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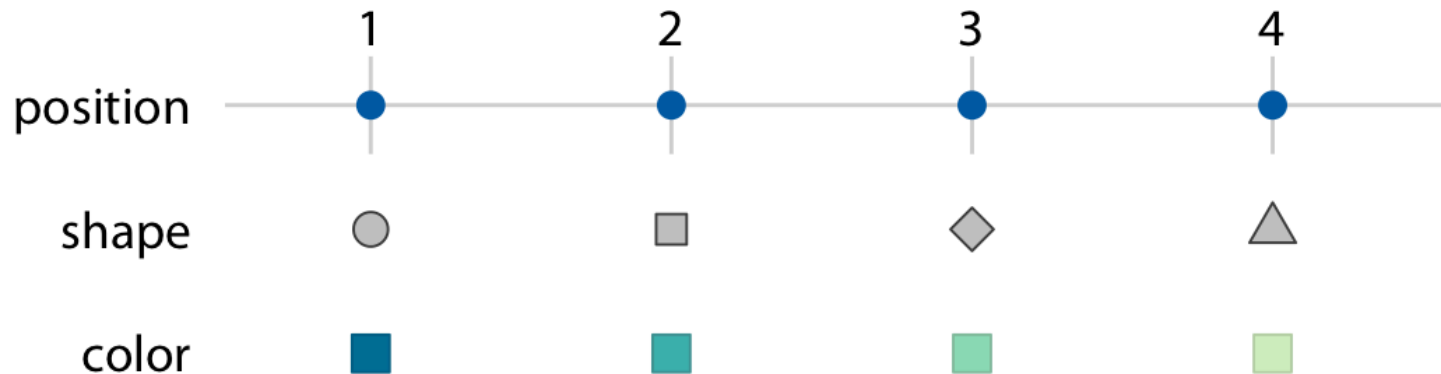
# **DATA VISUALIZATION AND VISUAL ANALYTICS**



# **SCALES FUNCTION**

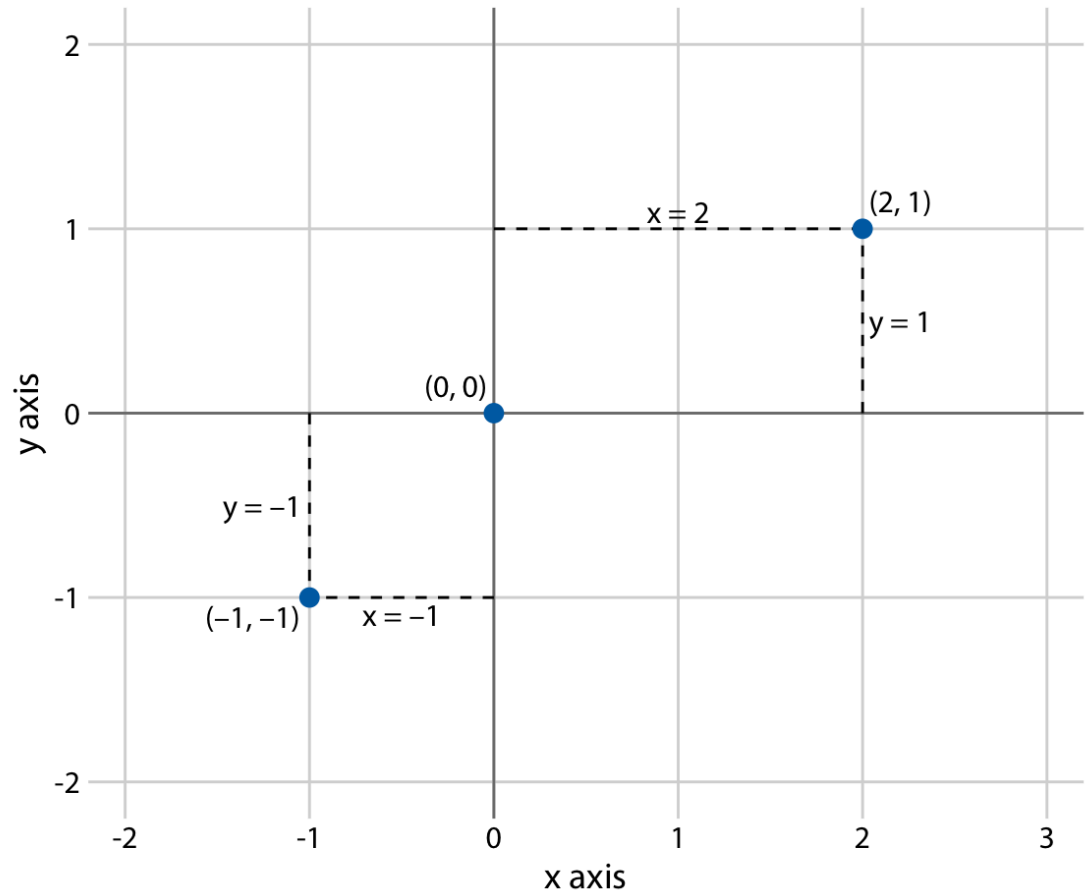
# Map data to VV

- We specify a scaling function to map data values to the visual representation
- A **scale** is a unique mapping between data and visual representation
- Scales are **functions** that map from an **input domain** to an **output range**

