BIMM 143

Data visualization with R

Lecture 5

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UC San Diego

http://thegrantlab.org/bimm143

Recap From Last Time:

- What is R and why should we use it?
- Familiarity with R's basic syntax.
- Familiarity with major R data structures namely **vectors** and **data.frames**.
- Understand the basics of using **functions** (arguments, vectorizion and re-cycling).
- Appreciate how you can use R scripts to aid with reproducibility.

[MPA Link]

Today’s Learning Goals

- Appreciate the major elements of **exploratory data analysis** and why it is important to visualize data.

- Be conversant with **data visualization best practices** and understand how good visualizations optimize for the human visual system.

- Be able to generate informative graphical displays including **scatterplots**, **histograms**, **bar graphs**, **boxplots**, **dendrograms** and **heatmaps** and thereby gain exposure to the extensive graphical capabilities of R.

- Appreciate that you can build even more complex charts with **ggplot** and additional R packages such as **rgl**.

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Why visualize at all?

Over-the-Counter

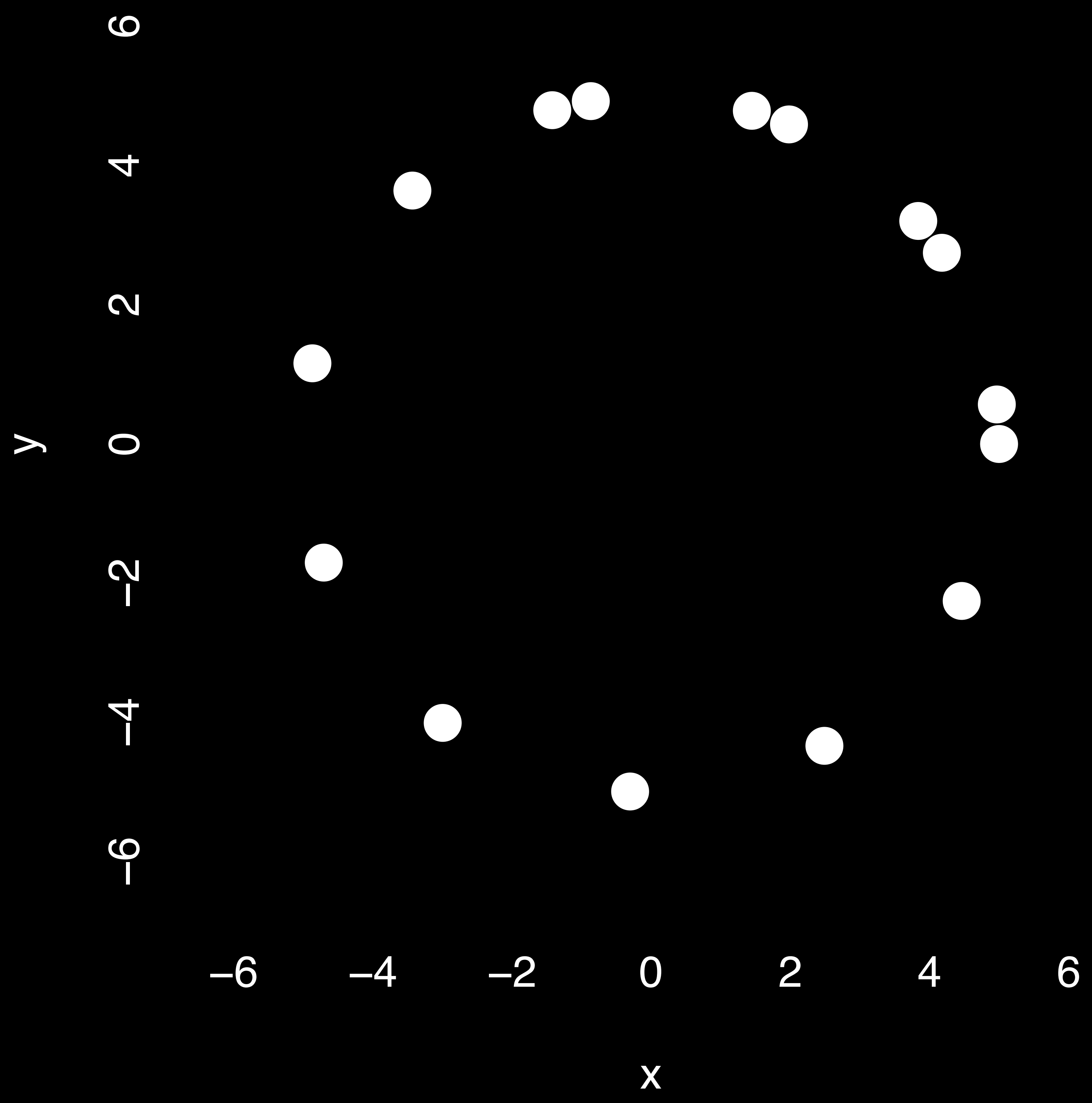
National Market System

The companies listed below reflect the volume in 100's of shares on a daily basis and the closing price and net change are reflected from the previous day based on trades as provided under the NASD National Market System.

Stock		Sales	Close	Chg.
[illegible]	[illegible]	[illegible]	[illegible]	[illegible]

	x	y
1	5.00	0.00
2	4.18	2.75
3	1.98	4.59
4	-0.86	4.92
5	-3.43	3.64
6	-4.86	1.16
7	-4.70	-1.70
8	-2.99	-4.01
9	-0.30	-4.99
10	2.49	-4.34
11	4.46	-2.25
12	4.97	0.57
13	3.84	3.20
14	1.45	4.79
15	-1.42	4.79

	x	y
Min.	-4.86	-4.99
1st Qu.	-2.21	-1.98
Median	1.45	1.16
Mean	0.65	0.87
3rd Qu.	4.01	4.12
Max.	5.00	4.92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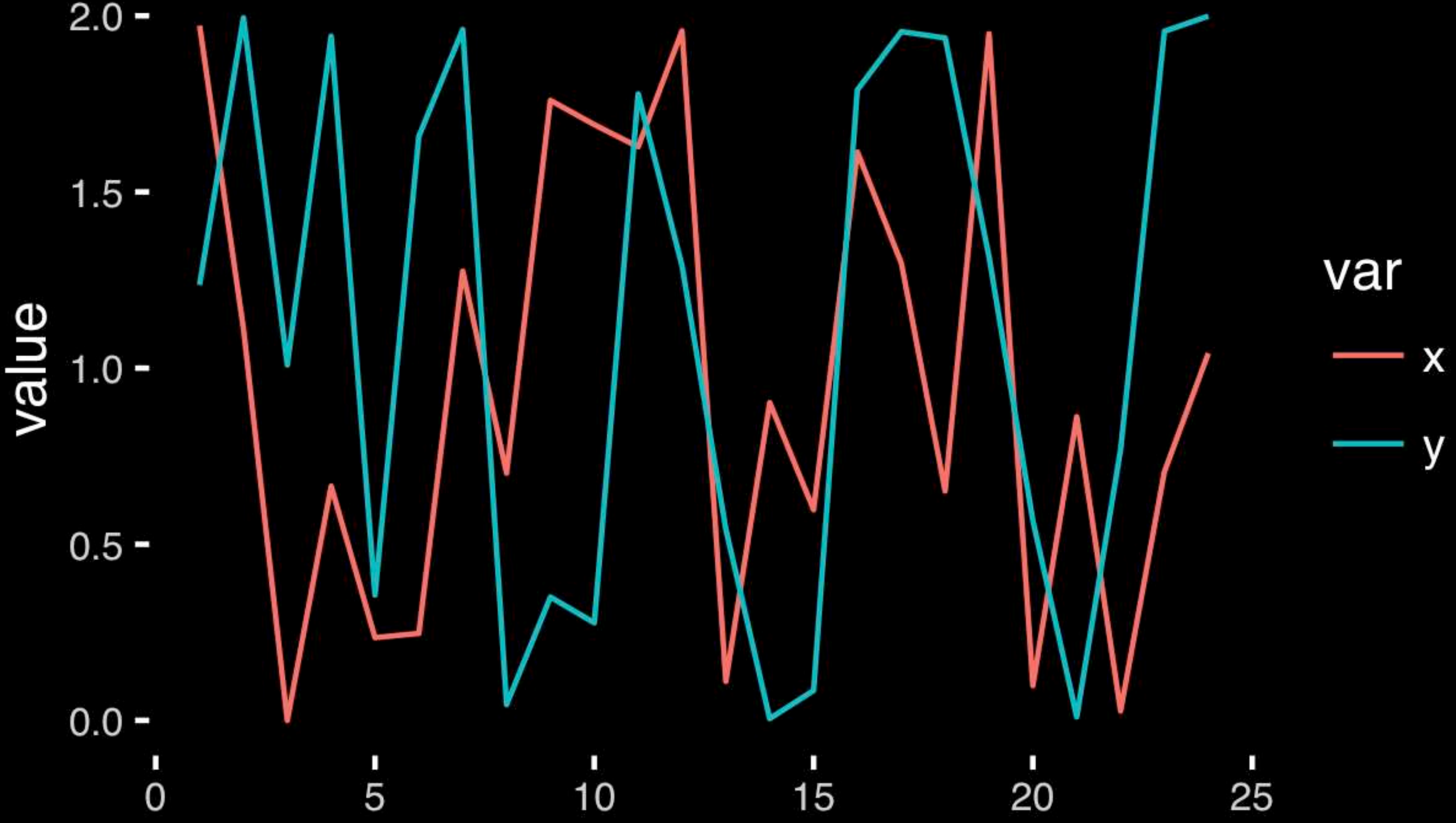
2.0
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1.0
0.5
0.0
value
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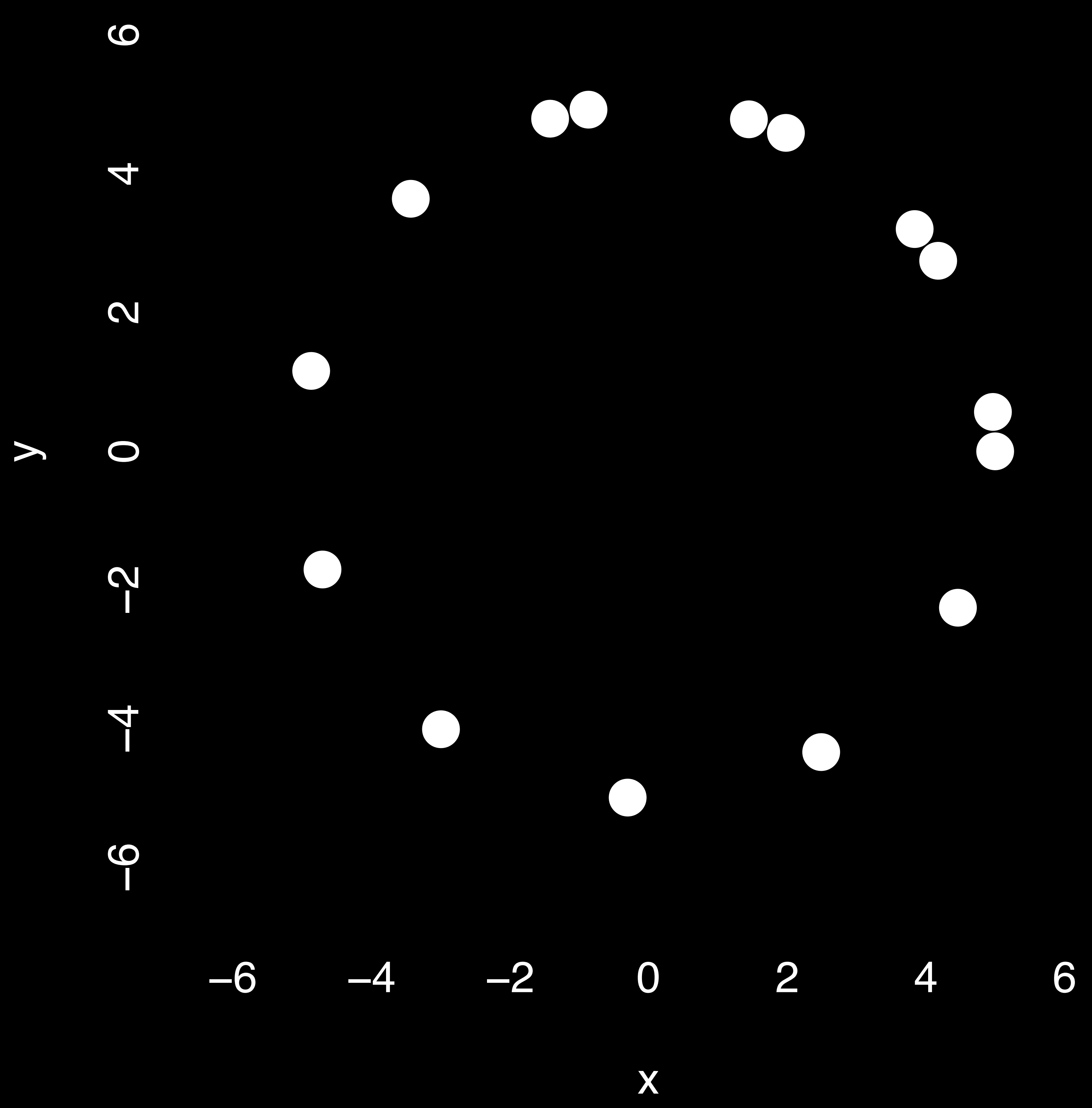
10
15
20
25
var
x
y

https://bioboot.github.io/bimm143_F18/class-material/05_draw_circle_points/

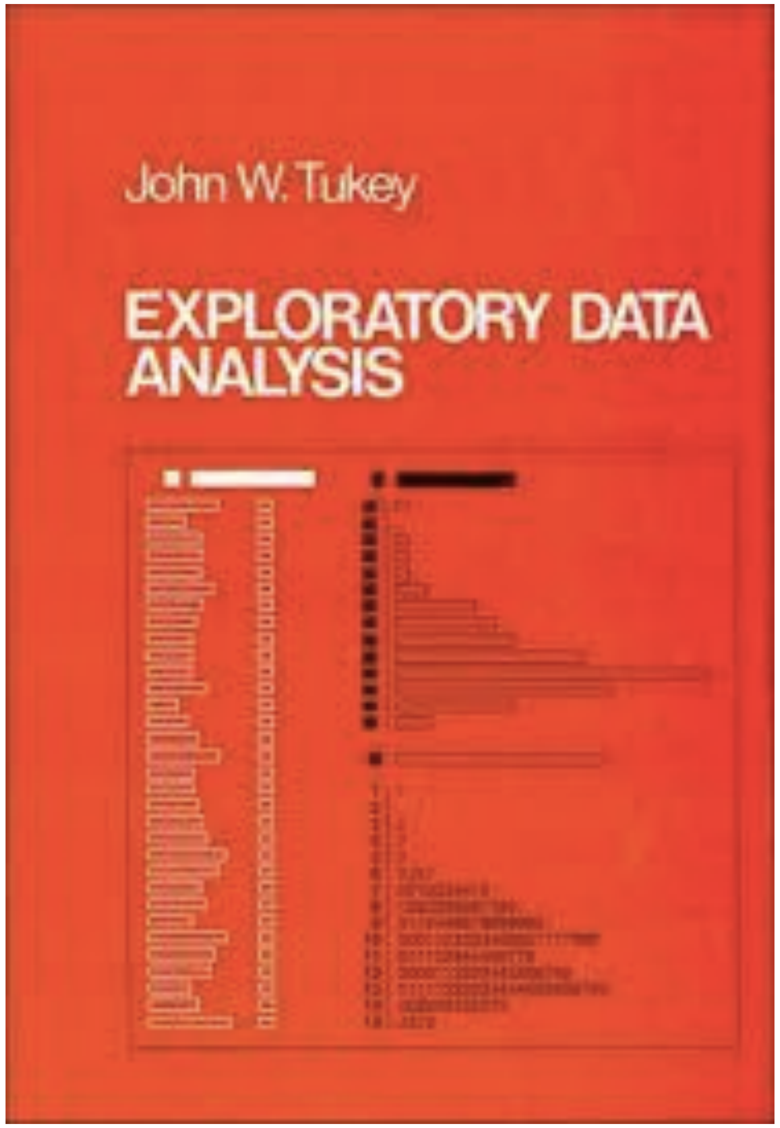


Exploratory Data Analysis

- ALWAYS look at your data!
- If you can't see it, then don't believe it!
- Exploratory Data Analysis (EDA) allows us to:
    1. Visualize distributions and relationships
    2. Detect errors
    3. Assess assumptions for confirmatory analysis
- EDA is the first step of data analysis!

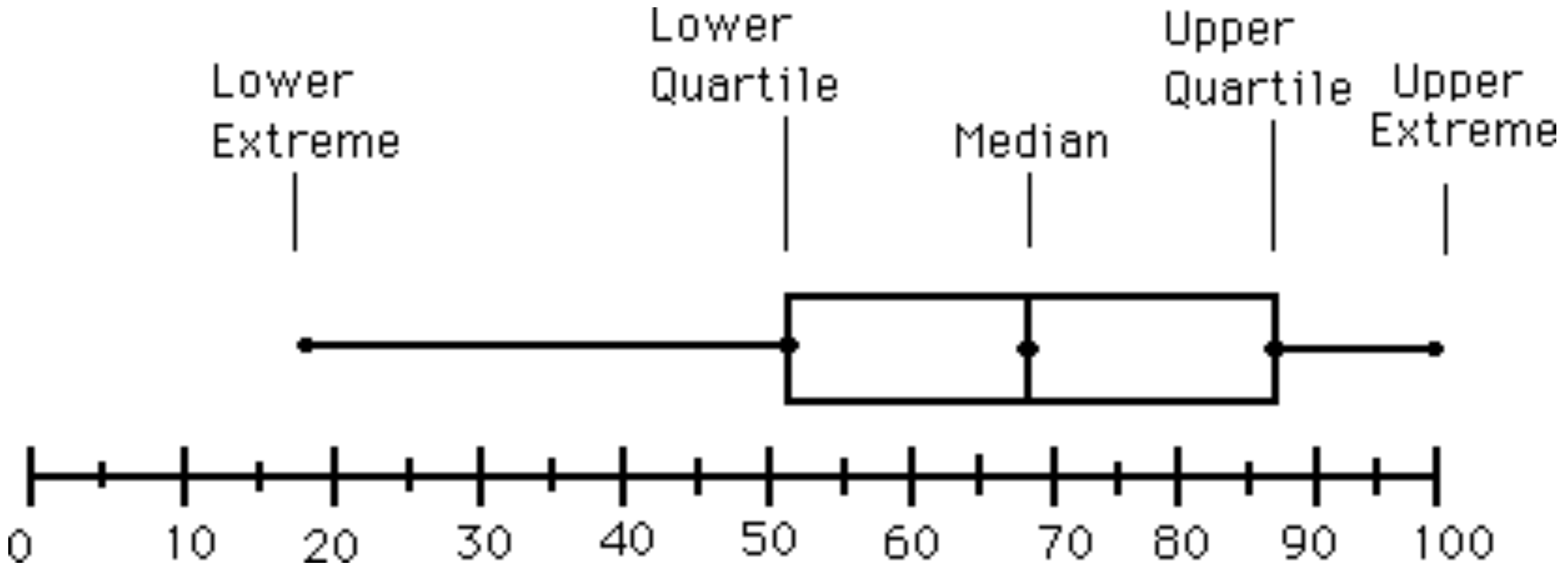
Exploratory Data Analysis 1977

- Based on insights developed at Bell Labs in the 60’s
- Techniques for visualizing and summarizing data
- What can the data tell us? (in contrast to “confirmatory” data analysis)
- Introduced many basic techniques:
  - 5-number summary, box plots, stem and leaf diagrams,…
- 5 Number summary:
  - extremes (min and max)
  - median & quartiles
  - More robust to skewed & longtailed distributions



Side-note: **boxplots**

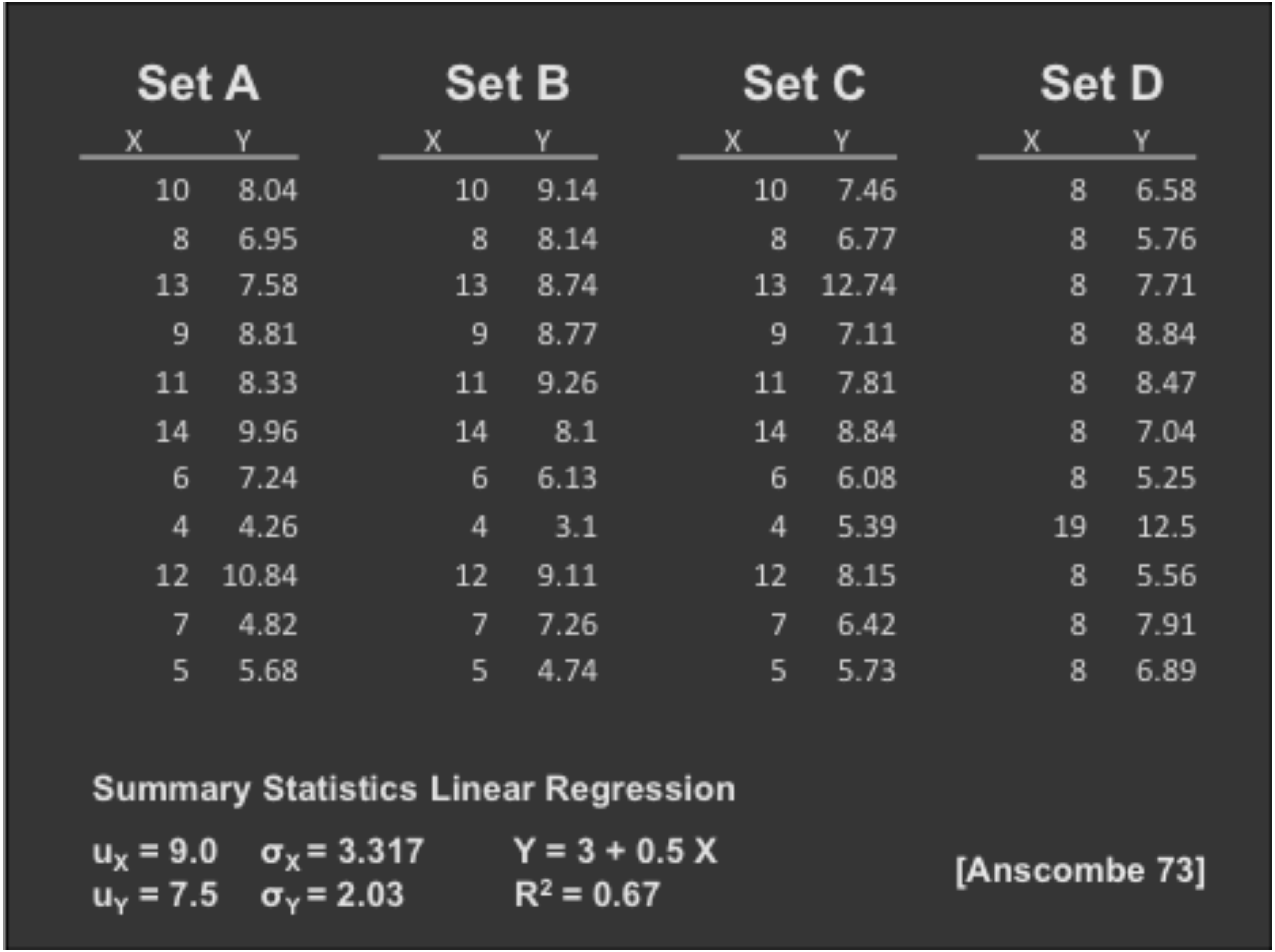
- **Box-and-whisker plot** : a graphical form of 5-number summary (Tukey)



boxplot(rnorm(1000,0))

summary(); hist()

The Trouble with Summary Stats



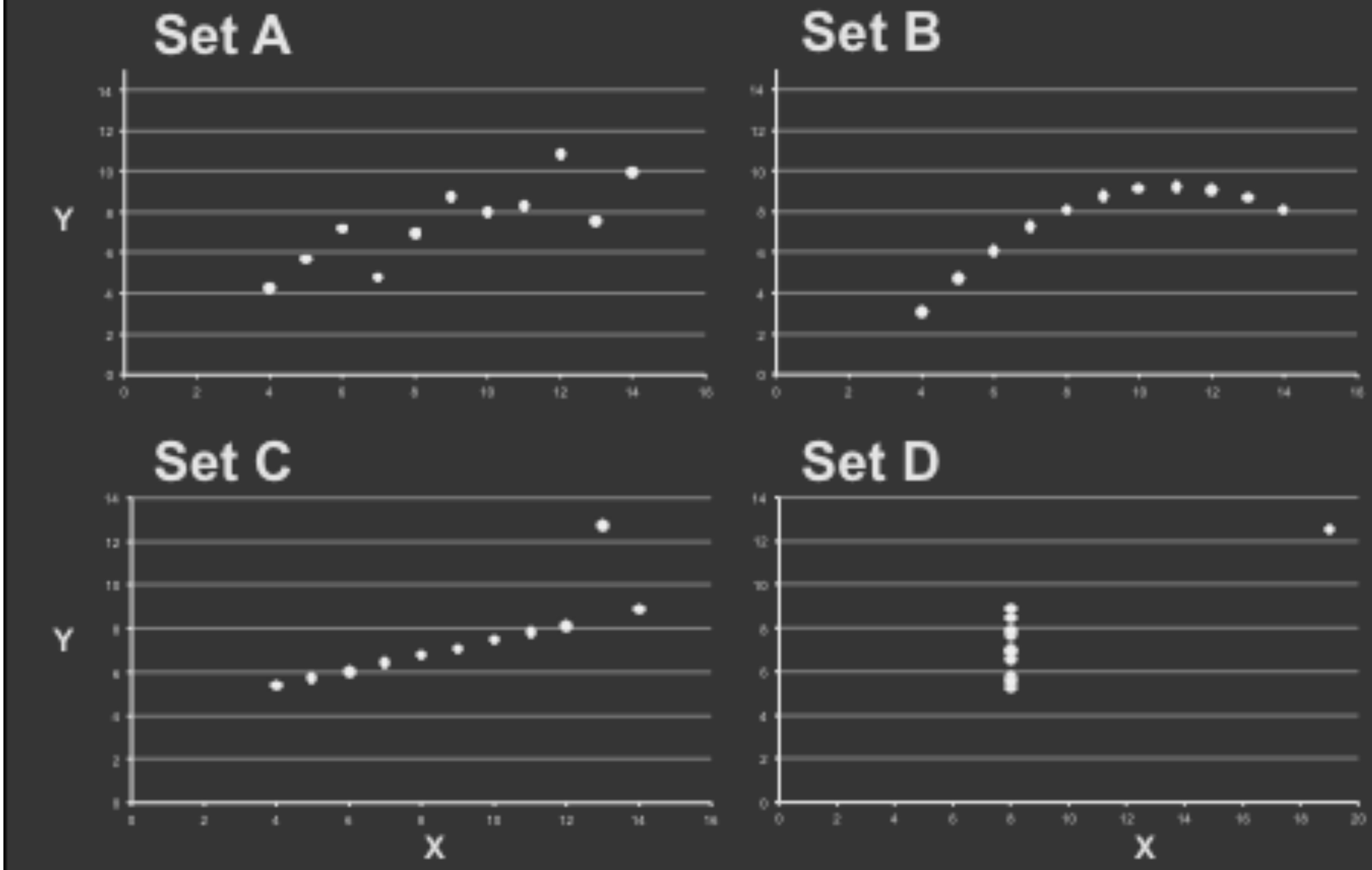
Set A		Set B		Set C		Set D	
X	Y	X	Y	X	Y	X	Y
10	8.04	10	9.14	10	7.46	8	6.58
8	6.95	8	8.14	8	6.77	8	5.76
13	7.58	13	8.74	13	12.74	8	7.71
9	8.81	9	8.77	9	7.11	8	8.84
11	8.33	11	9.26	11	7.81	8	8.47
14	9.96	14	8.1	14	8.84	8	7.04
6	7.24	6	6.13	6	6.08	8	5.25
4	4.26	4	3.1	4	5.39	19	12.5
12	10.84	12	9.11	12	8.15	8	5.56
7	4.82	7	7.26	7	6.42	8	7.91
5	5.68	5	4.74	5	5.73	8	6.89

