### **BGGN 213**

Data visualization with R

Barry Grant UC San Diego

http://thegrantlab.org/bggn213

## 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).
- Be able to use R to read and parse comma-separated (.csv) formatted files ready for subsequent analysis.
- Appreciate how you can use R scripts to aid with reproducibility.



# 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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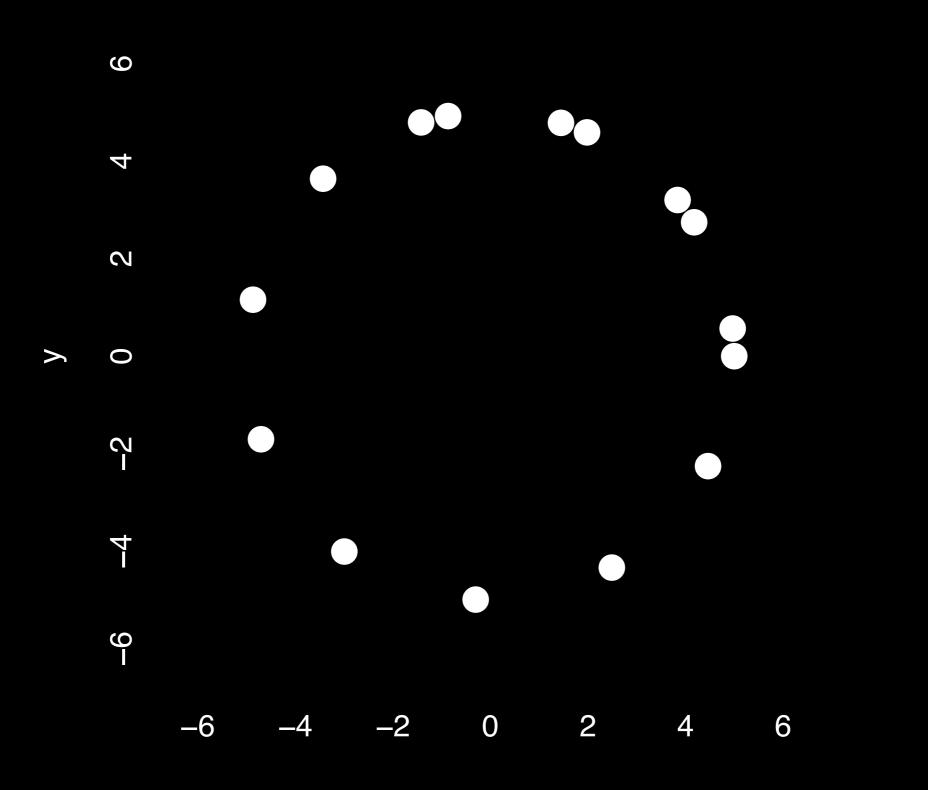
## Why visualize at all?

THE HERALD

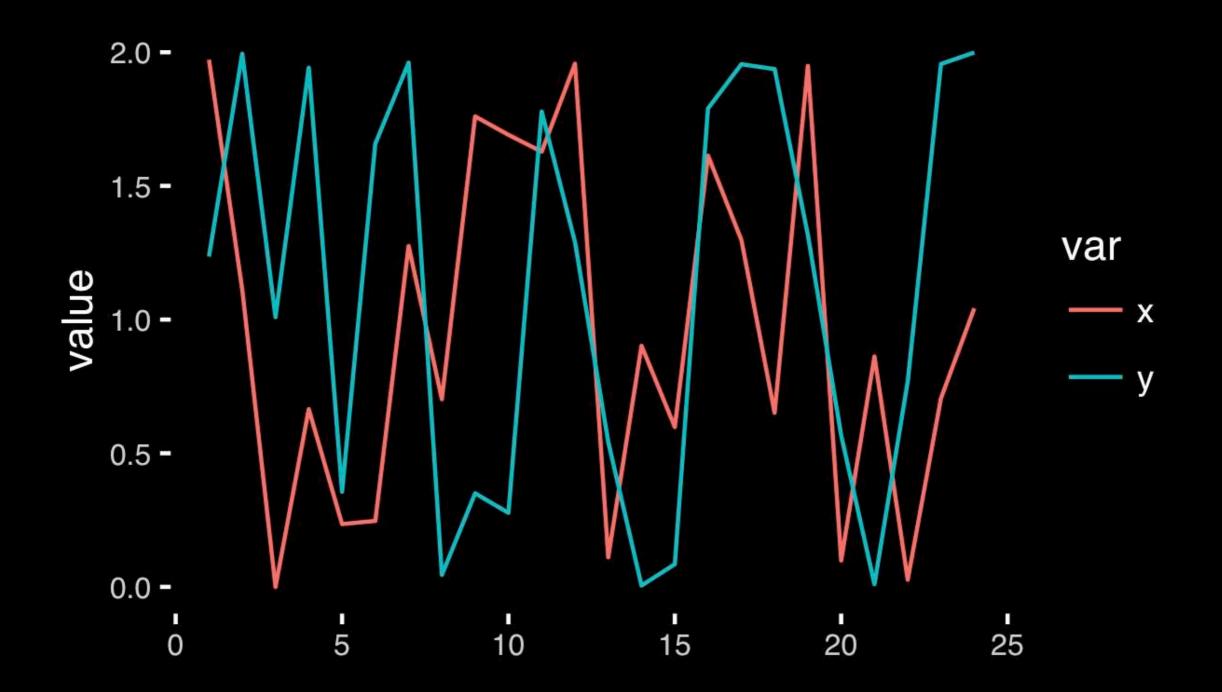
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	X	У		
1	5.00	0.00		
2	4.18	2.75		
3	1.98	4.59		
4	-0.86	4.92		
5	-3.43	3.64		
6	-4.86	1.16		
7	-4.70	-1.70		
8	-2.99	-4.01		
9	-0.30	-4.99		
10	2.49	-4.34		
11	4.46	-2.25		
12	4.97	0.57		
13	3.84	3.20		
14	1.45	4.79		
15	-1.42	4.79		