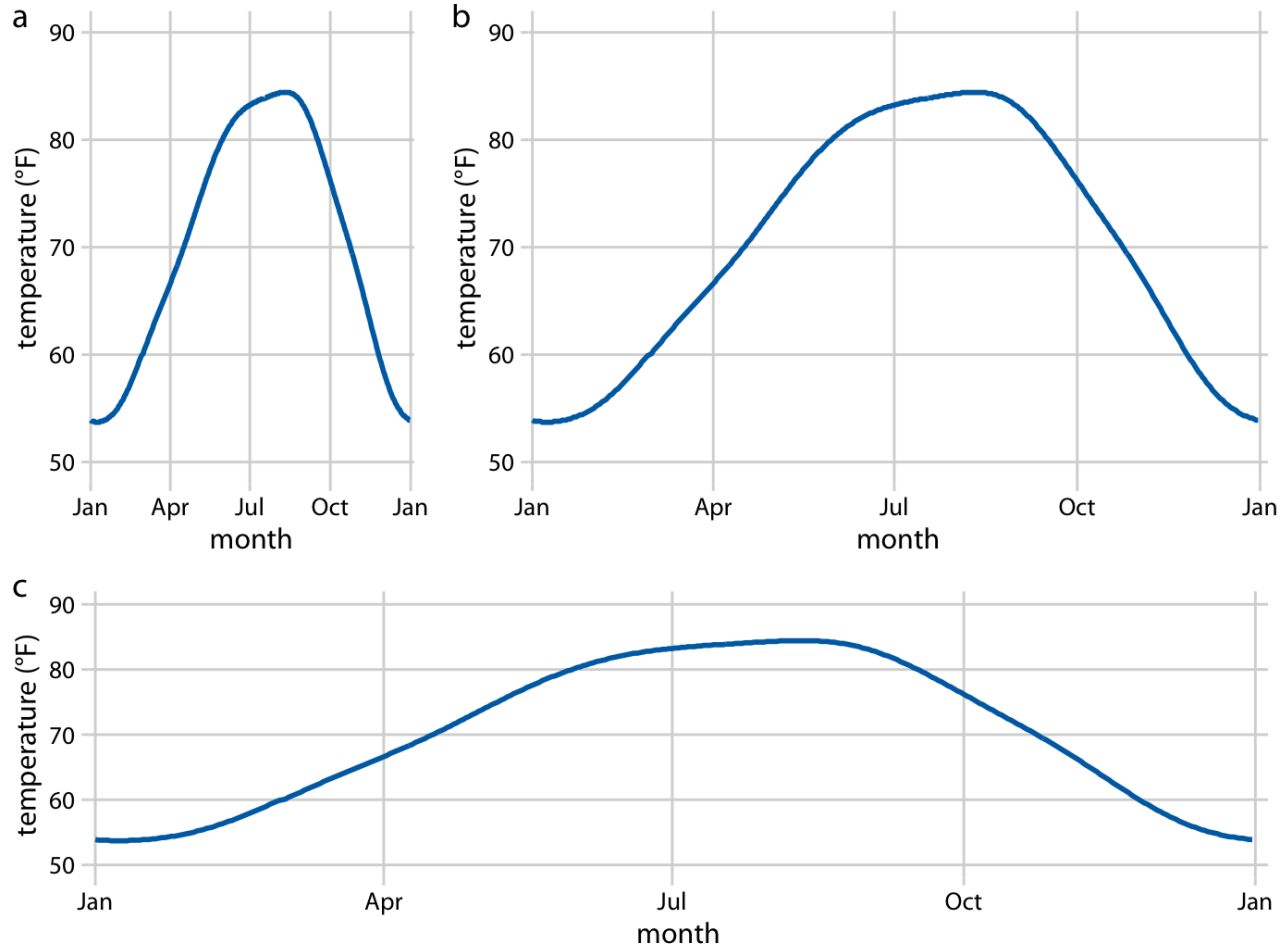


# Positional scales: axis

- Axis are at the base of many scientific plots
- Cartesian coordinate systems are composed of two orthogonal axis
- Values are positioned proportionally on the axes

