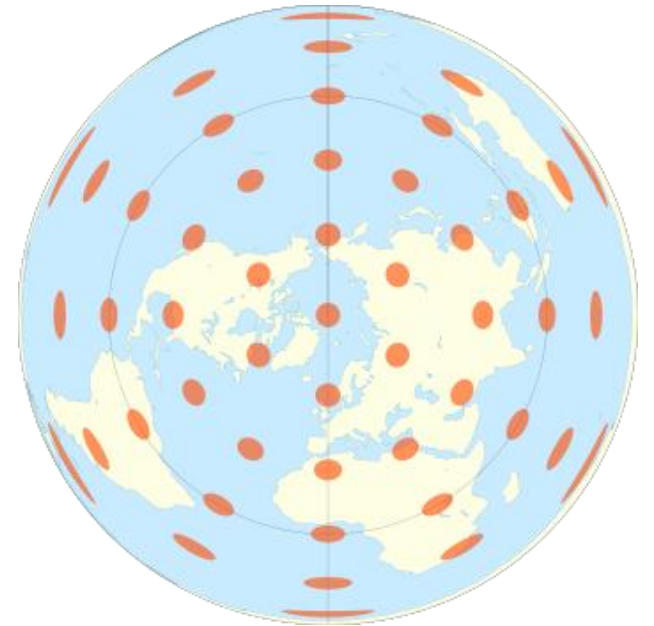
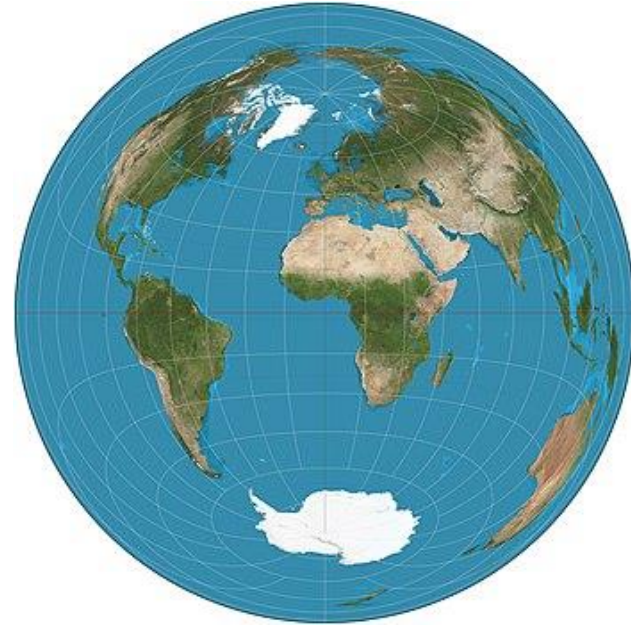
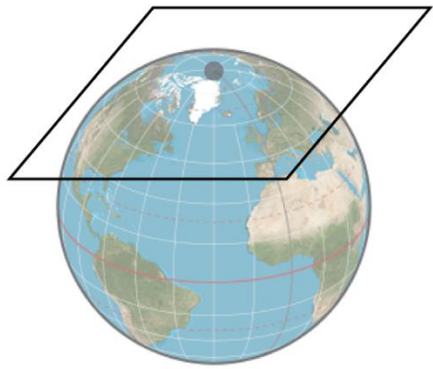
$$x = \rho \sin \theta, \quad y = -\rho \cos \theta$$

$$\cos \frac{\rho}{R} = \sin \varphi_0 \sin \varphi + \cos \varphi_0 \cos \varphi \cos(\lambda - \lambda_0)$$

$$\tan \theta = \frac{\cos \varphi \sin(\lambda - \lambda_0)}{\cos \varphi_0 \sin \varphi - \sin \varphi_0 \cos \varphi \cos(\lambda - \lambda_0)}$$

Other equal area

# Azimuthal equal area





Lambert cylindrical

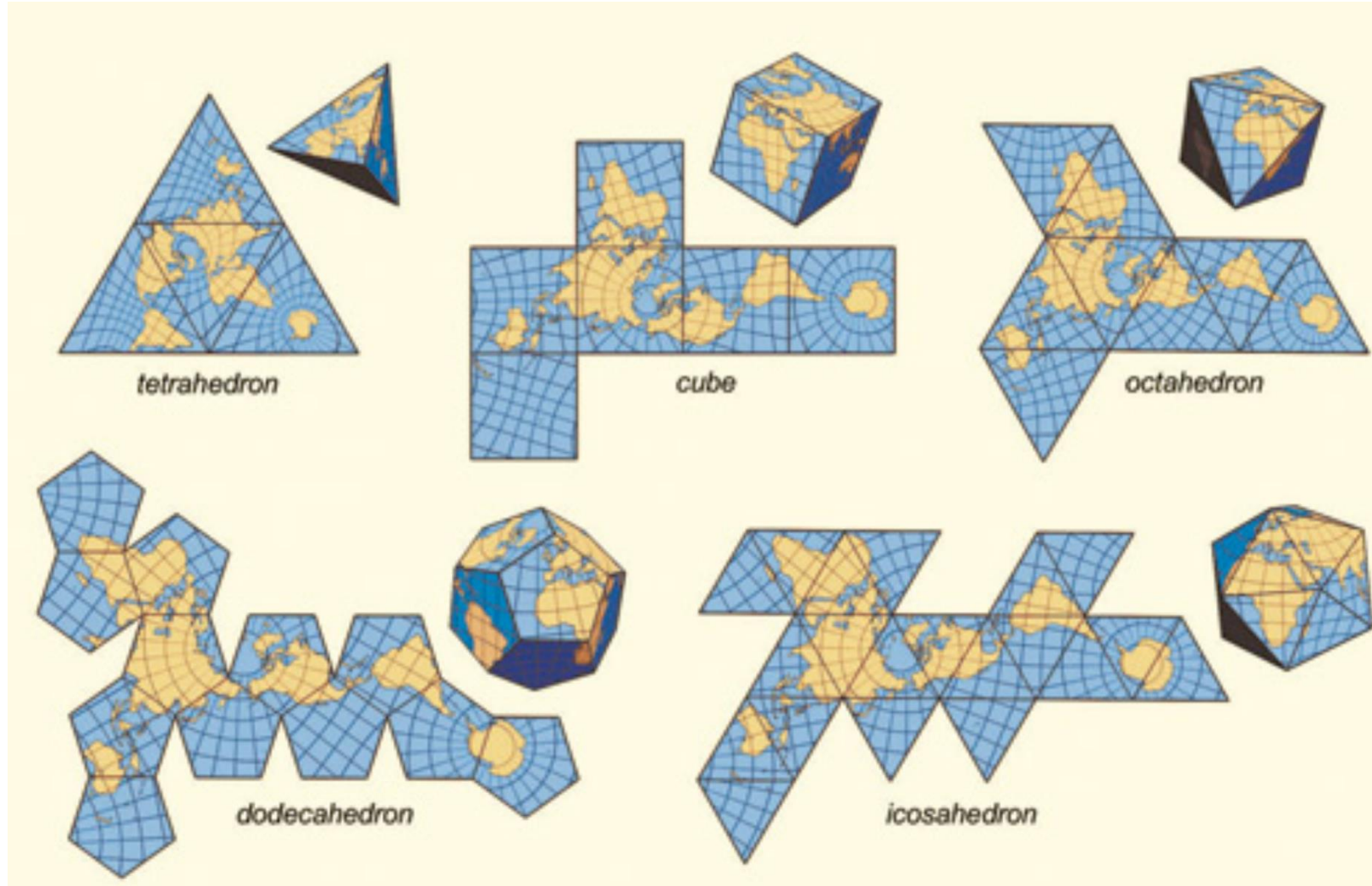


Equal Earth



Bottomley

# Unfolding a sphere



# Other projections

# Scatterplot maps

# GeoJSON

- Format for encoding geographic shapes
  - Shapes encoded as latitude and longitude coordinates

```
{
  "type": "FeatureCollection",
  "features": [
    {
      "type": "Feature",
      "geometry": {
        "type": "Point",
        "coordinates": [102.0, 0.5]
      },
      "properties": {
        "prop0": "value0"
      }
    },
    {
      "type": "Feature",
      "geometry": {
        "type": "LineString",
        "coordinates": [
          [102.0, 0.0],
          [103.0, 1.0],
          [104.0, 0.0],
          [105.0, 1.0]
        ]
      },
      "properties": {
        "prop0": "value0",
        "prop1": 0.0
      }
    },
    {
      "type": "Feature",
      "geometry": {
        "type": "Polygon",
        "coordinates": [
          [
            [100.0, 0.0],
            [101.0, 0.0],
            [101.0, 1.0],
            [100.0, 1.0],
            [100.0, 0.0]
          ]
        ]
      },
      "properties": {
        "prop0": "value0",
        "prop1": { "this": "that" }
      }
    }
  ]
}
```

# GeoJSON

```
Object {  
  type: "FeatureCollection"  
  totalFeatures: 214  
  features: ▼Array(214) [  
    0: ▶ Object {type: "Feature", id: "nyu_2451_34509.1", geometry: Object, geometry_name: "geom", properties:  
    1: ▶ Object {type: "Feature", id: "nyu_2451_34509.2", geometry: Object, geometry_name: "geom", properties:  
    2: ▶ Object {type: "Feature", id: "nyu_2451_34509.3", geometry: Object, geometry_name: "geom", properties:  
    3: ▶ Object {type: "Feature", id: "nyu_2451_34509.7", geometry: Object, geometry_name: "geom", properties:  
    4: ▶ Object {type: "Feature", id: "nyu_2451_34509.8", geometry: Object, geometry_name: "geom", properties:  
    5: ▶ Object {type: "Feature", id: "nyu_2451_34509.34", geometry: Object, geometry_name: "geom", properties:  
    6: ▶ Object {type: "Feature", id: "nyu_2451_34509.4", geometry: Object, geometry_name: "geom", properties:  
    7: ▶ Object {type: "Feature", id: "nyu_2451_34509.5", geometry: Object, geometry_name: "geom", properties:  
    8: ▶ Object {type: "Feature", id: "nyu_2451_34509.6", geometry: Object, geometry_name: "geom", properties:  
    9: ▶ Object {type: "Feature", id: "nyu_2451_34509.9", geometry: Object, geometry_name: "geom", properties:  
    10: ▶ Object {type: "Feature", id: "nyu_2451_34509.10", geometry: Object, geometry_name: "geom", properties:  
    11: ▶ Object {type: "Feature", id: "nyu_2451_34509.11", geometry: Object, geometry_name: "geom", properties:  
    12: ▶ Object {type: "Feature", id: "nyu_2451_34509.12", geometry: Object, geometry_name: "geom", properties:  
    13: ▶ Object {type: "Feature", id: "nyu_2451_34509.30", geometry: Object, geometry_name: "geom", properties:  
    14: ▶ Object {type: "Feature", id: "nyu_2451_34509.13", geometry: Object, geometry_name: "geom", properties:  
    15: ▶ Object {type: "Feature", id: "nyu_2451_34509.14", geometry: Object, geometry_name: "geom", properties:  
    16: ▶ Object {type: "Feature", id: "nyu_2451_34509.15", geometry: Object, geometry_name: "geom", properties:  
    17: ▶ Object {type: "Feature", id: "nyu_2451_34509.16", geometry: Object, geometry_name: "geom", properties:  
    18: ▶ Object {type: "Feature", id: "nyu_2451_34509.31", geometry: Object, geometry_name: "geom", properties:  
    19: ▶ Object {type: "Feature", id: "nyu_2451_34509.17", geometry: Object, geometry_name: "geom", properties:  
    ... more  
  ]  
  crs: ▶ Object {type: "name", properties: Object}  
  bbox: ▶ Array(4) [-74.258918762207, 40.4938278198242, -73.6479721069336, 40.9173812866211]
```

# GeoJSON

```
▼ Object {  
  type: "Feature"  
  id: "nyu_2451_34509.1"  
  geometry: ▼ Object {  
    type: "MultiPolygon"  
    coordinates: ▼ Array(1) [  
      0: ► Array(3) [Array(67), Array(6), Array(9)]  
    ]  
  }  
  geometry_name: "geom"  
  properties: ► Object {zcta: "10001", bcode: "36061", note: null, bbox: Array(4)}  
}
```

```
nycGeo.features[0]
```



```
projection = f(t)
```

```
projection = d3.geoAlbers()  
    .fitSize([mapWidth, mapHeight], nycGeo)
```

```
path = f(t)
```

```
path = d3.geoPath().projection(projection)
```

```
<svg width=${mapWidth} height=${mapHeight}>  
  <path  
    d=${path(nycGeo)}  
    fill="#d3d3d3"  
    stroke="white"  
  />  
</svg>
```



```
projection = f(t)
```

```
projection = d3.geoAlbers()  
    .fitSize([mapWidth, mapHeight], nycGeo)
```

```
path = f(t)
```

```
path = d3.geoPath().projection(projection)
```

```
<svg width=${mapWidth} height=${mapHeight}>  
  <path  
    d=${path(nycGeo.features[0])}  
    fill="#d3d3d3"  
    stroke="white"  
  />  
</svg>
```

```

{
  const svg = d3.create('svg')
    .attr('width', mapWidth)
    .attr('height', mapHeight);

  // draw one svg path per zip code
  svg.selectAll("path")
    .data(nycGeo.features)
    .join("path")
    .attr("d", path) // same as `d => path(d)`
    .attr("fill", '#d3d3d3')
    .attr("stroke", "white");

  return svg.node();
}

```



```

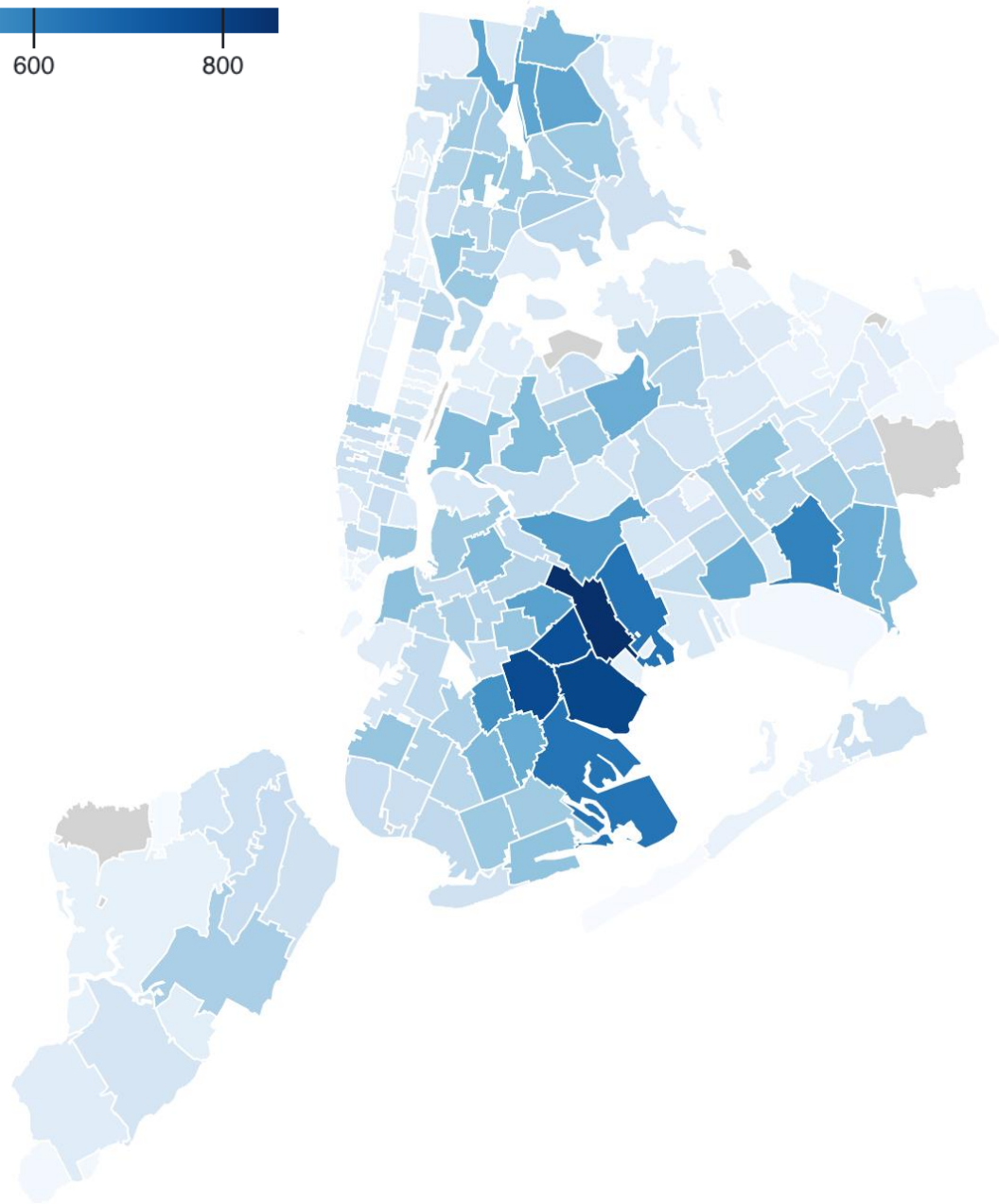
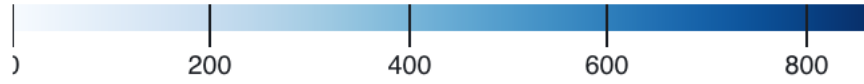
{
  const svg = d3.create('svg')
    .attr('width', mapWidth)
    .attr('height', mapHeight);

  // draw one svg path for the entire map
  svg.append("path")
    .attr("d", path(nycGeo))
    .attr("fill", '#d3d3d3')
    .attr("stroke", "white");

  return svg.node();
}