Summary Statistics Linear Regression

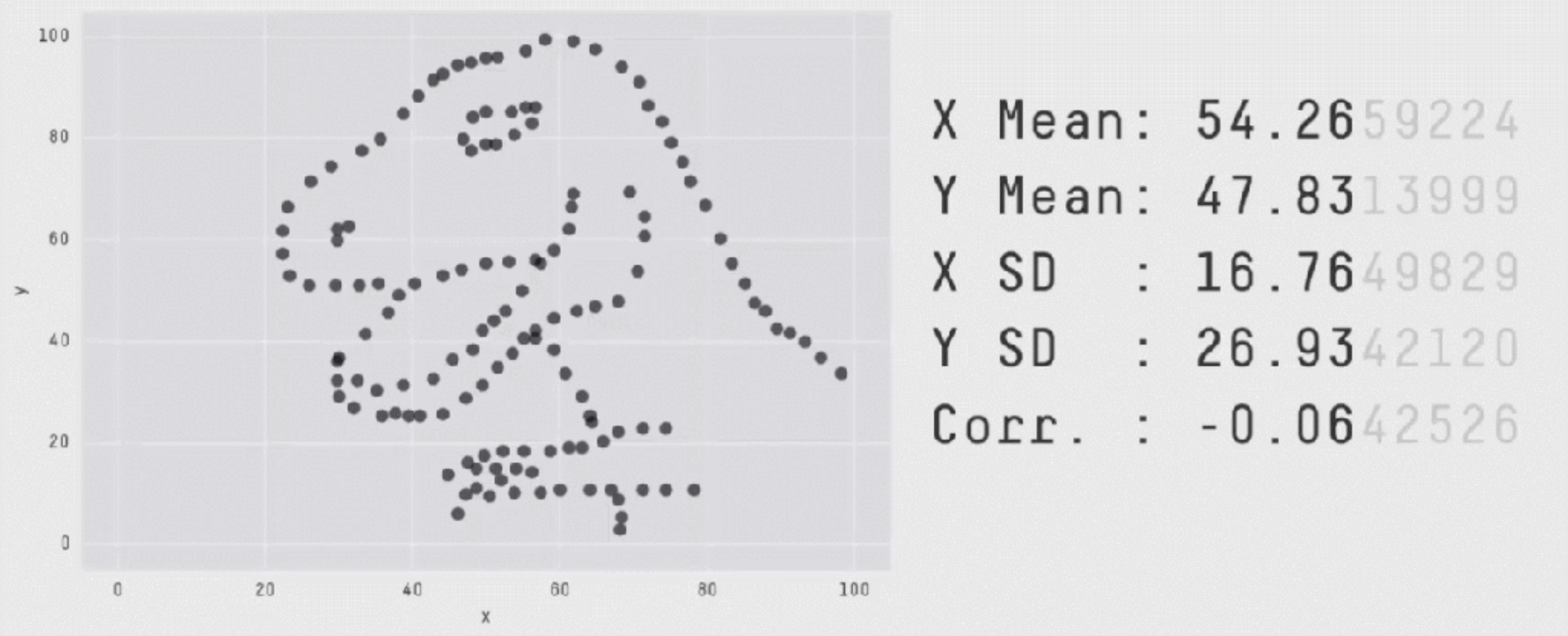
$u_X = 9.0$ $\sigma_X = 3.317$ $Y = 3 + 0.5\,X$
$u_Y = 7.5$ $\sigma_Y = 2.03$ $R^2 = 0.67$

[Anscombe 73]

Looking at Data



https://en.wikipedia.org/wiki/Anscombe%27s_quartet



```
X Mean: 54.2659224
Y Mean: 47.8313999
X SD  : 16.7649829
Y SD  : 26.9342120
Corr. : -0.0642526
```

Key point: You need to visualize your data!

https://github.com/stephlocke/datasauRus

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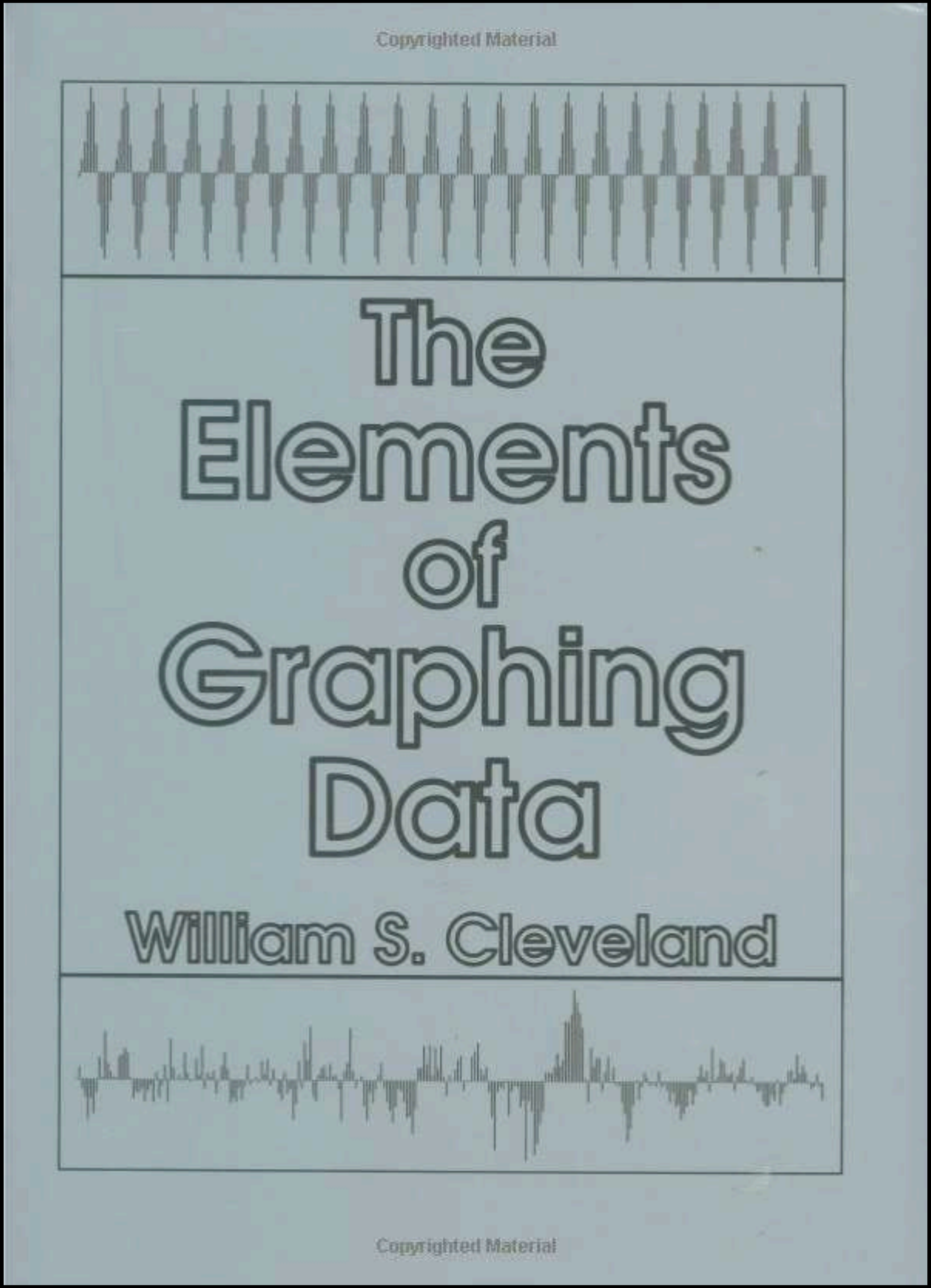
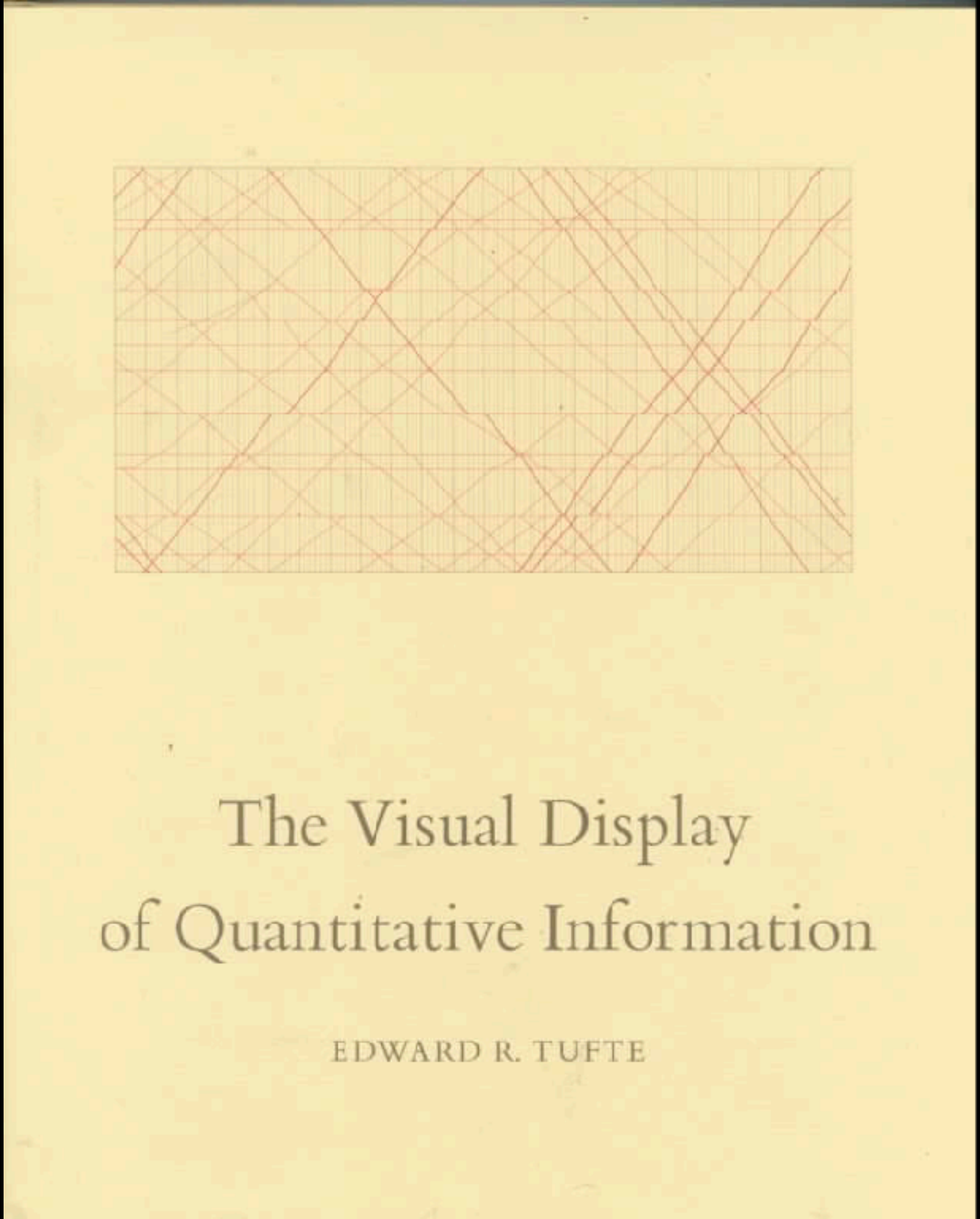


The Elements of Graphing Data

William S. Cleveland



The Visual Display of Quantitative Information

EDWARD R. TUFTE

Key Point:

Good visualizations optimize for the human visual system.

Key Point: The most important measurement should exploit the highest ranked encoding possible

- Position along a common scale
- Position on identical but nonaligned scales
- Length
- Angle or Slope
- Area
- Volume or Density or Color saturation/hue

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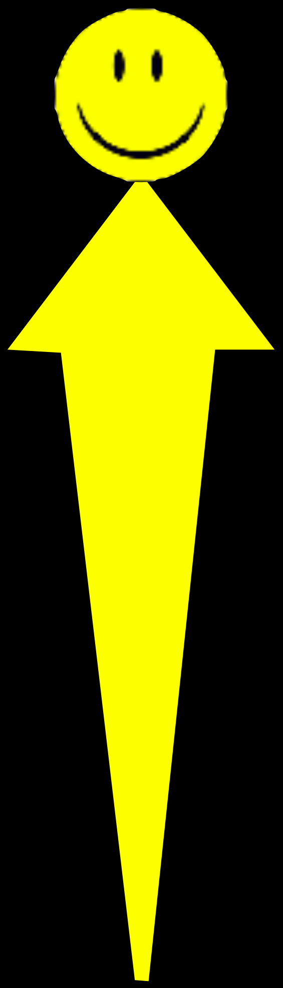

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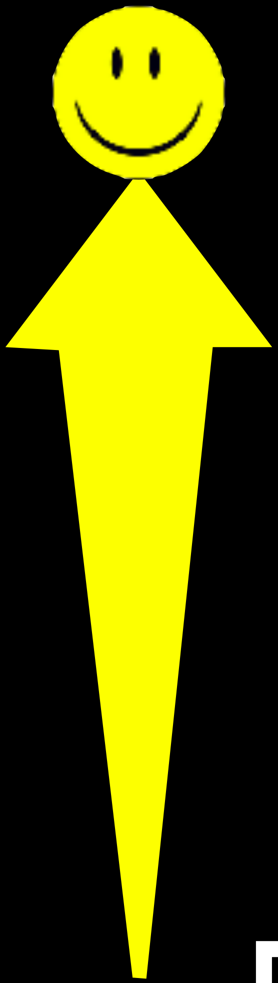

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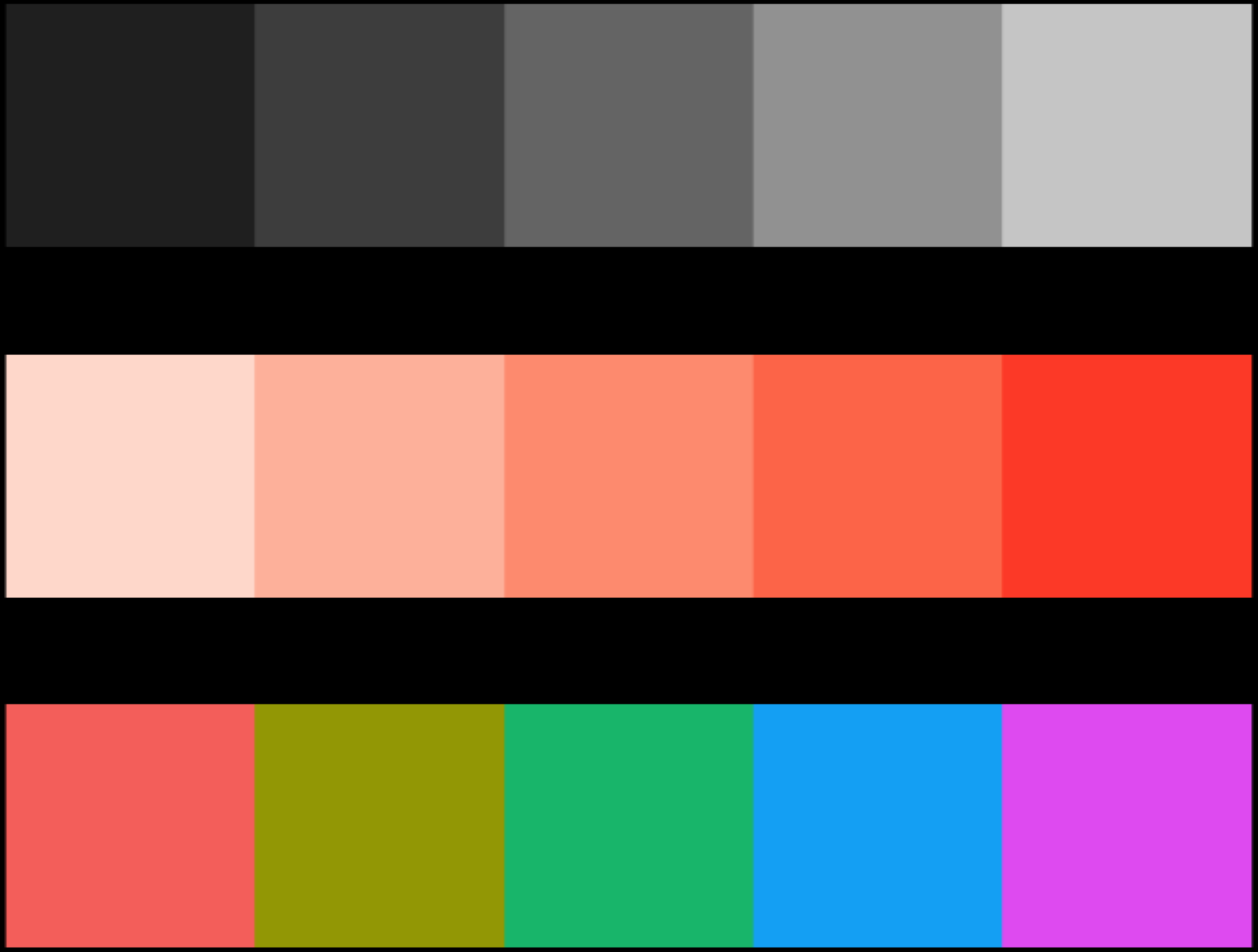
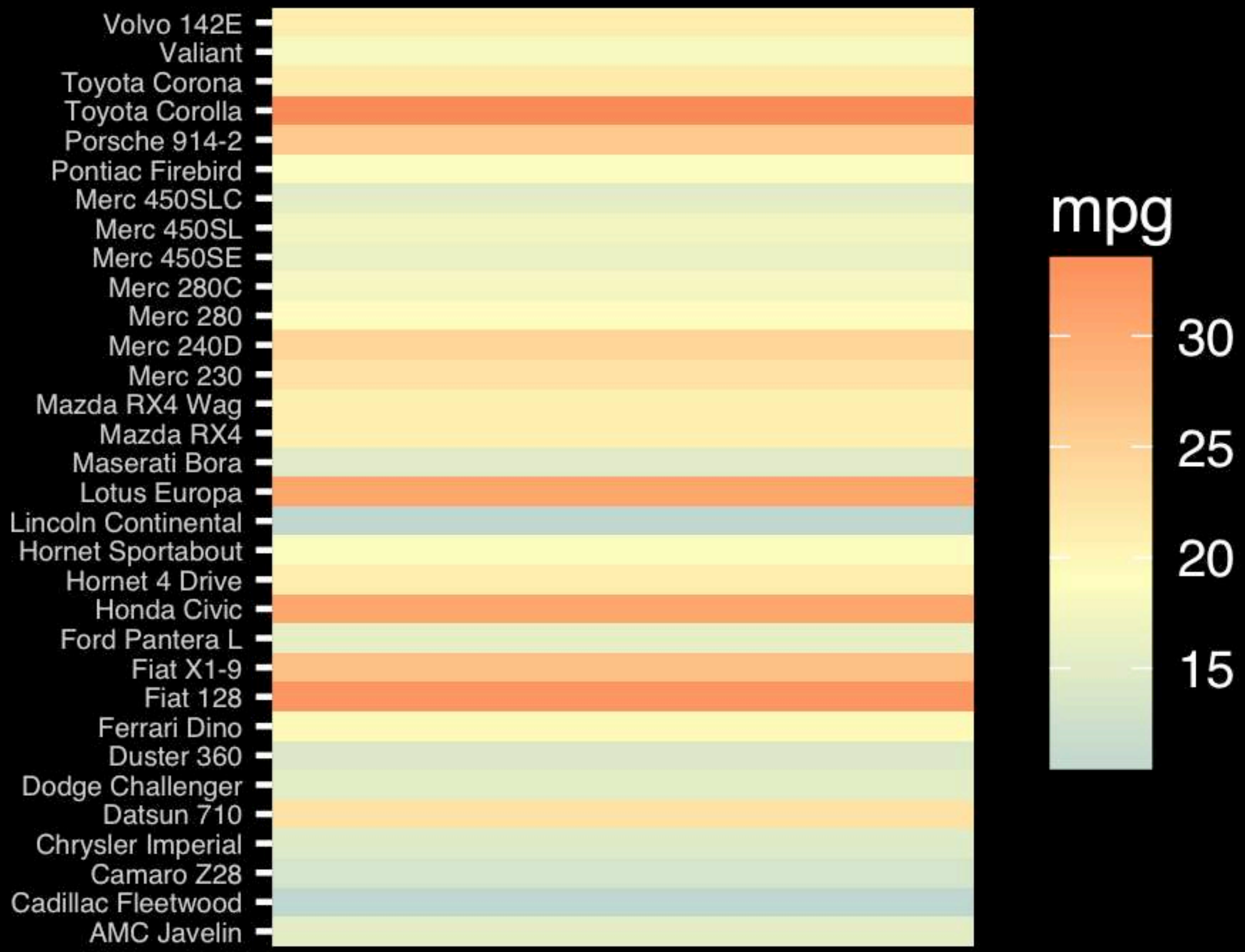