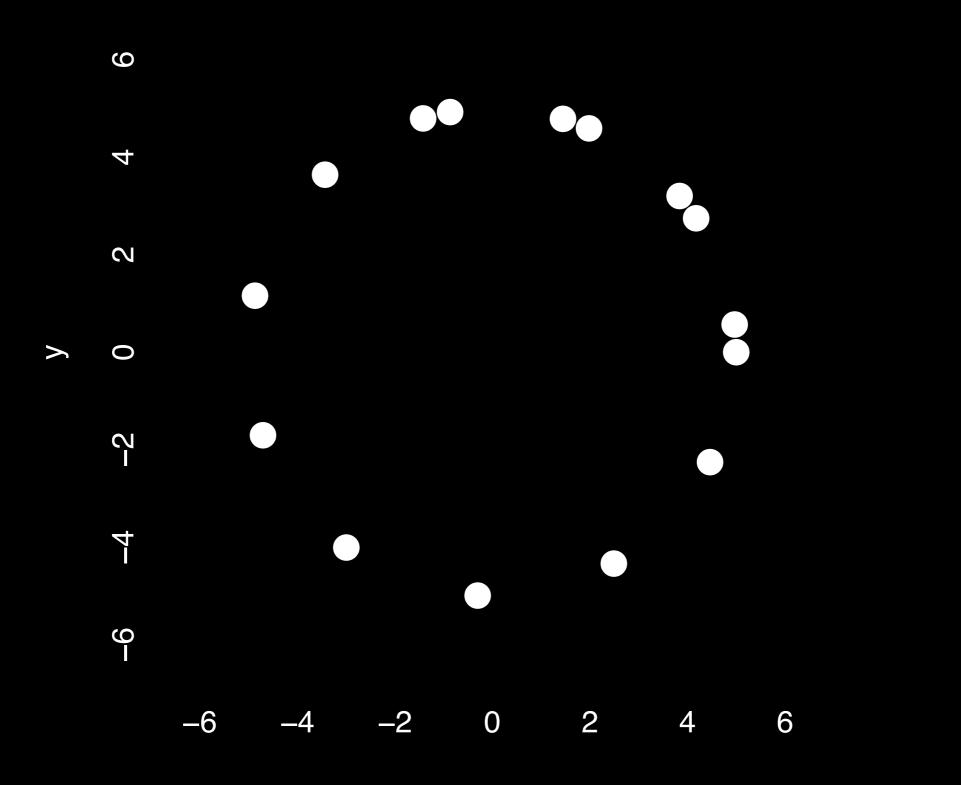
	X	У
Min.	-4.86	-4.99
1st Qu.	-2.21	-1.98
Median	1.45	1.16
Mean	0.65	0.87
3rd Qu.	4.01	4.12
Max.	5.00	4.92