```

Vehicle collision injuries, 2019



```
▼ Object {
  10000: 32
  10001: 187
  10002: 323
  10003: 203
  10004: 37
  10005: 16
  10006: 26
  10007: 62
  10009: 114
  10010: 183
  10011: 151
  10012: 135
  10013: 209
  10014: 71
  10016: 298
  10017: 153
  10018: 127
  10019: 294
  10020: 12
  10021: 90
  ... more
}
```

zipToInjuries

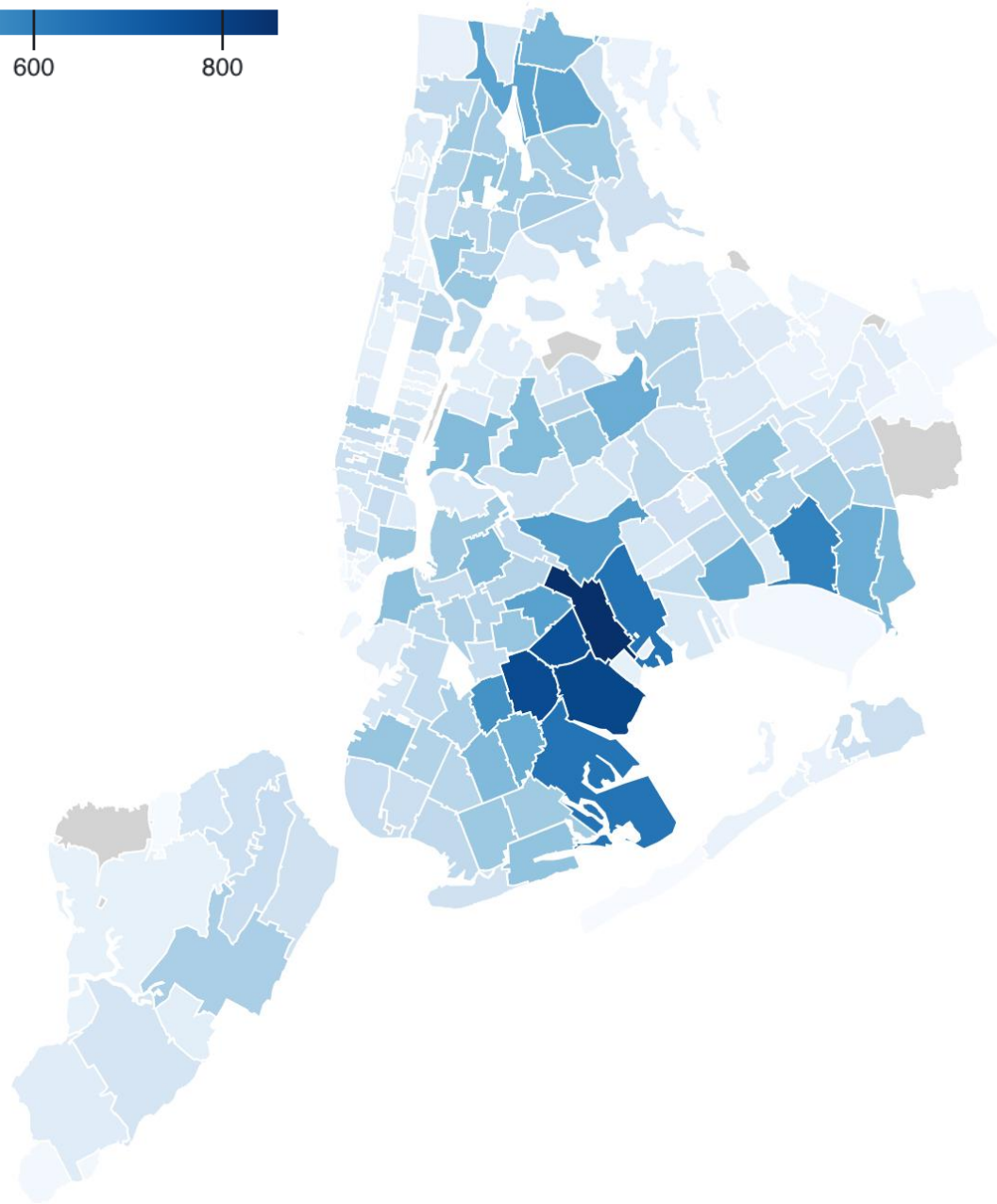
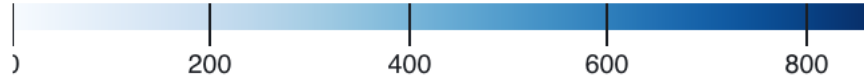
```
maxInjuriesForZip = 860
```

```
maxInjuriesForZip = d3.max(Object.values(zipToInjuries))
```

```
const color = d3.scaleSequential(d3.interpolateBlues);
```

```
color = d3.scaleSequential()
  .domain([0, maxInjuriesForZip])
  .interpolator(d3.interpolateBlues)
  .unknown('#d3d3d3')
```

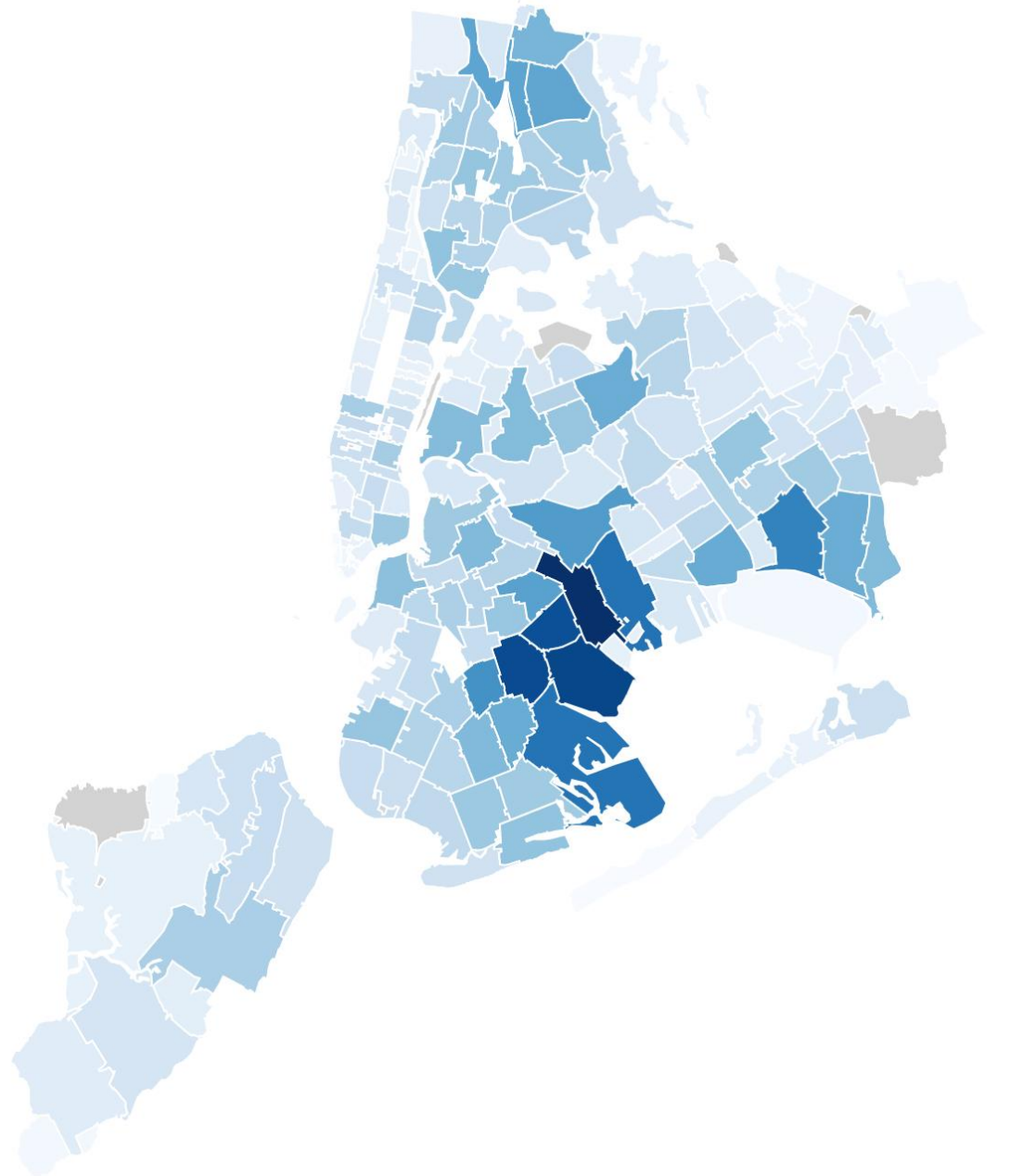
Vehicle collision injuries, 2019



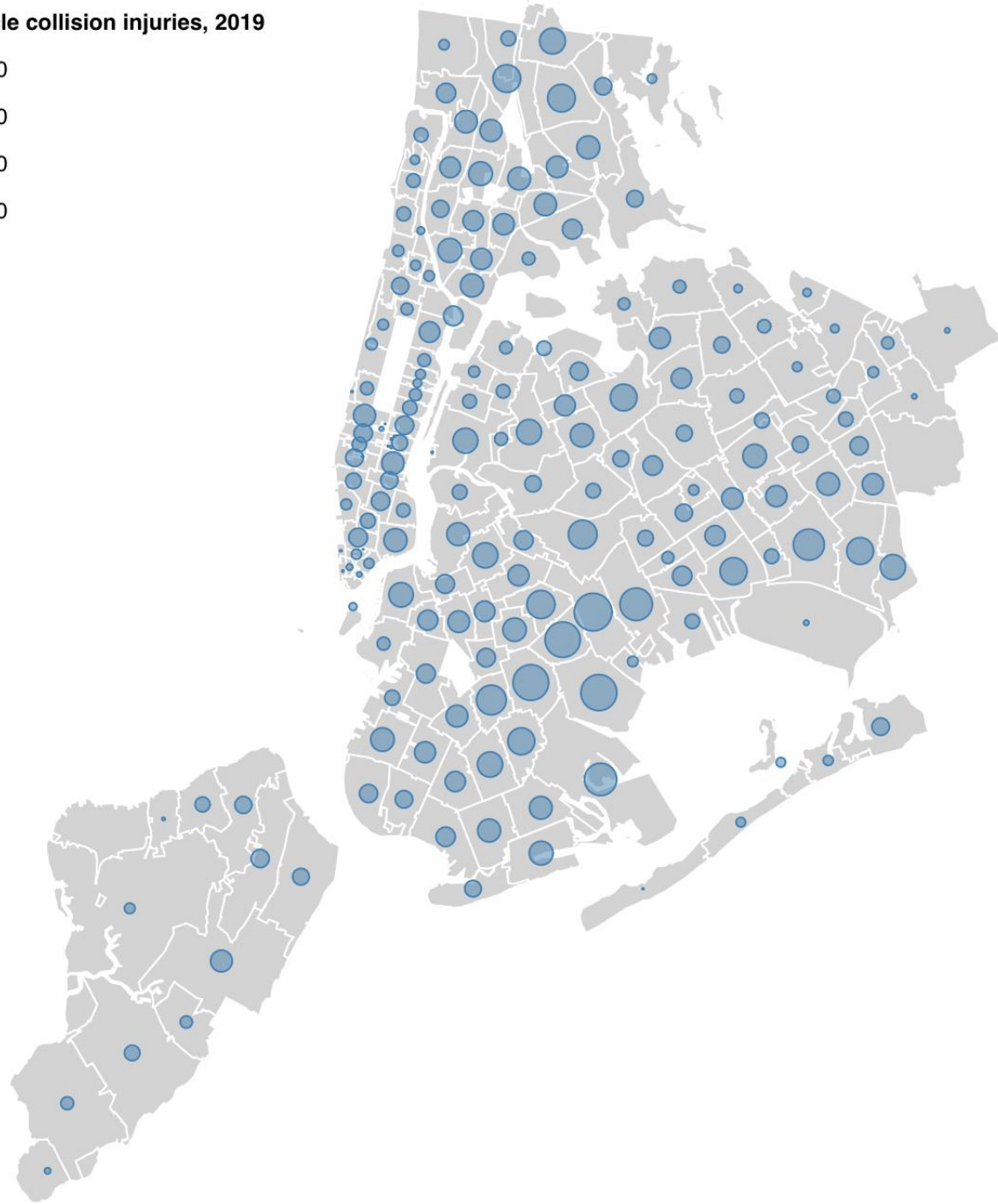
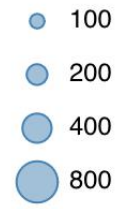
```
{
  const svg = d3.create('svg')
    .attr('width', mapWidth)
    .attr('height', mapHeight);

  // create one path for each zip code
  svg.selectAll("path")
    .data(nycGeo.features)
    .join("path")
    // use the geographic path generator to set the shape of the path
    .attr("d", path)
    // use the color scale to set the color of the path
    .attr("fill", d => color(zipToInjuries[d.properties.zcta]))
    .attr("stroke", "white");

  return svg.node();
}
```



**Vehicle collision injuries, 2019**



```
maxRadius = 10
```

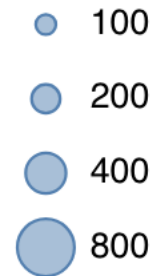
```
maxRadius = 10
```

```
radius = f(n)
```

```
// we want to map number of injuries to the area of the circle,  
// so we use scaleSqrt since the area of a circle is pi * (r^2)
```

```
radius = d3.scaleSqrt()  
  .domain([0, maxInjuriesForZip])  
  .range([0, maxRadius])
```