luminance

saturation

hue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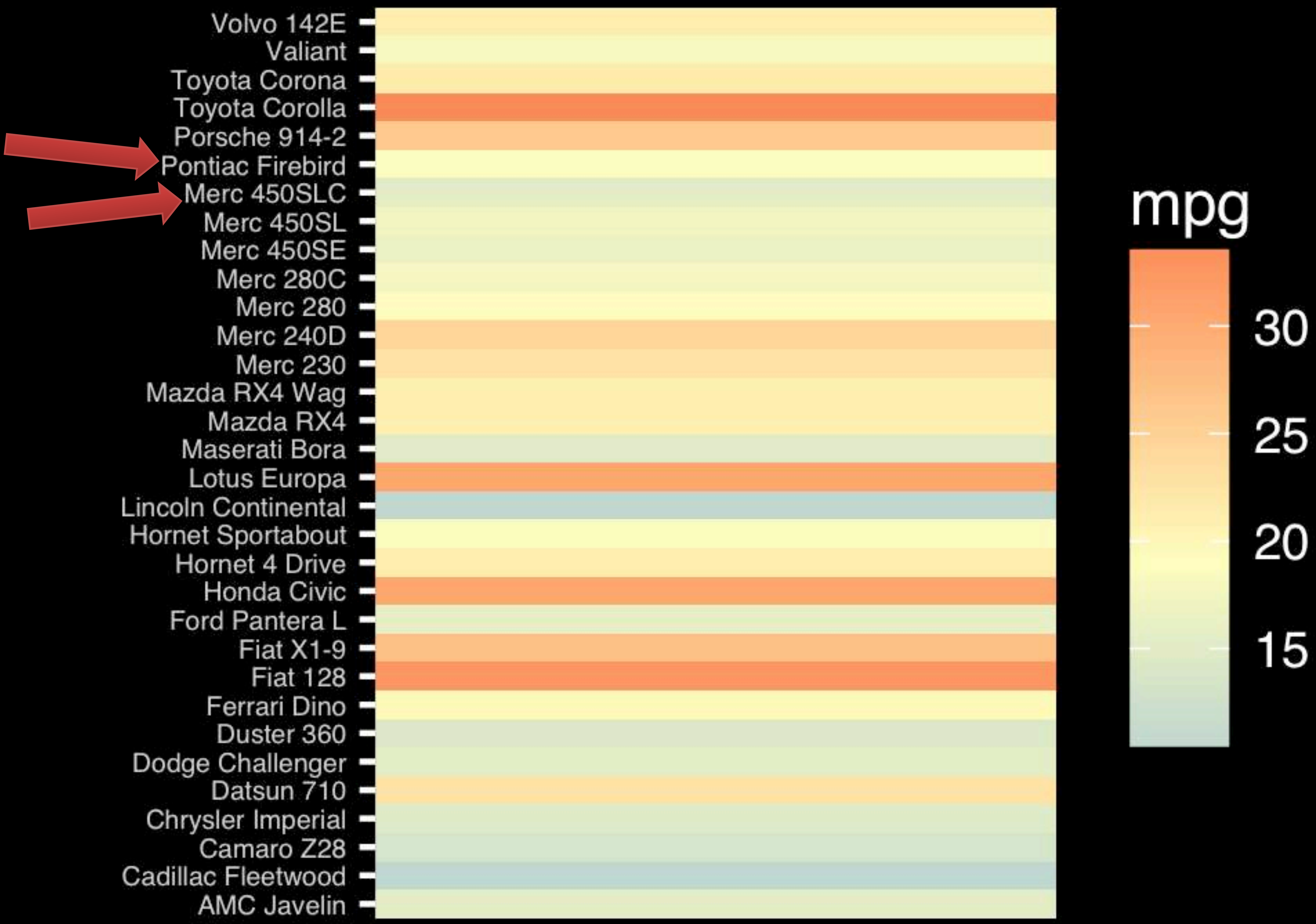
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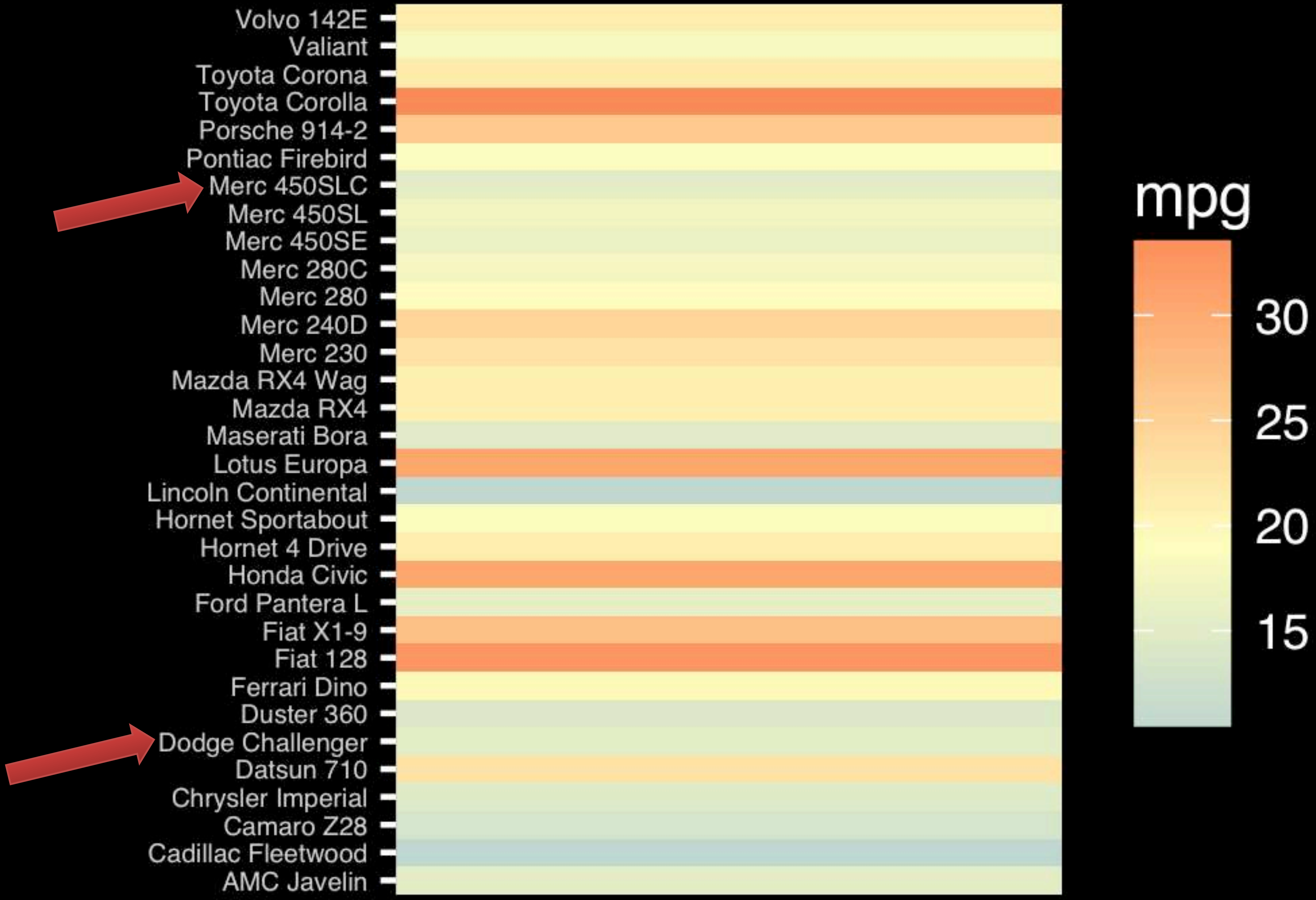
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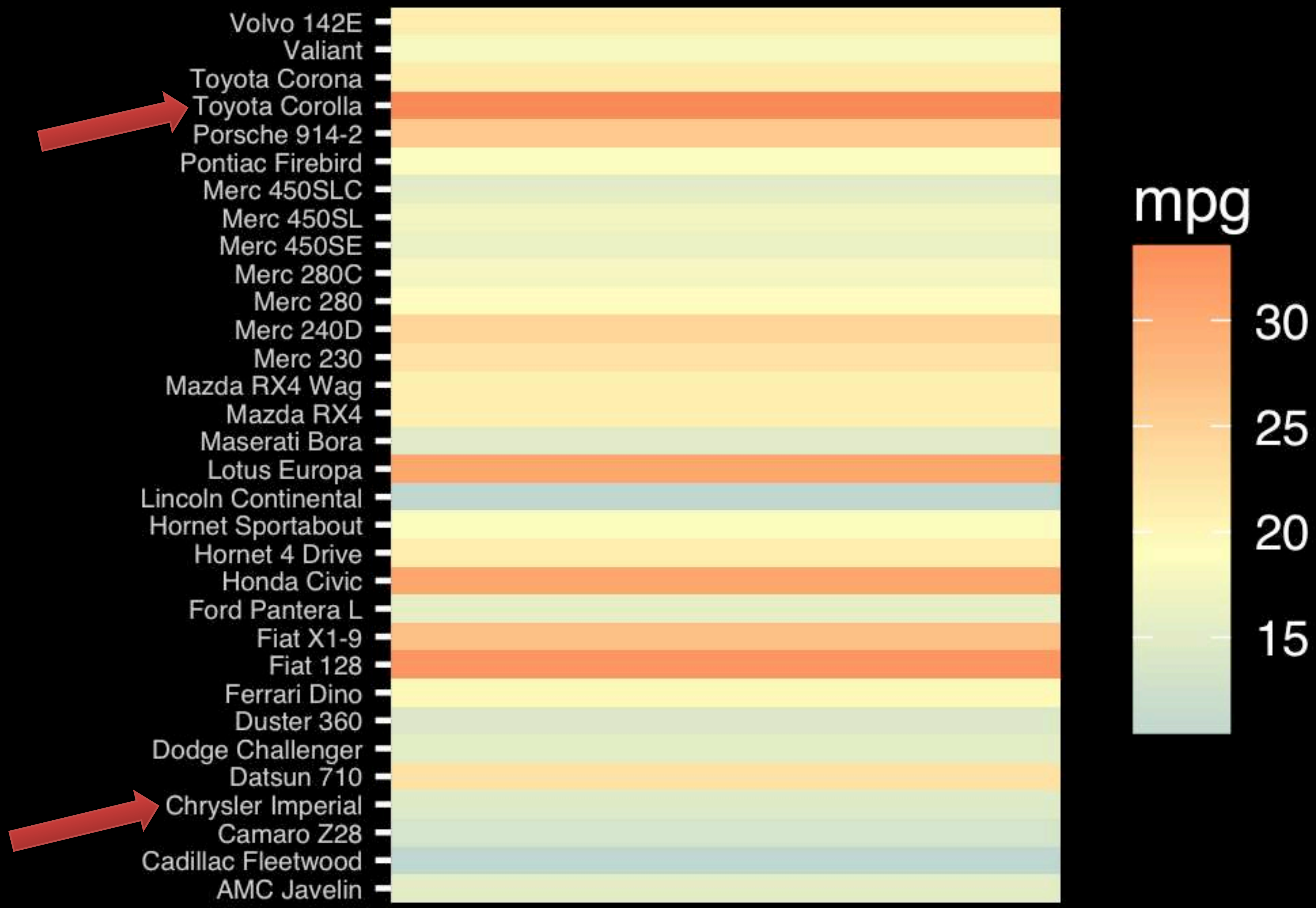
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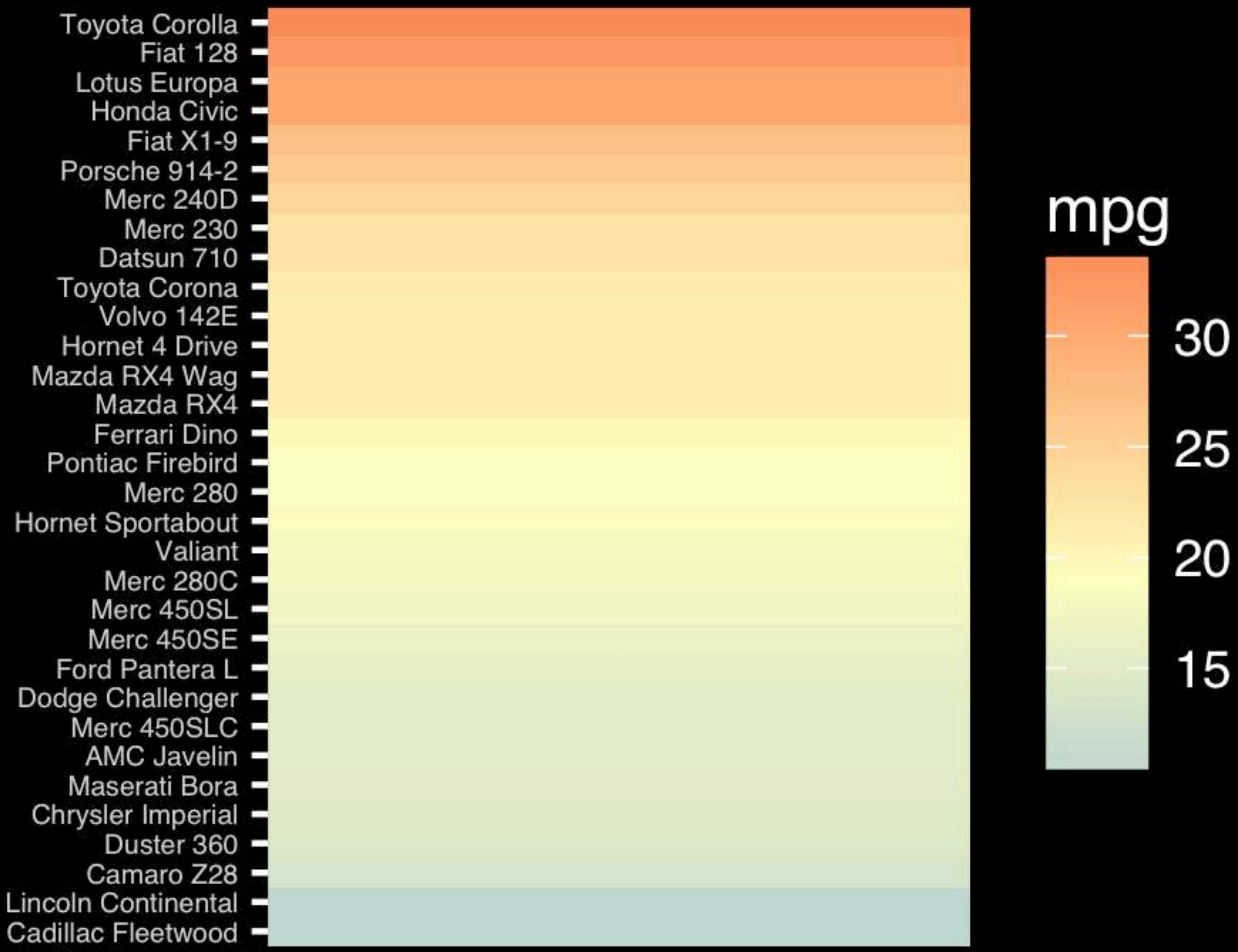
Observation: Alphabetical is almost never the correct ordering of a categorical variable.



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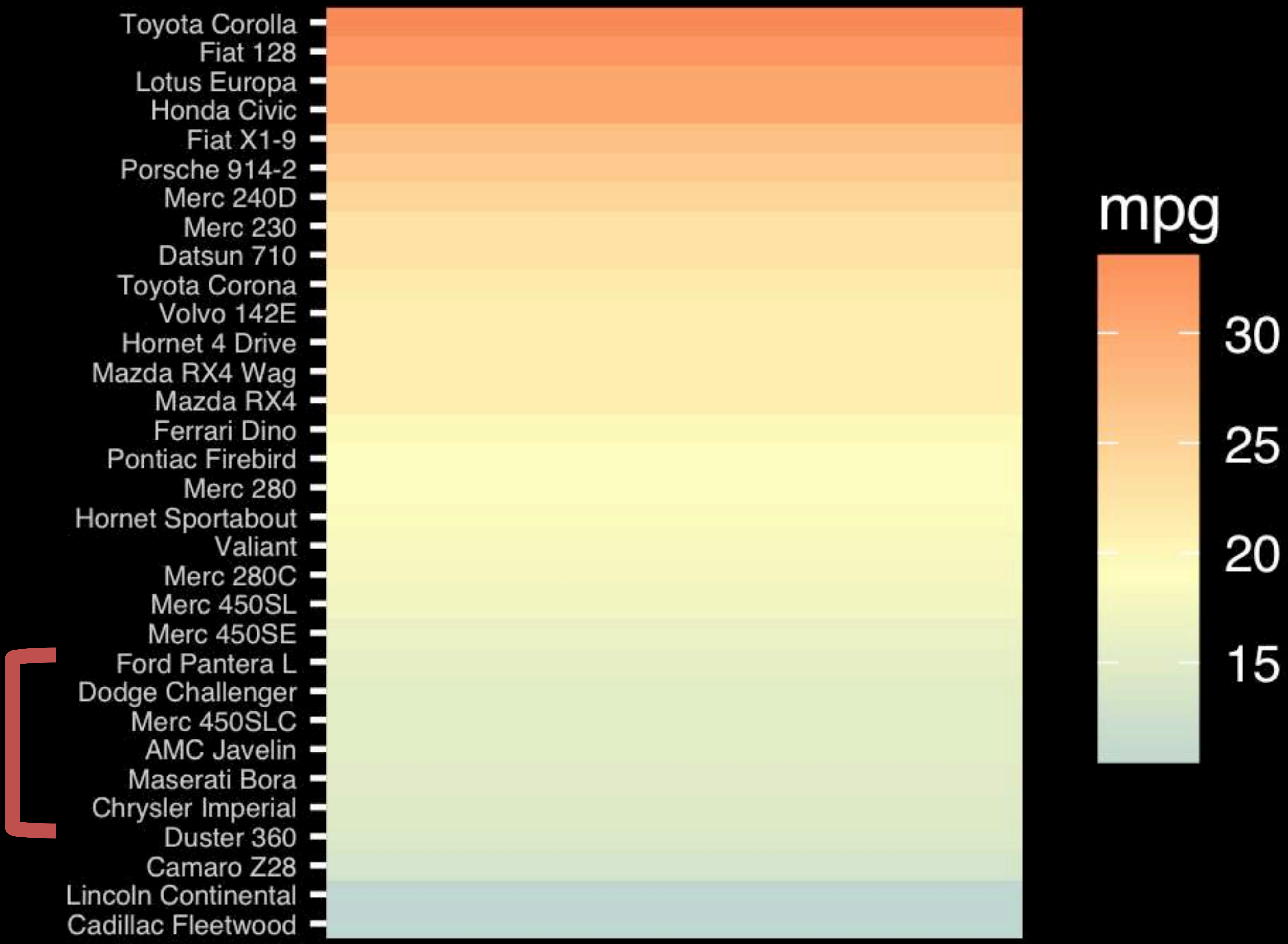
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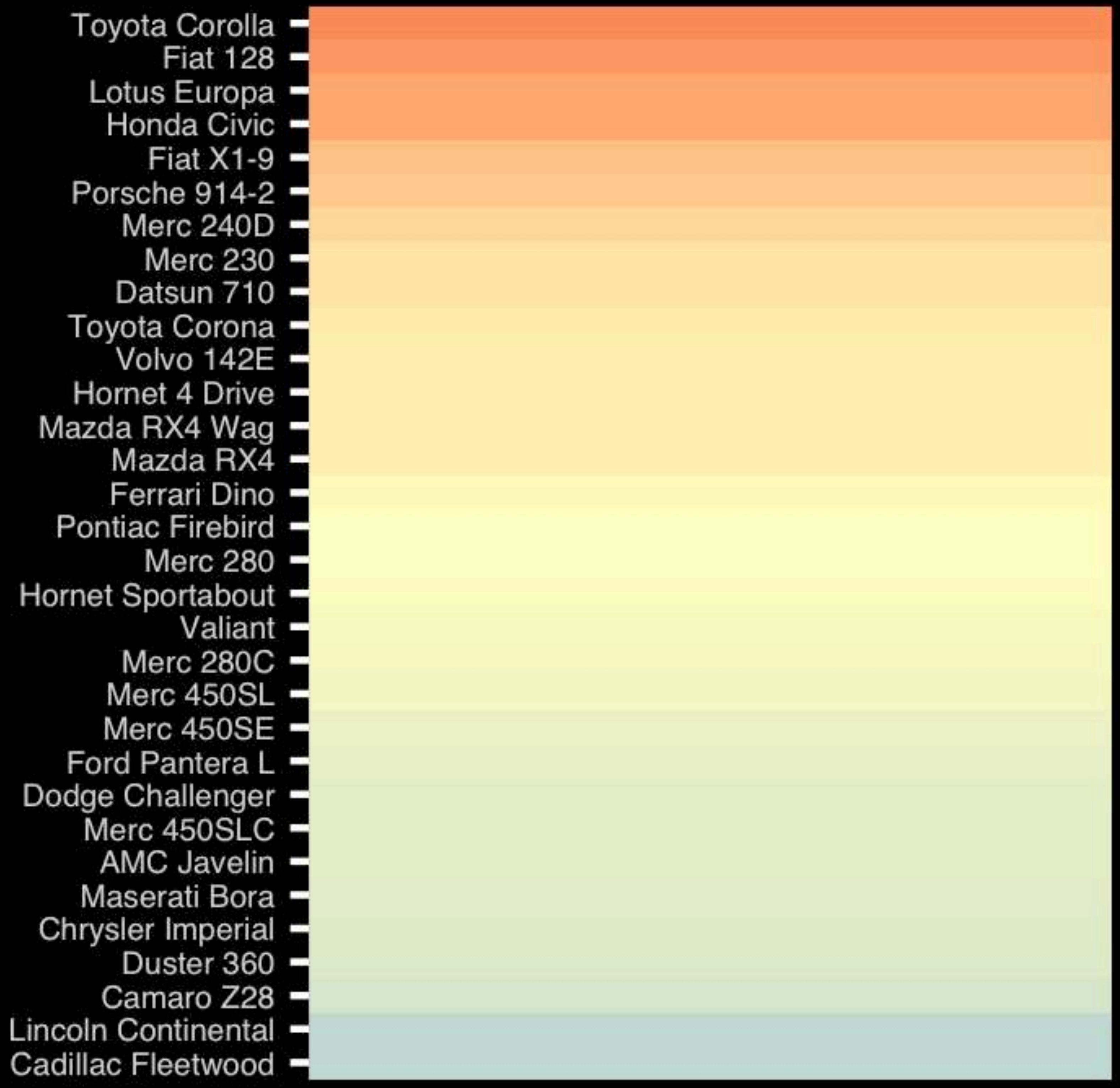
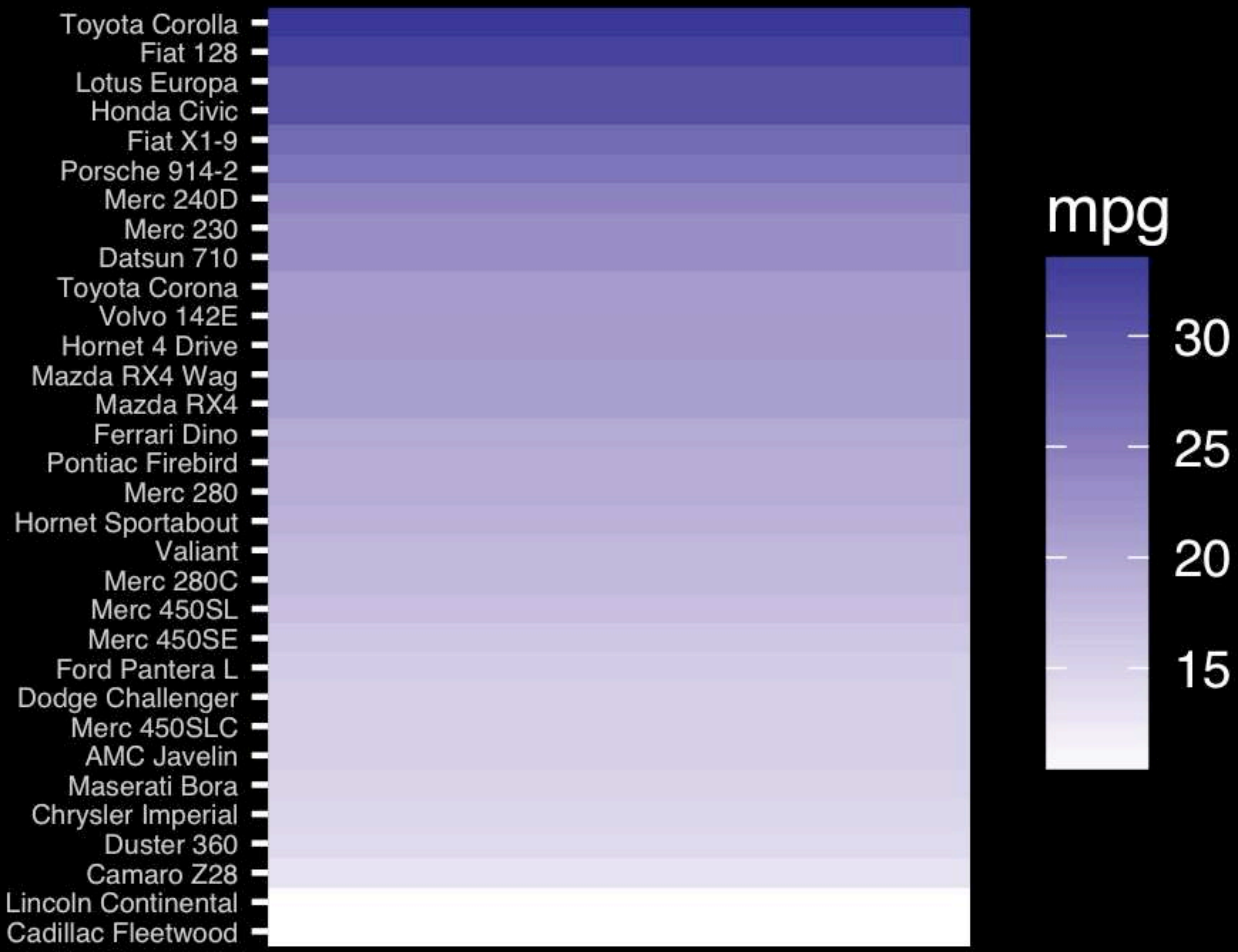