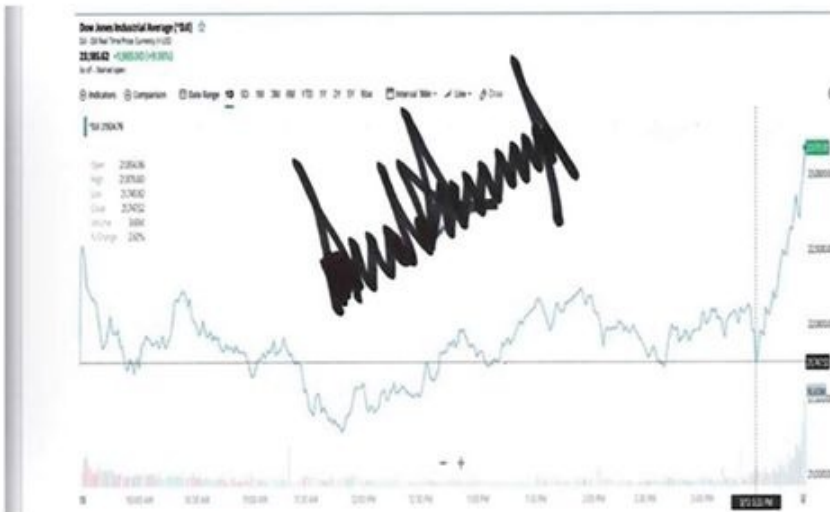


# Cartesian diagram with different scales

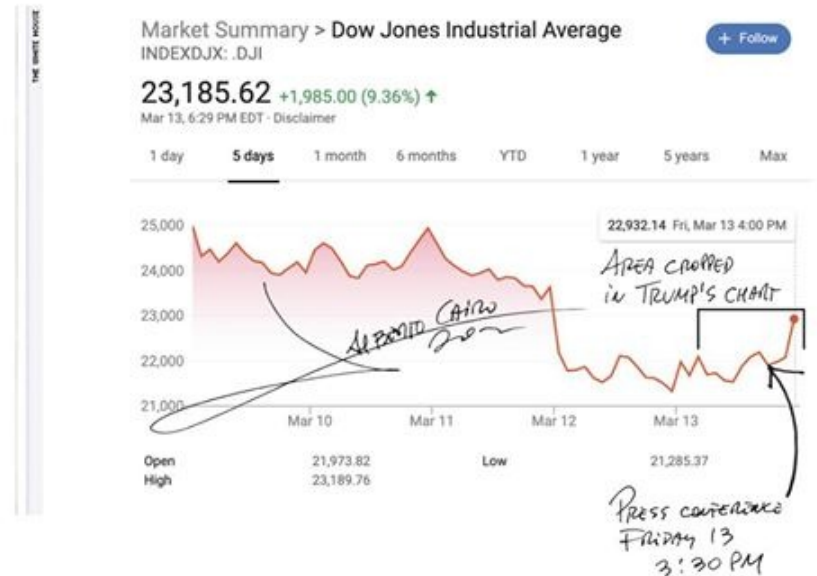


# Cartesian diagram with different scales

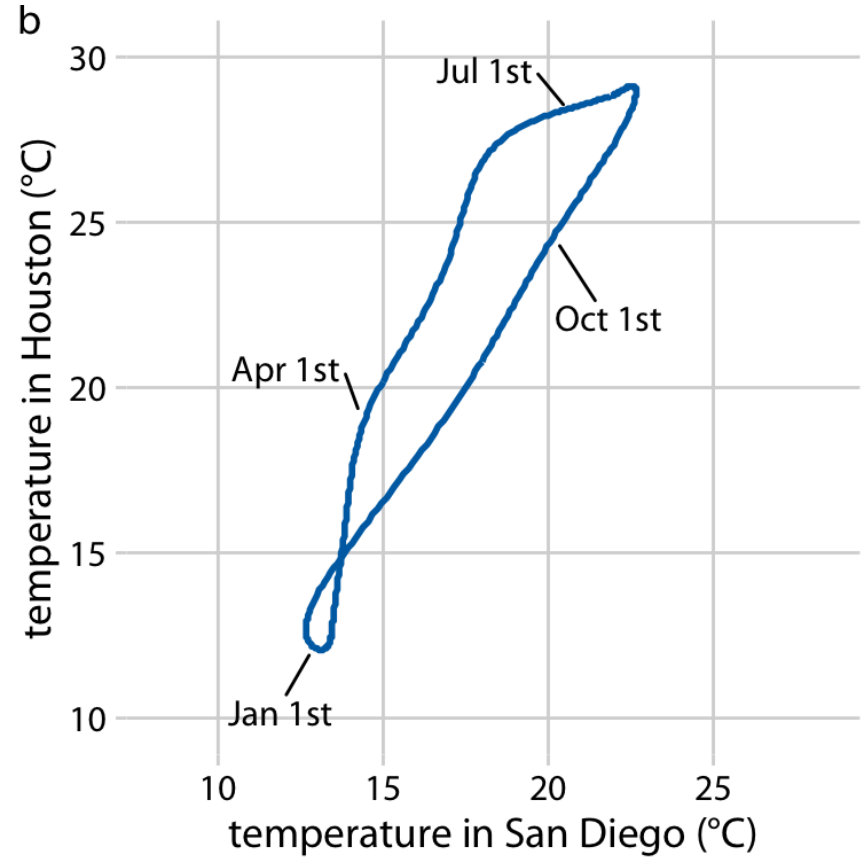
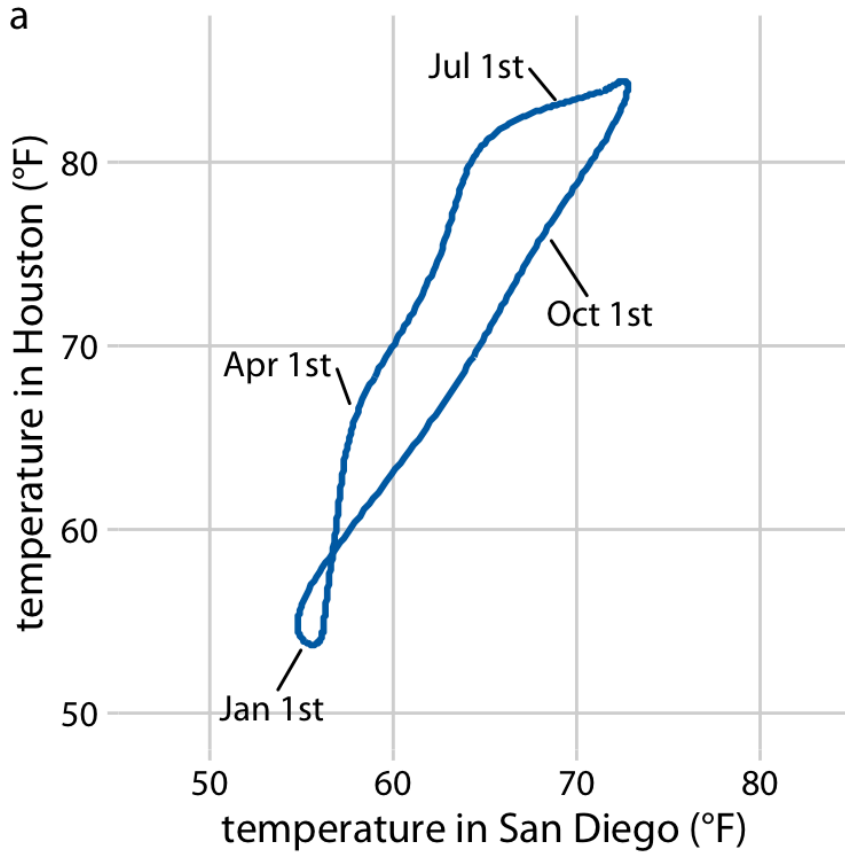
## CHART THAT TRUMP SIGNED AND SENT TO HIS FOLLOWERS