### Vehicle collision injuries, 2019



```
const svg = d3.create('svg')
  .attr('width', mapWidth)
  .attr('height', mapHeight);

// draw map

svg.selectAll("path")
  .data(nycGeo.features)
  .join("path")
  .attr("d", path)
  .attr("fill", '#d3d3d3')
  .attr("stroke", "white");
```



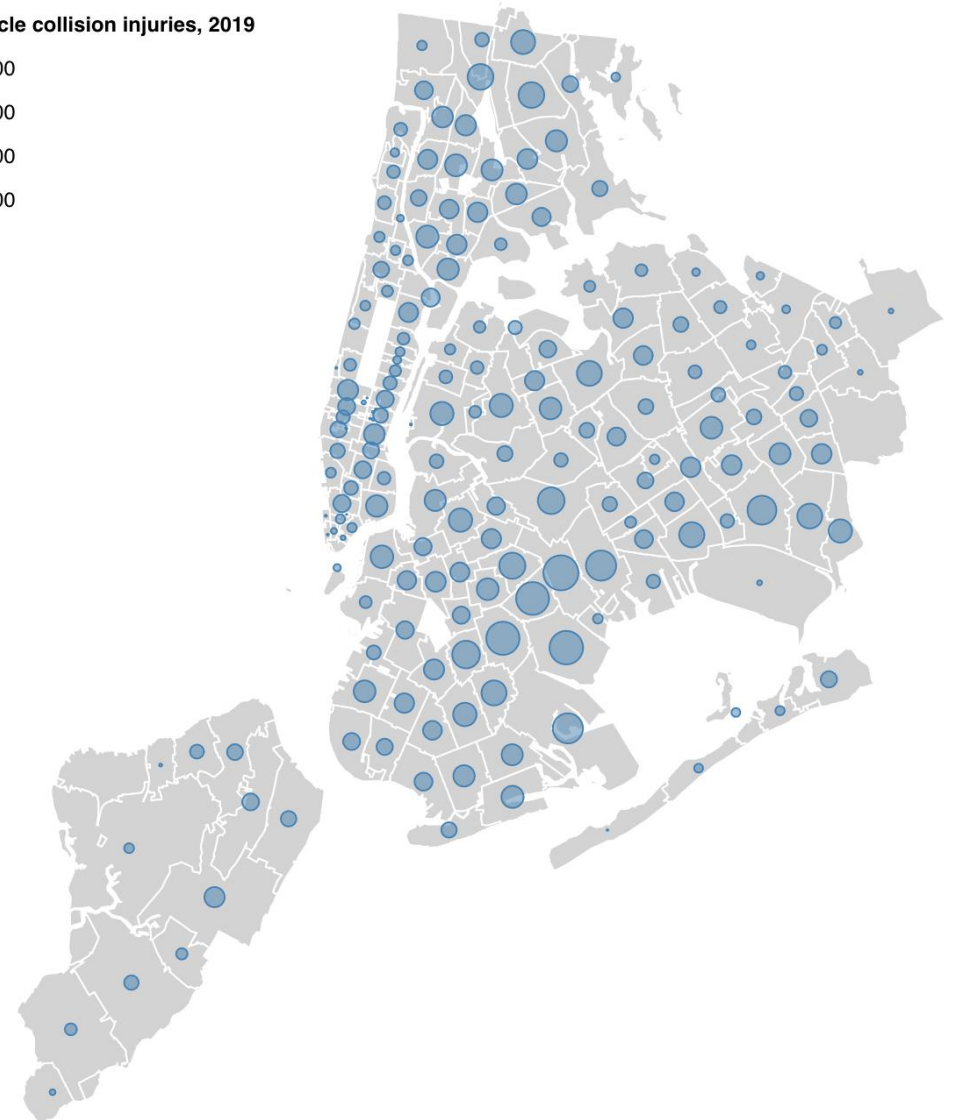
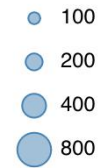
► Array(2) [238.85631317949486, 244.91887758343373]

```
path.centroid(nycGeo.features[0])
```

```
// draw circles
```

```
svg.append("g")
  .selectAll("circle")
  .data(nycGeo.features)
  .join("circle")
  .attr("stroke", "steelblue")
  .attr("fill", "steelblue")
  .attr("fill-opacity", 0.5)
  // set the position of the circles
  .attr("transform", d => `translate(${path.centroid(d)})`)
  // set the size of the circles
  .attr("r", d => radius(zipToInjuries[d.properties.zcta]));
```

Vehicle collision injuries, 2019



# Projections

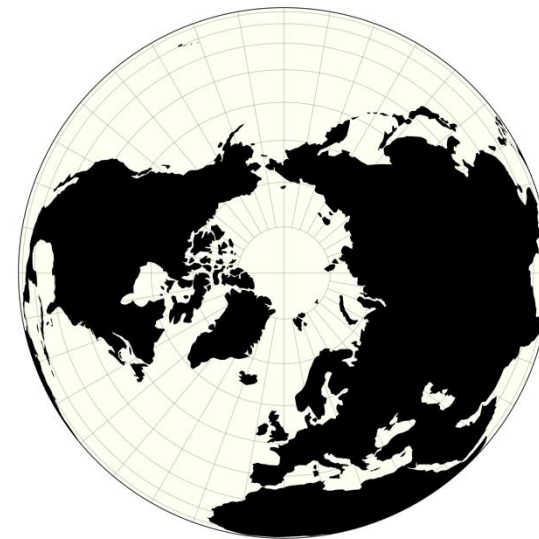
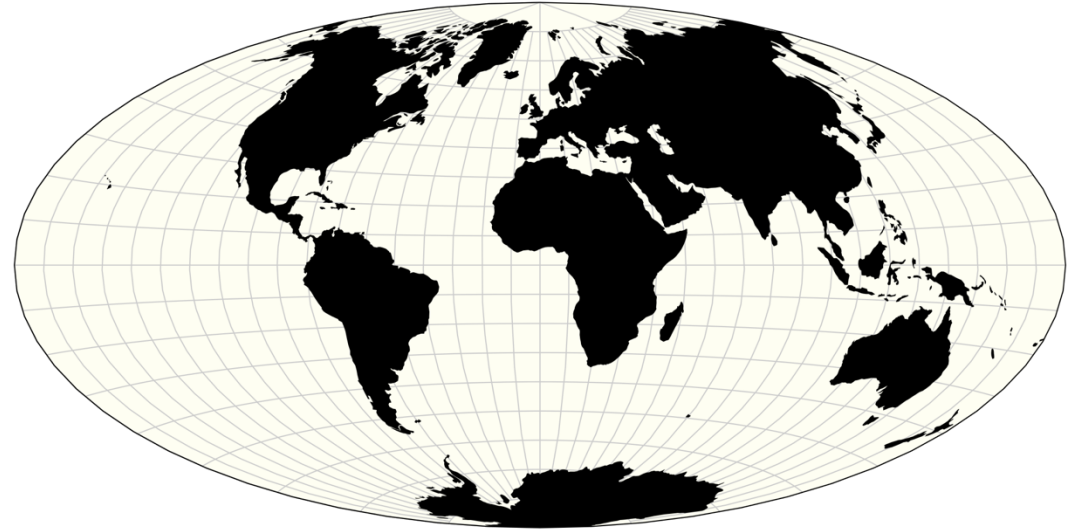
```
const path = d3.geoPath(projection);
```

svg

```
.append("g")  
.selectAll("path")  
.data(l.features)  
.enter()  
.append("path")  
.attr("id", d => d.properties.name)  
.attr("fill", "black");
```

```
graticule = ▼ Object {  
  type: "MultiLineString"  
  coordinates: ► Array(53) [Array(3),  
  ]  
}
```

```
graticule = d3.geoGraticule10()
```



# Approaches to Mapping Data

**Symbol Maps** → plot data over a map

**Choropleth Maps** → colored regions

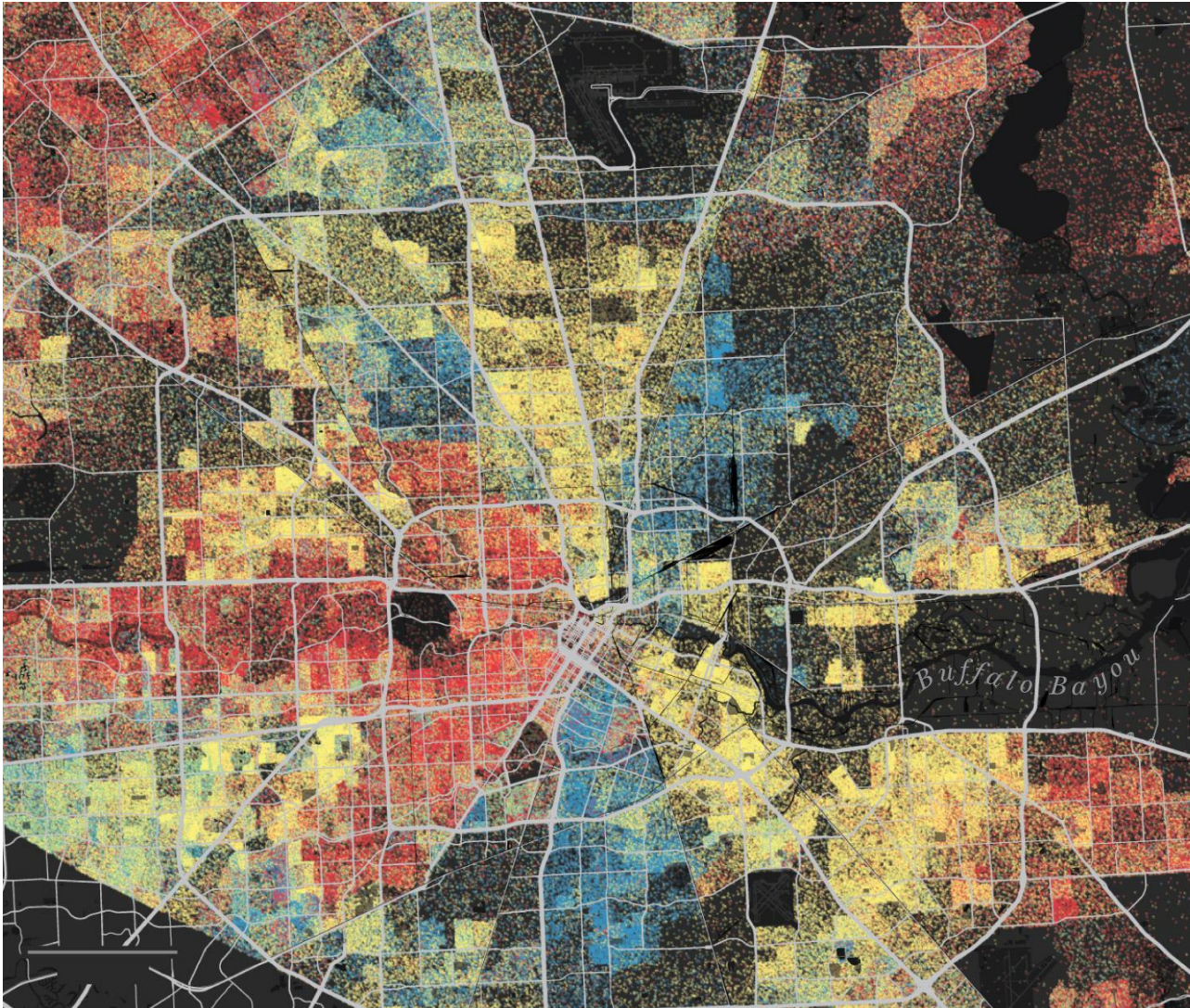
**Heatmaps & Contours** → show densities

**Cartograms** → distort to show quantities

**Flow Maps** → flux across regions

**Generalization** → distort/abstract to aid tasks

# Symbol map (scatterplot)



**226**  
Kamala Harris

74,109,194 votes (48.3%)

270  
TO WIN

153.0 million votes so far (Estimated 98.7% counted)

**312**

🗳️ Donald J. Trump

76,683,848 votes (50.0%)



By winner



Electoral votes



Margin by county



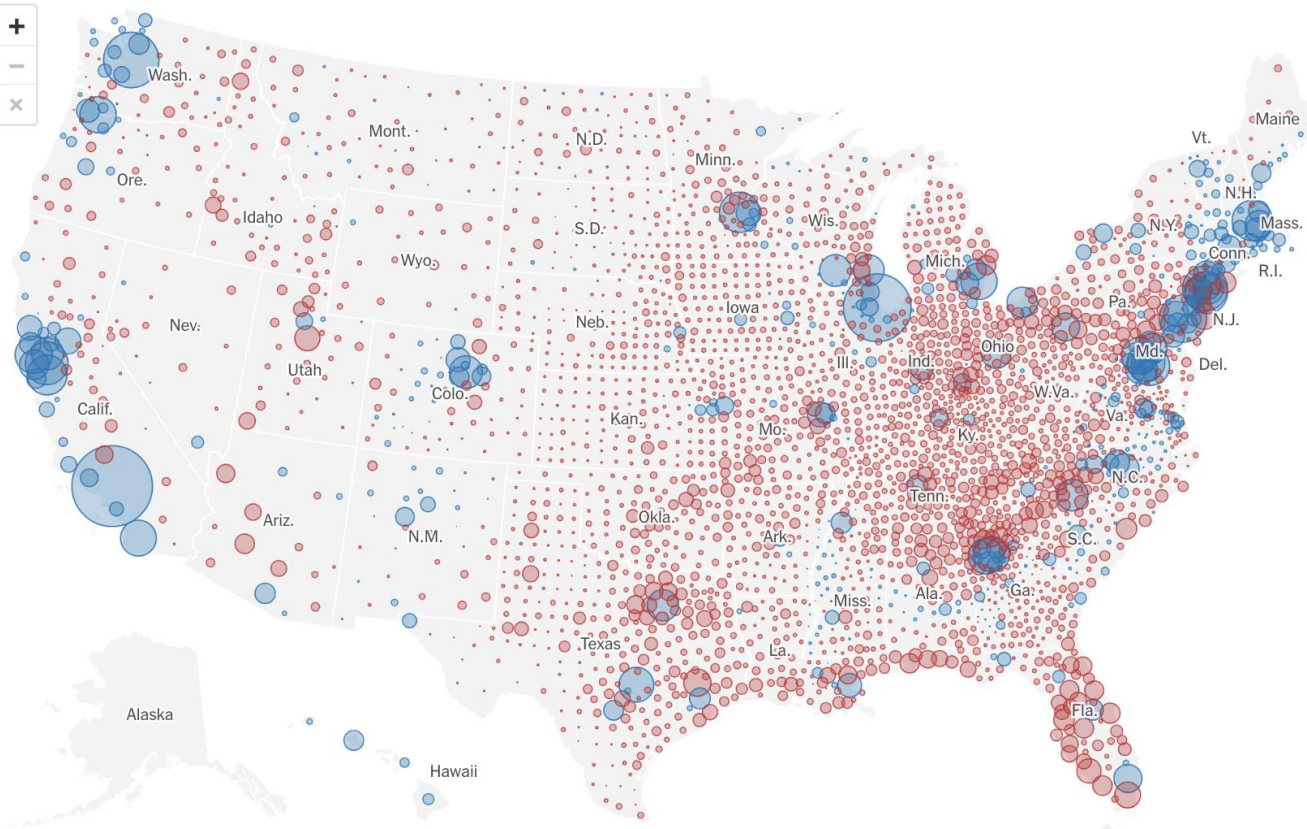
Shift from 2020

**LEADER**

● Dem.

● Rep.

Circle size is proportional to the amount each county's leading candidate is ahead.