Χ



#### https://bioboot.github.io/bggn213\_f17/class-material/08\_draw\_circle\_points/

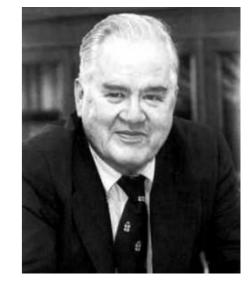


## **Exploratory Data Analysis**

- 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

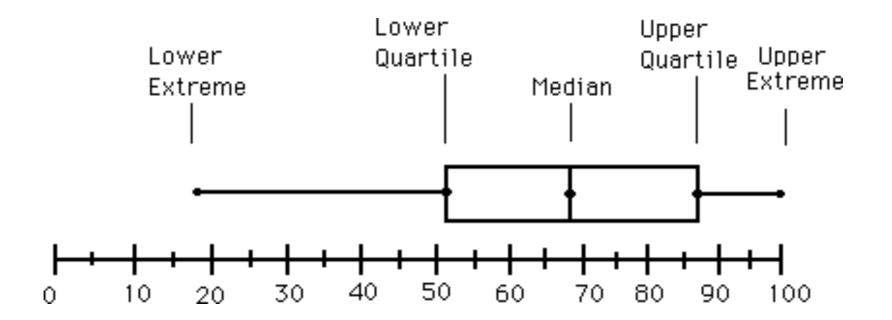


ORATORY DATA

John W. Tukey

### Chart types

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



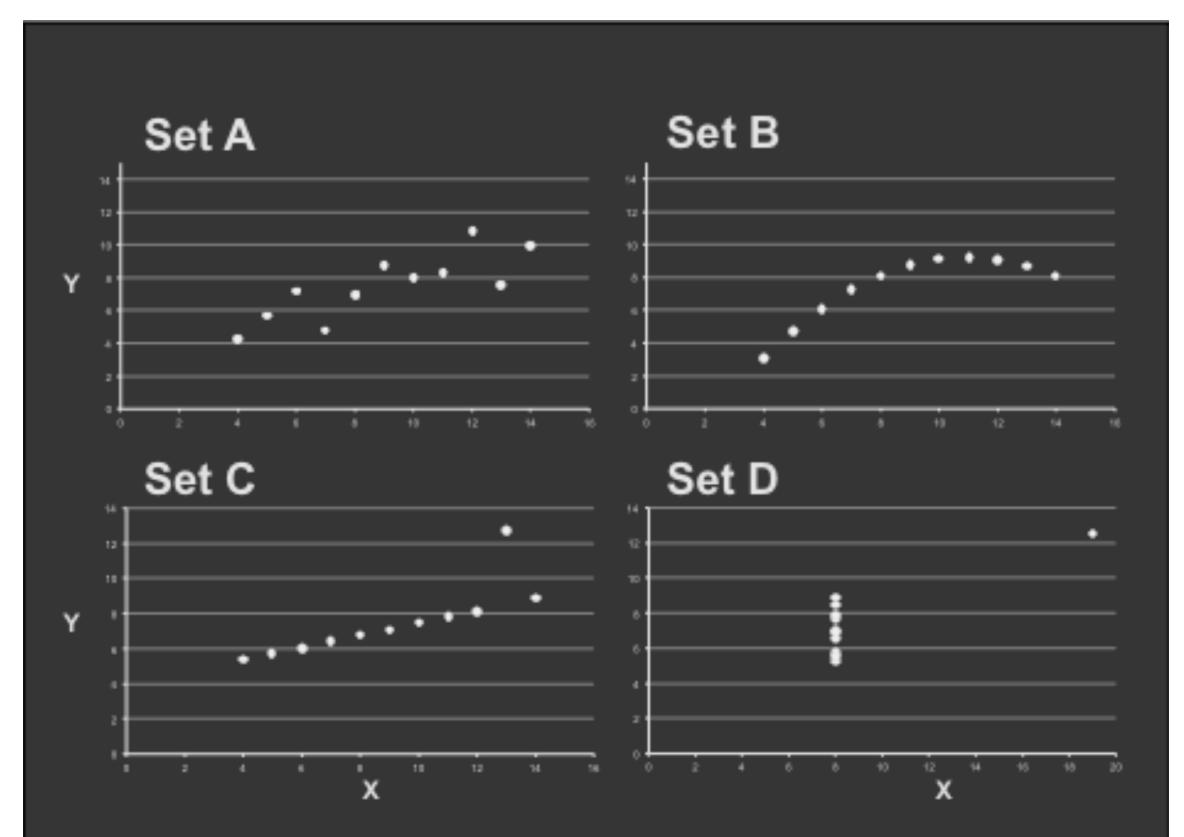
#### The Trouble with Summary Stats

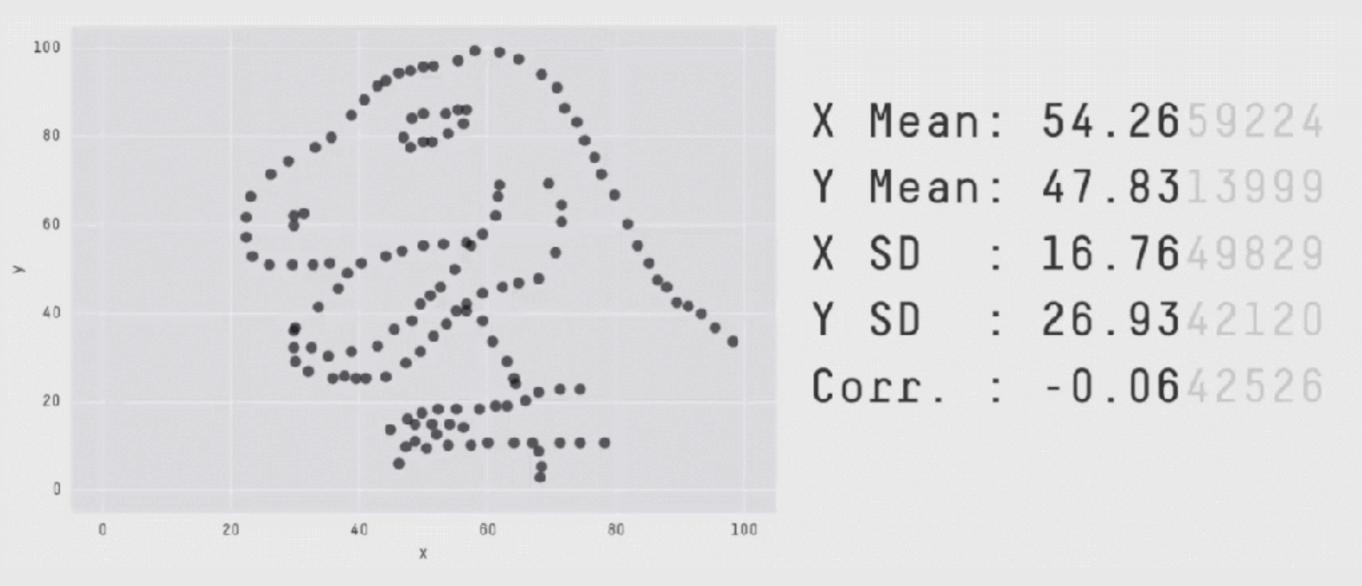
Set	t A	Se	t B	Se	t C	Se	t D
<u> </u>	<u>Y</u>	<u> </u>	<u>Y</u>	<u></u> X	Y	<u> </u>	<u>Y</u>
10	8.04	10	9.14	10	7.46	8	6.58
8	6.95	8	8.14	8	6.77	8	5.76
13	7.58	13	8.74	13	12.74	8	7.71
9	8.81	9	8.77	9	7.11	8	8.84
11	8.33	11	9.26	11	7.81	8	8.47
14	9.96	14	8.1	14	8.84	8	7.04
6	7.24	6	6.13	6	6.08	8	5.25
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 <sub>x</sub> = 9.0	σ <sub>x</sub> = 3.317	Y = 3 + 0.5 X	[Anscombe 73]
u <sub>y</sub> = 7.5	$\sigma_{\gamma} = 2.03$	R <sup>2</sup> = 0.67	