## CHART THAT I SIGNED AND SENT TO MY FOLLOWERS

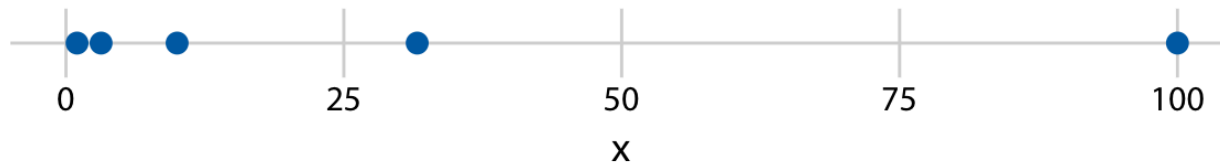


# Cartesian axes with same scale

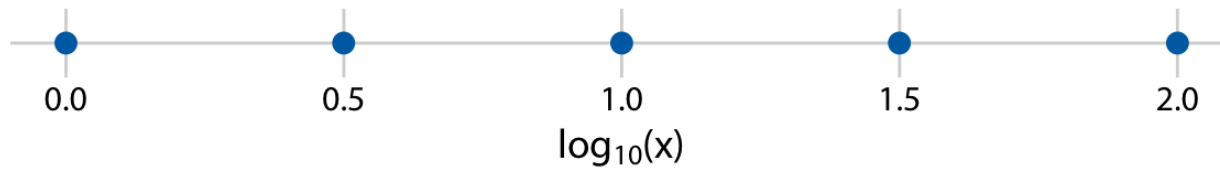


# Non linear axes

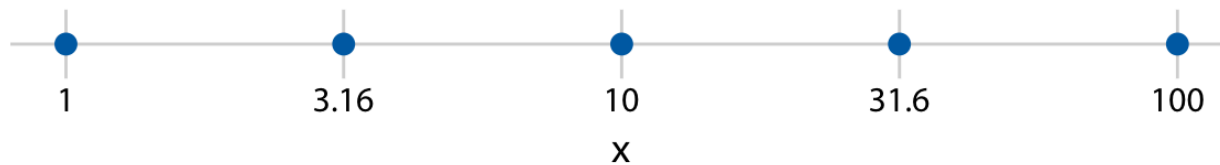
original data, linear scale



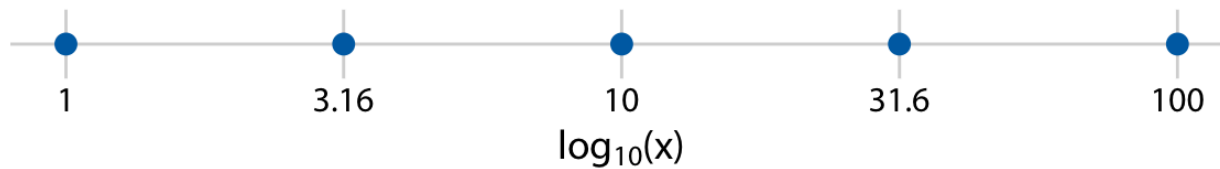
log-transformed data, linear scale



original data, logarithmic scale

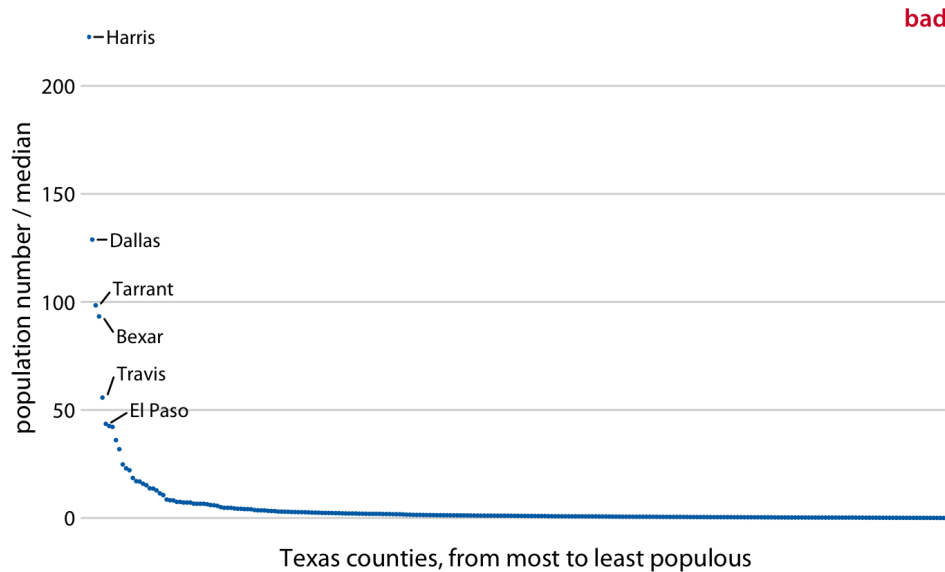
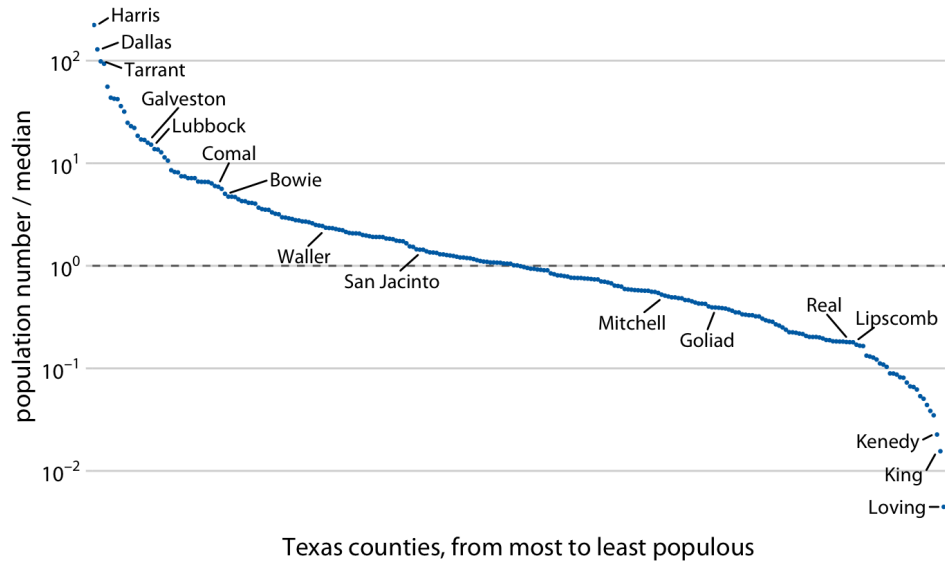


logarithmic scale with incorrect axis title

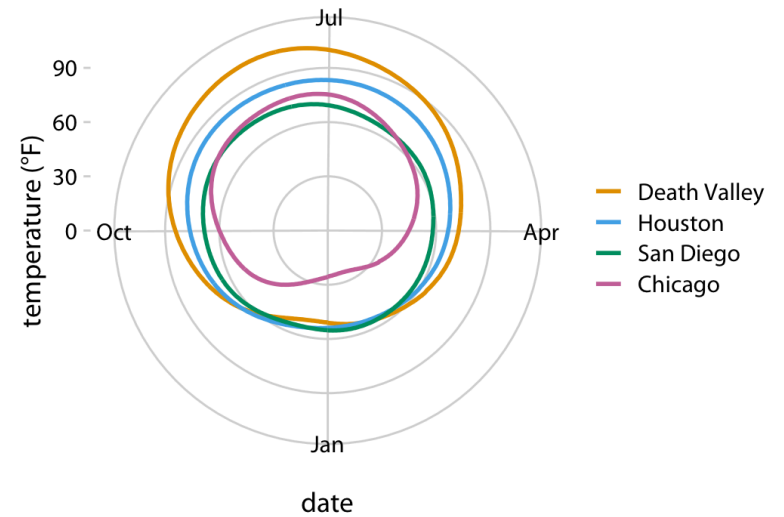
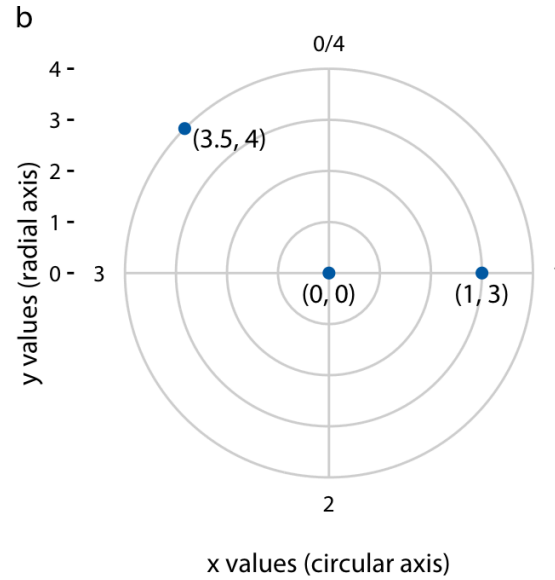
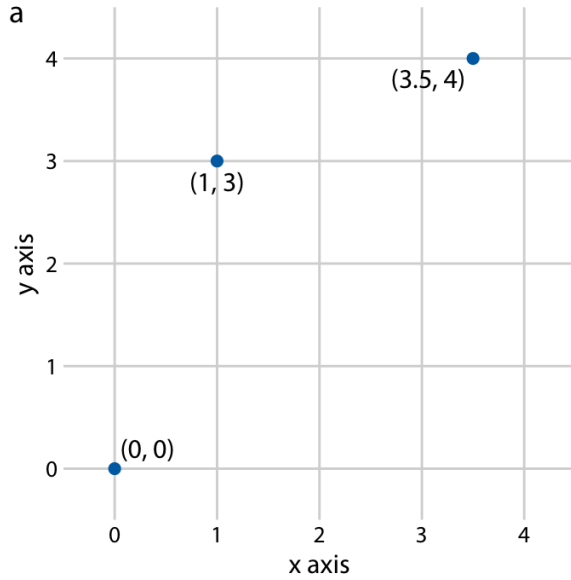