**226**  
Kamala Harris

74,109,194 votes (48.3%)

270  
TO WIN

153.0 million votes so far (Estimated 98.7% counted)

**312**  
✓ Donald J. Trump

76,683,848 votes (50.0%)



By winner



Electoral votes



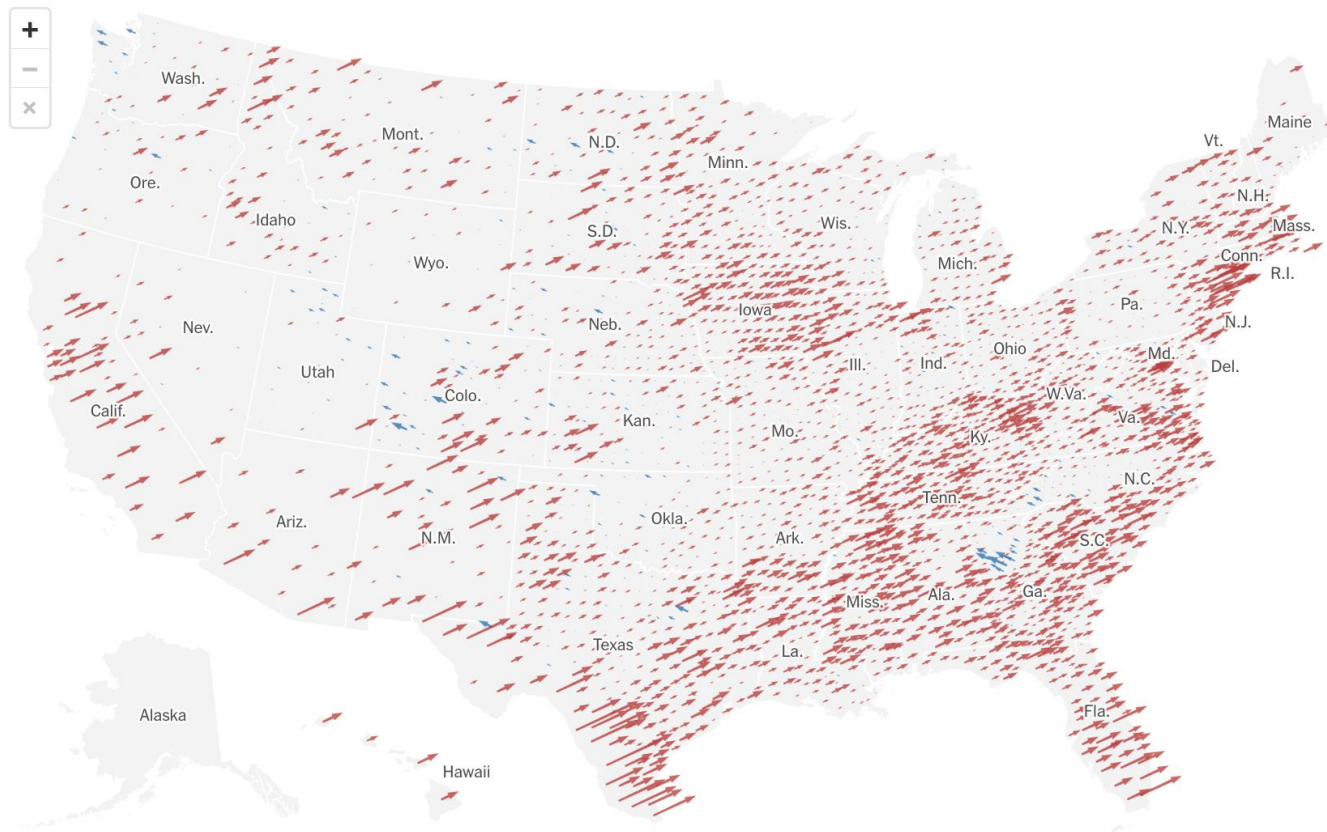
Margin by county



Shift from 2020

**SHIFT IN MARGIN**

More Dem. More Rep.  
Compared with 2020 presidential vote in places that have reported almost all of their votes.

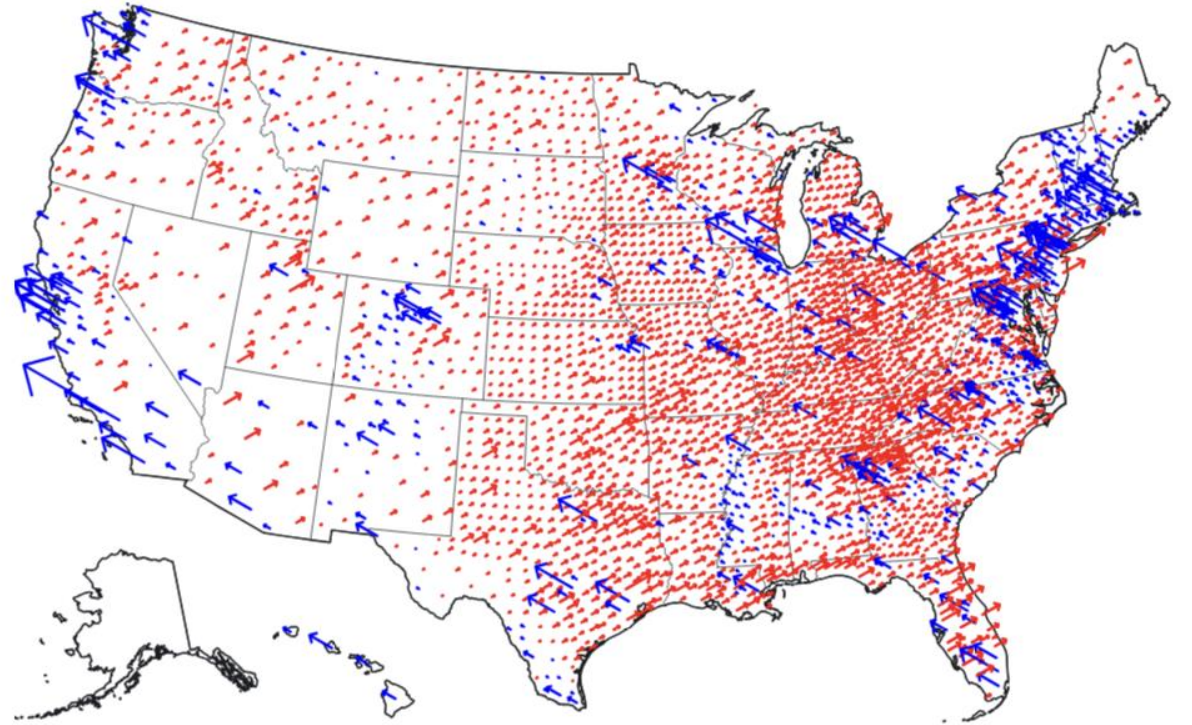


# In observable plot



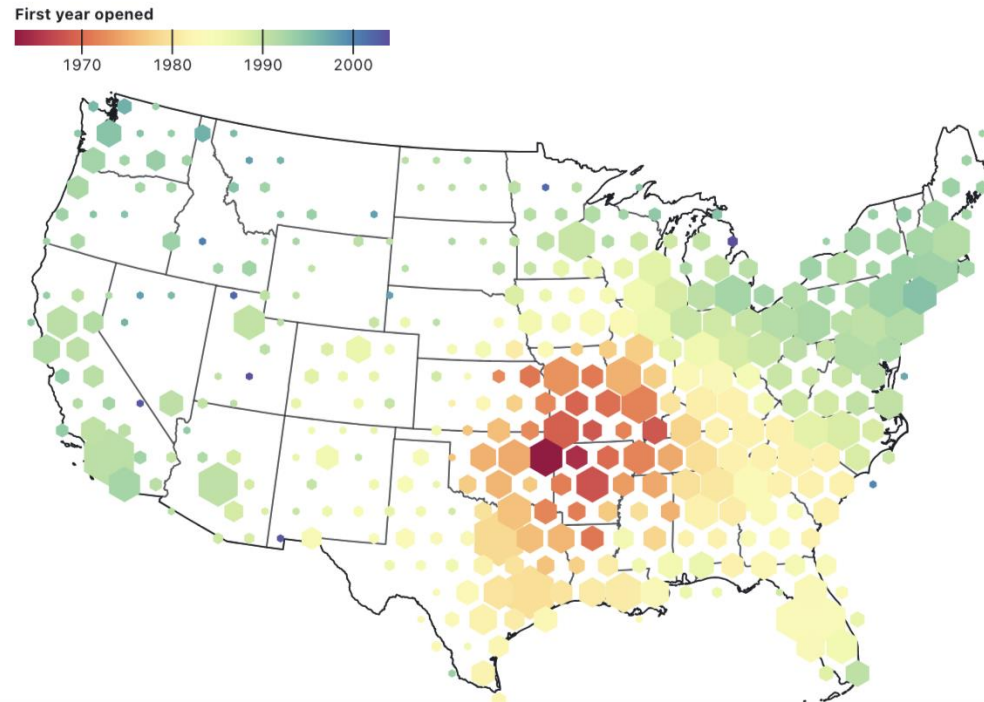
```
Plot.plot({
  projection: {type: "orthographic", rotate: [-longitude, -30]},
  r: {transform: (d) => Math.pow(10, d)}, // convert Richter to amplitude
  style: "overflow: visible;", // allow dots to escape
  marks: [
    Plot.geo(land, {fill: "currentColor", fillOpacity: 0.2}),
    Plot.sphere(),
    Plot.dot(earthquakes, {x: "longitude", y: "latitude", r: "magnitude", stroke: "red", fill: "red", fillOpacity: 0.2})
  ]
})
```

# In observable plot



```
Plot.plot({
  projection: "albers-usa",
  length: {type: "sqrt", transform: Math.abs},
  marks: [
    Plot.geo(statemesh, {strokeWidth: 0.5}),
    Plot.geo(nation),
    Plot.vector(
      counties,
      Plot.centroid({
        anchor: "start",
        length: (d) => d.properties.margin2020 * d.properties.votes,
        stroke: (d) => d.properties.margin2020 > 0 ? "red" : "blue",
        rotate: (d) => d.properties.margin2020 > 0 ? 60 : -60
      })
    )
  ]
})
```

# With binning



```
Plot.plot({
  projection: "albers",
  r: {range: [0, 16]},
  color: {scheme: "spectral", label: "First year opened", legend: true},
  marks: [
    Plot.geo(statemesh, {strokeOpacity: 0.5}),
    Plot.geo(nation),
    Plot.dot(walmarts, Plot.hexbin({r: "count", fill: "min"}, {x: "longitude", y: "latitude", fill: "date"}))
  ]
})
```

# Choropleth

**226**  
Kamala Harris

74,109,194 votes (48.3%)

270  
TO WIN

153.0 million votes so far (Estimated 98.7% counted)

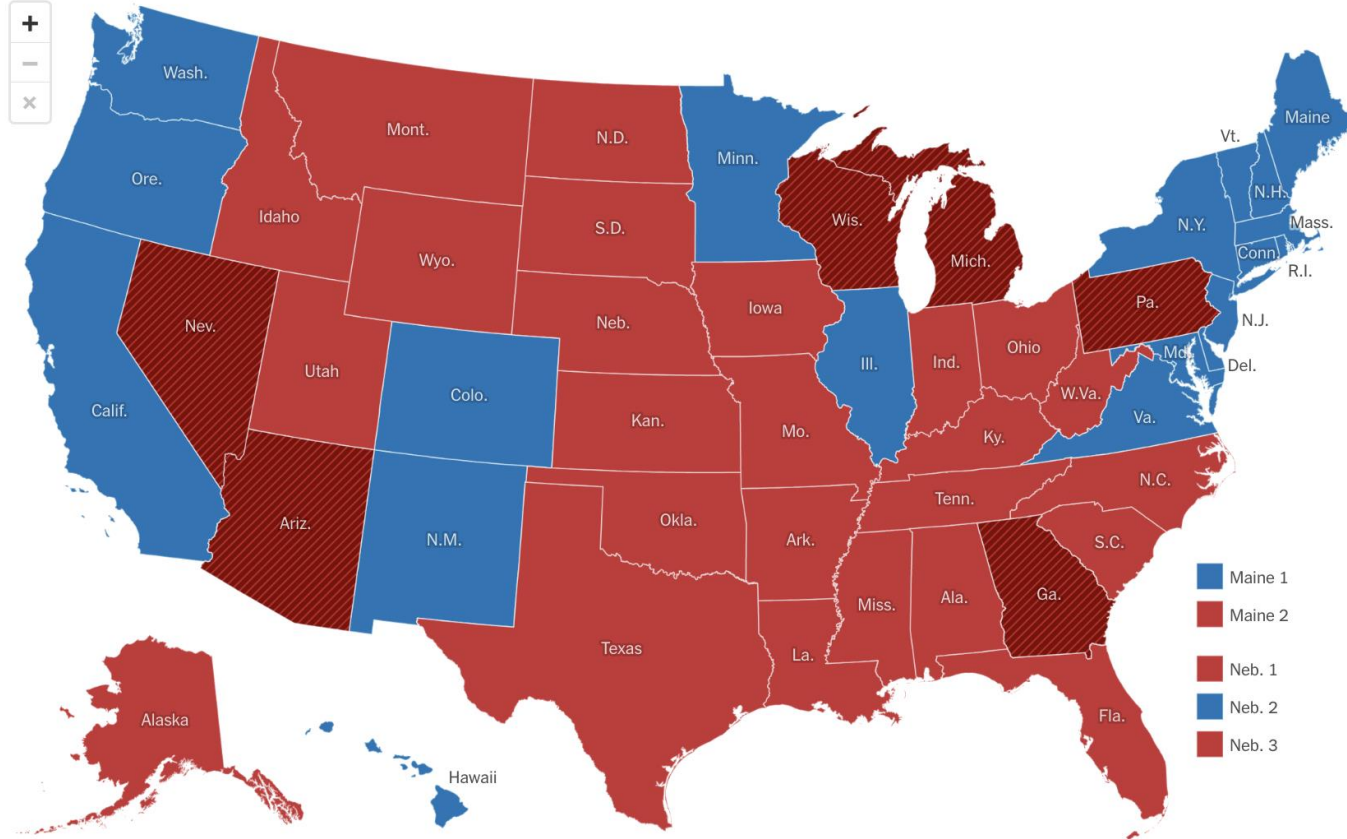
**312**

Donald J. Trump

76,683,848 votes (50.0%)

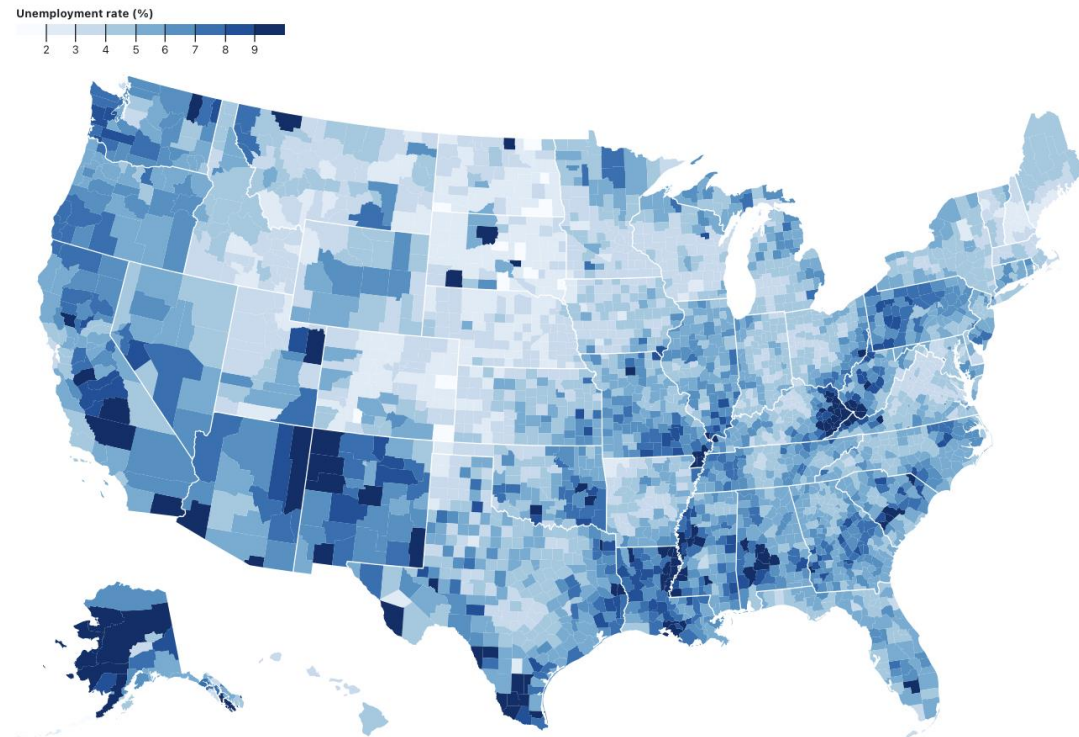


Dem. Win Flip Rep.