If we did not have the legend would you know which was low or high mpg?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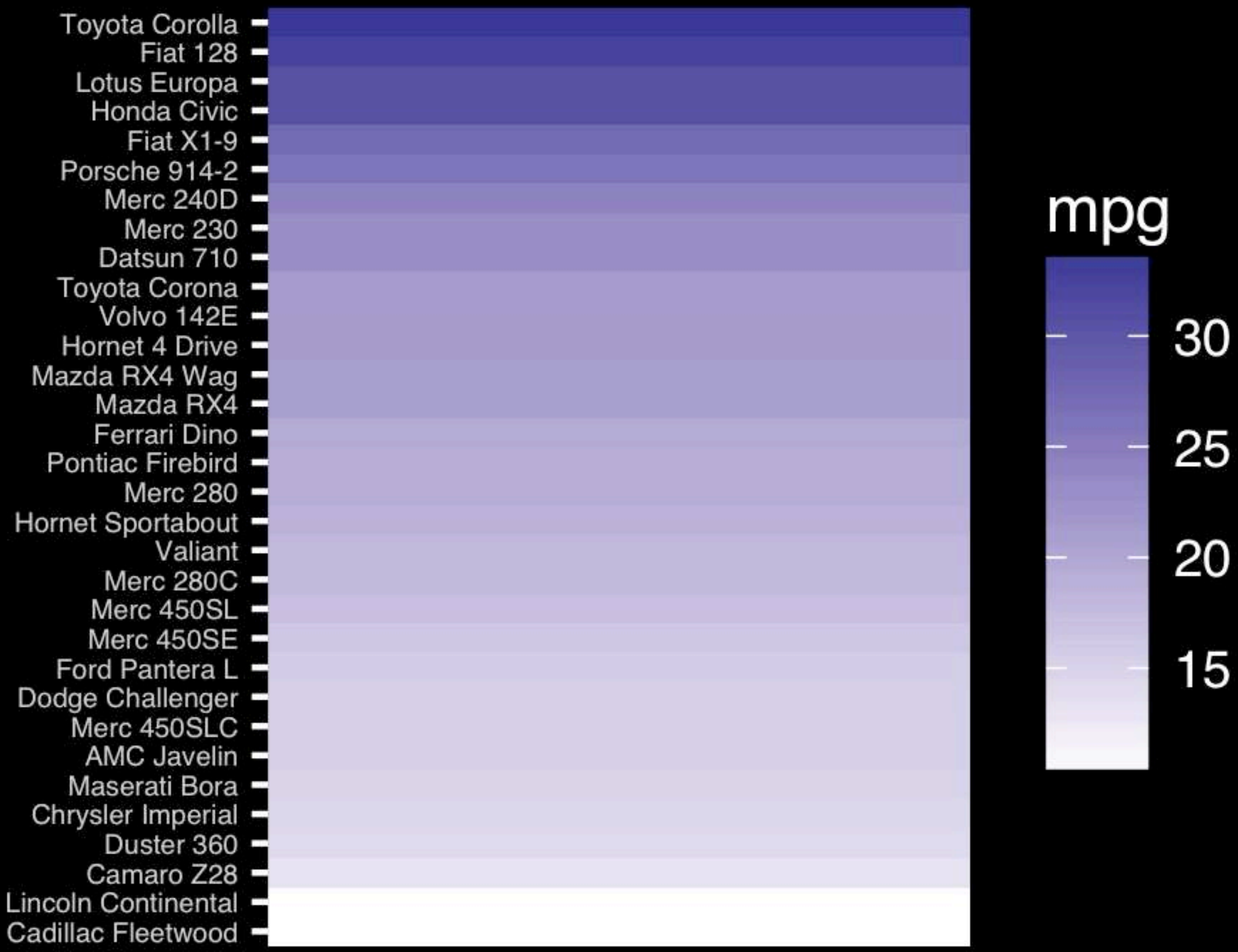
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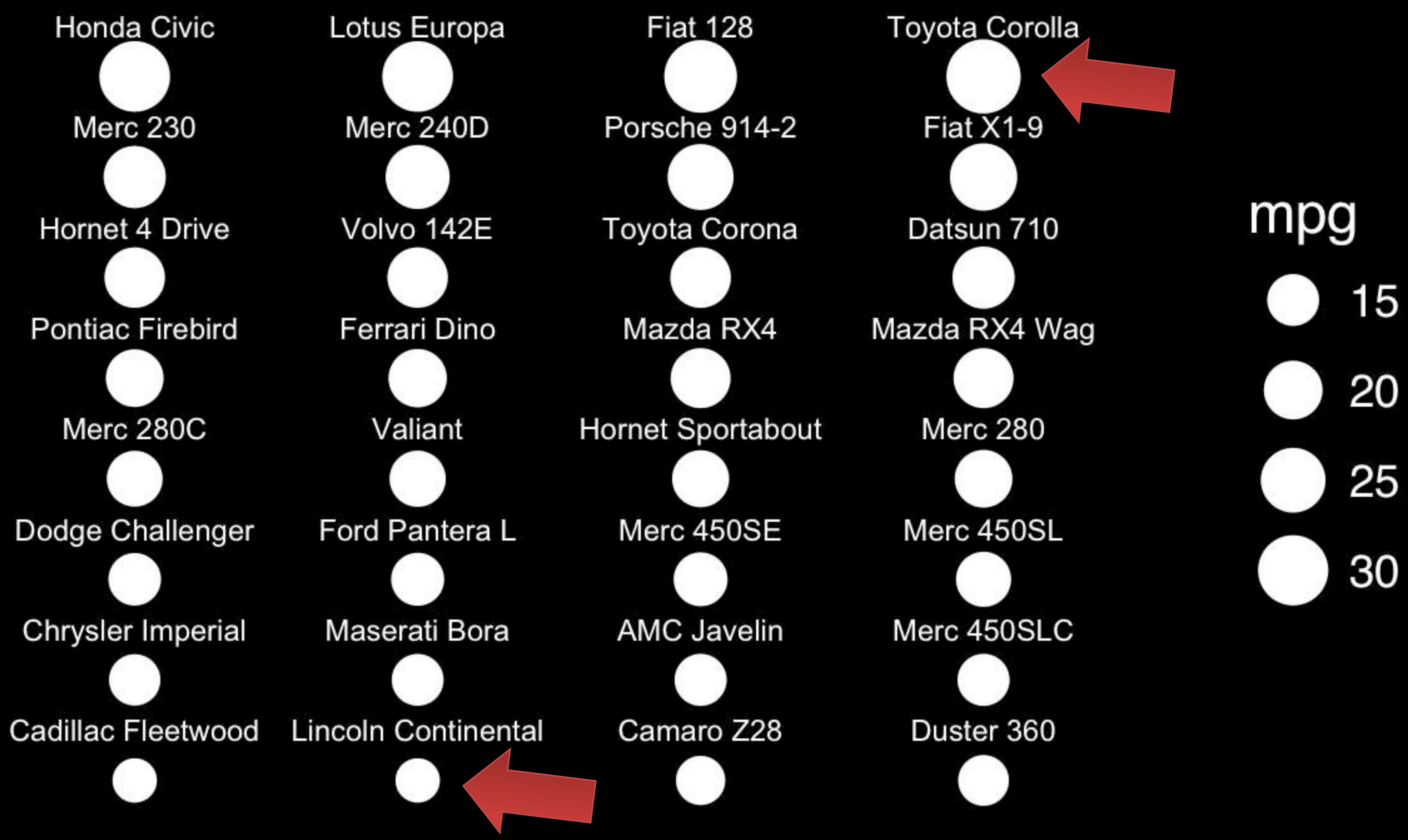
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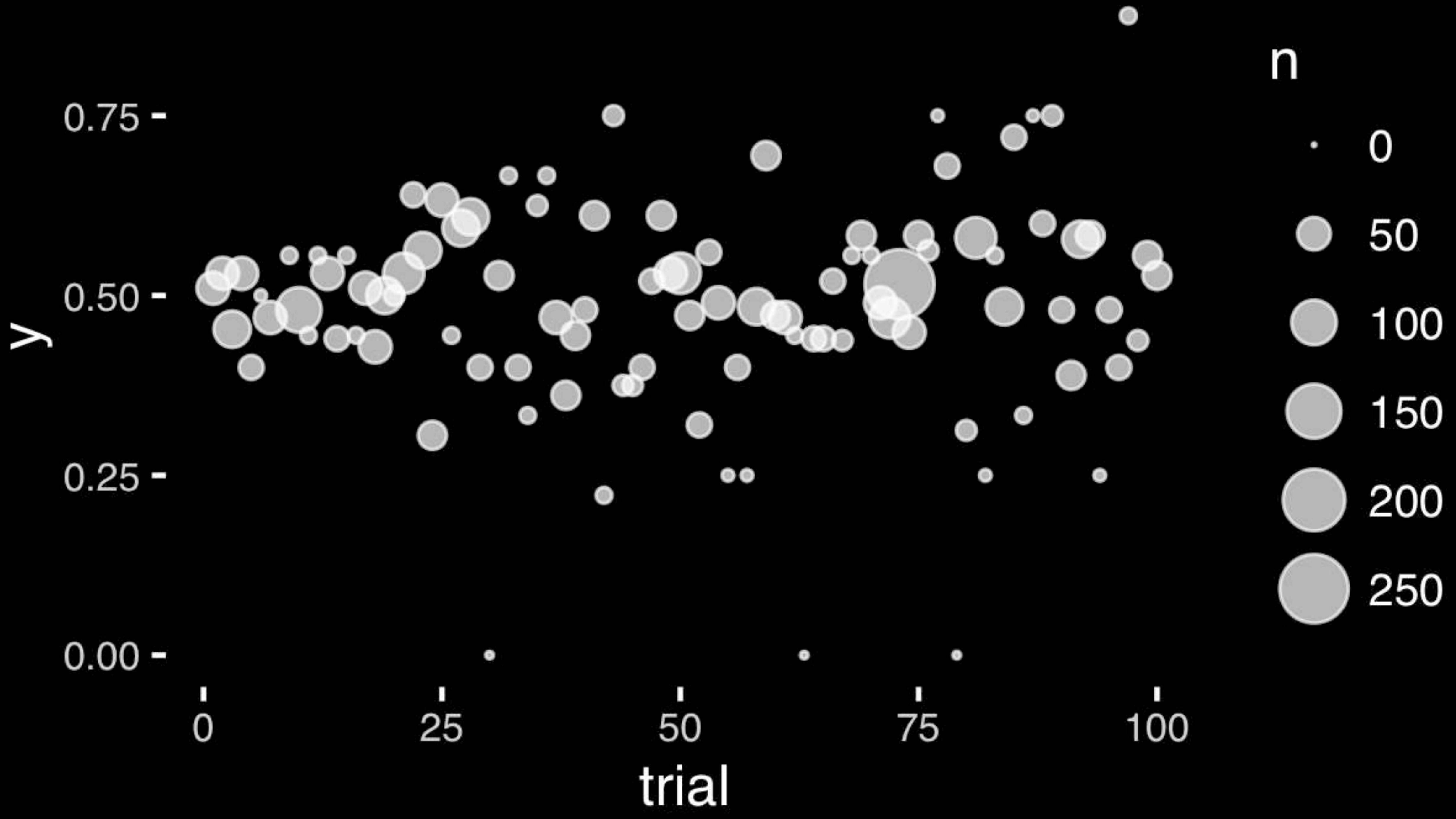
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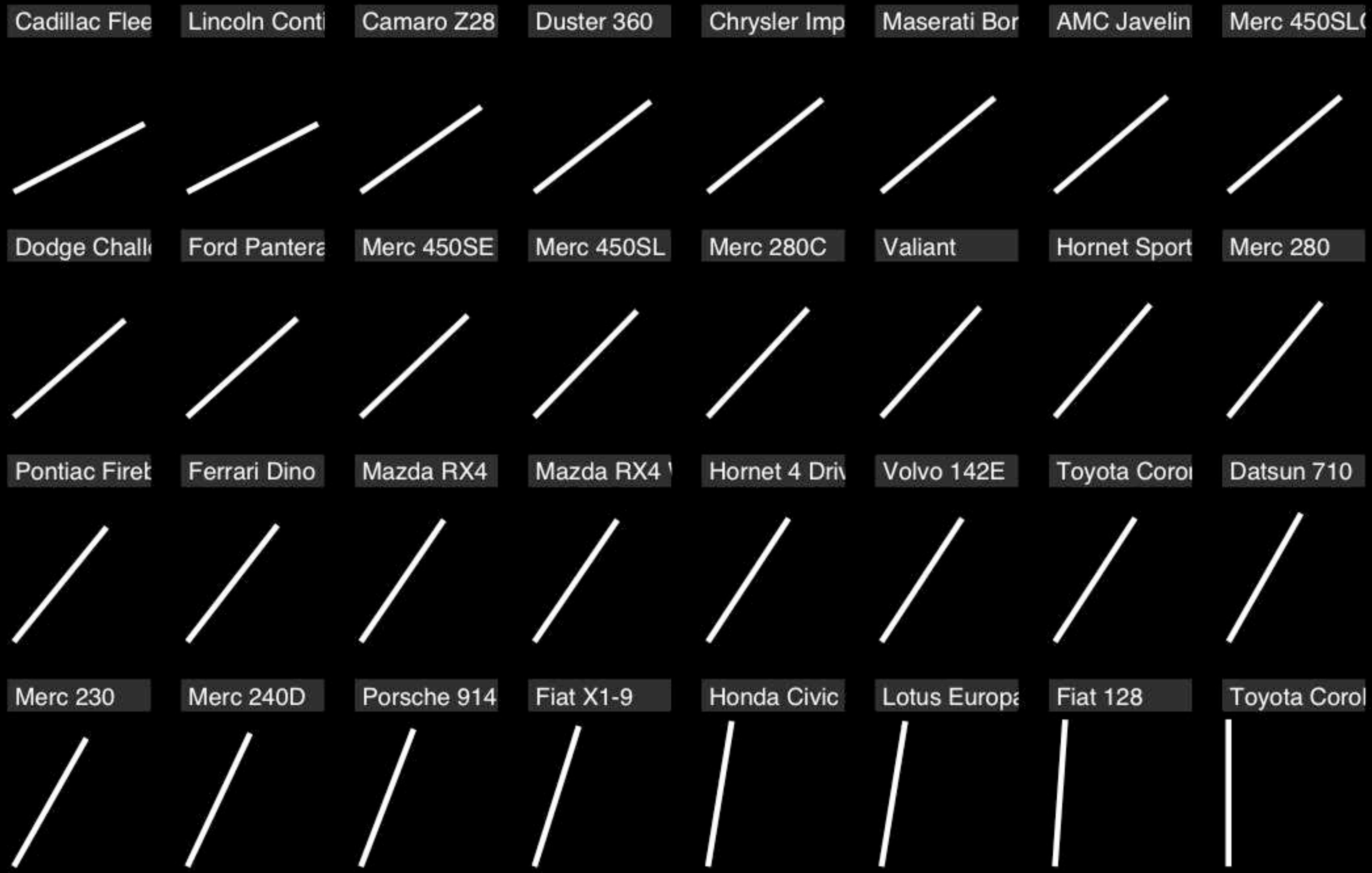
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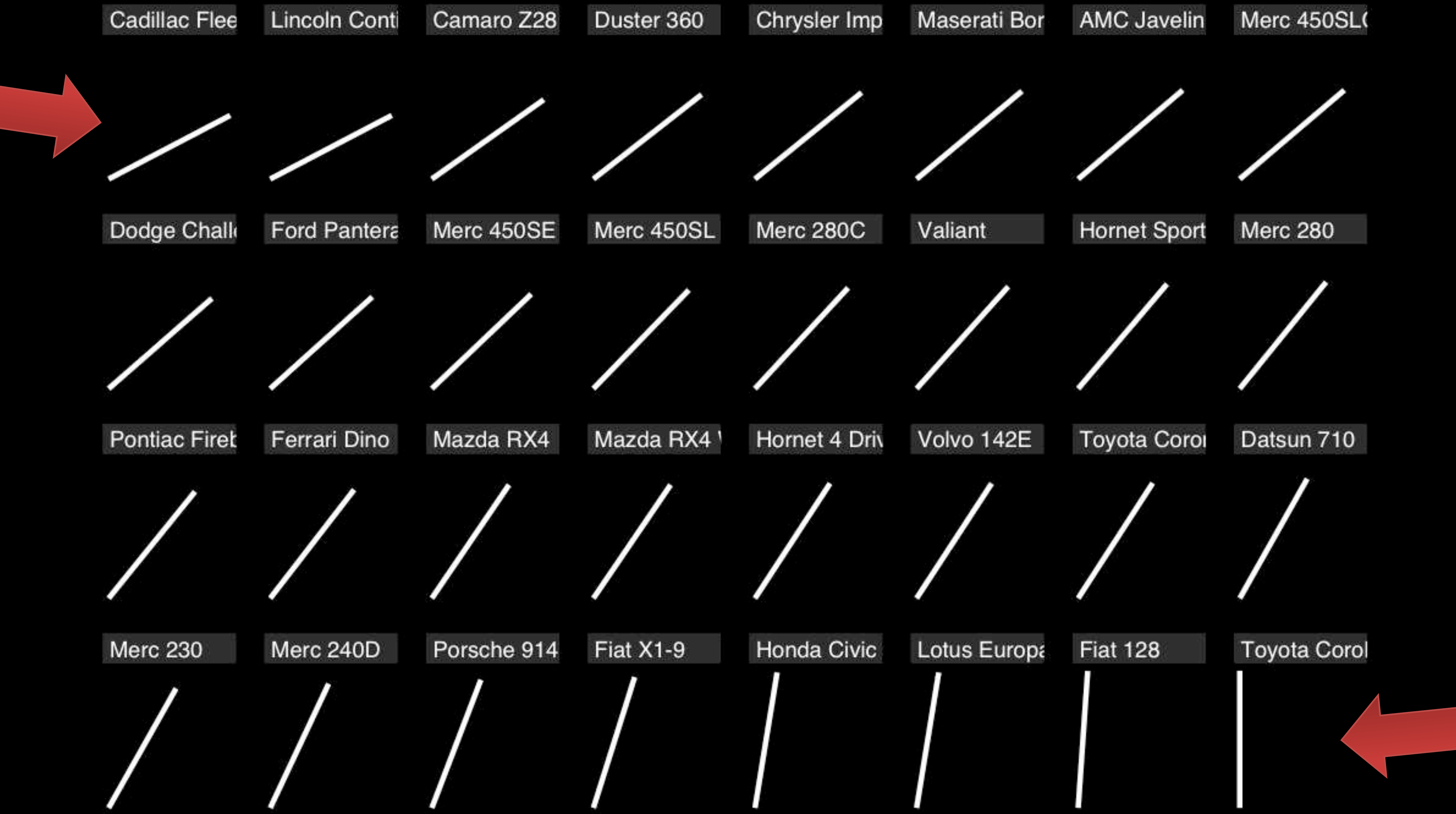
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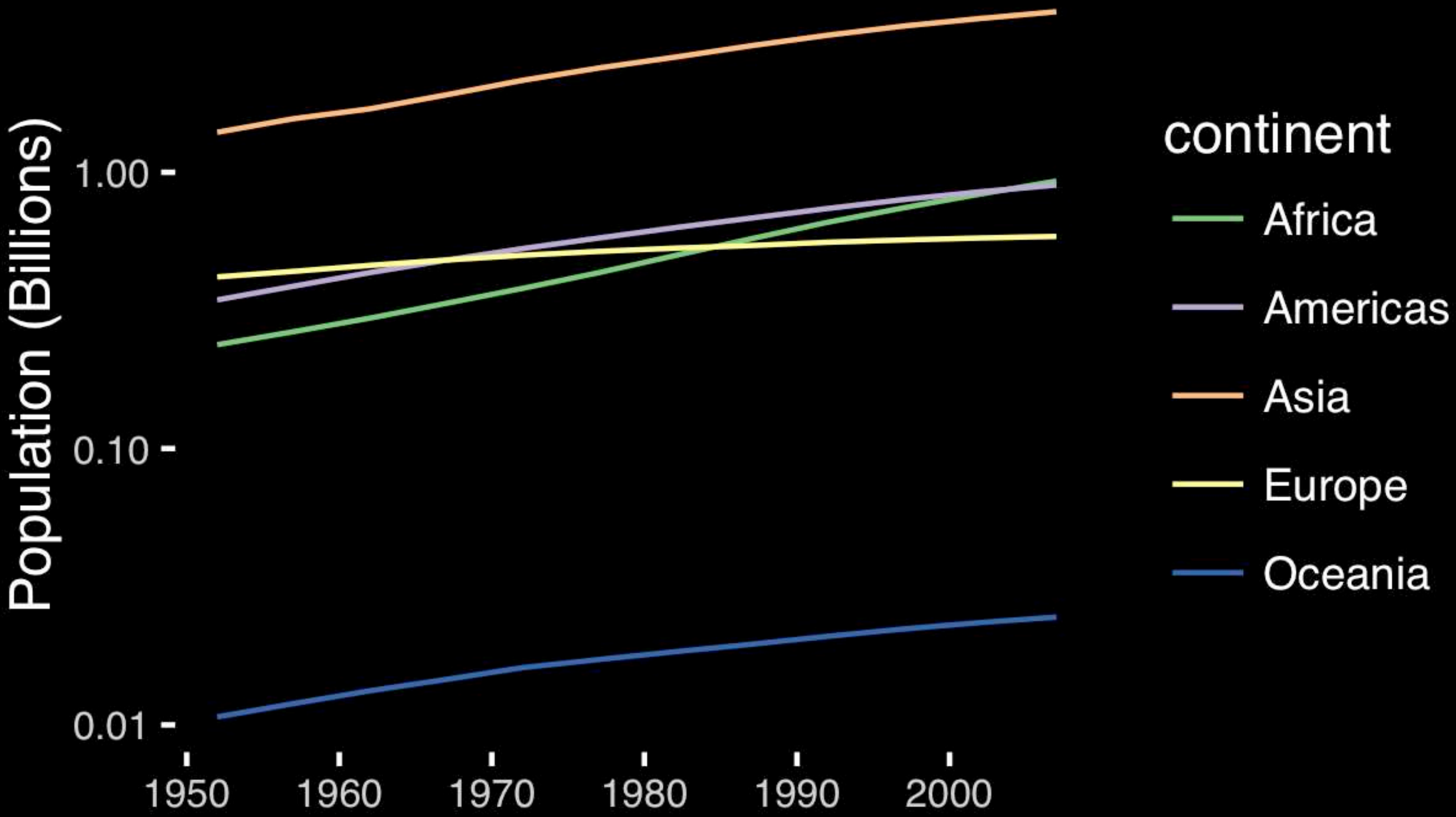
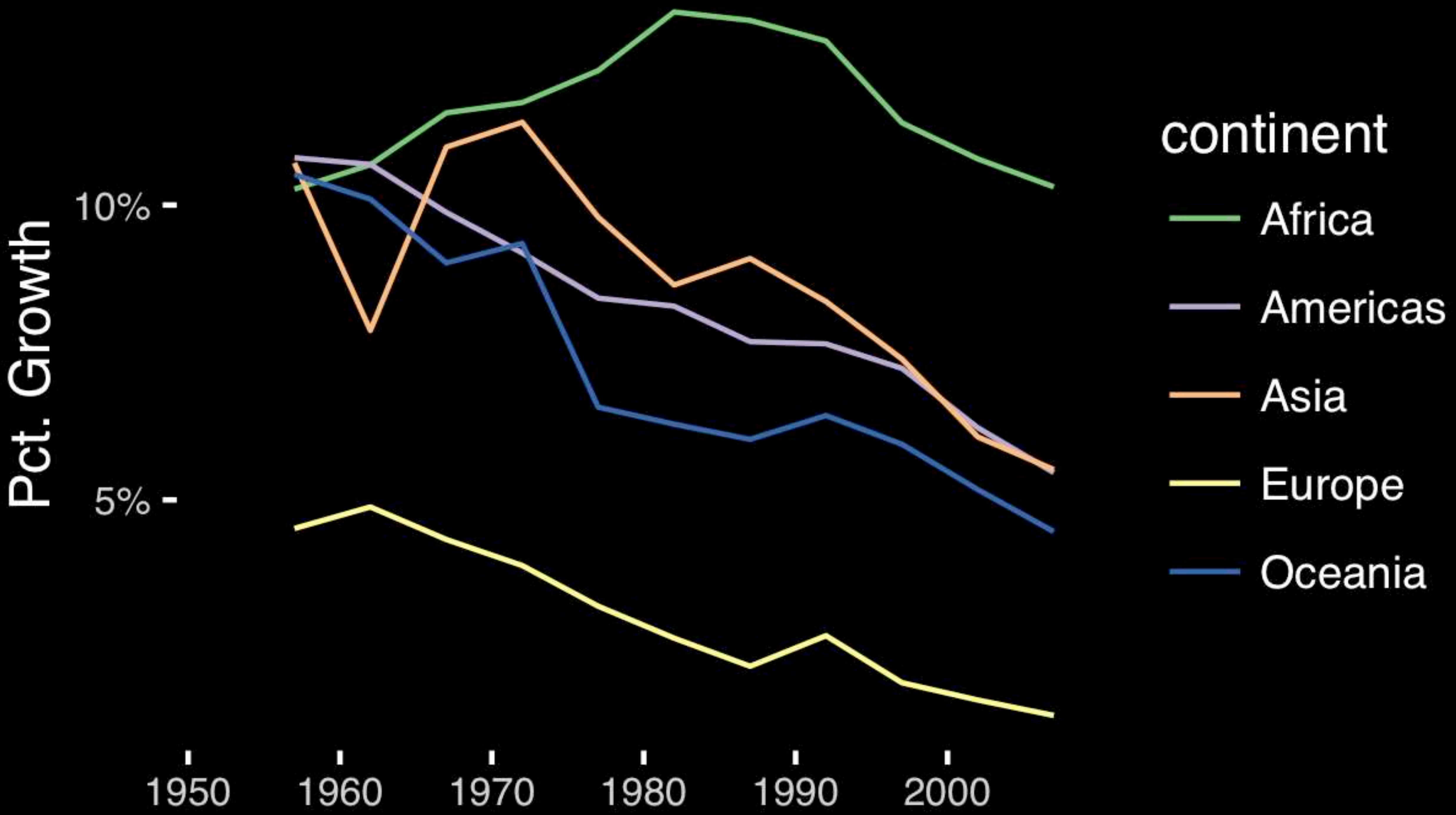

Population (Billions)
1.00
0.10
0.01
1950
1960
1970
1980
1990
2000
continent
Africa
Americas
Asia
Europe
Oceania

If growth (slope) is important, plot it directly.


Pct. Growth
10%
5%
1950
1960
1970
1980
1990
2000
continent
Africa
Americas
Asia
Europe
Oceania

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Observation: Pie charts are <u>ALWAYS</u> a mistake.

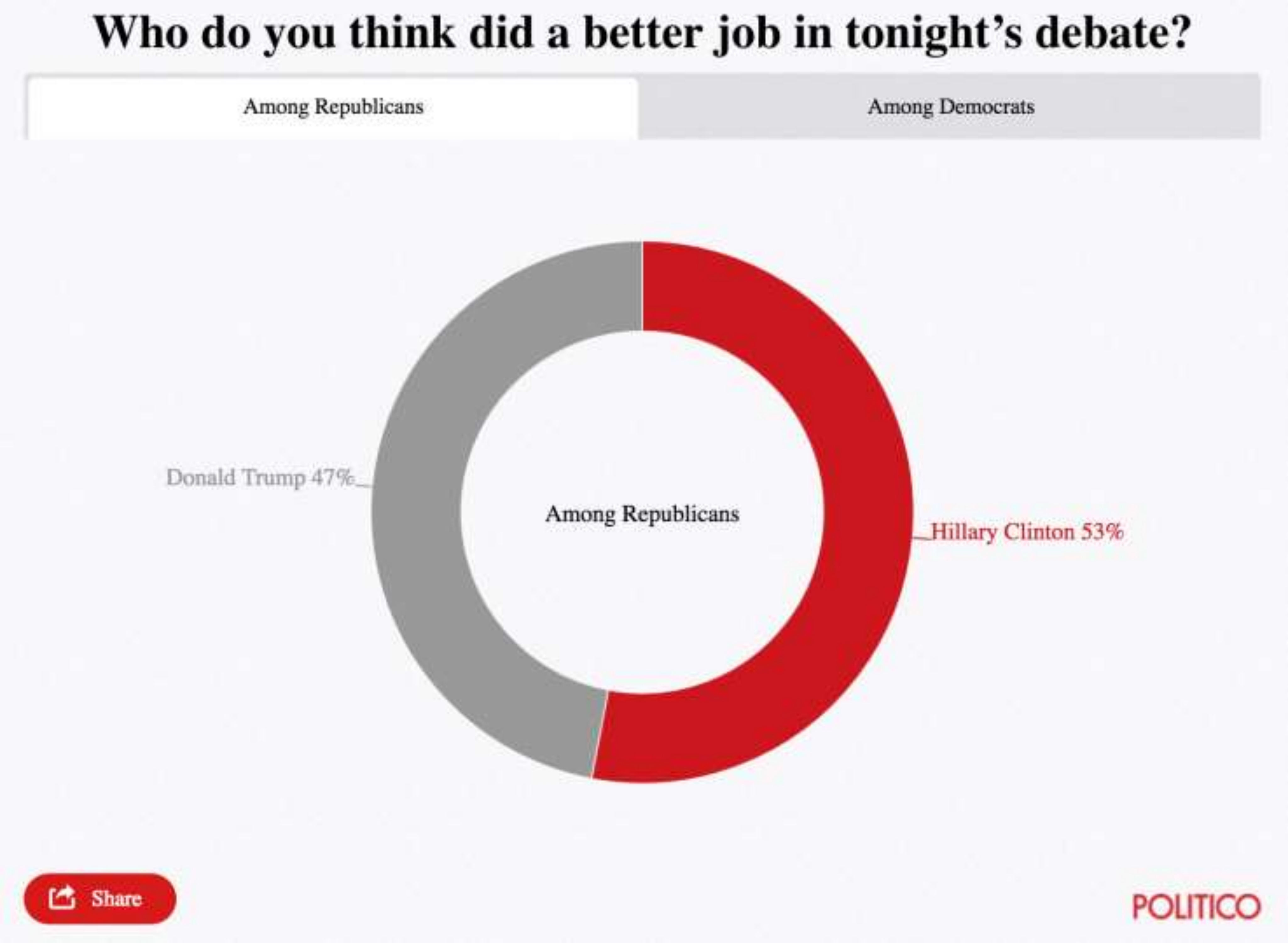
Apart from MPAs :-)

Piecharts are the information visualization equivalent of a roofing hammer to the frontal lobe. They have no place in the world of grownups, and occupy the same semiotic space as short pants, a runny nose, and chocolate smeared on one’s face. They are as professional as a pair of assless chaps.