# Non linera axes



# Curved axes



# Example

Table 2.2: First 12 rows of a dataset listing daily temperature normals for four weather stations. Data source: NOAA.

Month	Day	Location	Station ID	Temperature
Jan	1	Chicago	USW00014819	25.6
Jan	1	San Diego	USW00093107	55.2
Jan	1	Houston	USW00012918	53.9
Jan	1	Death Valley	USC00042319	51.0
Jan	2	Chicago	USW00014819	25.5
Jan	2	San Diego	USW00093107	55.3
Jan	2	Houston	USW00012918	53.8
Jan	2	Death Valley	USC00042319	51.2
Jan	3	Chicago	USW00014819	25.3
Jan	3	San Diego	USW00093107	55.3
Jan	3	Death Valley	USC00042319	51.3
Jan	3	Houston	USW00012918	53.8

Ordinal

Ordinal

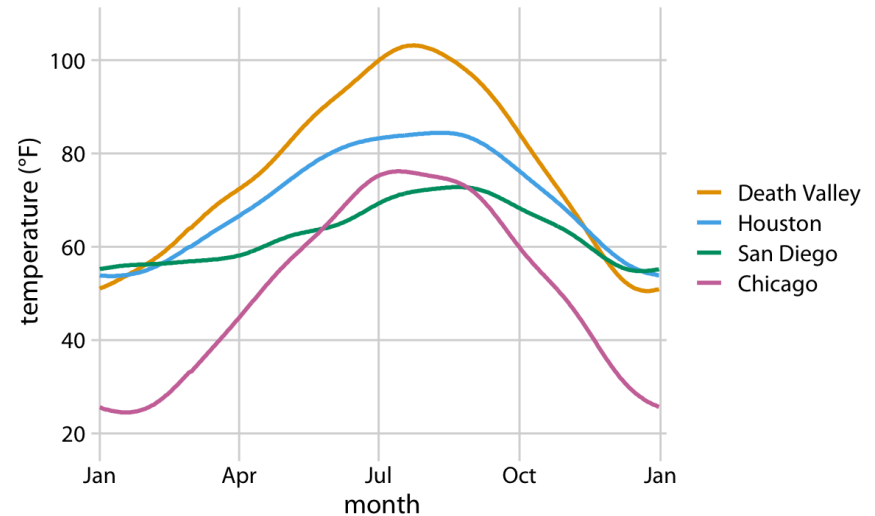
Nominal

Nominal

Quantitative

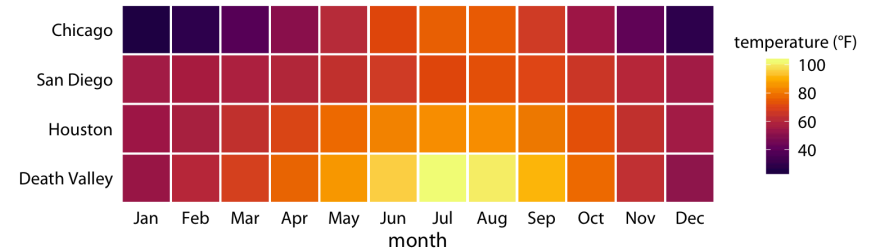
# Example

- Temperature (quantitative) on a linear axis (y)
- Month and day (ordinal) on a linear axis (x)
- City (nominal) on a color hue scale