Maine 1  
Maine 2  
Neb. 1  
Neb. 2  
Neb. 3

# Choropleth in observable plot



```
Plot.plot({
  projection: "identity",
  width: 975,
  height: 610,
  color: {scheme: "Blues", type: "quantize", n: 9, domain: [1, 10], label: "Unemployment rate (%)", legend: true},
  marks: [
    Plot.geo(topojson.feature(us, us.objects.counties), {fill: (map => d => map.get(d.id))(new Map(data.map(d => [d.id, d.rate])))}),
    Plot.geo(topojson.mesh(us, us.objects.states, (a, b) => a !== b), {stroke: "white"})
  ]
})
```

## READING, WRITING, AND EARNING MONEY

The latest data from the U.S. Census's American Community Survey paints a fascinating picture of the United States at the county level. We've looked at the educational achievement and the median income of the entire nation, to see where people are going to school, where they're earning money, and if there is any correlation.



1 HIGH SCHOOL GRADUATES 65% 71% 82% 88%



2 COLLEGE GRADUATES 15% 22% 30% 40%

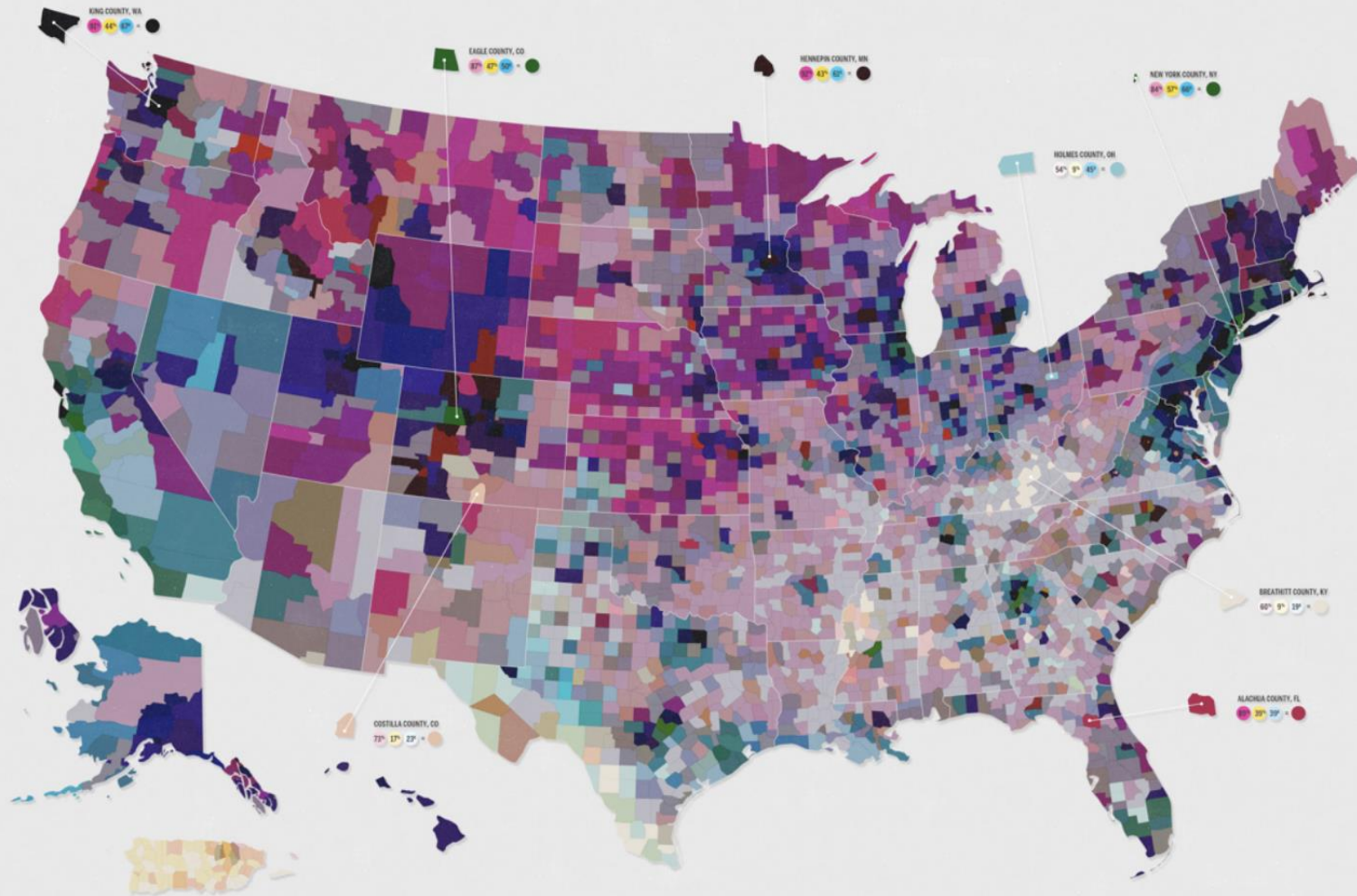


3 MEDIAN HOUSEHOLD INCOME 20K 40K 50K 60K

The map at right is a product of overlaying the three sets of data. The variations in hue and value has been produced from the data shown above. In general, darker counties represent a more educated, better paid population while lighter areas represent communities with lower graduates and lower incomes.



A collaboration between GOOD and Gregory Hebrich, SOURCE: US Census

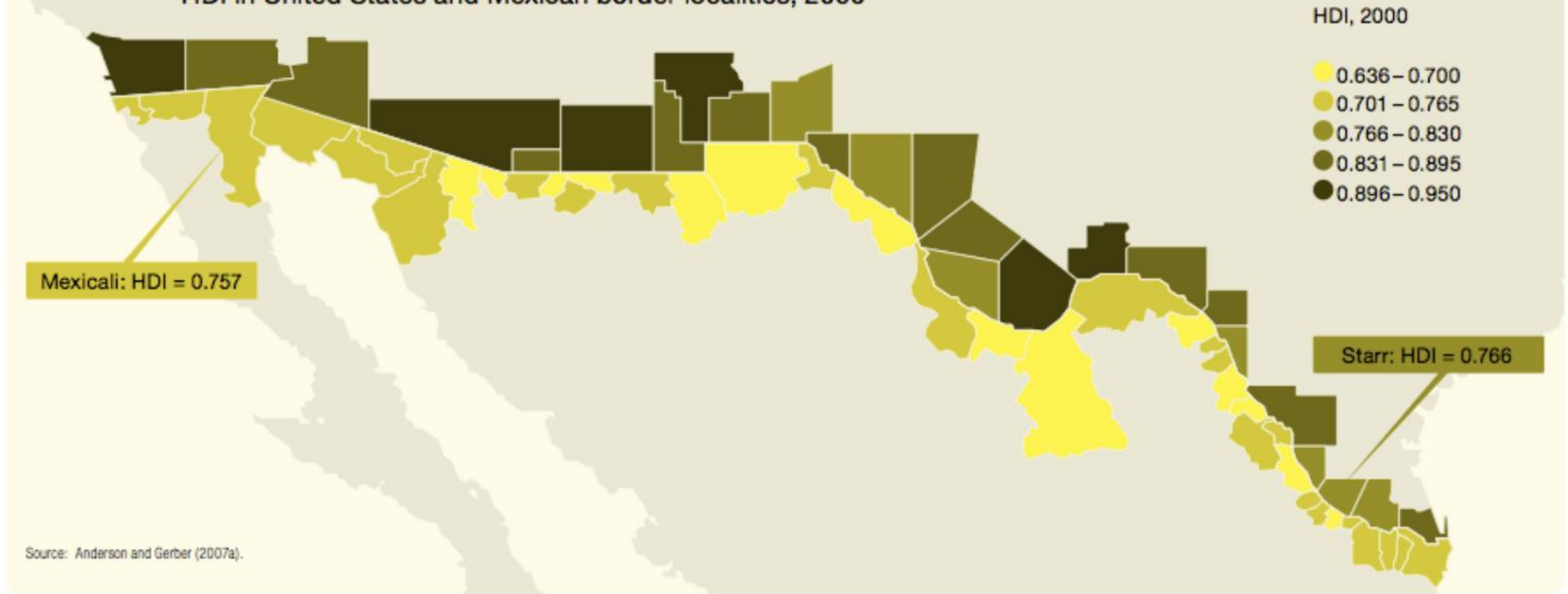


# Choose colors with care

Map 1.1

### Borders matter

HDI in United States and Mexican border localities, 2000



# Focus on the foreground

# Cartogram

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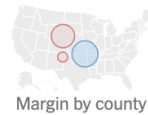
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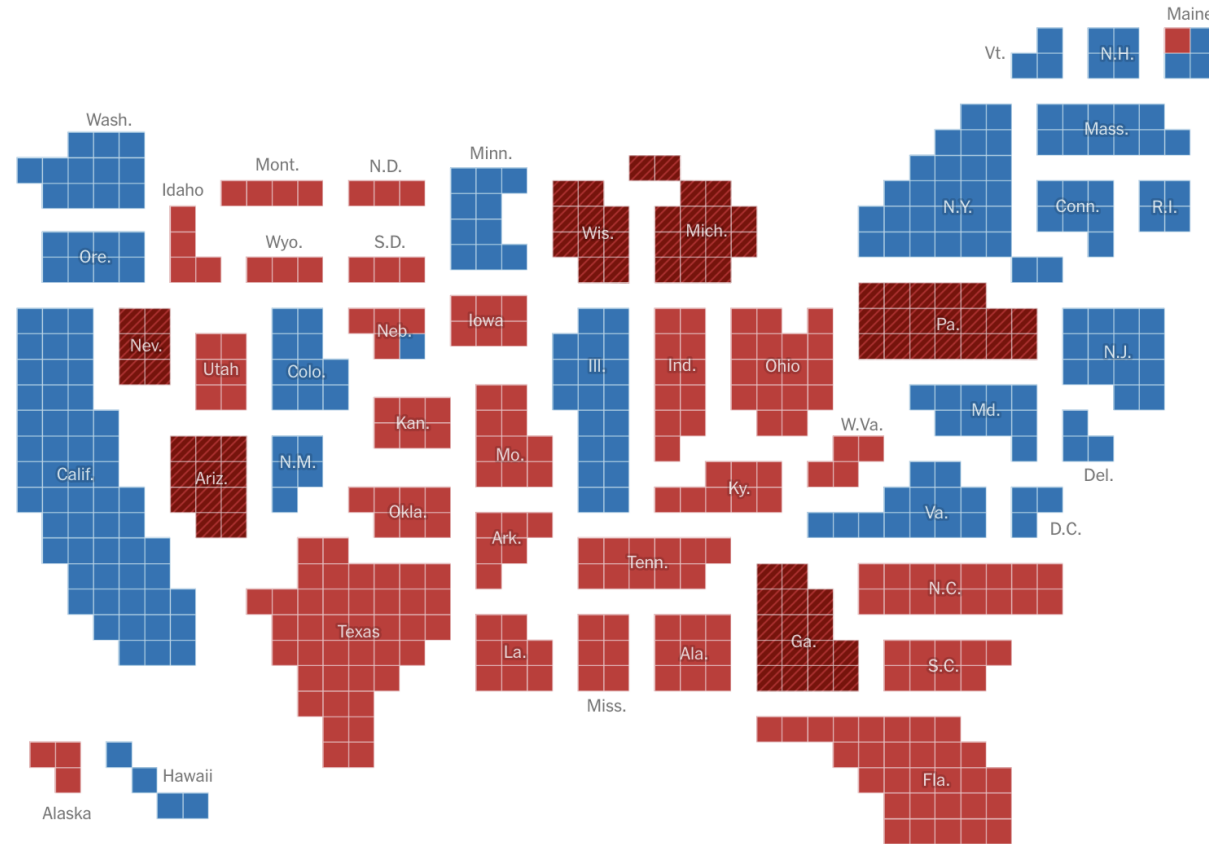
**312**

Donald J. Trump

76,683,848 votes (50.0%)



Dem. Win  
Rep. Flip



# Cartogram

## Economic Output

In this map, geography is distorted so that each country is **sized according to its economic output** in 2012. The countries are colored by their rate of growth; more established economies tend to grow more slowly.

China is both highly productive and growing rapidly. Considering individual provinces conveys its impressive scale: Guangdong, just one of 31 Chinese provinces, has an economic output greater than Indonesia.

Each hexagon represents \$2.7 billion in G.D.P.