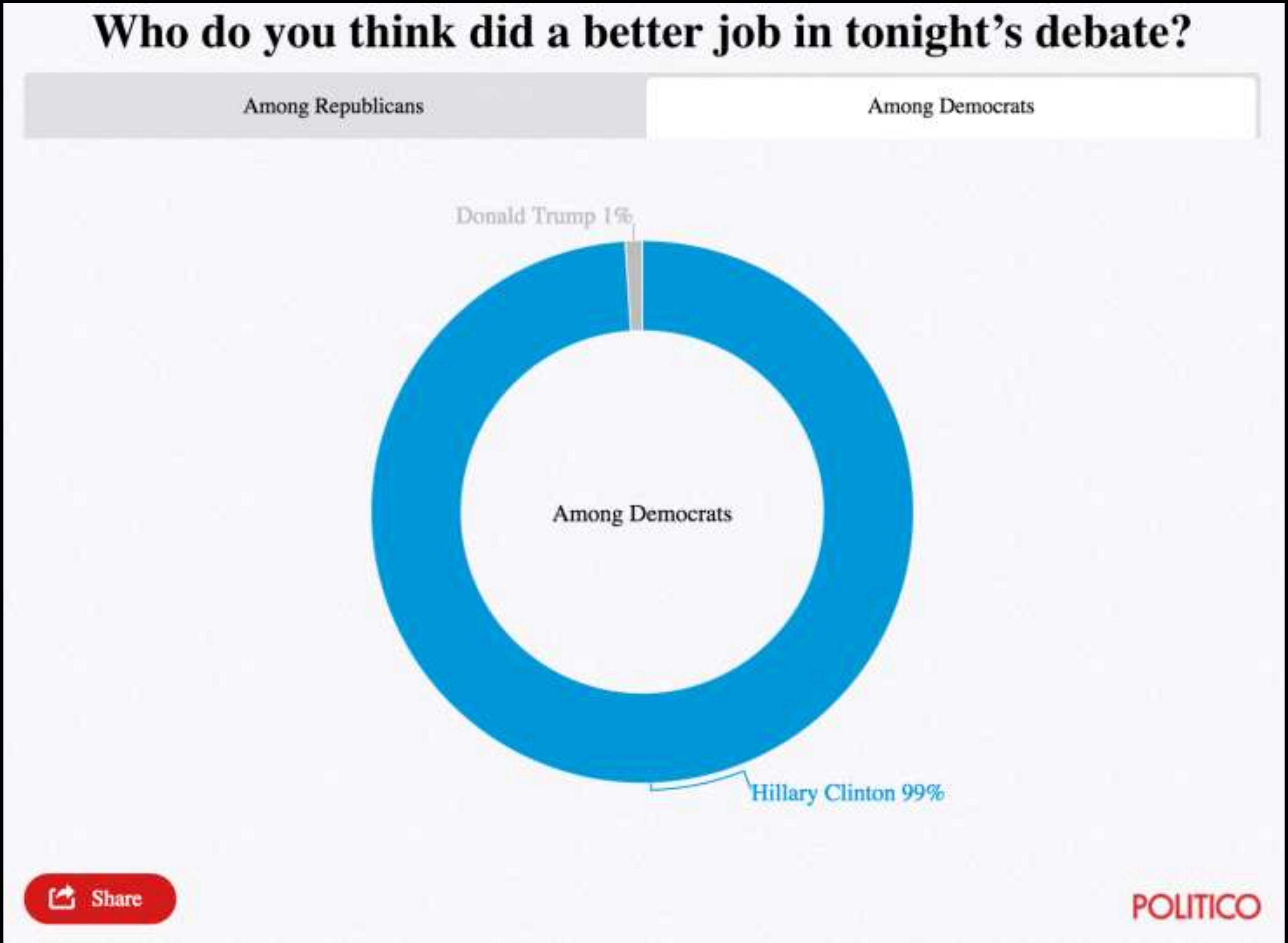
http://blog.codahale.com/2006/04/29/google-analytics-the-goggles-they-do-nothing/

Piecharts are the information visualization equivalent of a roofing hammer to the frontal lobe. They have no place in the world of grownups, and occupy the same semiotic space as short pants, a runny nose, and chocolate smeared on one’s face. **They are as professional as a pair of assless chaps.**

http://blog.codahale.com/2006/04/29/google-analytics-the-goggles-they-do-nothing/



Who do you think did a better job in tonight’s debate?



Who do you think did a better job in tonight’s debate?

Tables are preferable to graphics for many small data sets. A table is nearly always better than a dumb pie chart; the only thing worse than a pie chart is several of them, for then the viewer is asked to compare quantities located in spatial disarray both within and between pies... Given their low data-density and failure to order numbers along a visual dimension, **pie charts should never be used.**

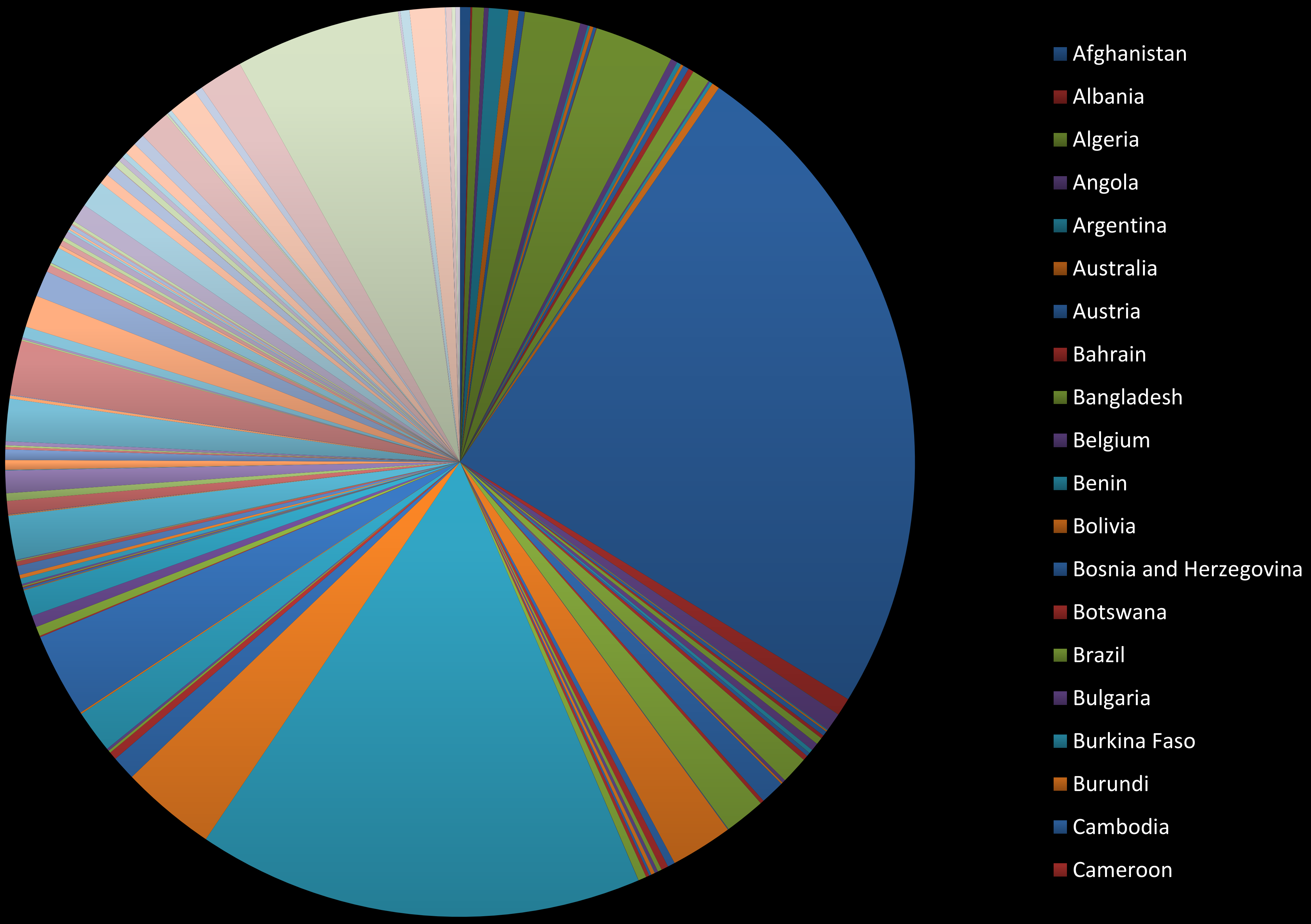
-Edward Tufte, The Visual Display of Quantitative Information

Tables are preferable to graphics for many small data sets. A table is nearly always better than a dumb pie chart; the only thing worse than a pie chart is several of them, for then the viewer is asked to compare quantities located in spatial disarray both within and between pies... Given their low data-density and failure to order numbers along a visual dimension, pie charts should never be used.

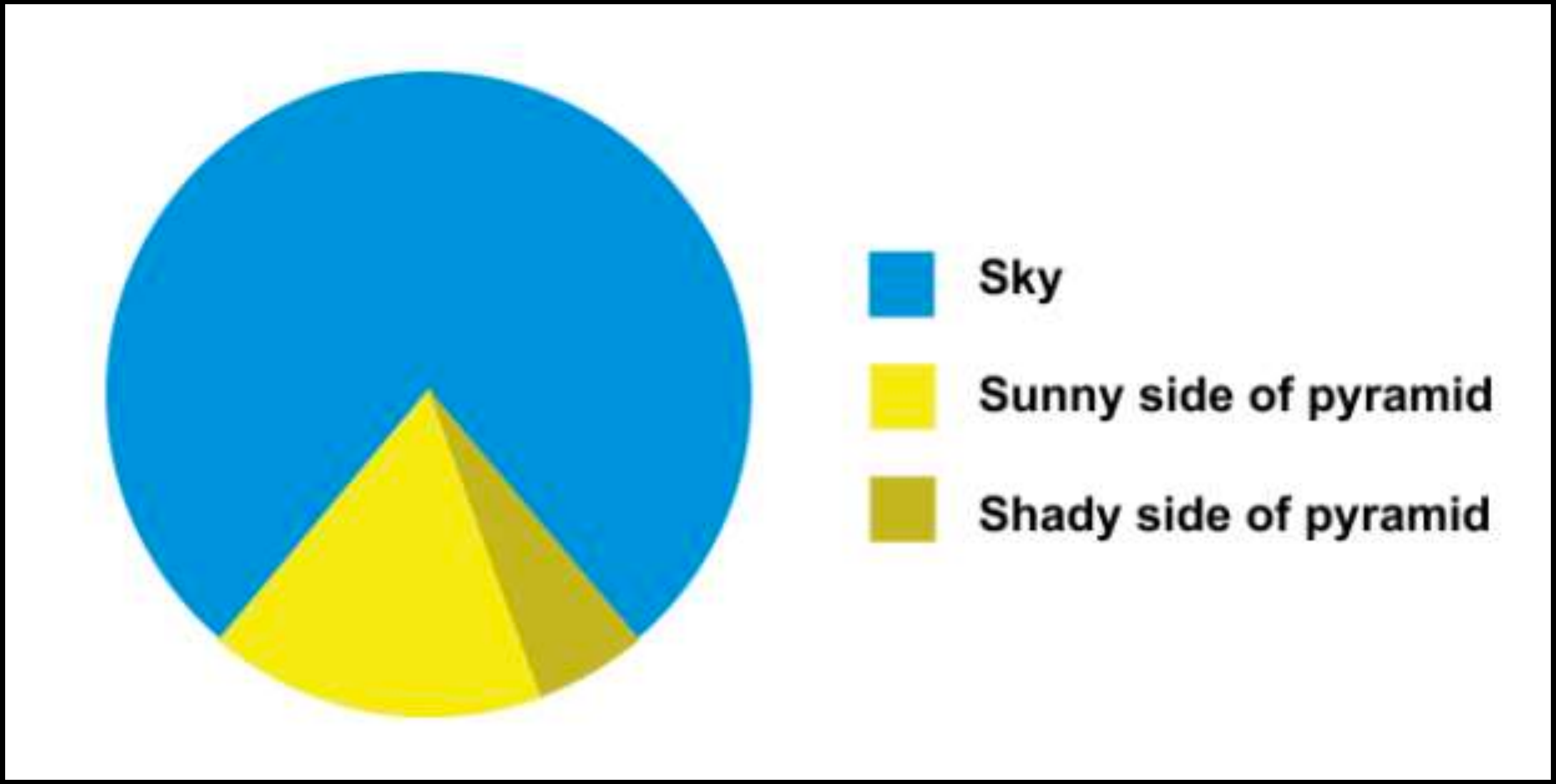
-Edward Tufte, The Visual Display of Quantitative Information

Who do you think did a better job in tonight’s debate?

	Clinton	**Trump**
Among Democrats	99%	1%
Among Republicans	53%	47%

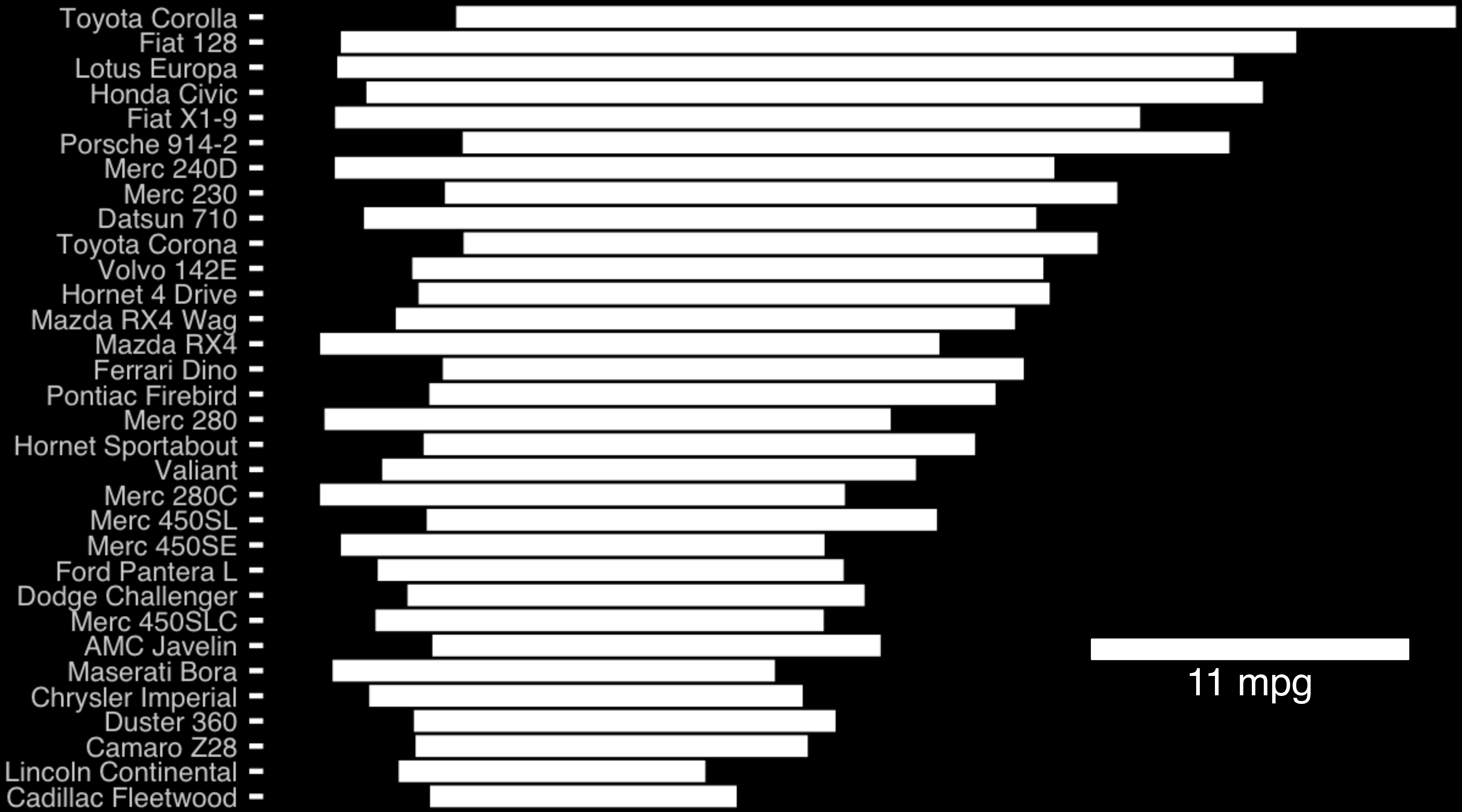

Afghanistan
Albania
Algeria
Angola
Argentina
Australia
Austria
Bahrain
Bangladesh
Belgium
Benin
Bolivia
Bosnia and Herzegovina
Botswana
Brazil
Bulgaria
Burkina Faso
Burundi
Cambodia
Cameroon

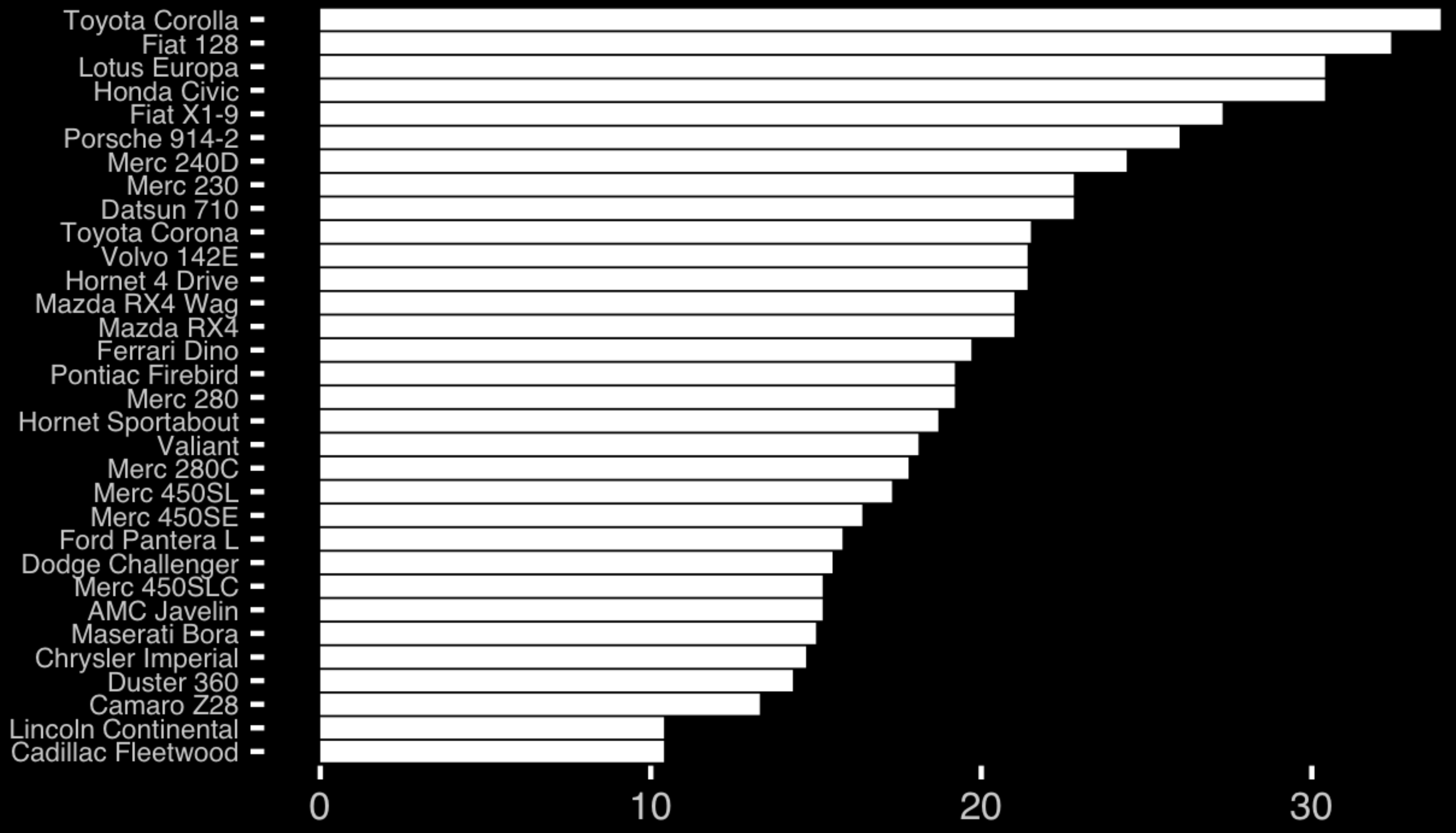
All good pie charts are jokes…



The most important measurement should exploit the highest ranked encoding possible.

- Position along a common scale
- Position on identical but nonaligned scales
- Length
- Angle or Slope
- Area
- Volume or Density or Color saturation/hue


Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
Toyota Corona
Volvo 142E
Hornet 4 Drive
Mazda RX4 Wag
Mazda RX4
Ferrari Dino
Pontiac Firebird
Merc 280
Hornet Sportabout
Valiant
Merc 280C
Merc 450SL
Merc 450SE
Ford Pantera L
Dodge Challenger
Merc 450SLC
AMC Javelin
Maserati Bora
Chrysler Imperial
Duster 360
Camaro Z28
Lincoln Continental
Cadillac Fleetwood
11 mpg


Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
Toyota Corona
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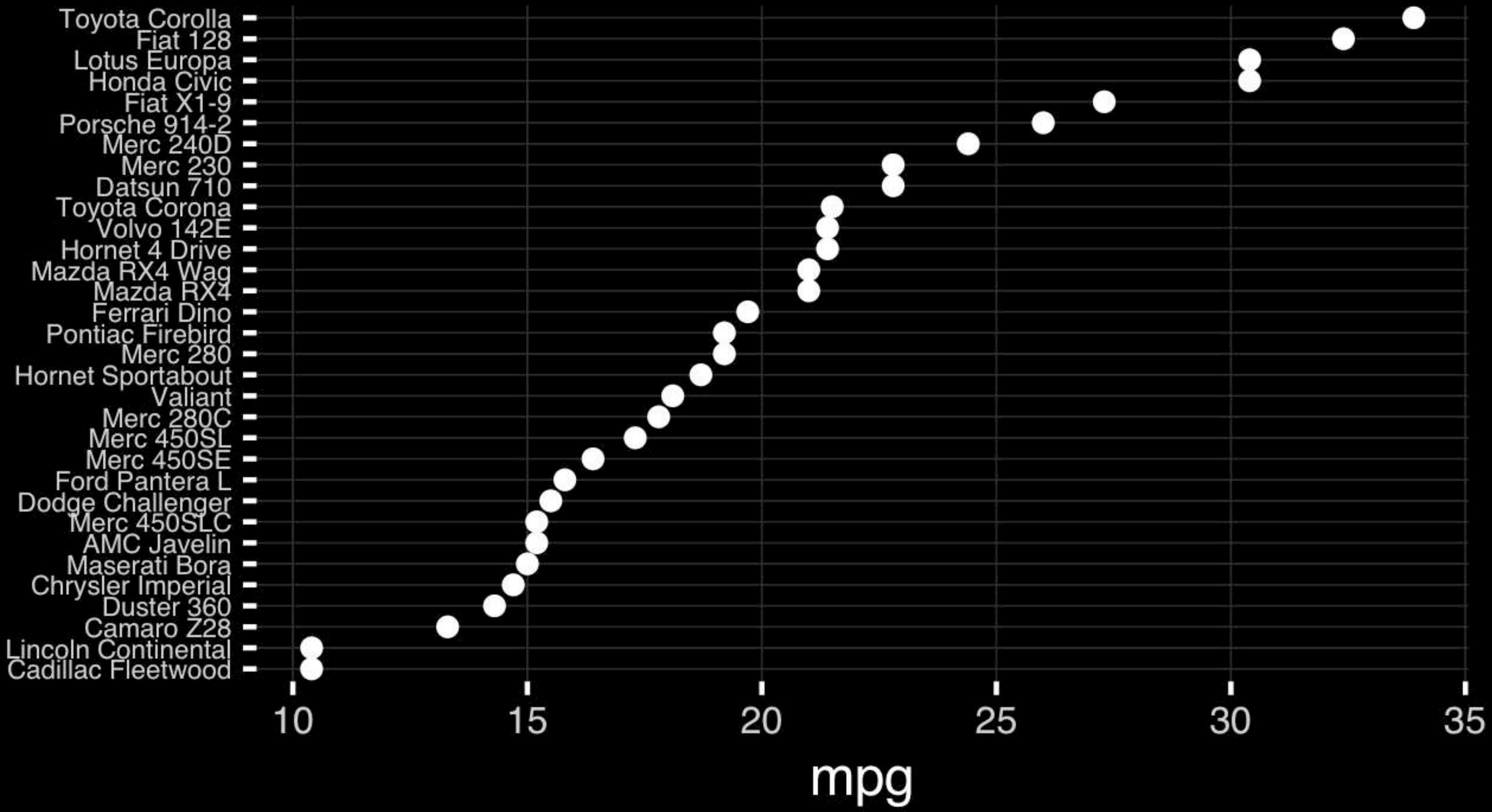
Mazda RX4
Ferrari Dino
Pontiac Firebird
Merc 280
Hornet Sportabout
Valiant
Merc 280C
Merc 450SL
Merc 450SE
Ford Pantera L
Dodge Challenger
Merc 450SLC
AMC Javelin
Maserati Bora
Chrysler Imperial
Duster 360
Camaro Z28
Lincoln Continental
Cadillac Fleetwood
0
10
20
30


Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
Toyota Corona
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Dodge Challenger
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AMC Javelin
Maserati Bora
Chrysler Imperial
Duster 360
Camaro Z28
Lincoln Continental
Cadillac Fleetwood
10
15
20
25
30
35
mpg