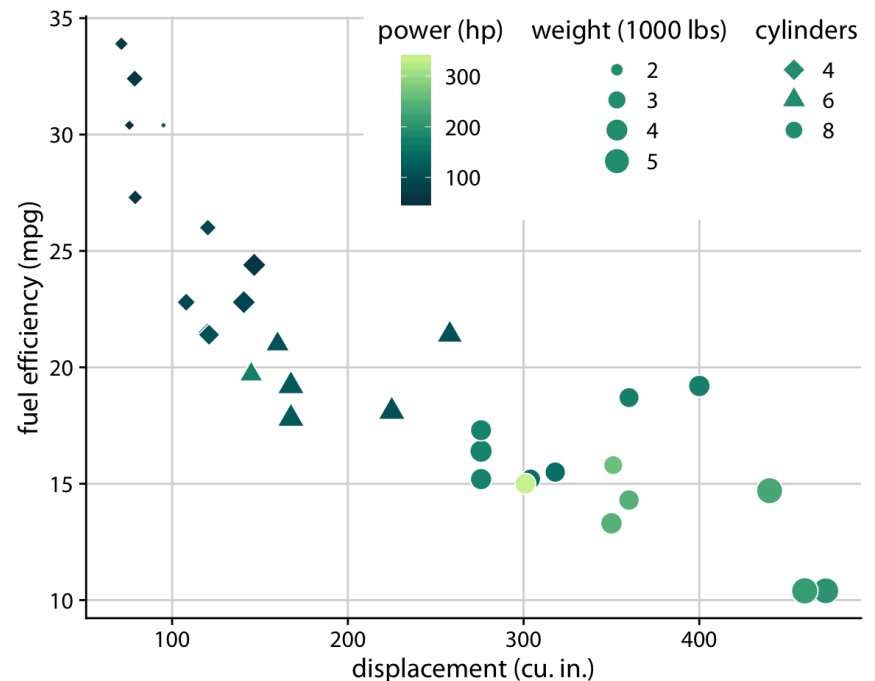
# Example

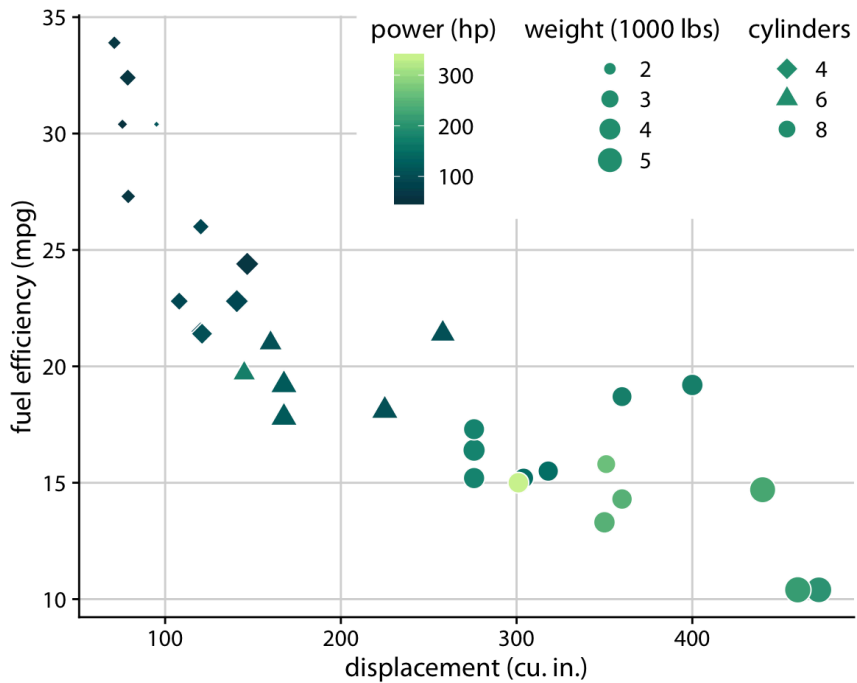
- Month (ordinal) on a ordinal axis (x)
- City (nominal) on a ordinal axis (y) (order determined on sum of temperatures on the line)
- Temperature (quantitative) on a color scale



# Example

- Displacement (quantitative) on linear axis (x)
- Fuel efficiency (quantitative) on linear axis (y)
- Power (quantitative) on lineal color scale
- Weight (quantitative -> ordinal) on ~~linear~~ squared size scale
- Cylinders (ordinal -> nominal) on shape scale

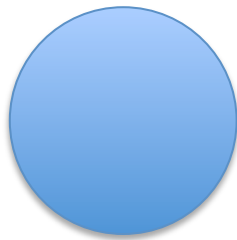




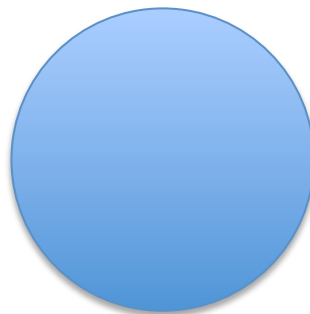
weight (1000 lbs)