### Looking at Data

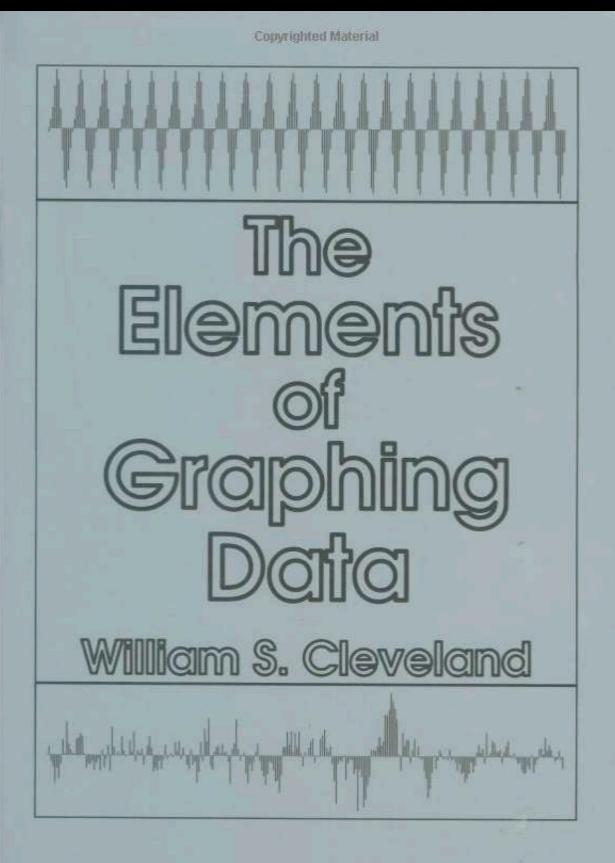




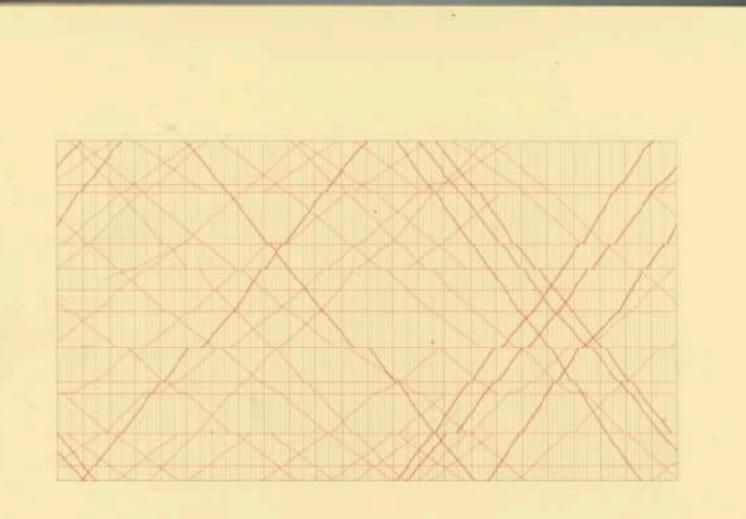
Key point: You need to visualize your data!

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### 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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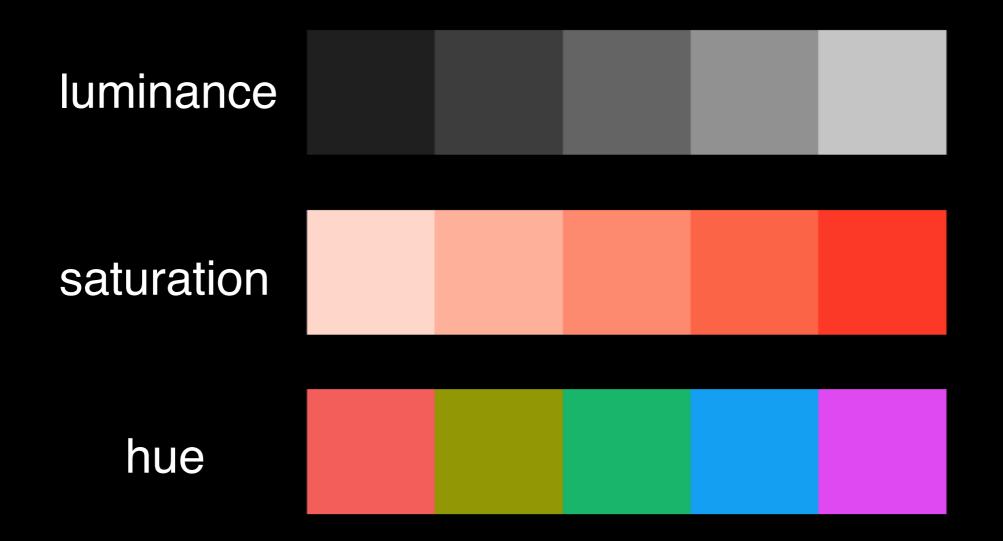
11