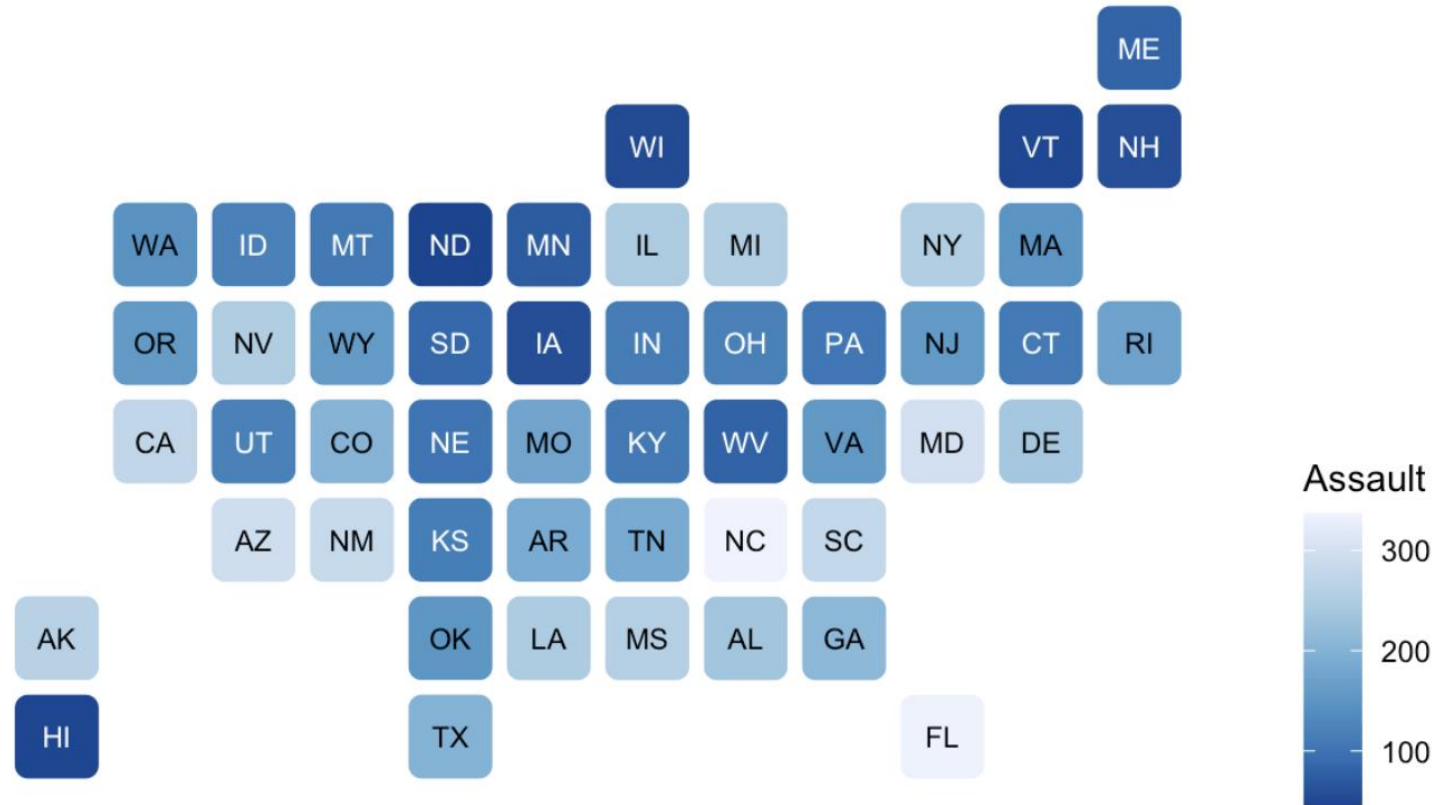
G.D.P. growth, 2011 to 2012



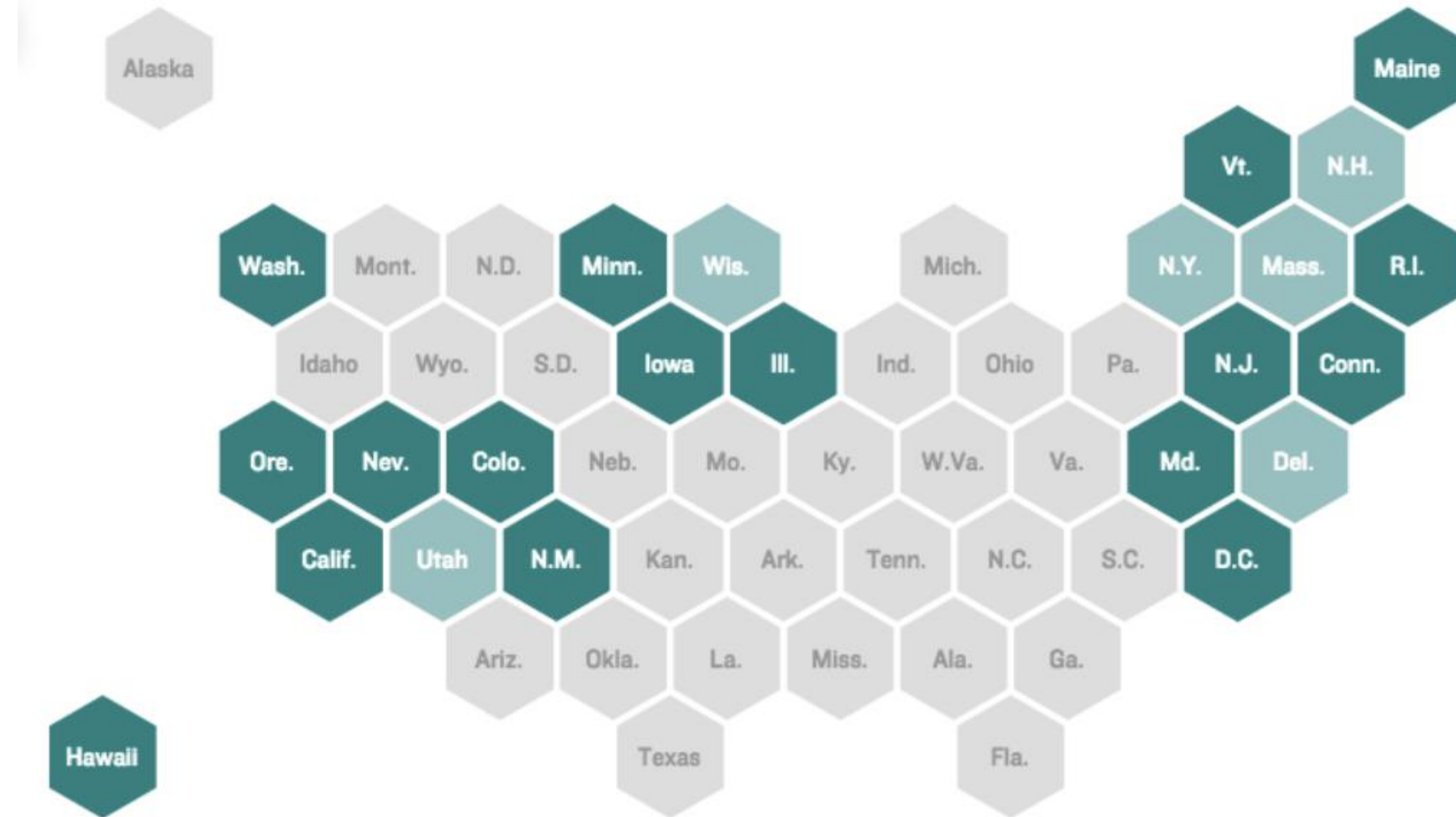
Japan and South Korea have large economic output, but growth has slowed as they have caught up with the West and innovation becomes more difficult.

New York shown for comparison.

# Cartogram



# Cartogram



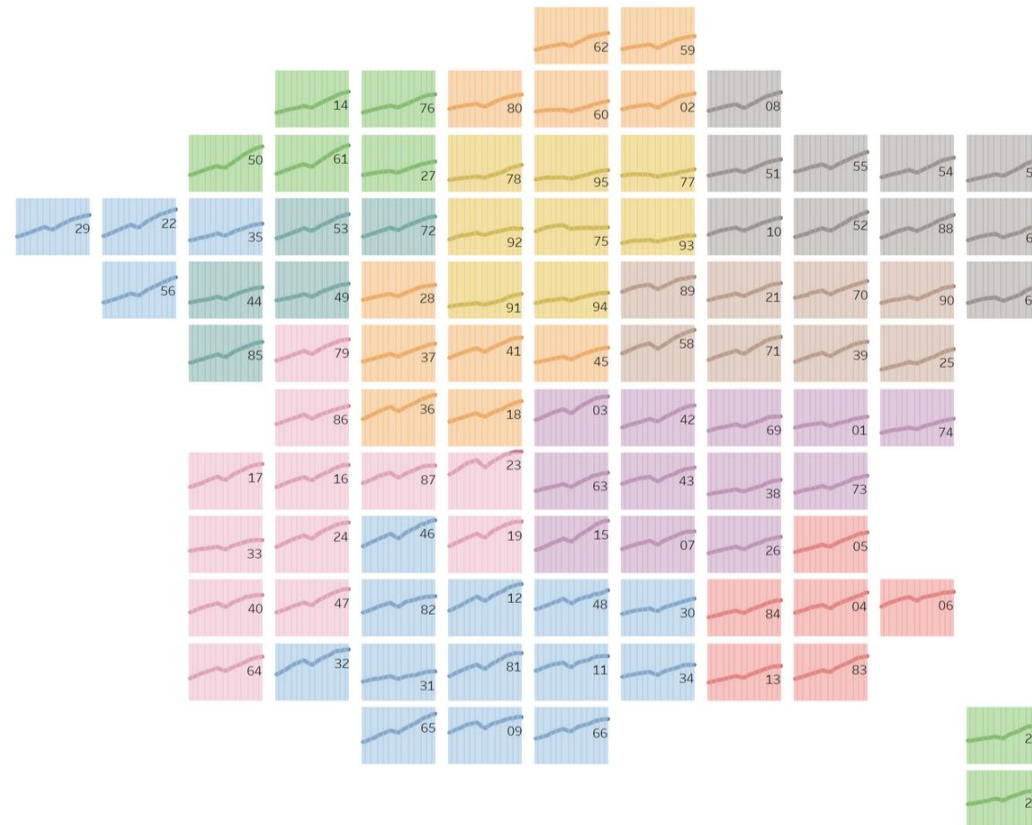
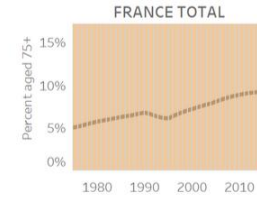


# Cartogram

## France's Ageing Population

The number of people aged 75 and over in France has risen from 4% in 1975 to 10% in 2015.

See below how this has changed in each of France's numbered departments





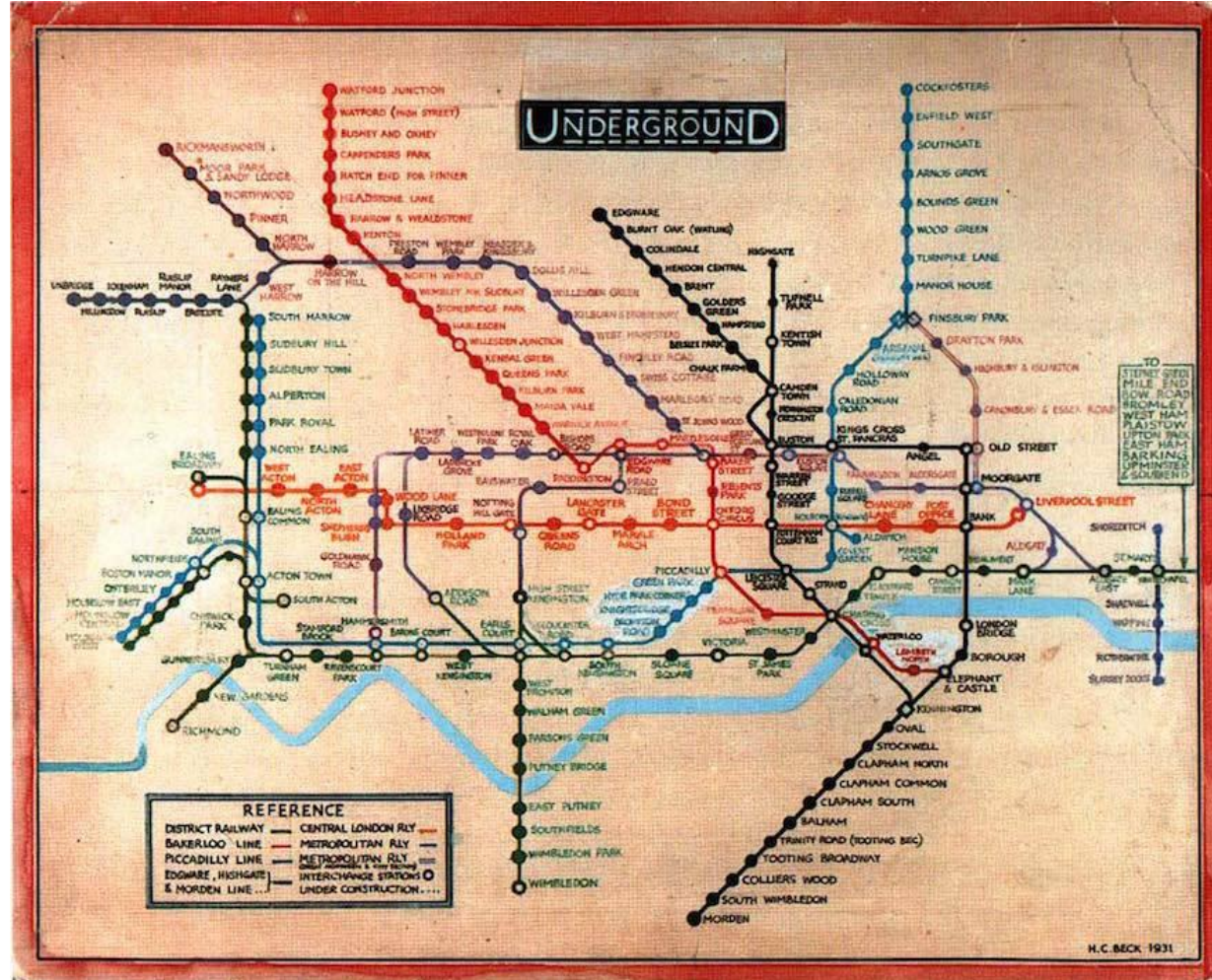
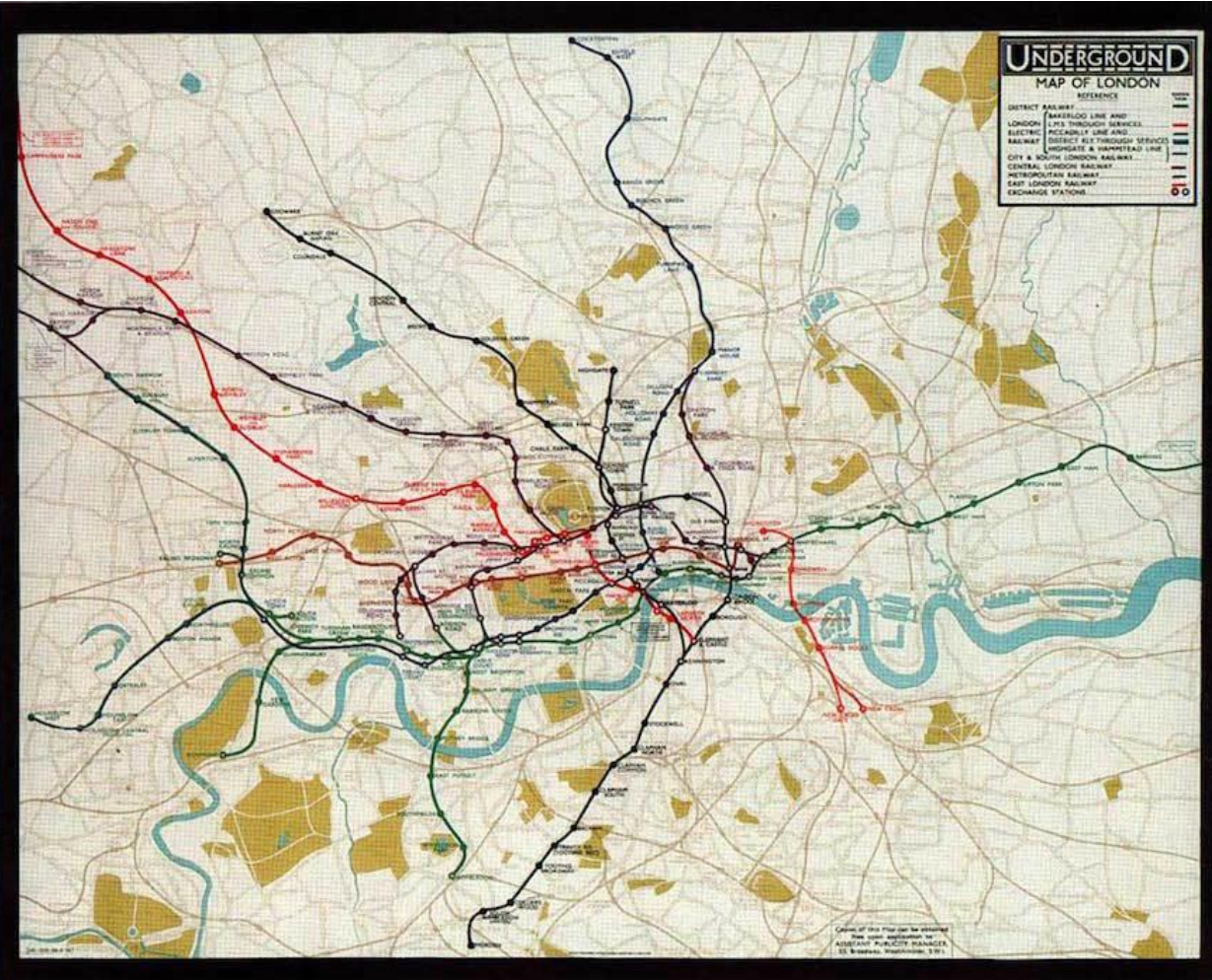
# General abstraction