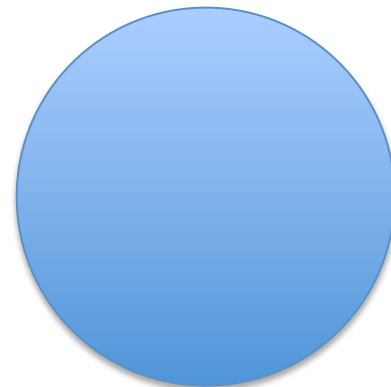
$r=2$   
 $A = 4 \cdot \pi$



$r=3$   
 $A = 9 \cdot \pi$



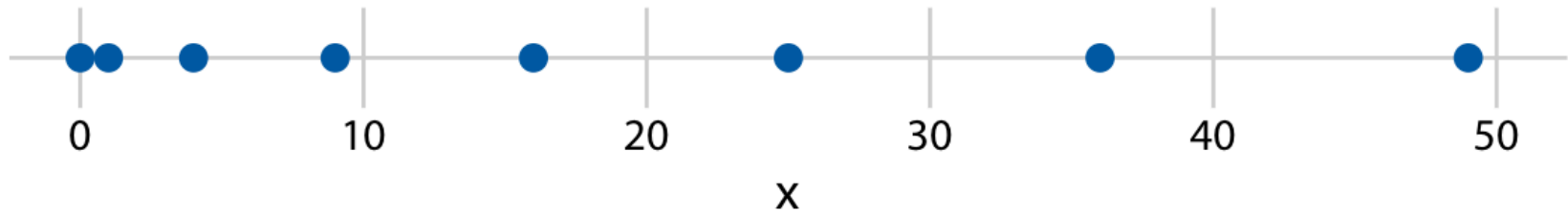
$r=4$   
 $A = 16 \cdot \pi$



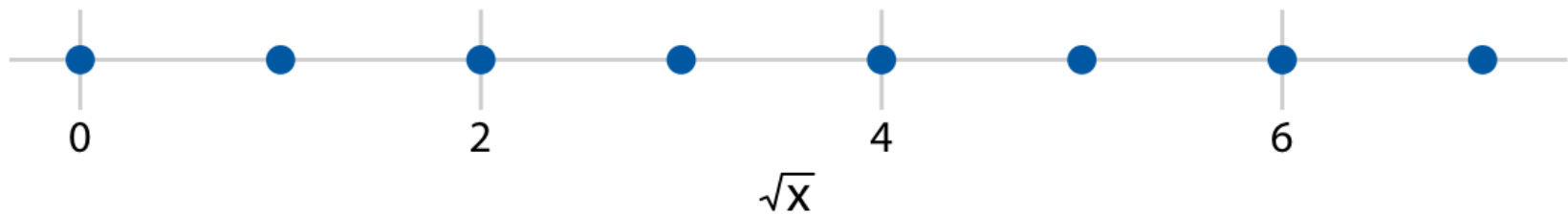
$r=5$   
 $A = 25 \cdot \pi$

# Non linear axes

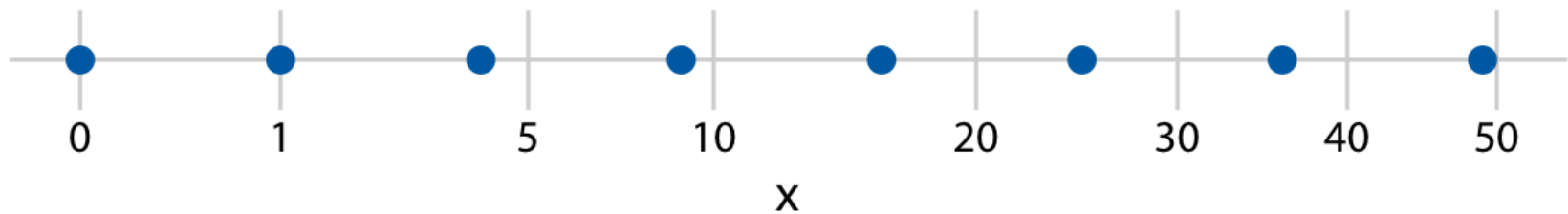
original data, linear scale



square-root-transformed data, linear scale



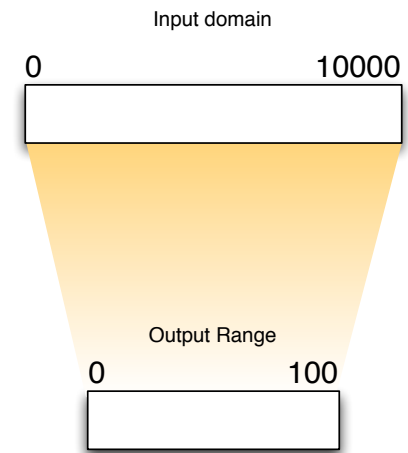
original data, square-root scale





# Manual Mapping

1. For input domain
  1. Select the largest number in original interval (10000)
  2. Select the smallest number in original interval (0)
  3. Select the difference of the two values (10000)
2. For output range
  1. Select the largest number in the new interval (100)
  2. Select the minimum number in the new interval (0)
  3. Select the difference of the two values (100)
3. Compute the ratio of the two intervals' range (10000/100 = 100)
  - This is an example of a linear scaling
    - $y = mx + b$ , where  $b=0$  and  $m=1/100$
    - 100 units in the original interval correspond to 1 unit in the destination interval



# An example – an alternative solution

```
join.enter()  
  .append("rect")  
  .attr("x", function(d,i){  
    return i*barw;  
  })  
  .attr("y",function(d){  
    return height - d*4;  
  })  
  .attr("width", barw)  
  .attr("height",function(d){  
    return d*4;  
  });
```

```
join.enter()  
  .append("rect")  
  .attr("x", function(d,i){  
    return x(i);  
  })  
  .attr("y",function(d){  
    return y(d);  
  })  
  .attr("width", barw)  
  .attr("height",function(d){  
    return h(d);  
  });  
function x(d){return m*d + b};  
function y(d){return m'*d + b'};  
function h(d){return m''*d + b''};
```

# D3.js Scales generator

- D3 provides several scale types
  - Quantitative
    - Continuous
      - Identity
      - Linear ( $y=mx+b$ )
      - Power ( $y=mx^k+b$ )
      - Log ( $y=m \log(x) + b$ )
    - Discrete
      - Quantize
      - Quantile
      - Threshold
  - Ordinal
  - Time

## Creating a scale

```
var scale = d3.scaleLinear();
```

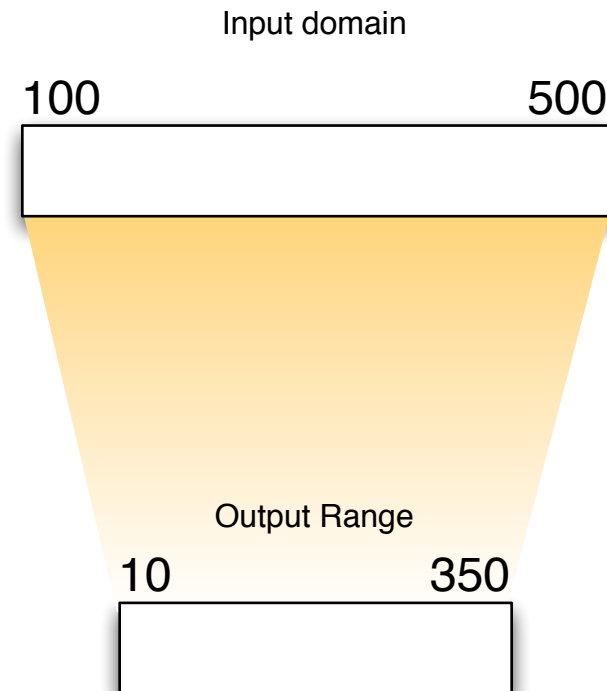
- Default scale uses
  - Domain is [0,1]
  - Range is [0,1]
  - Function is Identity
  - `scale(2.5); //returns 2.5`

# Creating a scale – setting domain and range

```
var scale = d3.scaleLinear()  
  .domain([100, 500])  
  .range([10, 350]);
```

- Default scale uses

- `scale(100); //returns 10`
- `scale(300); //returns 180`
- `scale(500); //returns 350`



# Quantitative power scale – circle radius

## Previous example

```
g.append("circle")
  .attr("fill", "pink")
  .attr("stroke", "red")
  .attr("r", function(d){
    return Math.sqrt(d*100);
  })
```

## Refined solution

```
var r = d3.scaleSqrt()
  .domain([0,20])
  .range([0,30]);

g.append("circle")
  .attr("fill", "pink")
  .attr("stroke", "red")
  .attr("r", function(d){
    return r(d);
  })
```

## Utility functions: `d3.min`, `d3.max`, `d3.extent`

- To determine the domain and range interval we should know min and max of the two intervals
- D3.js provides utility functions to access such values
  - `d3.min(array[, accessor])`
  - `d3.max(array[, accessor])`
  - `d3.extent(array[, accessor])`

# Utility Functions: examples

```
d3.min([10, 30, 40, 70, 100]) //returns 10  
d3.max([10, 30, 40, 70, 100]) //returns 100  
d3.extent([10, 30, 40, 70, 100]) //returns  
[10, 100]
```



# Domains & Ranges

- Typically, domains are derived from data while ranges are constant.

```
var x = d3.scaleLinear()  
  .domain([0, d3.max(numbers)])  
  .range([0, 720]);
```

```
var x = d3.scaleLog()  
  .domain(d3.extent(numbers))  
  .range([0, 720]);
```

```
function value(d) { return d.value; }  
  
var x = d3.scaleLog()  
  .domain(d3.extent(objects, value))  
  .range([0, 720]);
```

# ObservableHQ – Introduction to D3.js Scales

Observable

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New



D3 d3js.org  
Bring your data to life.  
By Fil

♡  
5

Published Jun 24, 2019 1 Fork Listed in d3-scale

## Introduction to D3's scales

When, on a print map, 1 cm figures a real distance of 1 km on the terrain, we say that the map has a 1:100,000 scale.