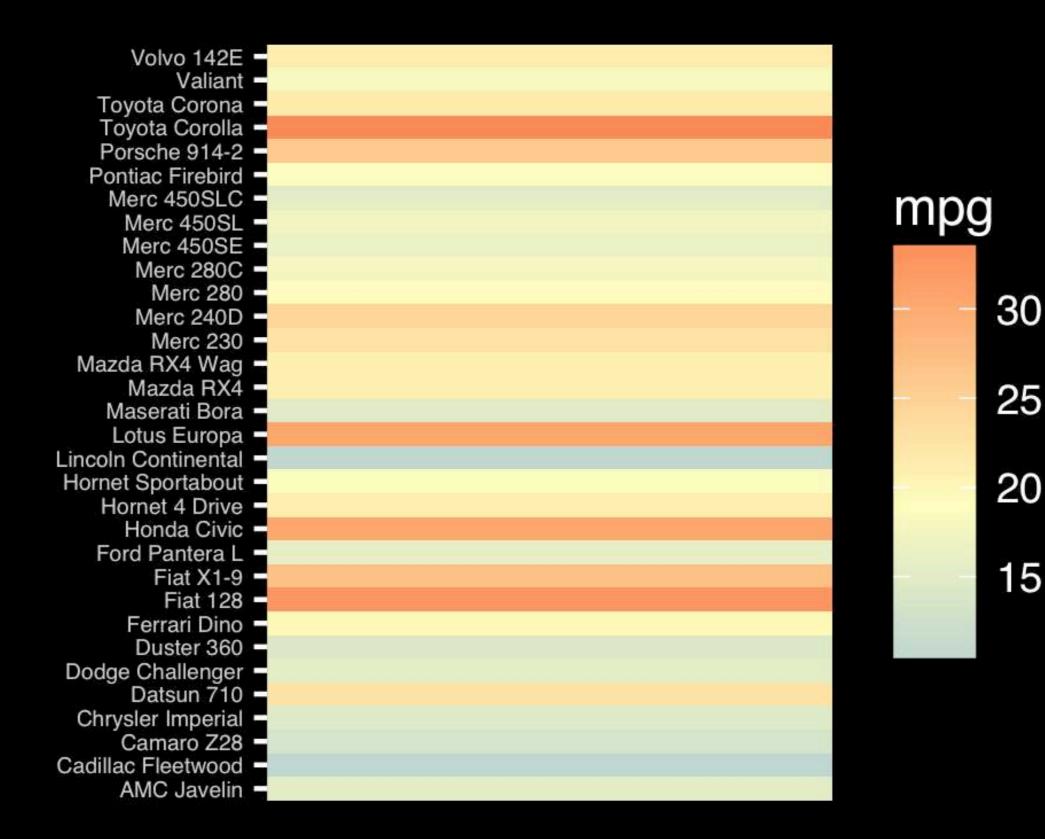
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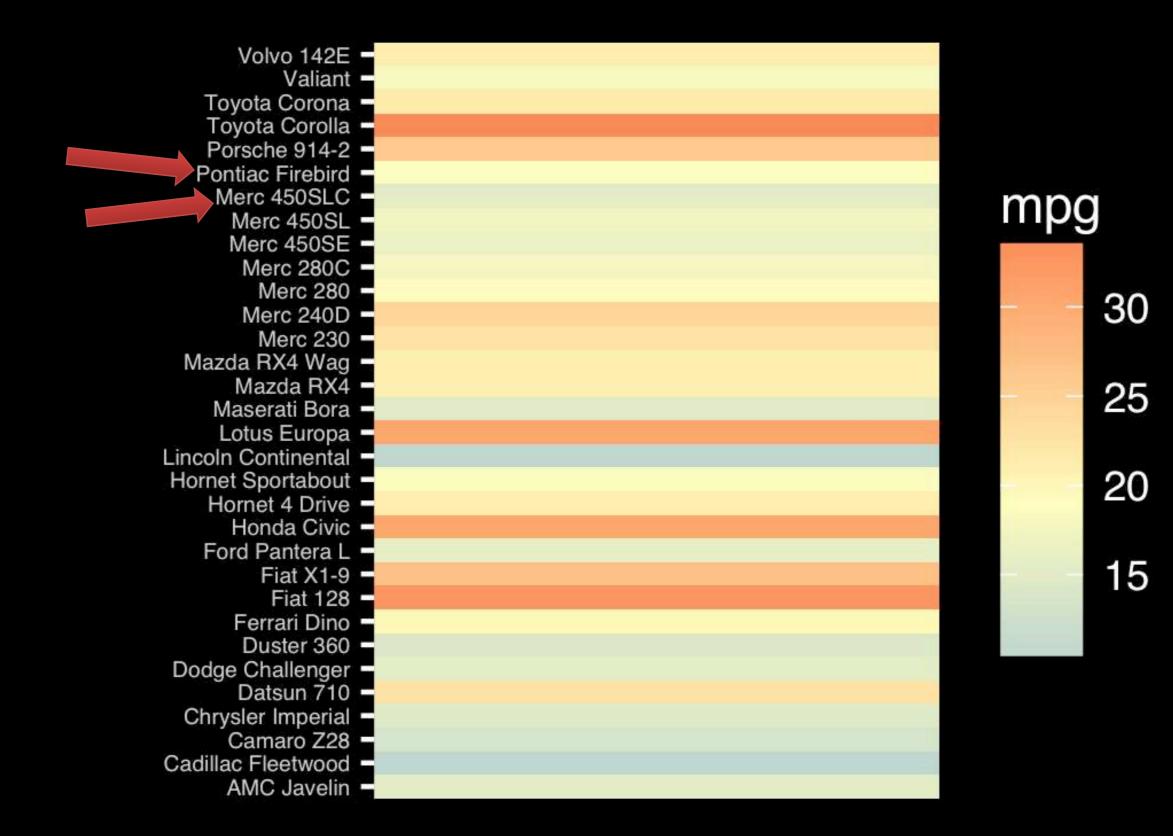
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