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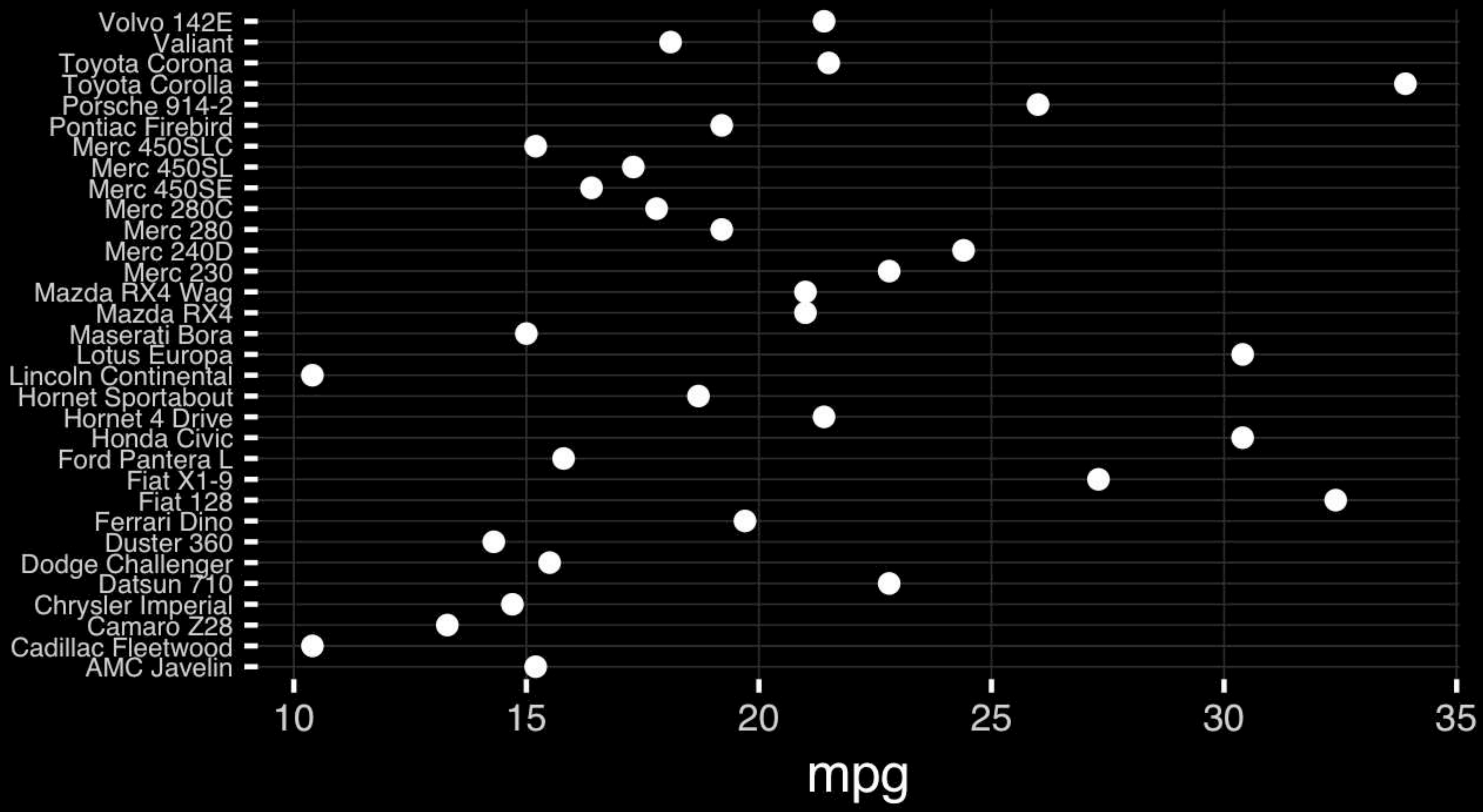
- Volume or Density or Color saturation/hue


Volvo 142E
Valiant
Toyota Corona
Toyota Corolla
Porsche 914-2
Pontiac Firebird
Merc 450SLC
Merc 450SL
Merc 450SE
Merc 280C
Merc 280
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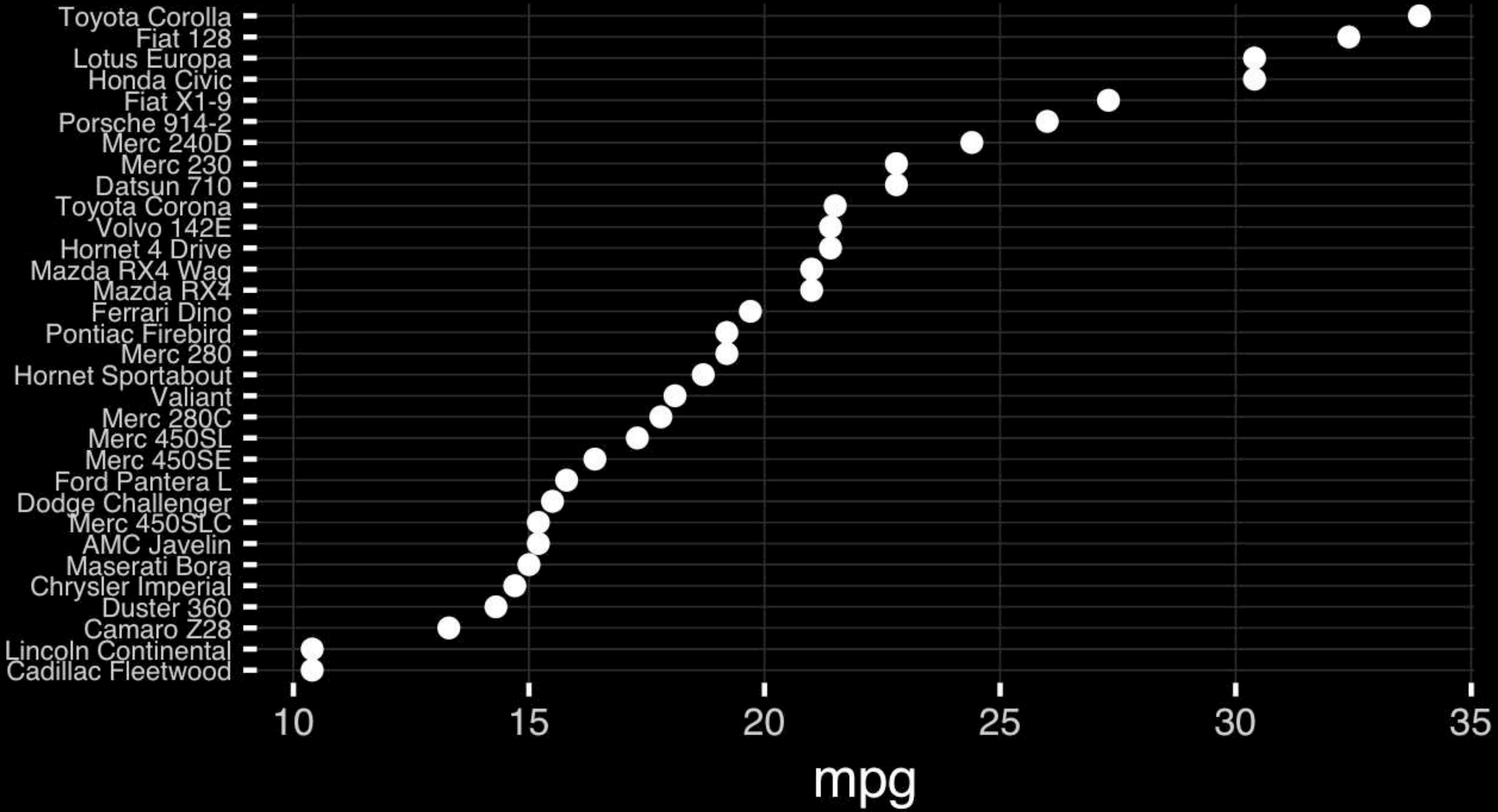
mpg


Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
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Merc 450SL
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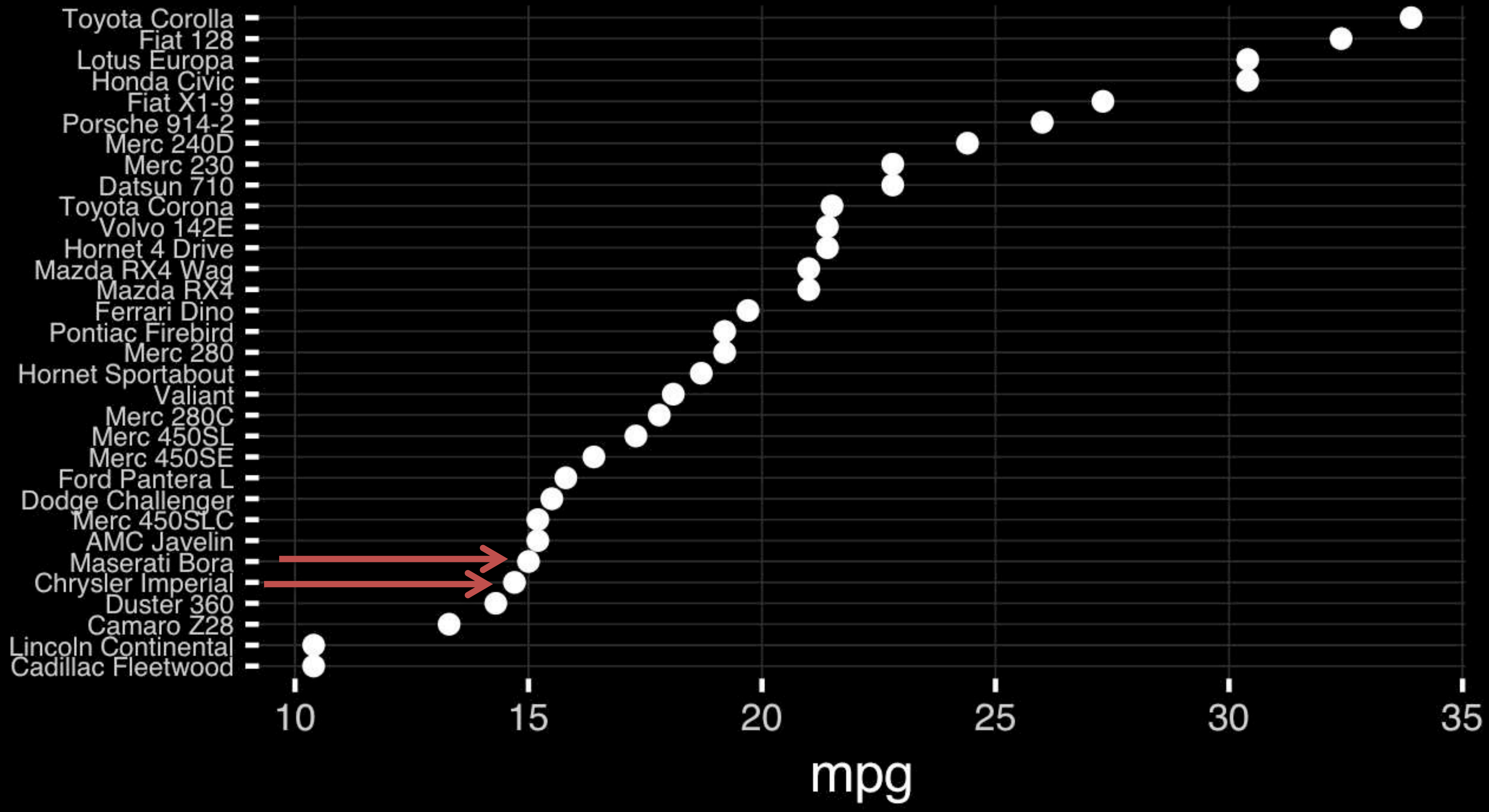

Toyota Corolla
Fiat 128
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AMC Javelin
Maserati Bora
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Duster 360
Camaro Z28
Lincoln Continental
Cadillac Fleetwood
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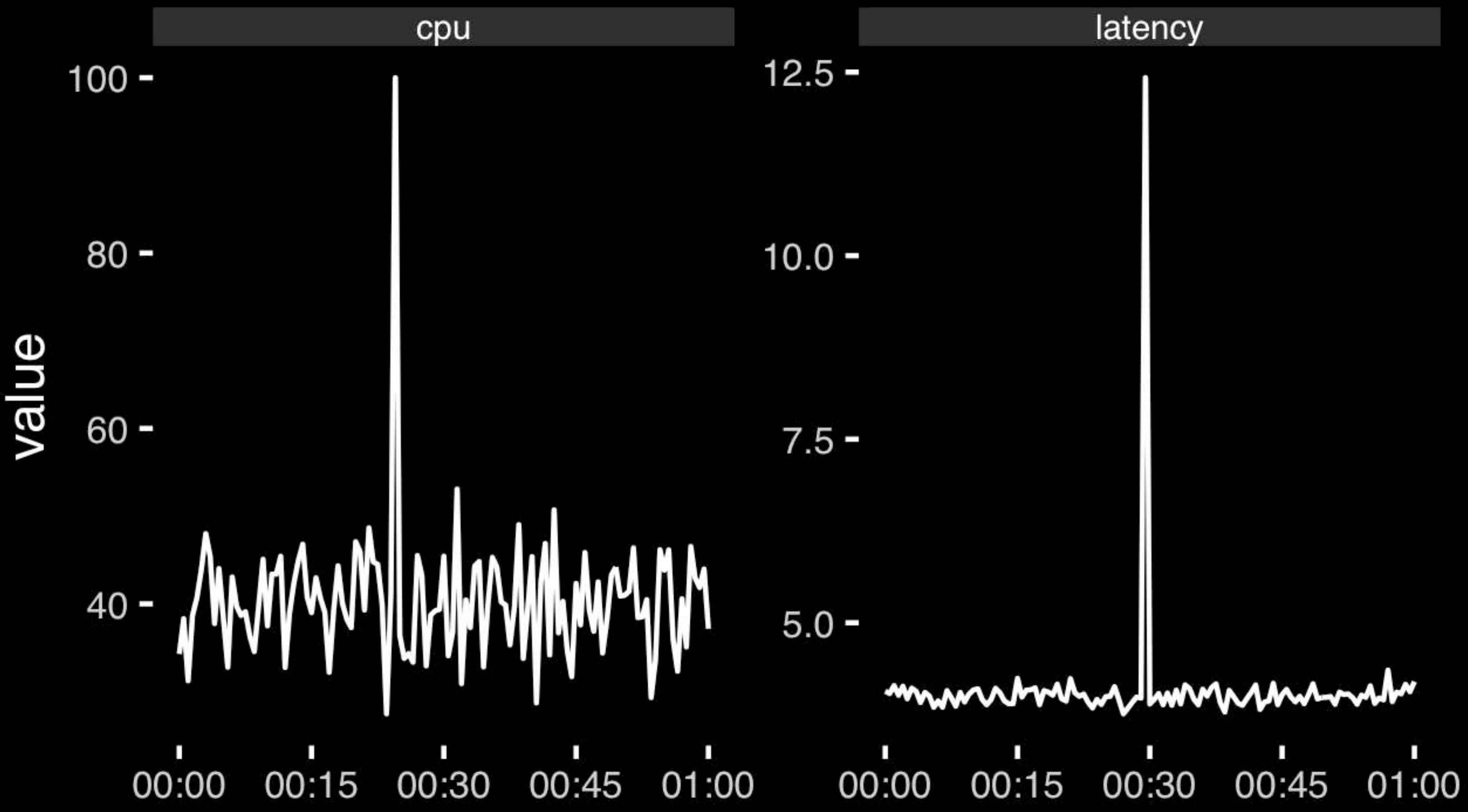
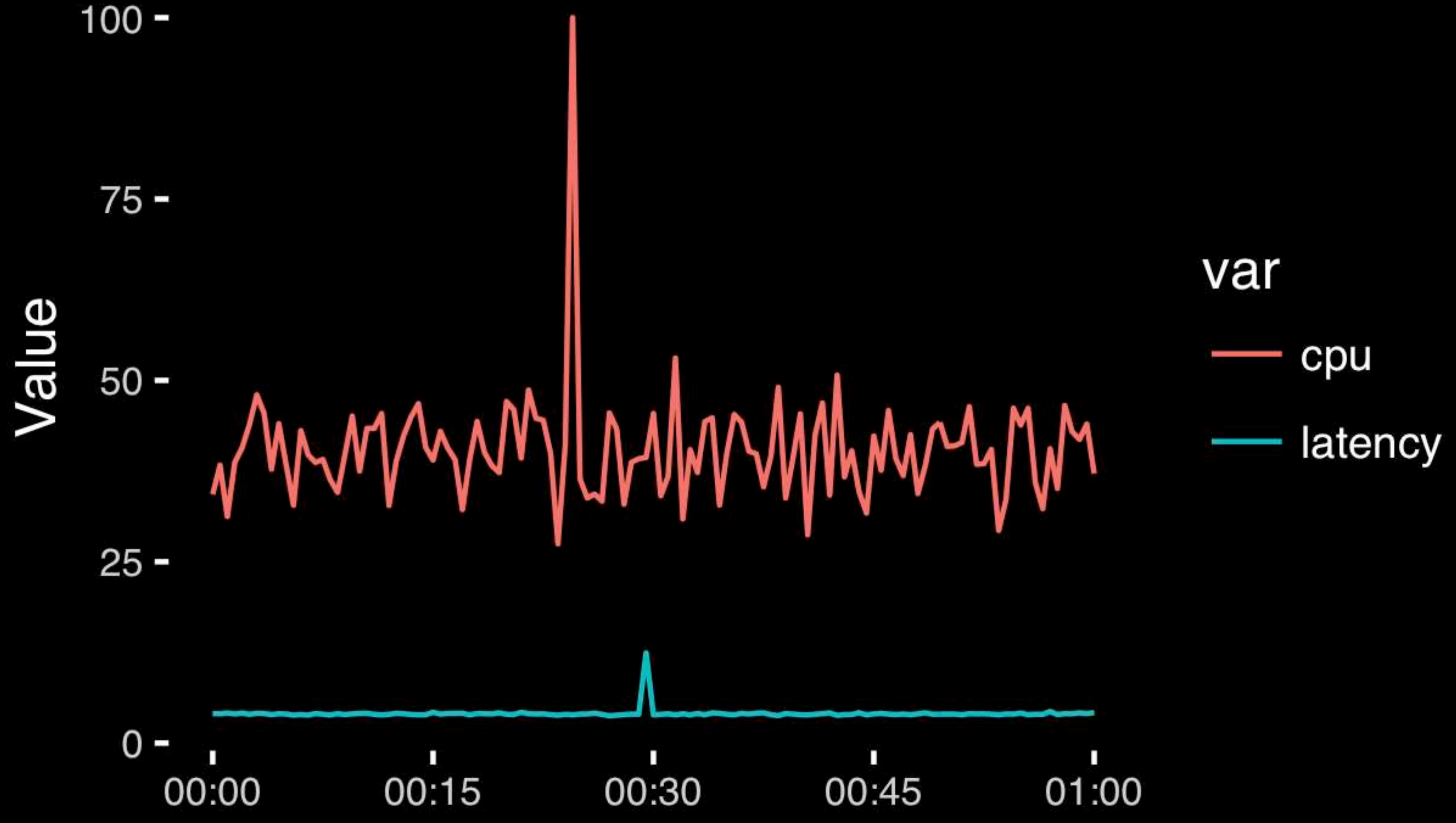
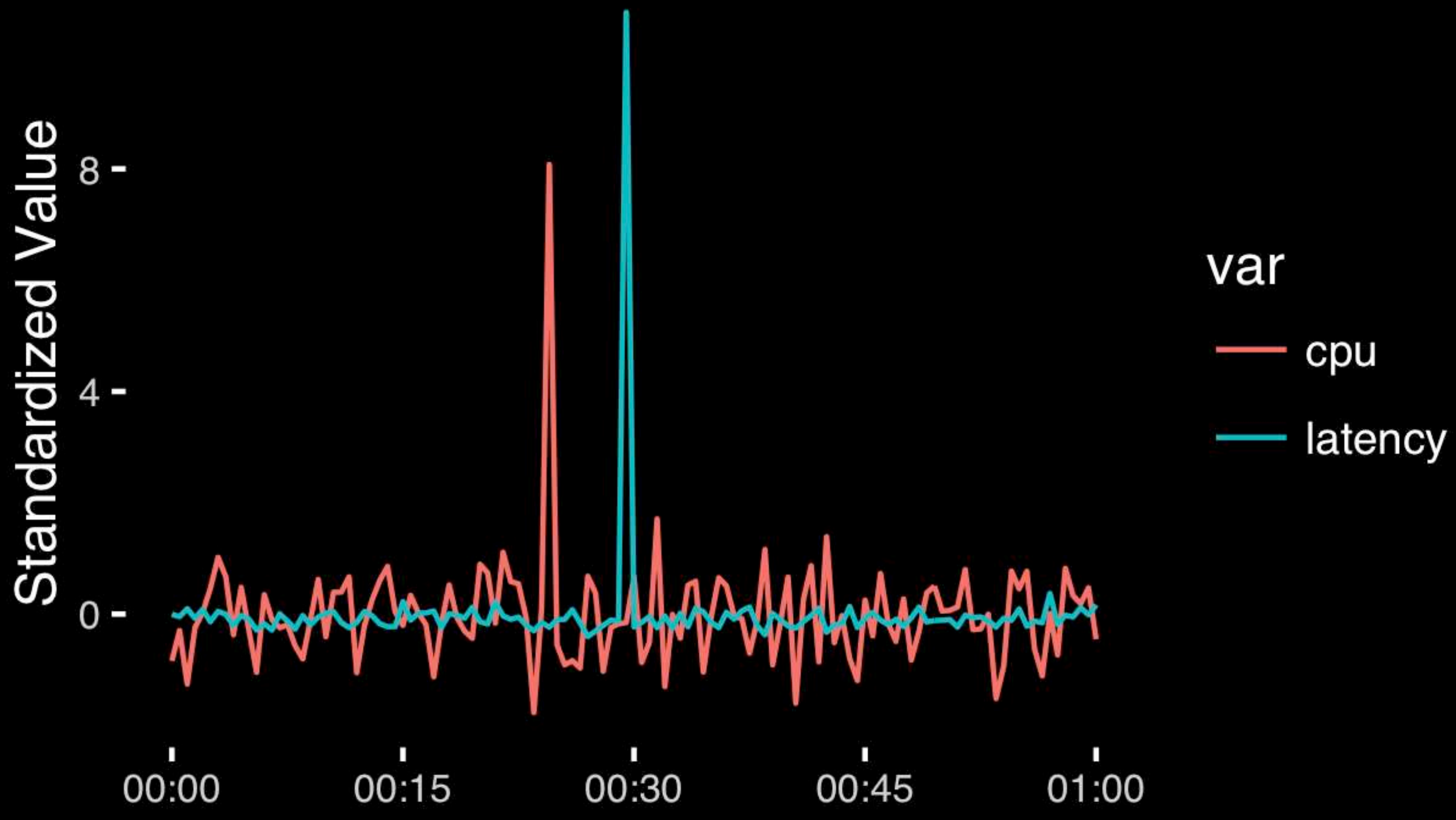
mpg

Observation: Comparison is trivial on a common scale.


cpu
latency
value
100
80
60
40
12.5
10.0
7.5
5.0
00:00
00:15
00:30
00:45
01:00
00:00
00:15
00:30
00:45
01:00


100
75
50
25
0
Value
00:00
00:15
00:30
00:45
01:00
var
cpu
latency


Standardized Value
8
4
0
00:00
00:15
00:30
00:45
01:00
var
cpu
latency

Today’s Learning Goals

- Appreciate the major elements of **exploratory data analysis** and why it is important to visualize data.
- Be conversant with **data visualization best practices** and understand how good visualizations optimize for the human visual system.
- Be able to generate informative graphical displays including **scatterplots**, **histograms**, **bar graphs**, **boxplots**, **dendrograms** and **heatmaps** and thereby gain exposure to the extensive graphical capabilities of R.
- Appreciate that you can build even more complex charts with **ggplot** and additional R packages such as **rgl**.

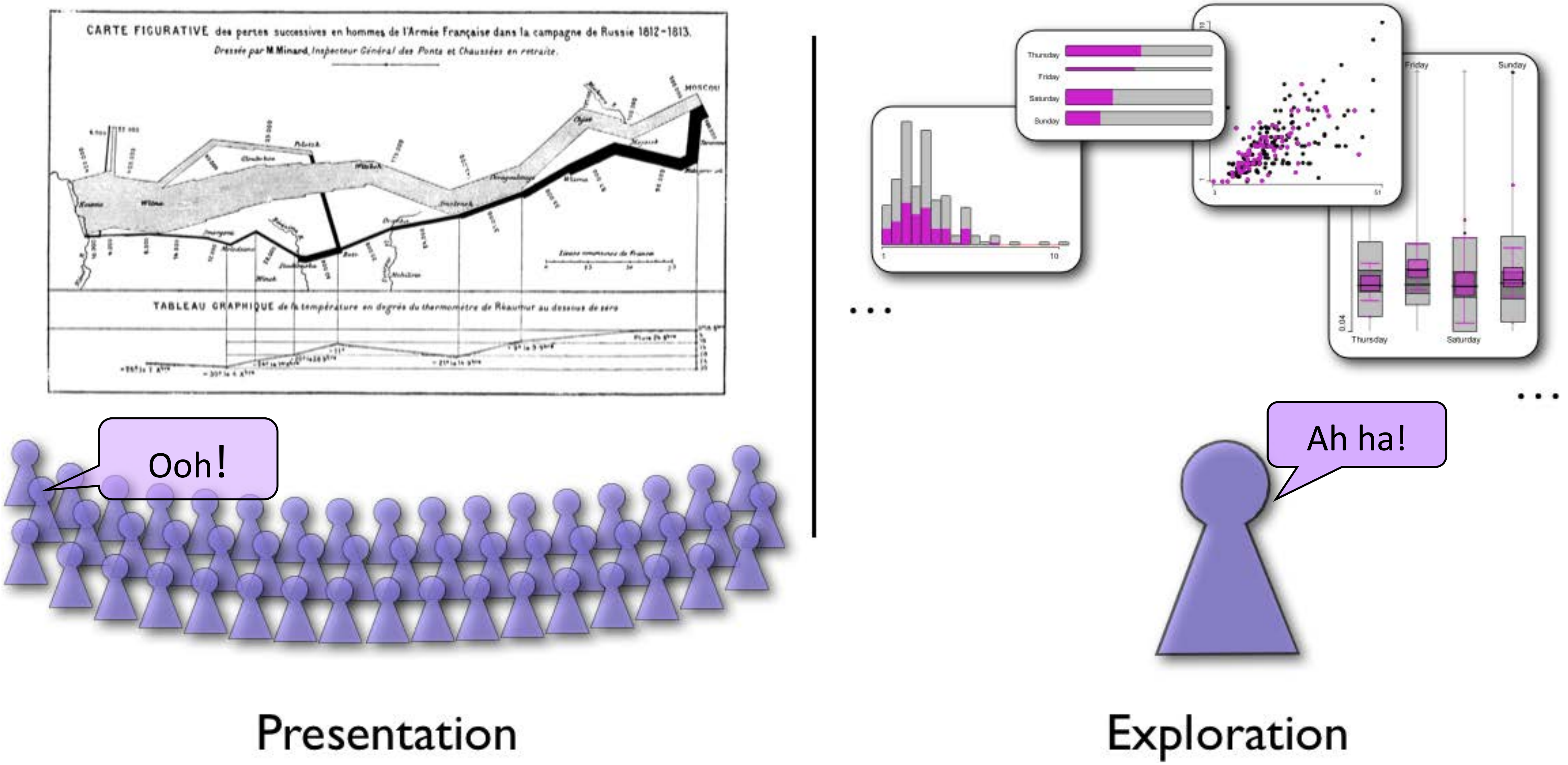
Do it Yourself!