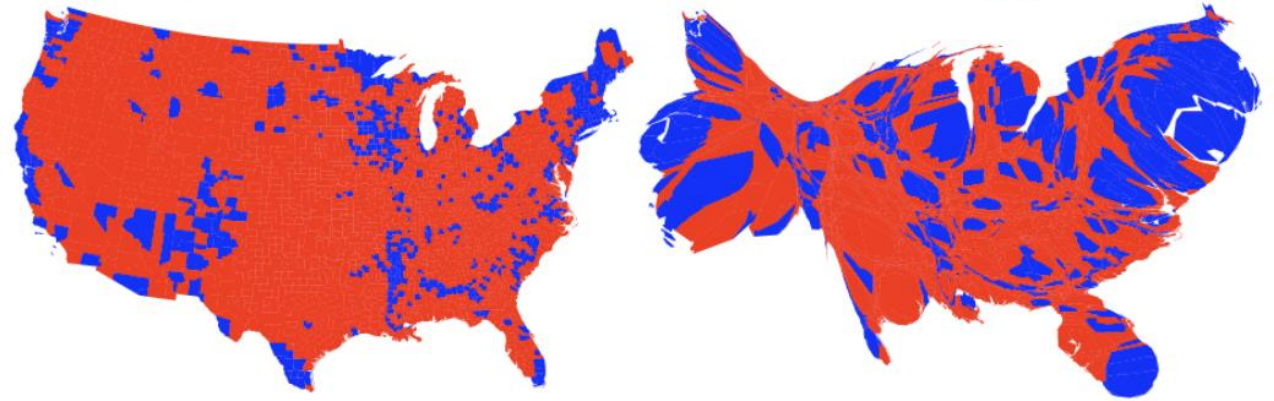
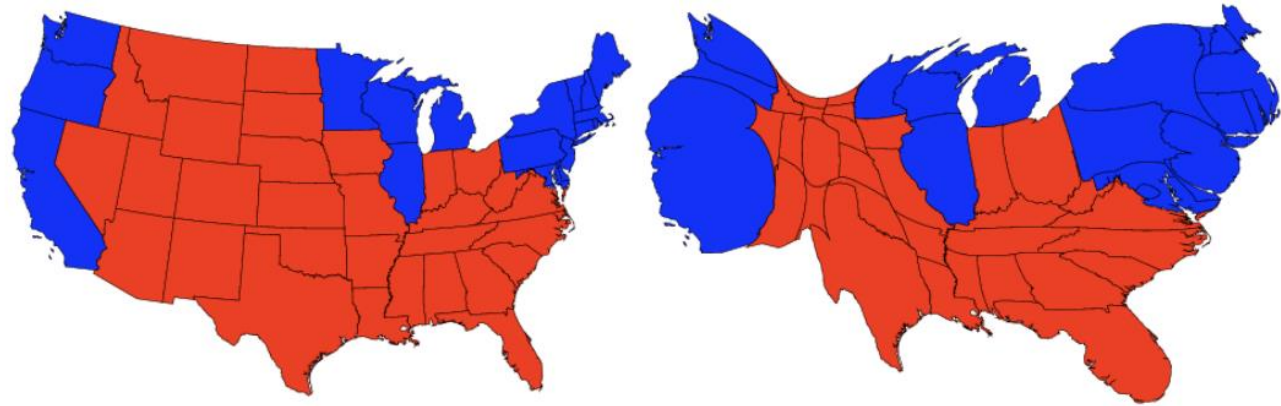
# General abstraction



# Abstraction



# Cartogram



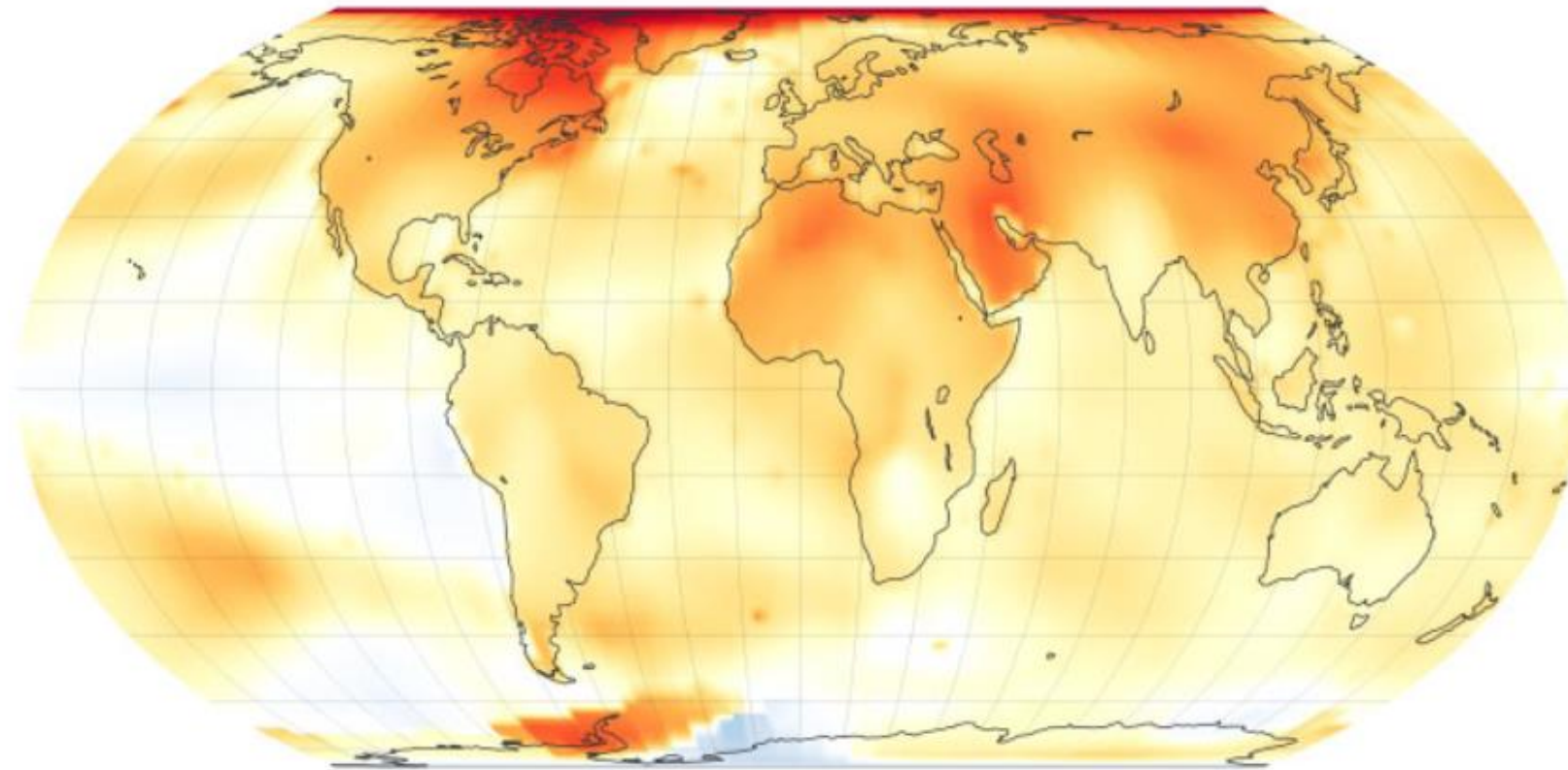
**Physical  
Diffusion  
Model**



[Newman 2004]

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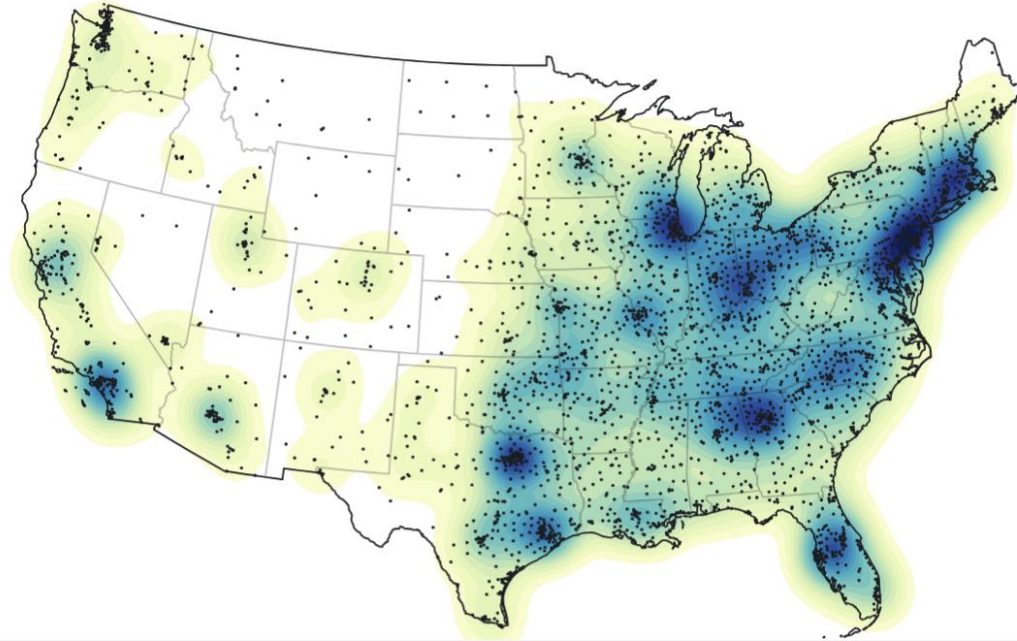
# Heatmaps



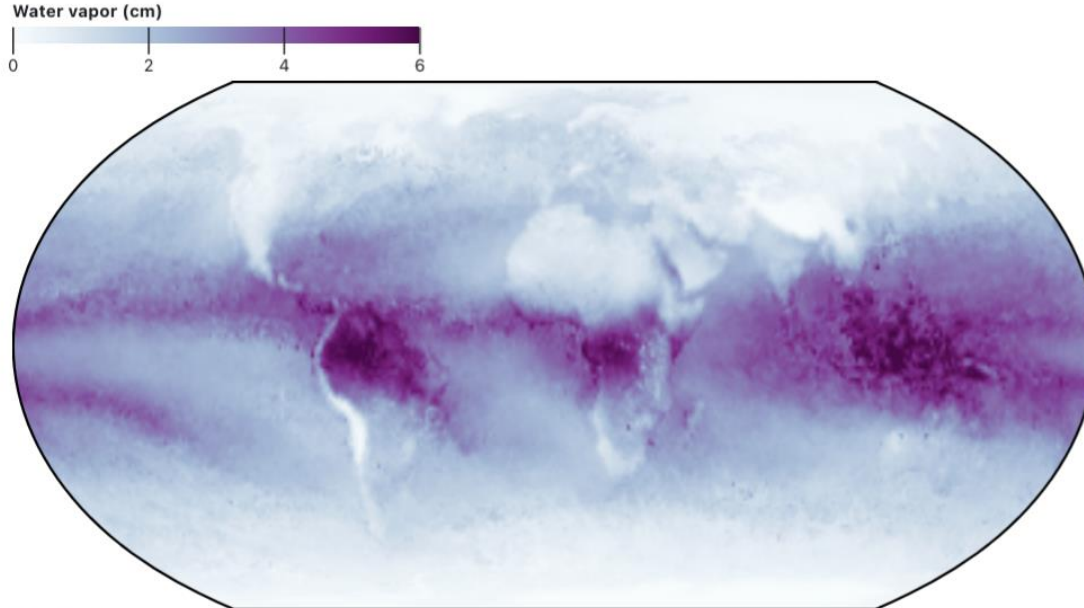
Temperature Anomaly (°C compared to the 1951-1980 average)



# Heatmaps



```
Plot.plot({
  projection: "albers",
  color: {scheme: "YlGnBu"},
  style: "overflow: visible;",
  marks: [
    Plot.density(walmarts, {x: "longitude", y: "latitude", bandwidth: 10, fill: "density"}),
    Plot.geo(statemesh, {strokeOpacity: 0.3}),
    Plot.geo(nation),
    Plot.dot(walmarts, {x: "longitude", y: "latitude", r: 1, fill: "currentColor"})
  ]
})
```



```
Plot.plot({
  projection: "equal-earth",
  color: {
    scheme: "BuPu",
    domain: [0, 6],
    legend: true,
    label: "Water vapor (cm)"
  },
  marks: [
    Plot.raster(vapor, {
      fill: Plot.identity,
      width: 360,
      height: 180,
      x1: -180,
      y1: 90,
      x2: 180,
      y2: -90,
      interpolate: "barycentric",
      clip: "sphere"
    }),
    Plot.sphere({stroke: "black"})
  ]
})
```

# Break data into buckets

DATAVIZ

## CRIMESPOTTING

The brazen 2007 murder of journalist Chauncey Bailey in Oakland, California, led Stamen partner Mike Migurski to

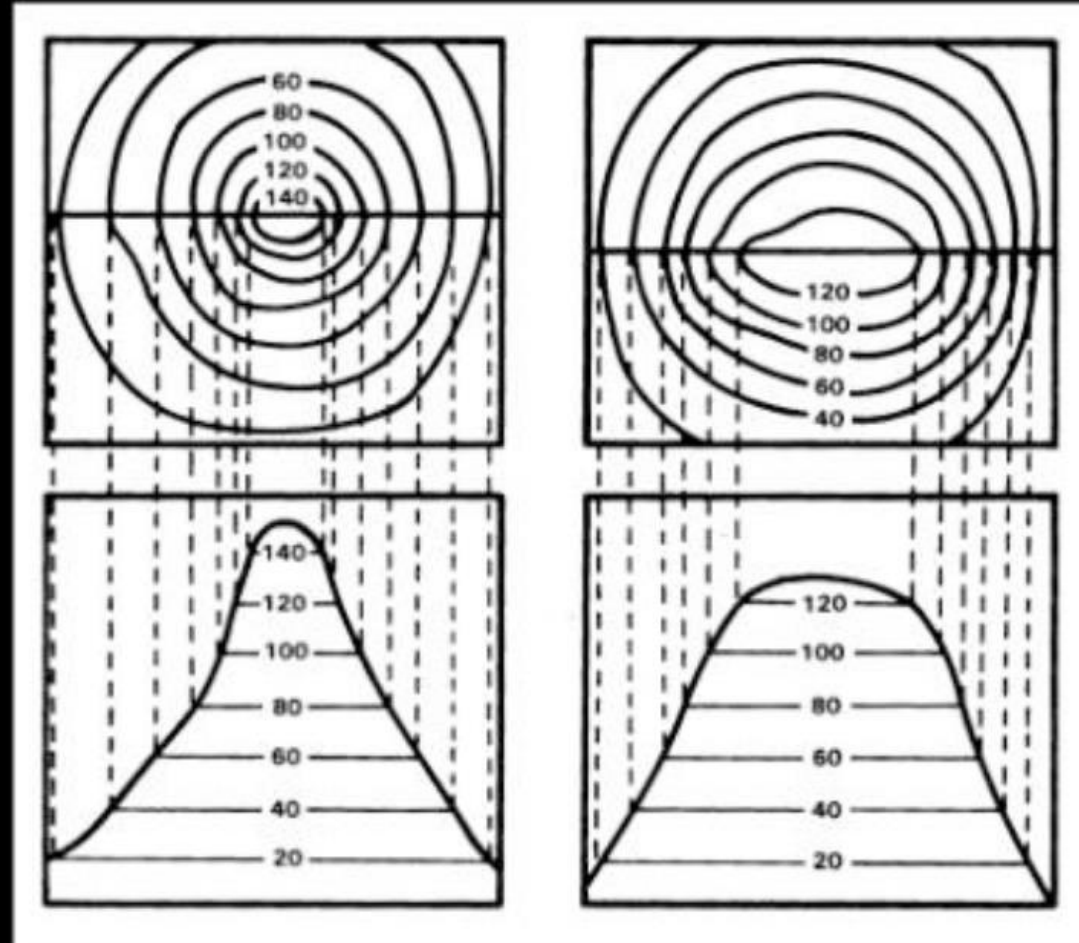
make the city's crime data more accessible. This heat map of downtown uses data from CrimeWatch, a community website,