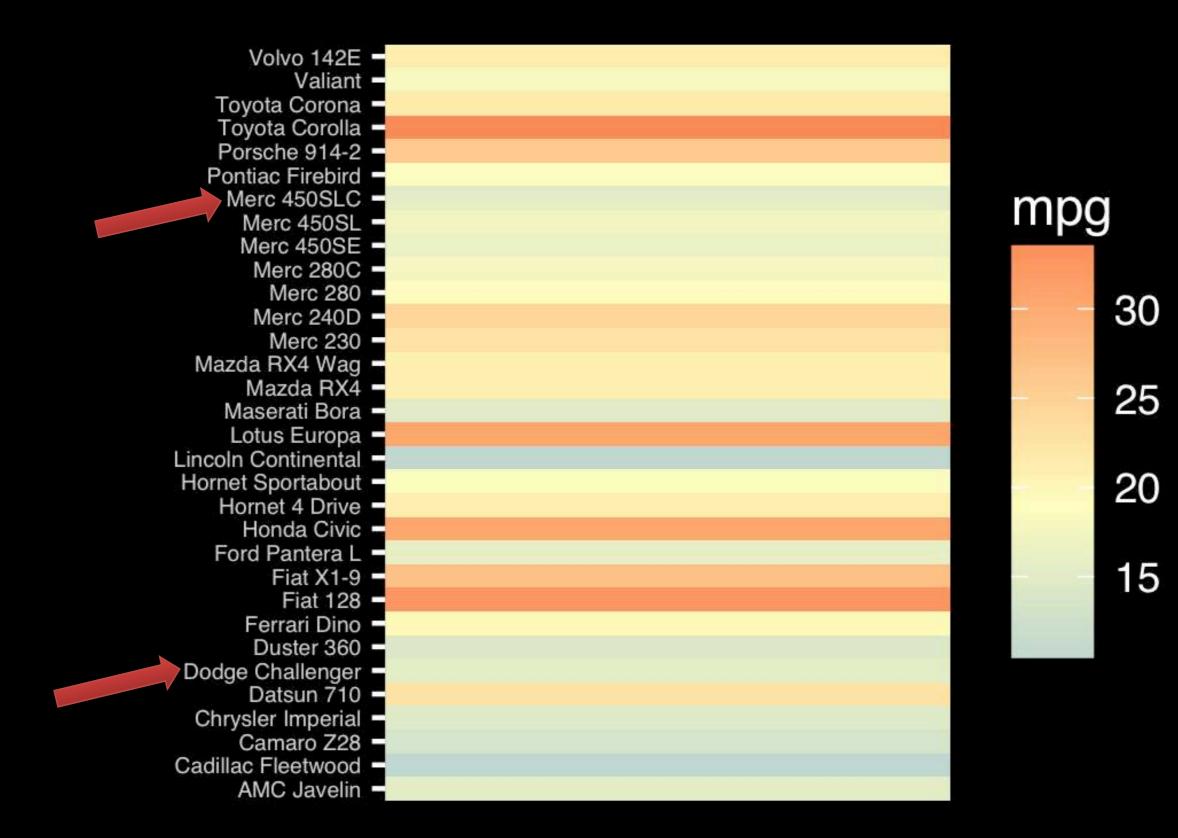
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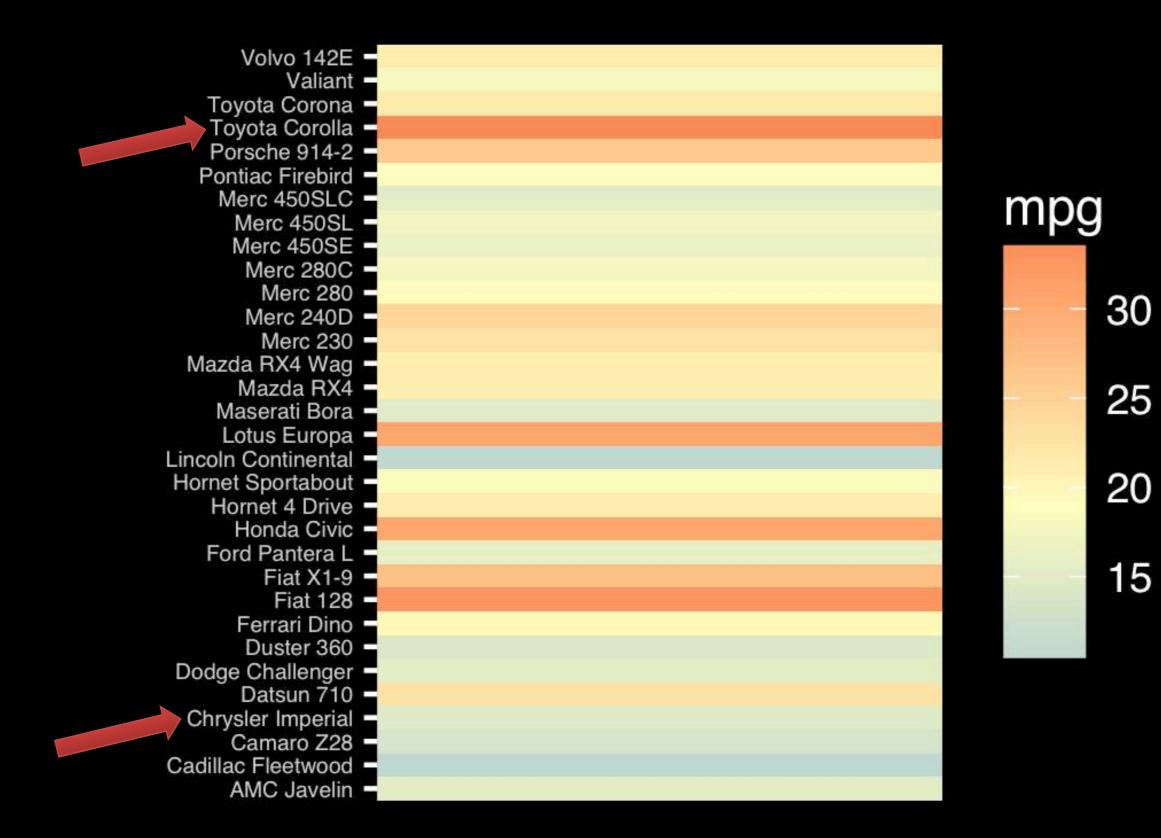
Volume or Density or Color saturation/hue