Hands-on

Section 1 only please

Different graphs for different purposes

Exploratory graphs: many images for a narrow audience (you!)
Presentation graphs: single image for a large audience

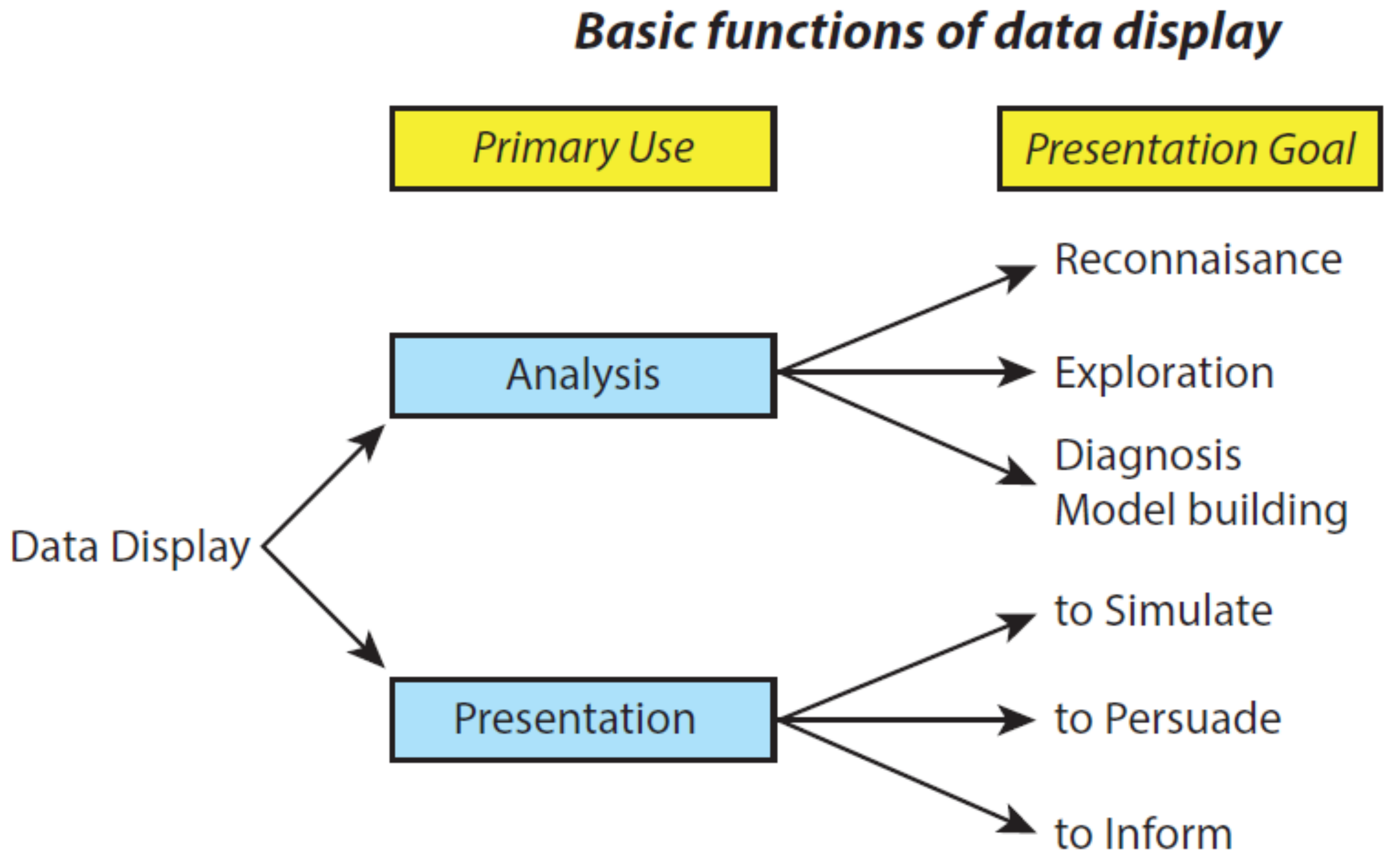

CARTE FIGURATIVE des pertes successives en hommes de l'Armée Française dans la campagne de Russie 1812-1813.
Dressée par M. Minard, Inspecteur Général des Ponts et Chaussées en retraite.
TABLEAU GRAPHIQUE de la température en degrés du thermomètre de Réaumur au dessous de zero
Ooh!
Thursday
Friday
Saturday
Sunday
...
Ah ha!
Presentation
Exploration

Roles of graphics in data analysis

- Graphs (& tables) are forms of communication:
  - What is the audience?
  - What is the message?

Analysis graphs: design to see patterns, trends, aid the process of data description, interpretation

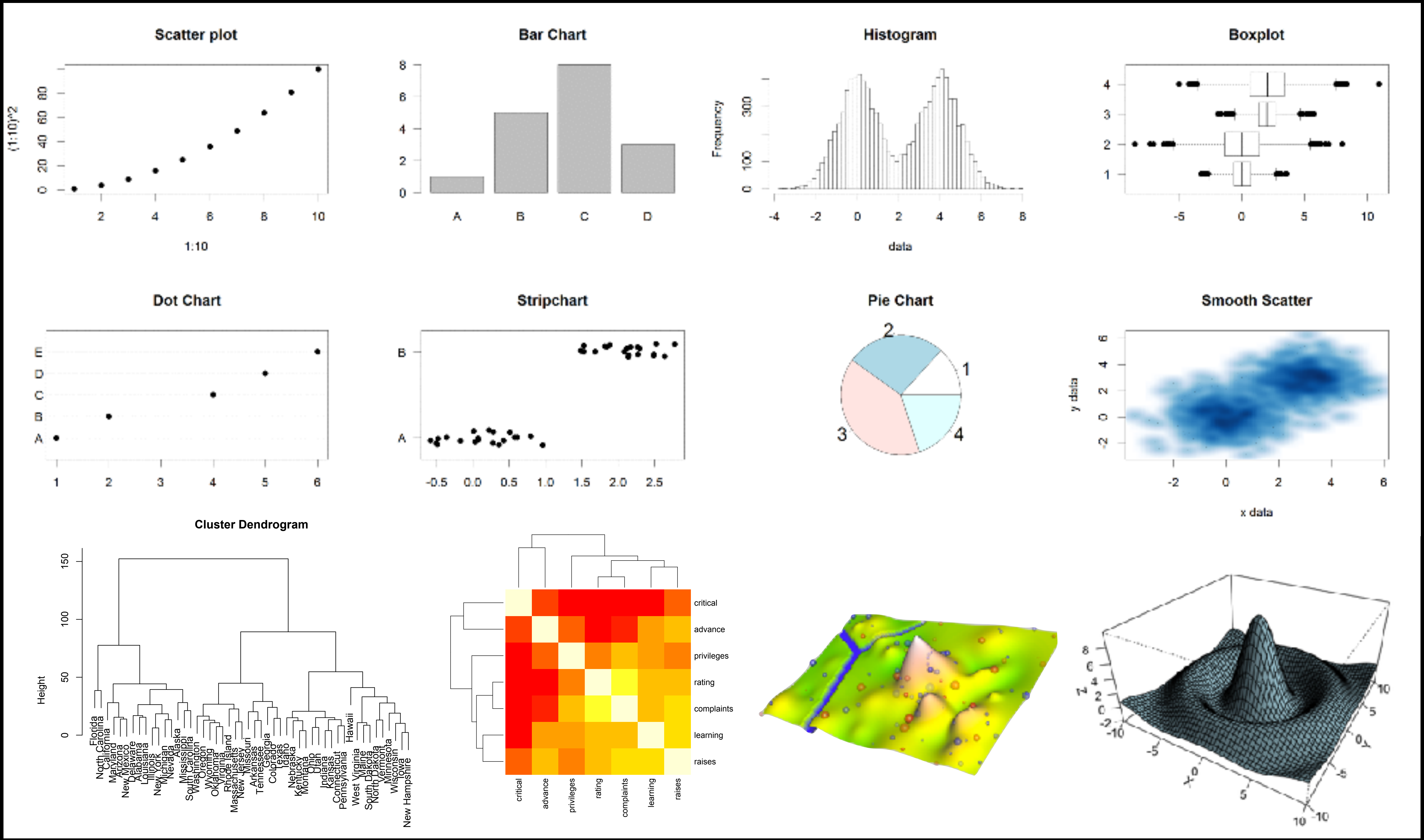
Presentation graphs: design to attract attention, make a point, illustrate a conclusion



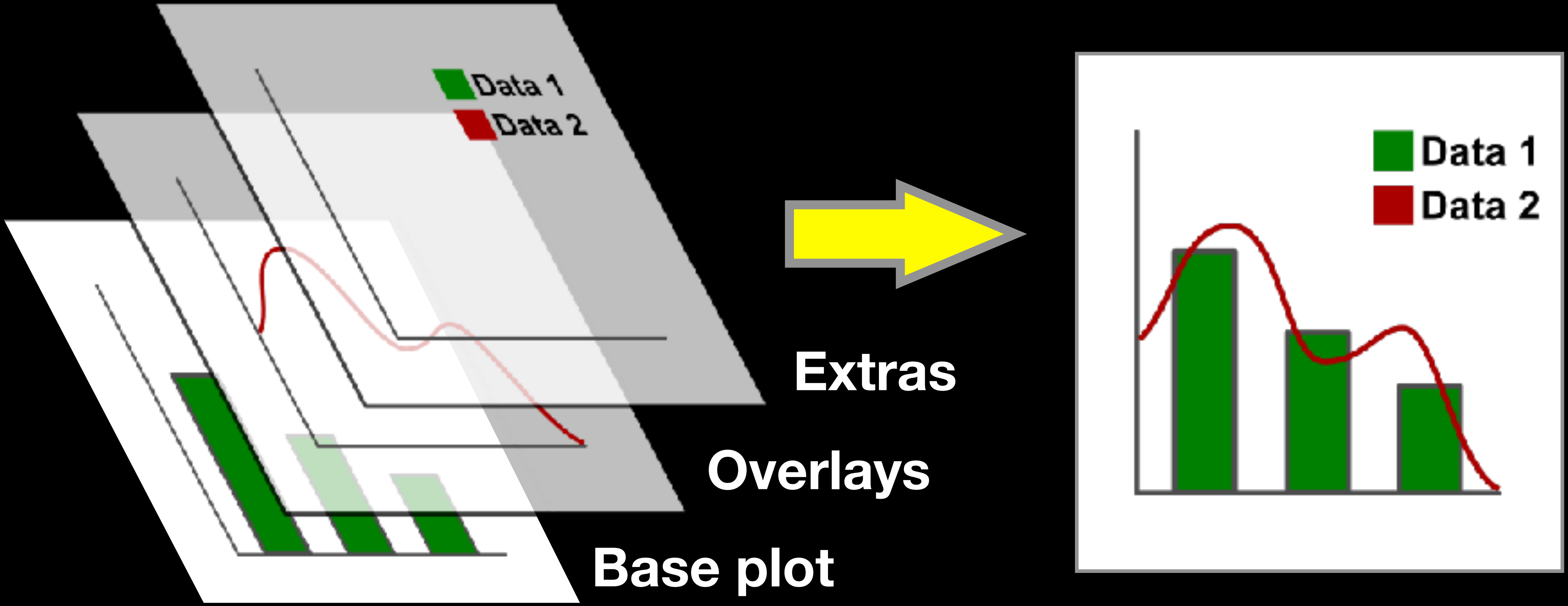
Exploratory (for you!)

Info for others, publications & sharing etc.

Core R Graph Types



The R Painters Model



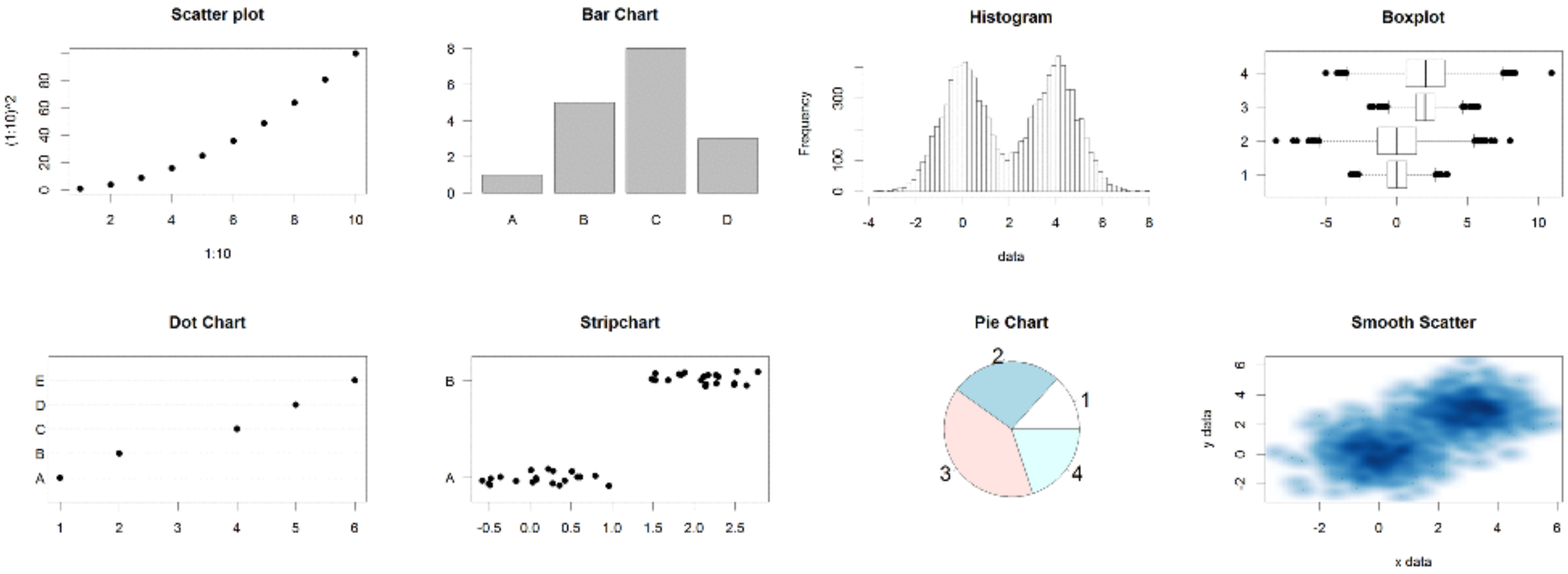
Extras

Overlays

Base plot

Side-Note: “Red and green should never be seen”

Core Graph Types



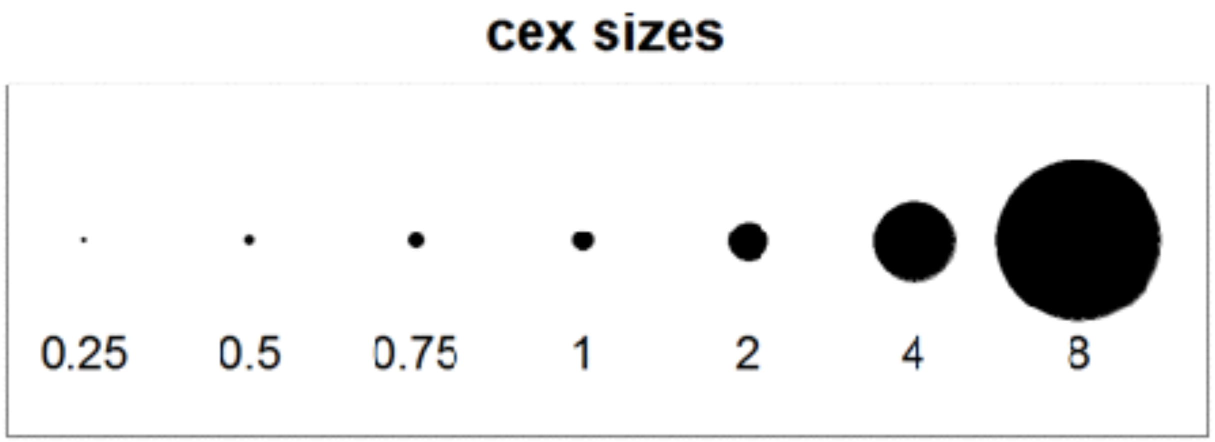
- Local options to change a specific plot
- Global options to affect all graphs

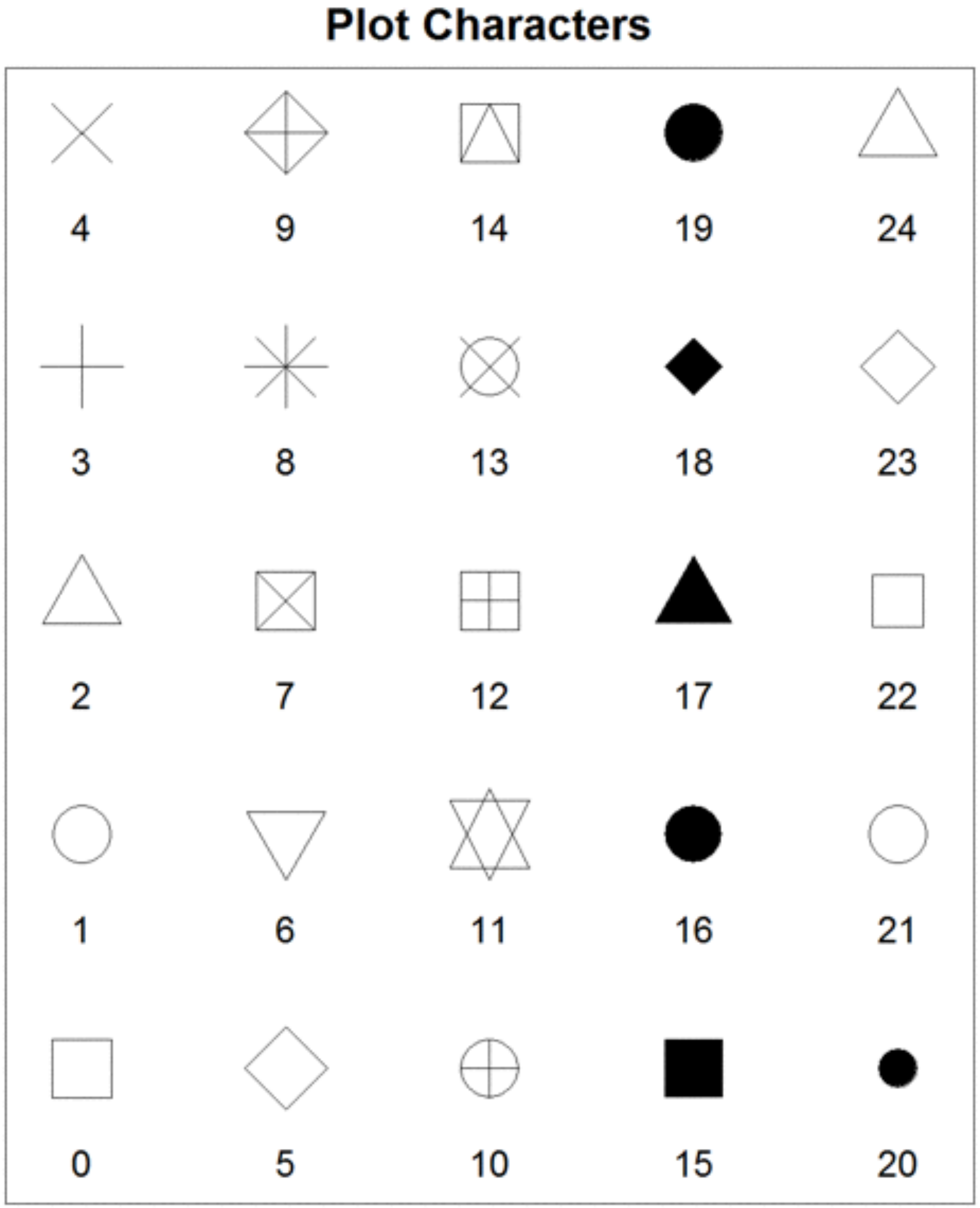
Common Options

- Axis scales
  - `xlim c(min,max)`
  - `ylim c(min,max)`
- Axis labels
  - `xlab(text)`
  - `ylab(text)`
- Plot titles
  - `main(text)`
  - `sub(text)`
- Plot characters
  - `pch(number)`
  - `cex(number)`

- Local options to change a specific plot
- Global options to affect all graphs

Plot Characters

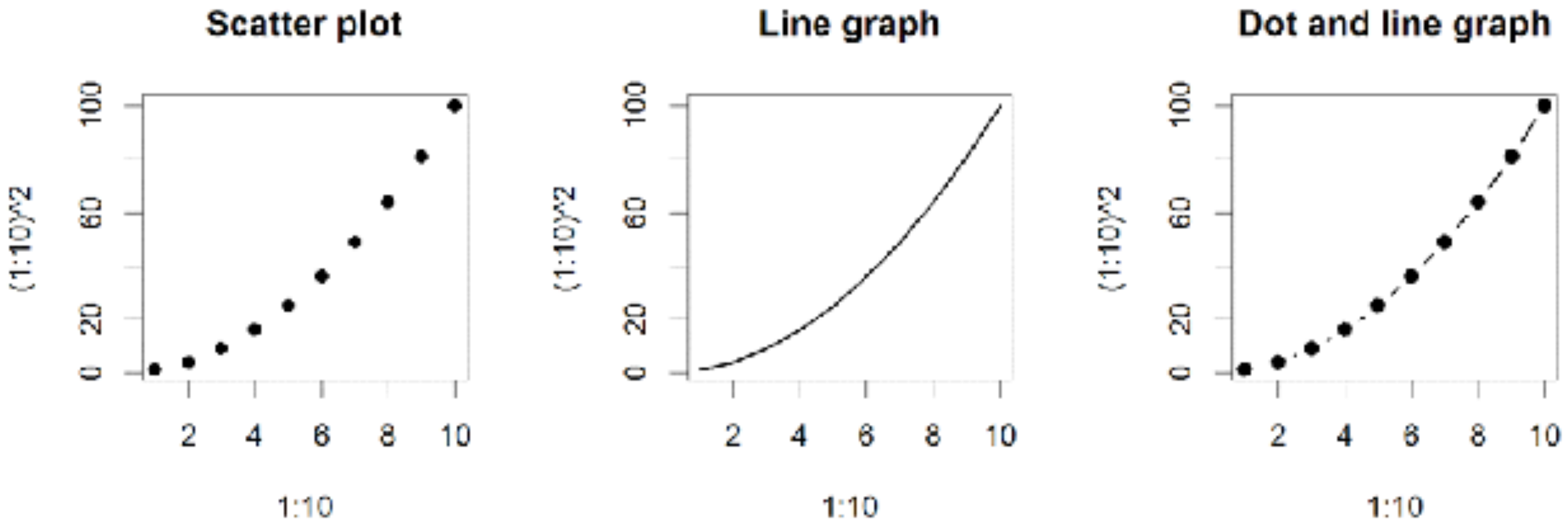




```
plot( 1:5, pch=1:5, cex=1:5 )
```

Plot Type Specific Options

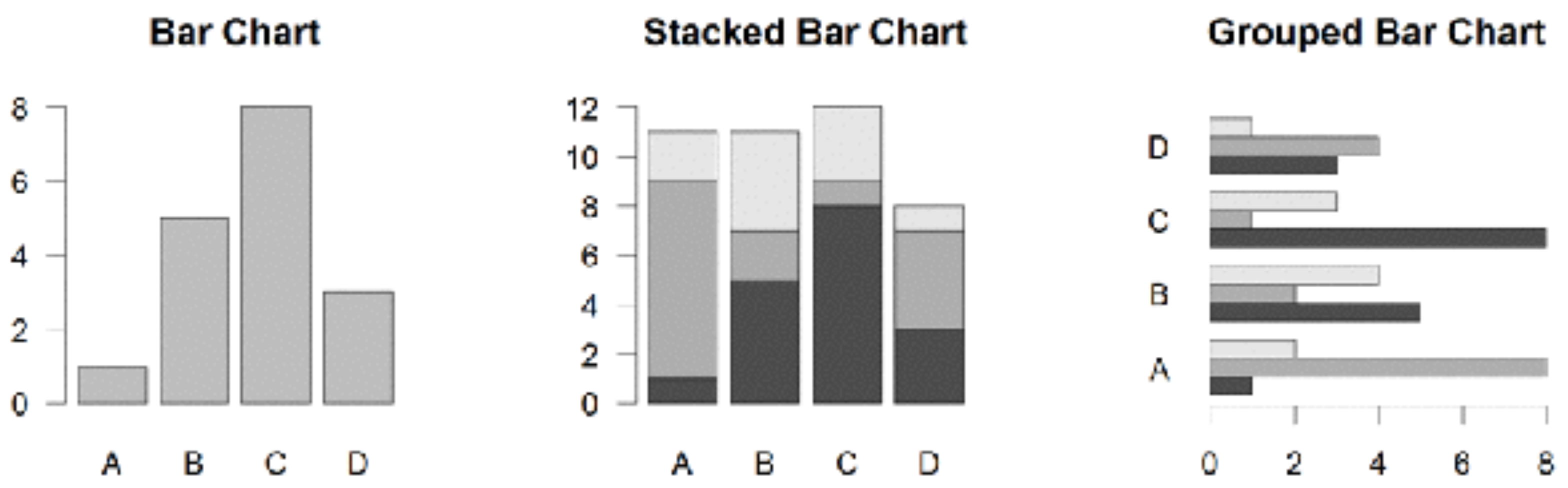
Plot (scatterplots and line graphs)



- Input: Almost anything. 2 x Vectors
- Output: Nothing
- Options:
  - `type` l=line, p=point, b=line+point
  - `lwd` line width (thickness)
  - `lty` line type (1=solid,2=dashed,3=dotted etc.)

```
plot( c(1:10)^2, typ="b", lwd=4, lty=3 )
```

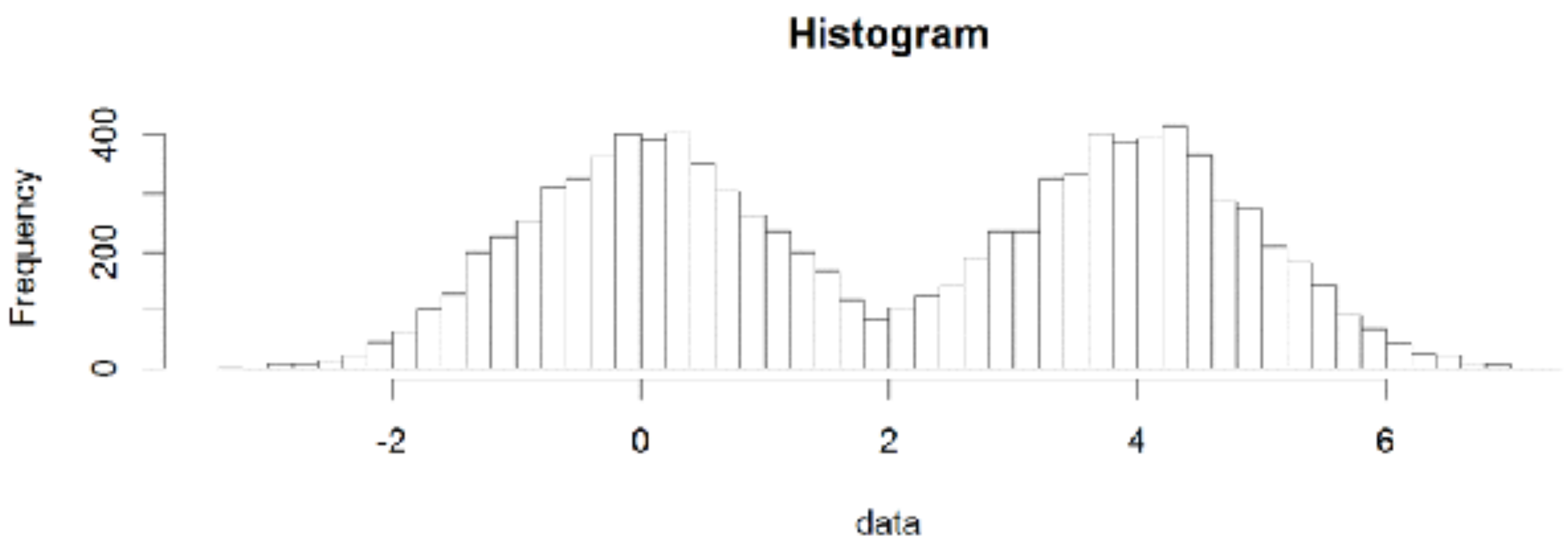
Barplot (bar graphs)