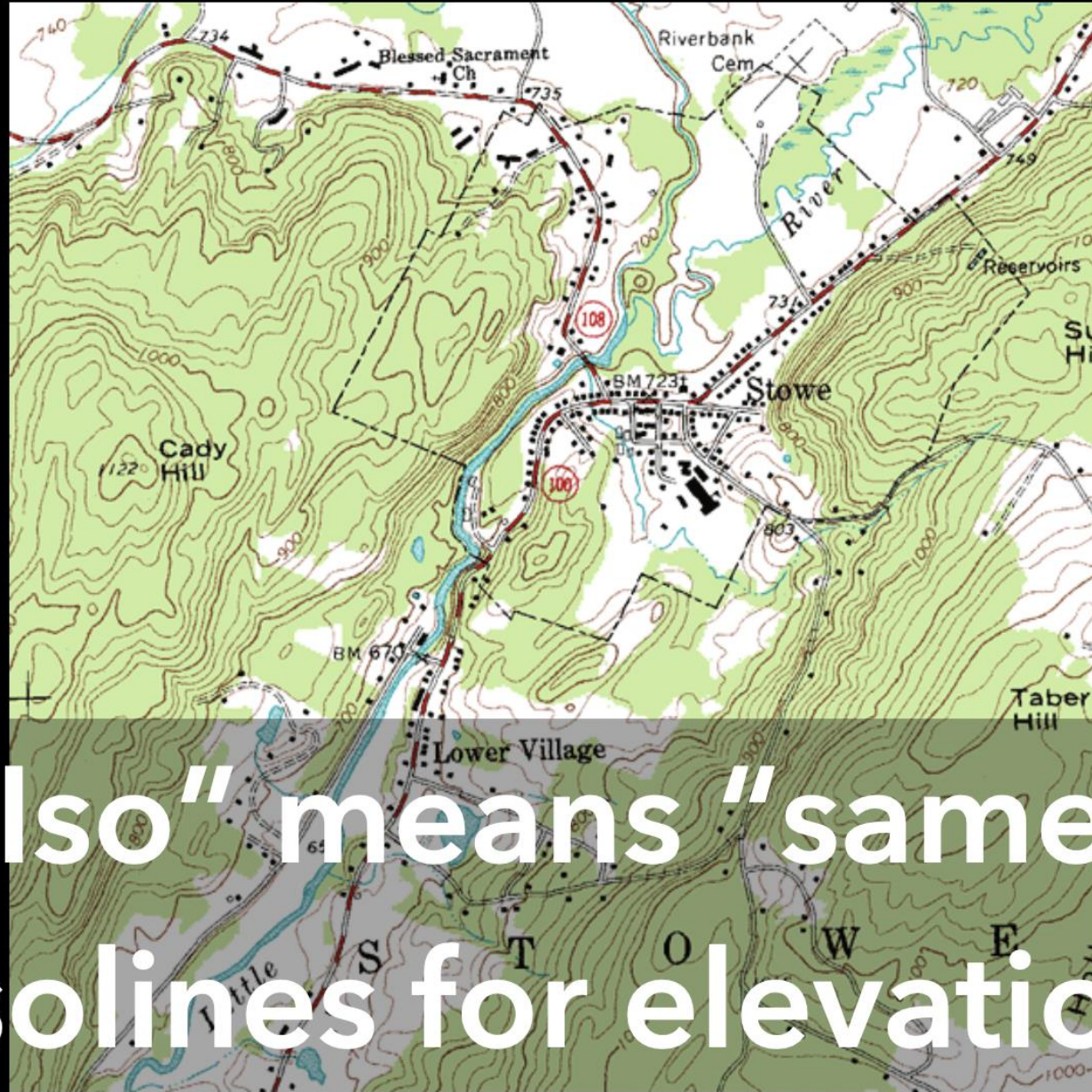
to show the gaps between crimes at a given intersection: white is high-crime; darker areas are safe. [stamen.com](http://stamen.com)

**KEY**  
Colours show how recently a crime was reported in a given part of Oakland

- A week ago
- Two weeks ago
- A month ago
- Two months ago
- Three months ago
- Four months ago
- Five months ago

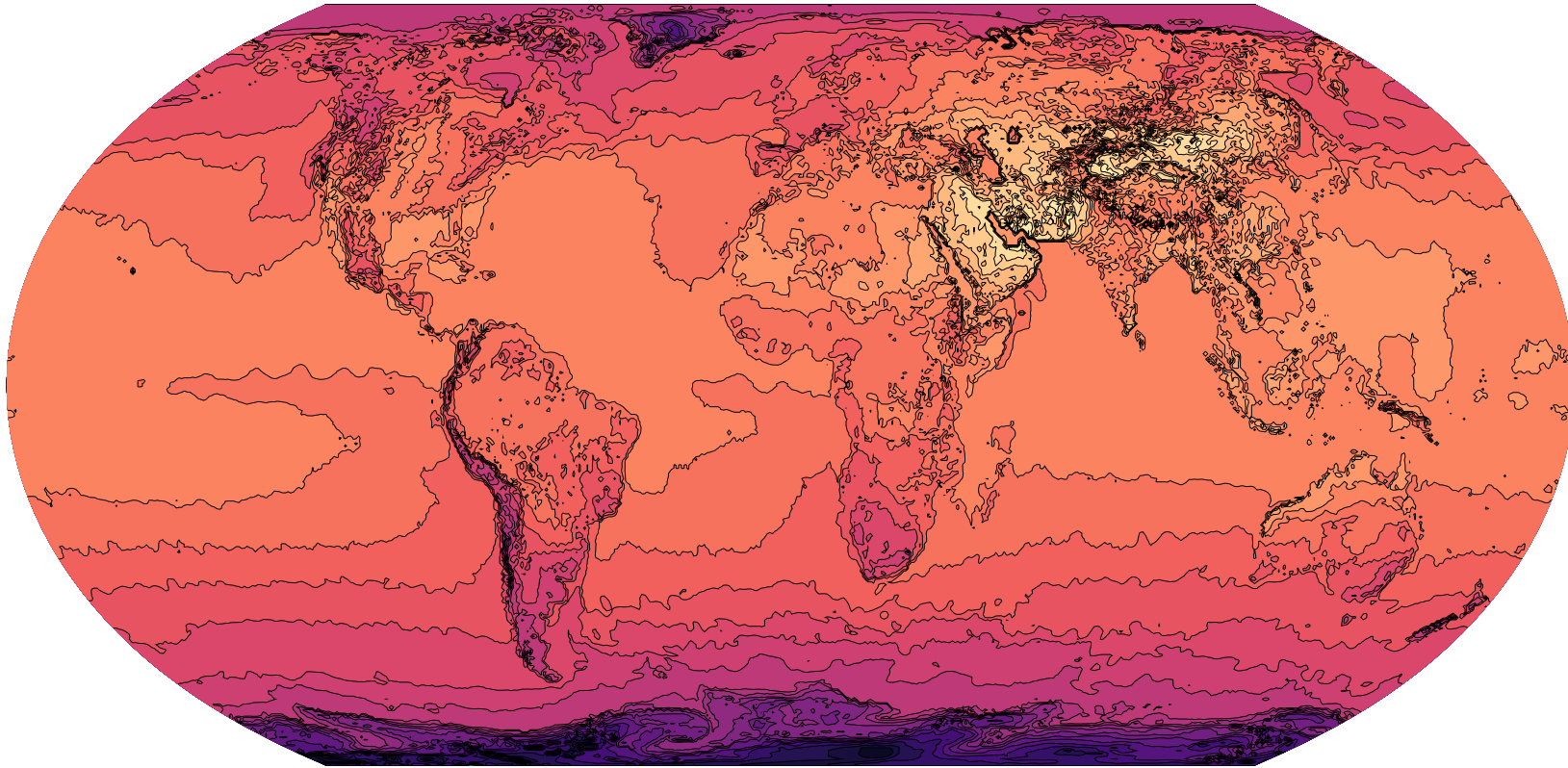
# Meaningful buckets, isolines



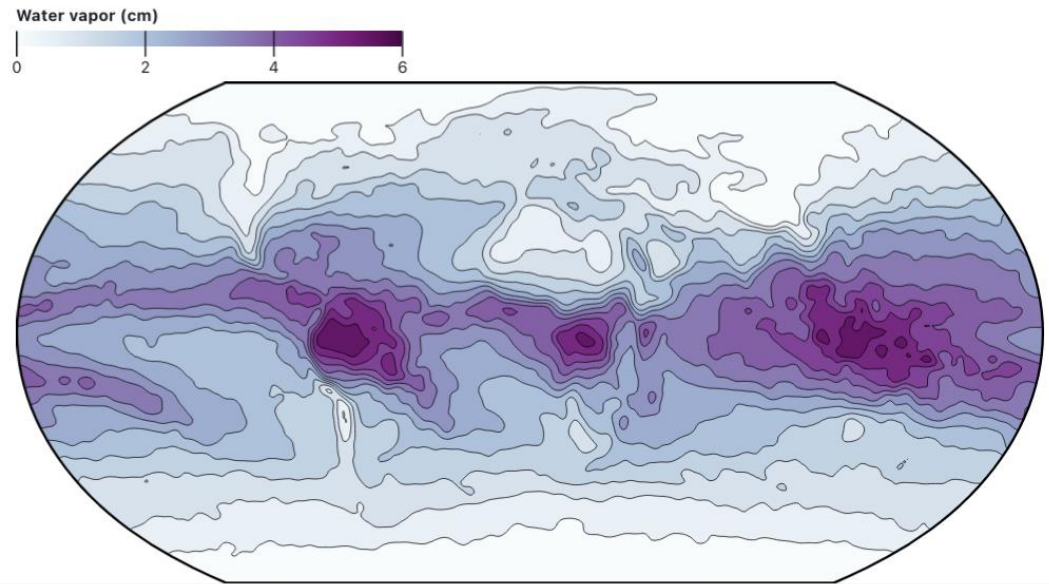


“Iso” means “same”  
Isolines for elevation

# Contour plots

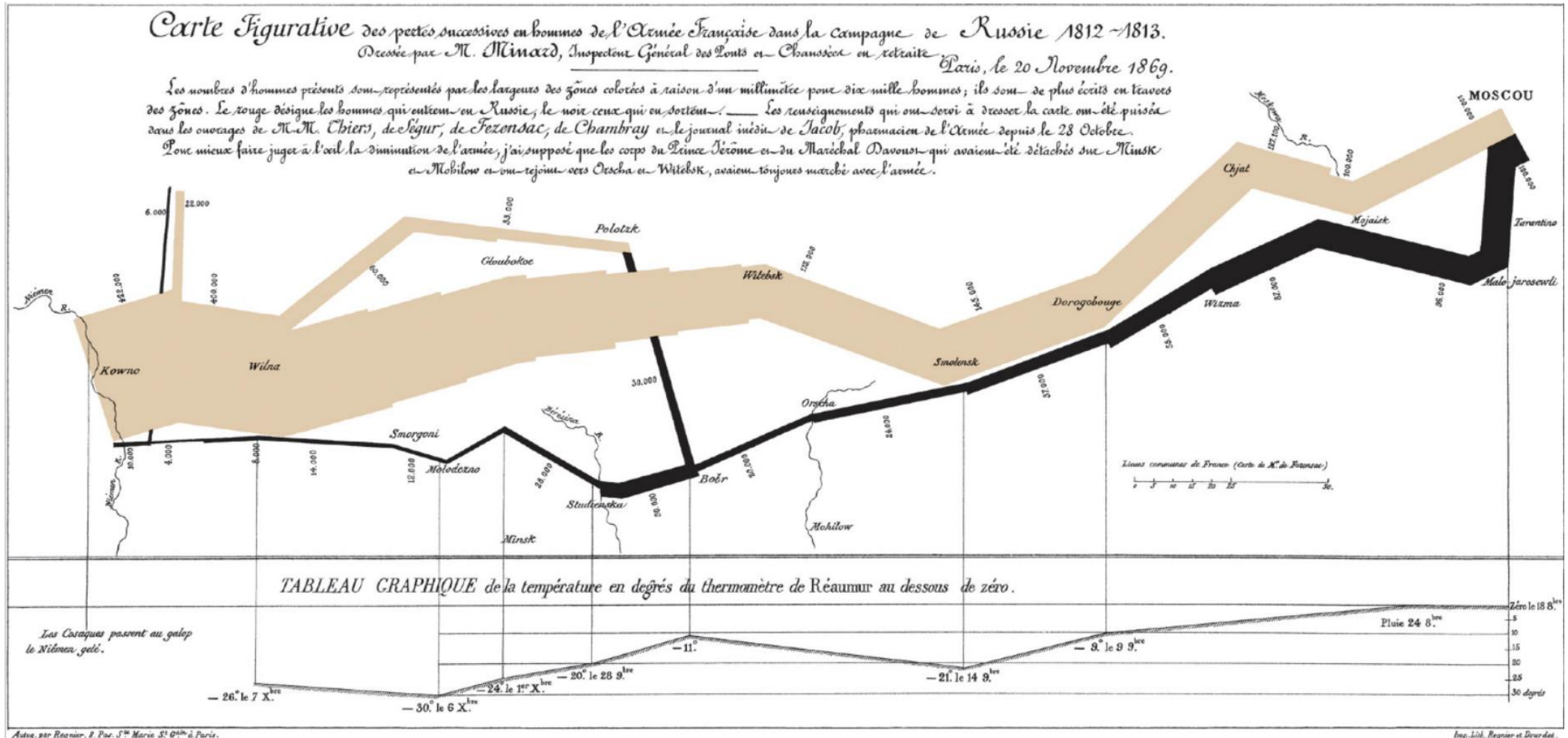


# Contour plots



```
Plot.plot({
  projection: "equal-earth",
  color: {
    scheme: "BuPu",
    domain: [0, 6],
    legend: true,
    label: "Water vapor (cm)"
  },
  marks: [
    Plot.contour(vapor, {
      fill: Plot.identity,
      width: 360,
      height: 180,
      x1: -180,
      y1: 90,
      x2: 180,
      y2: -90,
      blur: 1,
      stroke: "black",
      strokeWidth: 0.5,
      clip: "sphere"
    }),
    Plot.sphere({stroke: "black"})
  ]
})
```

# Flow map



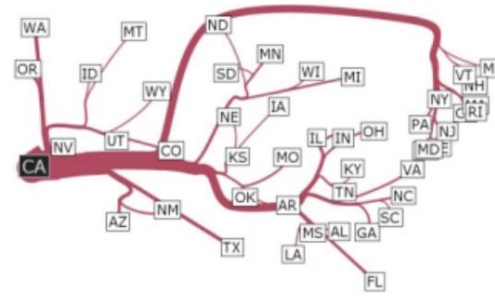
# Flow map

## Migration from California, '95-'00

Tobler 1987



Phan et al. 2005



Verbeek et al. 2011



Cui et al. 2008



Holten & van Wijk 2009

# Flow Map

**Nov. 20, 2024**

1:12 pm EST

(time of forecast download)

top speed: **41.7 mph**

average: **11.1 mph**