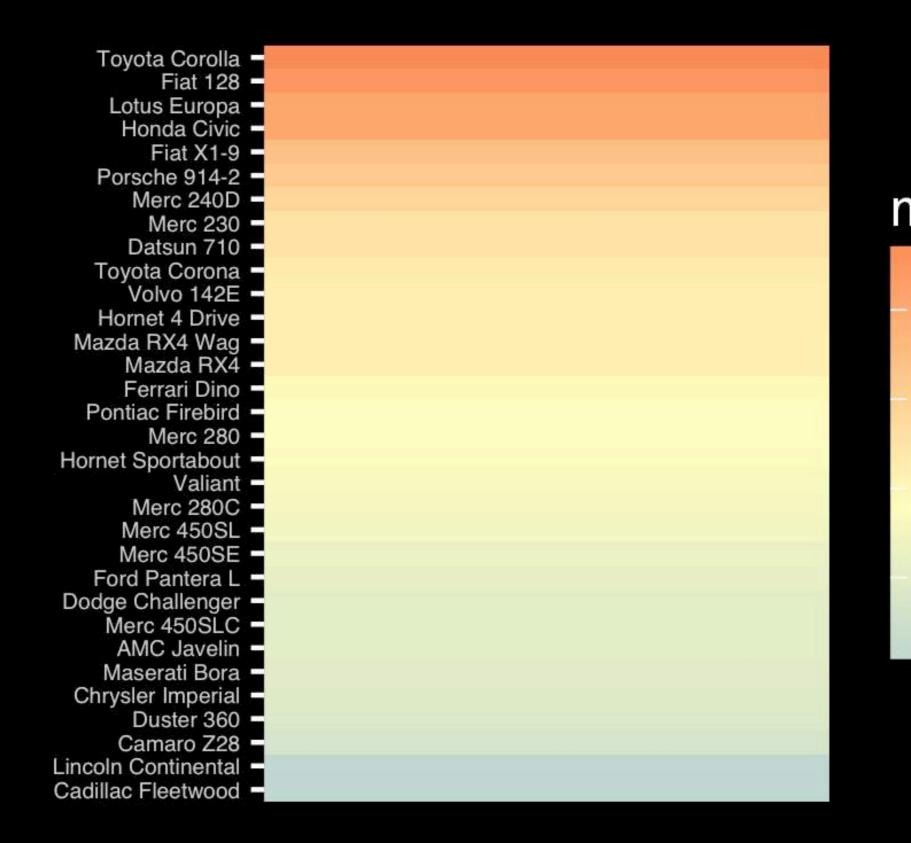


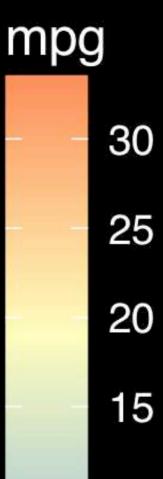


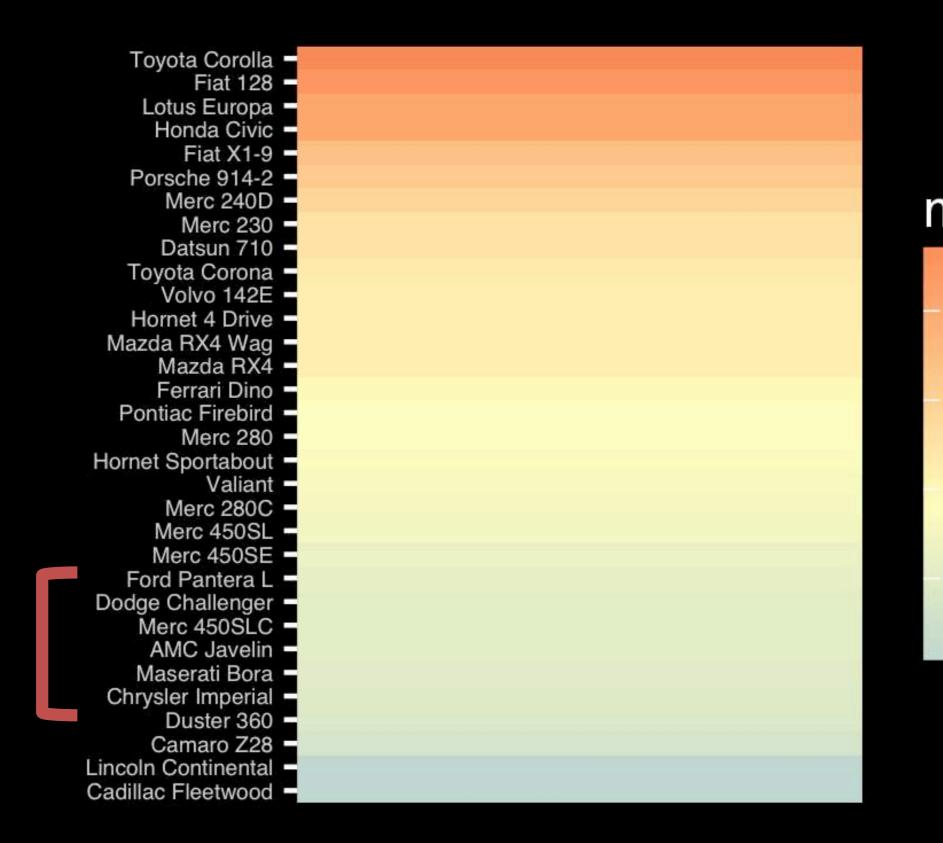


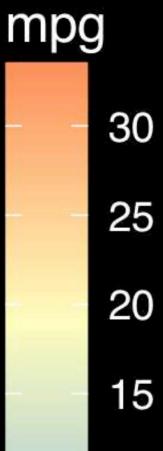


Observation: Alphabetical is almost never the correct ordering of a categorical variable.