- Input: Vector (single) or Matrix (stack or group)
- Output: Bar centre positions
- Options:
  - – `names.arg` Bar labels (if not from data)
  - – `horiz=TRUE` Plot horizontally
  - – `beside=TRUE` Plot multiple series as a group not stacked

```
barplot(VADeaths, beside = TRUE)
```

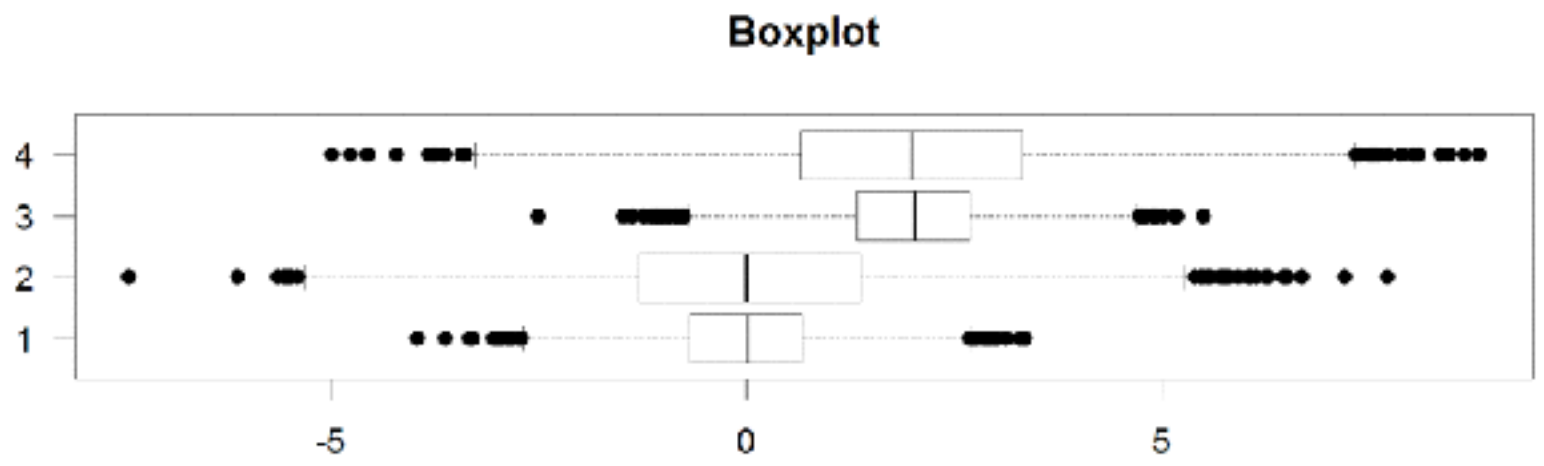
Hist (histograms)



- Input: Vector
- Output: Summary of binned data
- Options:
  - `breaks` Number or limits of bins
  - `probability` Y axis is probability, not freq
  - `labels` Per bin text labels

```
hist( c( rnorm(1000,0), rnorm(1000,4) ), breaks=20 )
```

Boxplot



- Input: Vector, List or formula (`data~factor`)
- Output: Summary of the boxplot parameters
- Options:
  - `range` Sensitivity of whiskers
  - `varwidth` Width represents total observations
  - `horizontal` Plot horizontally

```
boxplot( cbind( rnorm(1000,0), rnorm(1000,4) ) )
```

Controlling plot area options with `par`

Par

- The `par` function controls global parameters affecting all plots in the current plot area

- Changes affect all subsequent plots

- Many par options can also be passed to individual plots

?par

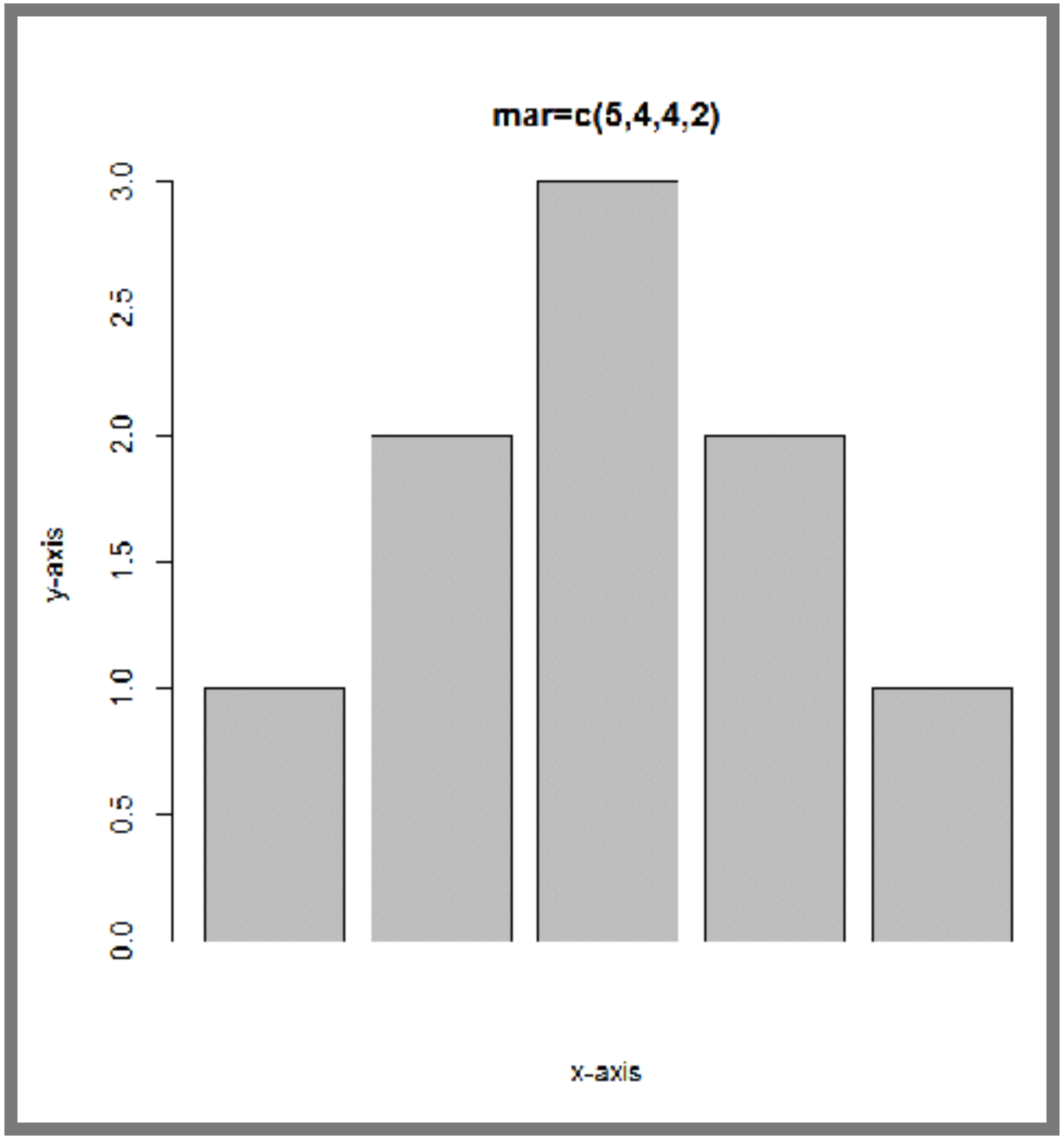
Par examples

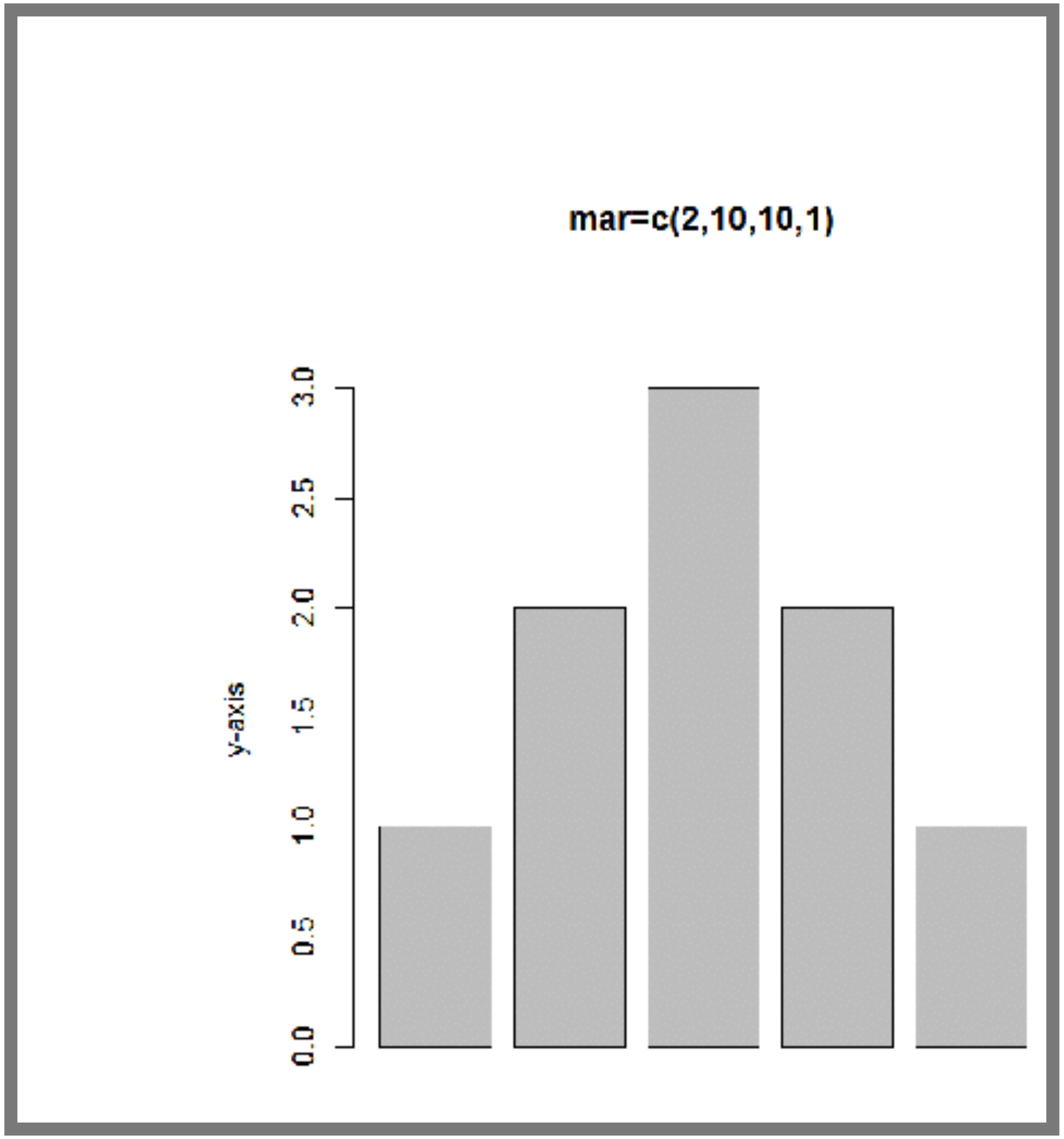
- Reading current value
  - `par()$cex`
- Setting a value
  - `par(cex=1.5) -> old.par`
- Restoring a value
  - `par(old.par)`
  - `dev.off()`

Par options

- Margins
  - `mai` (set margins in inches)
  - `mar` (set margins in number of lines)
  - `mex` (set lines per inch)
  - 4 element vector (bottom, left, top, right)
- Warning
  - `Error in plot.new() : figure margins too large`

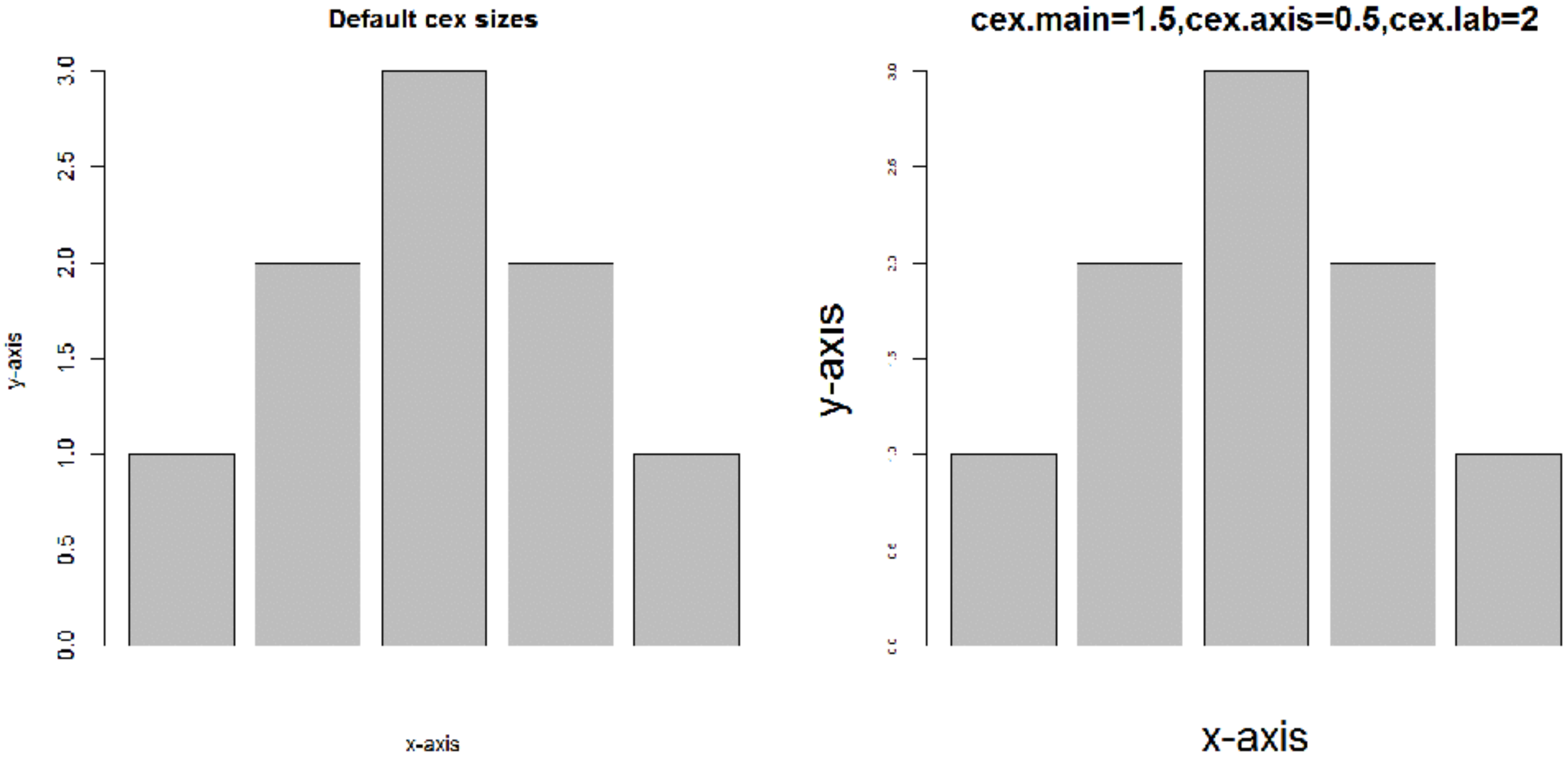
```
par( mar=c(2, 10, 1, 1) )
```


mar=c(5,4,4,2)
y-axis
0.0
0.5
1.0
1.5
2.0
2.5
3.0
x-axis


mar=c(2,10,10,1)
y-axis
0.0
0.5
1.0
1.5
2.0
2.5
3.0

Par options

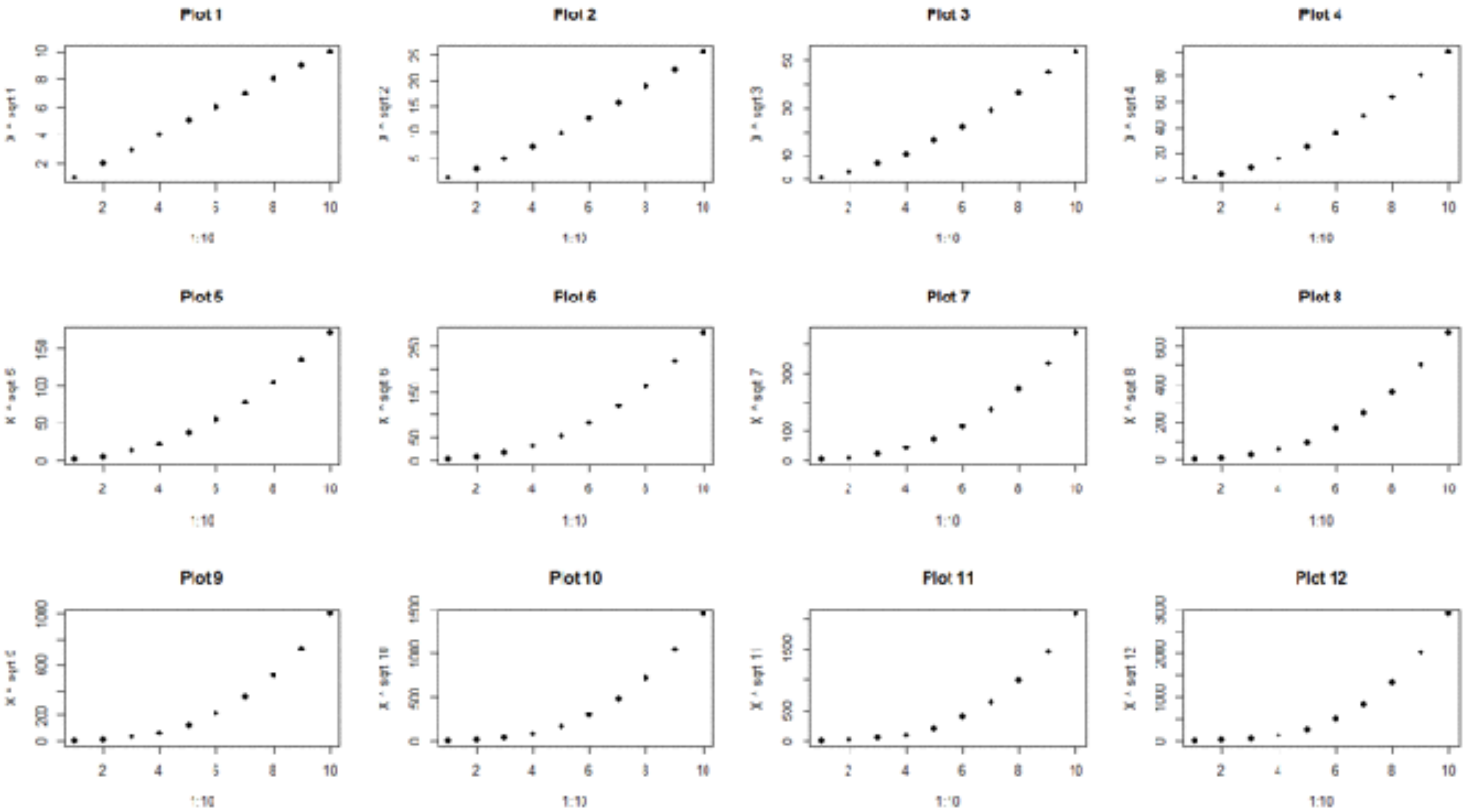
- Fonts and labels
  - `cex` - global char expansion
    - `cex.axis`
    - `cex.lab`
    - `cex.main`
    - `cex.sub`



par(cex.main=1.5, cex.axis=0.5, cex.lab=2)

Par options

- Multi-panel
  - `par( mfrow(rows,cols) )`



par(mfrow(3 , 4))

Do it Yourself!

Hands-on

Section 2 only please

Using Color

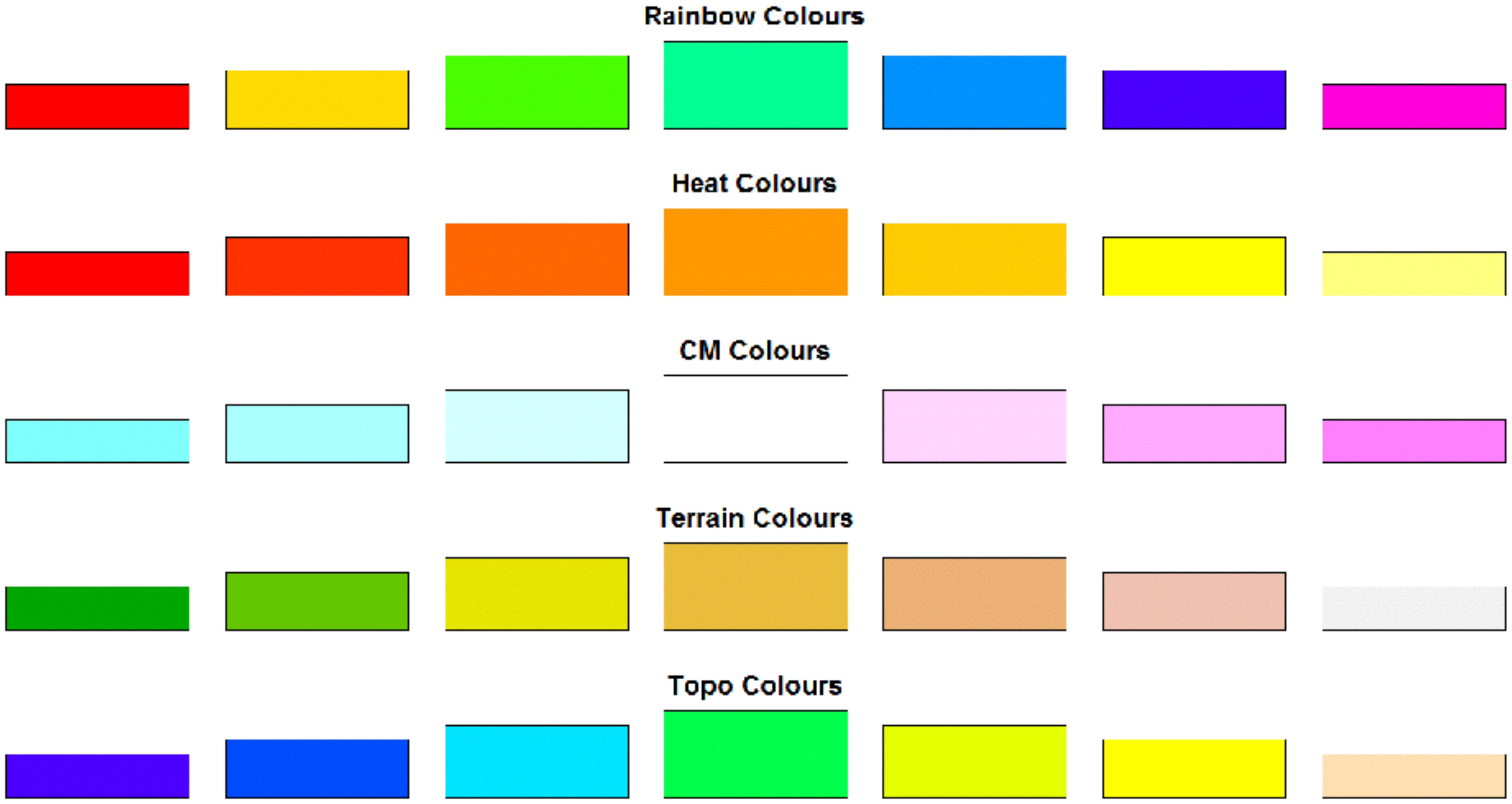
Specifying colors

- Hexadecimal strings
  - `#FF0000` (red)
  - `#0000FF` (blue)
  - `#CC00CC` (purple)
- Controlled names
  - `“red” “green”` etc.
  - `colors()`

Built in color schemes

- Functions to generate colors
- Pass in number of colors to make
- Functions:
  - `rainbow()`
  - `heat.colors()`
  - `cm.colors()`
  - `terrain.colors()`
  - `topo.colors()`

rainbow(7)


Rainbow Colours
Heat Colours
CM Colours
Terrain Colours
Topo Colours

```
rainbow( 7 )
```

Color Packages

- Color Brewer
  - Set of pre-defined, optimized palettes
  - `library(RColorBrewer)`
  - `brewer.pal(n_colours, palette)`
- ColorRamps
  - Create smooth palettes for ramped color
  - Generates a function to make actual color vectors
  - `colorRampPalette(c(“red”,”white”,”blue”))`
  - `colorRampPalette(c(“red”,”white”,”blue”))(5)`

Applying Color to Plots

- Vector of numbers or specified colors passed to the `col` parameter of a plot function
- Vector of **factors** used to divide the data
  - Colors will be taken from the set color **palette**
  - Can read or set using **pallete** function
    - `palette()`
    - `palette(brewer.pal(9,”Set1”)`

```
plot( 1:5, col=1:5, pch=15, cex=2)
```

Dynamic use of color

- Coloring by density
  - Pass data and palette to `densCols()`
  - Vector of colors returned

- Coloring by value
  - Need function to map values to colors

https://www.rdocumentation.org/packages/grDevices/versions/3.4.3/topics/densCols

Do it Yourself!

Hands-on

Section 3 only please

For next day:
Section 4 Revisited

- Open your previous Lecture5 RStudio **project** (and your saved **R script**)

- Locate and open in RStudio the downloaded file color_to_value_map.r

- This is an example of a poorly written function typical of something you might get from a lab mate that knows some R…

(POOR!) Color Mapping Function

```
map.colors <- function (value,high.low,palette) {
  proportion <- ((value-high.low[1])/(high.low[2]-high.low[1]))
  index <- round ((length(palette)-1)*proportion)+1
  return (palette[index])
}
```

Talking point:

- Can you figure out what this function it is supposed to do?
- What format should the inputs be in order to work?
- How could we improve this function?

Homework!

New **DataCamp** Assignments

- Introduction to R Markdown
- Functions
- Loops

Muddy Point Assessment Form Link