But scales are not limited to a proportional ratio (or rule of three) between an actual distance and a length on paper. More generally, they describe how an actual dimension of the original data is to be represented as a visual variable. In this sense, scales are one of the most fundamental abstractions of data visualization.

Scales from the [d3-scale](#) module are functions that take as input the actual value of a measurement or property. Their output can in turn be used to encode a relevant representation.

$f(t)$

```
d3.scaleLinear()
```

A scale thus maps a physical quantity (or, more generally, an observation), which might be expressed in meters, kilograms, years or seconds, number of horses in a field... to a length or a radius (in screen pixels or print centimeters), a [color](#) (in CSS representation), a [shape](#)...

### Domain and range

A scale has to know from whence this observation comes — and this is called its

<https://observablehq.com/@d3/introduction-to-d3s-scales>

# ObservableHQ – Discrete Scales

```
jenks = f(n)
```

```
jenks = d3  
  .scaleThreshold()  
  .domain([53440, 135096])  
  .range(["white", "pink", "red"])
```

Data

scaleQuantile

scaleThreshold

scaleJenks

scaleQuantize

scaleQuantizeNice

```
showScaleGrouping(data, {  
  scaleQuantile: quantile,  
  scaleThreshold: threshold,  
  scaleJenks: jenks,  
  scaleQuantize: quantize,  
  scaleQuantizeNice: quantize.copy().nice()  
})
```

<https://observablehq.com/@d3/quantile-quantize-and-threshold-scales>

# ObservableHQ – Sequential Scales

```
angryRainbow(t0,
```



```
viewof t0 = ramp(angryRainbow)
```

*Note.* Besides demonstration purposes, this color scheme is not recommended. Its is not perceptually uniform (most people will see spikes around the yellow, light blue and pink colors – hence the “angry rainbow” nickname); [d3-scale-chromatic](#) provides better alternatives for [cyclical color scales](#), such as `d3.interpolateSinebow`:

```
"rgb(255, 64, 64)"
```

```
d3.interpolateSinebow(t1)
```



```
viewof t1 = ramp(d3.scaleSequential(d3.interpolateSinebow))
```


(All manners of sequential color scales are available in the [d3-scale-chromatic](#) module: [diverging](#), [single-hue](#), [multi-hue](#) and [cyclical](#).)

[`sequential.interpolator`](#) allows to read or modify the interpolator function  $f$  in an existing sequential scale. It can be useful to construct a scale iteratively... see [Curran Kelleher's block](#) for an example. We use it below to read a scale's interpolator and create a mirror image (by applying it to  $1-t$  instead of  $t$ ):



```
{
```

# ObservableHQ – Diverging Scales



```
chart(scaleAnomalyPuOr) // chart is defined in the Annex, below
```

While the “PuOr” (purple-orange) interpolator looks good, we’ll prefer in this case a blue (for negative) to red (for positive) color interpolator, passing through white (for neutral). The `interpolator` is a function that takes its inputs in  $[0,1]$ , and we’re free to create our own.

As D3 offers a standard “RdBu” diverging color interpolator, that goes from red to white to blue. Almost what we needed: we’ll just reverse it to blue-white-red, by applying it to  $(1-t)$  instead of  $t$ .

The interpolator can be given, in a shorthand notation, as an argument to `d3.scaleDiverging`, so our final code is:

```
scaleAnomaly = f(n)
scaleAnomaly = d3.scaleDiverging(t => d3.interpolateRdBu(1 - t))
  .domain([extent[0], 0, extent[1]])
```



```
chart(scaleAnomaly)
```

To be complete, the shorthand notation also accepts the domain as an optional first argument:



```
chart(d3.scaleDiverging([extent[0], 0, extent[1]], t => d3.interpolateRdBu(1 - t)))
```

Variations are another typical use case to visualize a value change on a map (in that

# ObservableHQ – Qualitative Scales

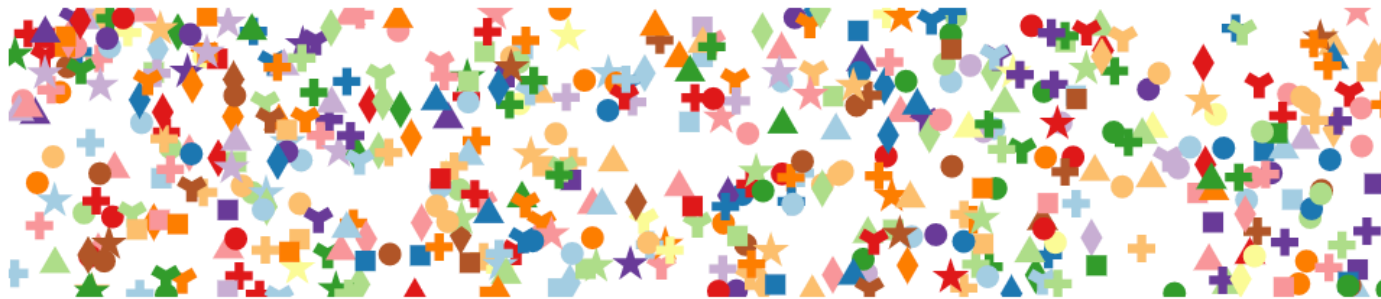
↕

## Colors & Symbols

Color palettes are a quite common use case for ordinal ranges. You are encouraged to create your own, by hand or using [one of the many tools available](#), but you can also use the list of color schemes provided by [d3-scale-chromatic](#).

A list of twelve words to explore d3 schemePaired color scheme !

A useful range for an ordinal scale can be a set of symbols that will be used to draw shapes, like for instance [d3.symbols](#).



```
{
  const symbols = d3.scaleOrdinal().range(d3.symbols),
    color = d3.scaleOrdinal(d3.schemePaired),
    height = 200,
    symbol = d3.symbol().size(200),
    data = d3.range(500).map(i => ({
      x: width * Math.random(),
      y: height * Math.random(),
      s: Math.floor(9 * Math.random()),
    }));
}
```

<https://observablehq.com/@d3/d3-scaleordinal>