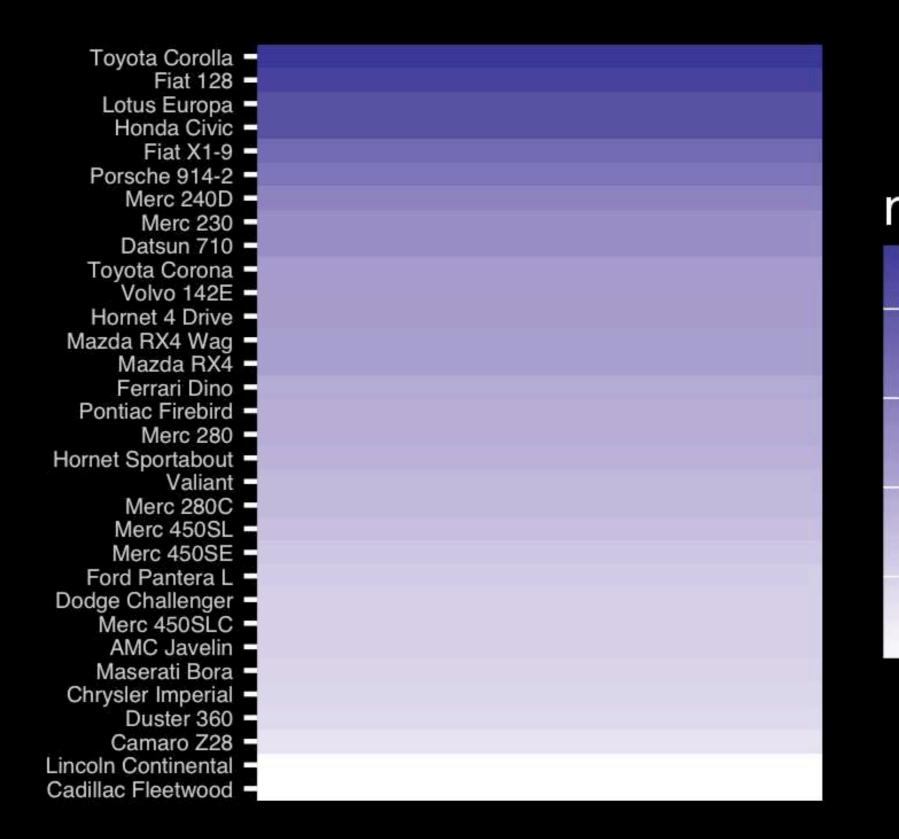


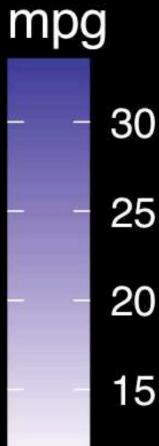




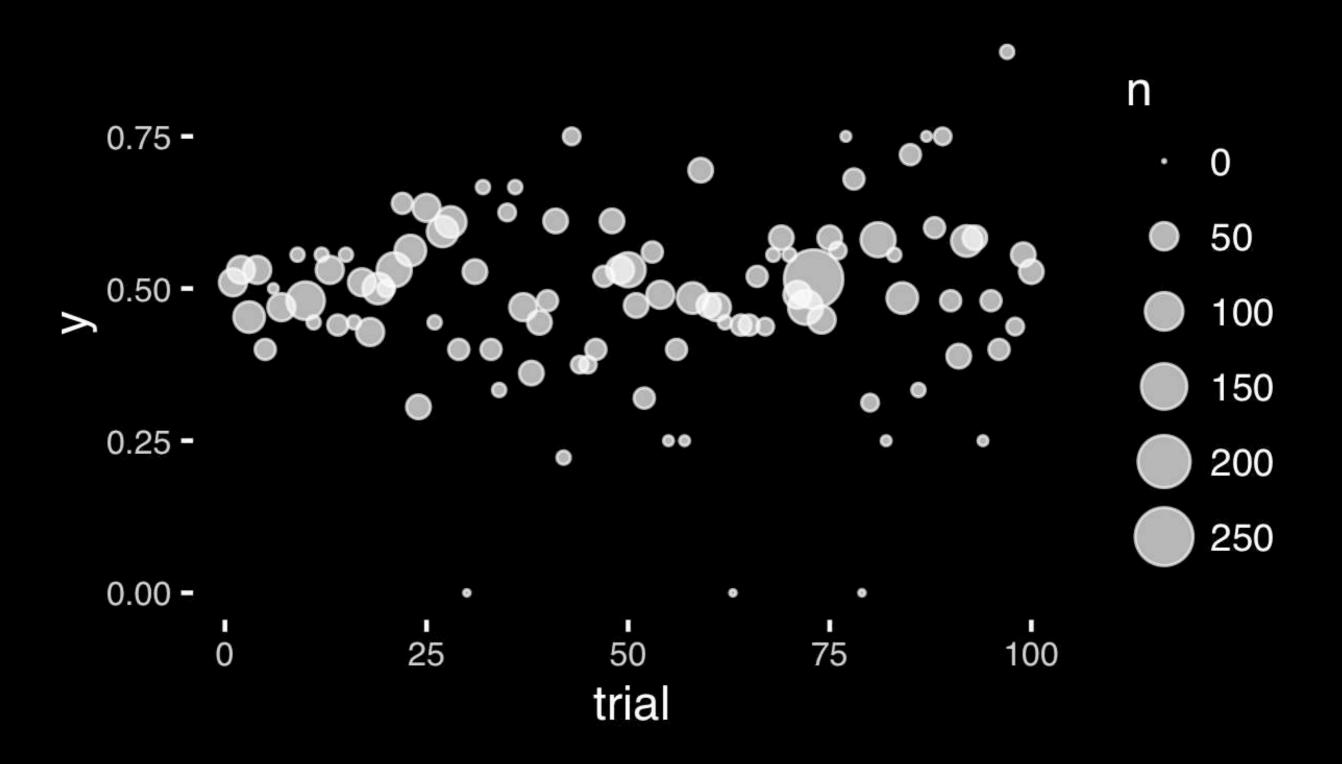
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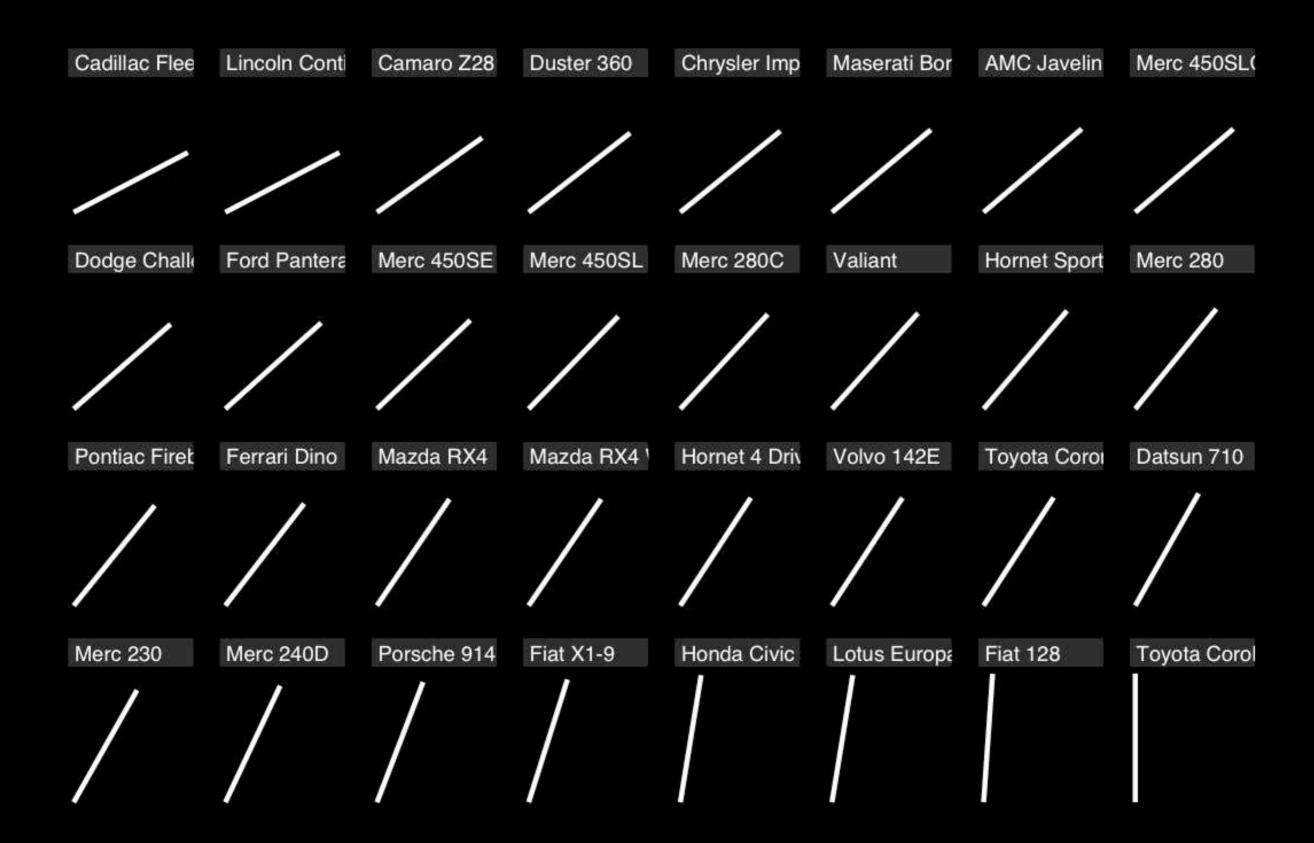


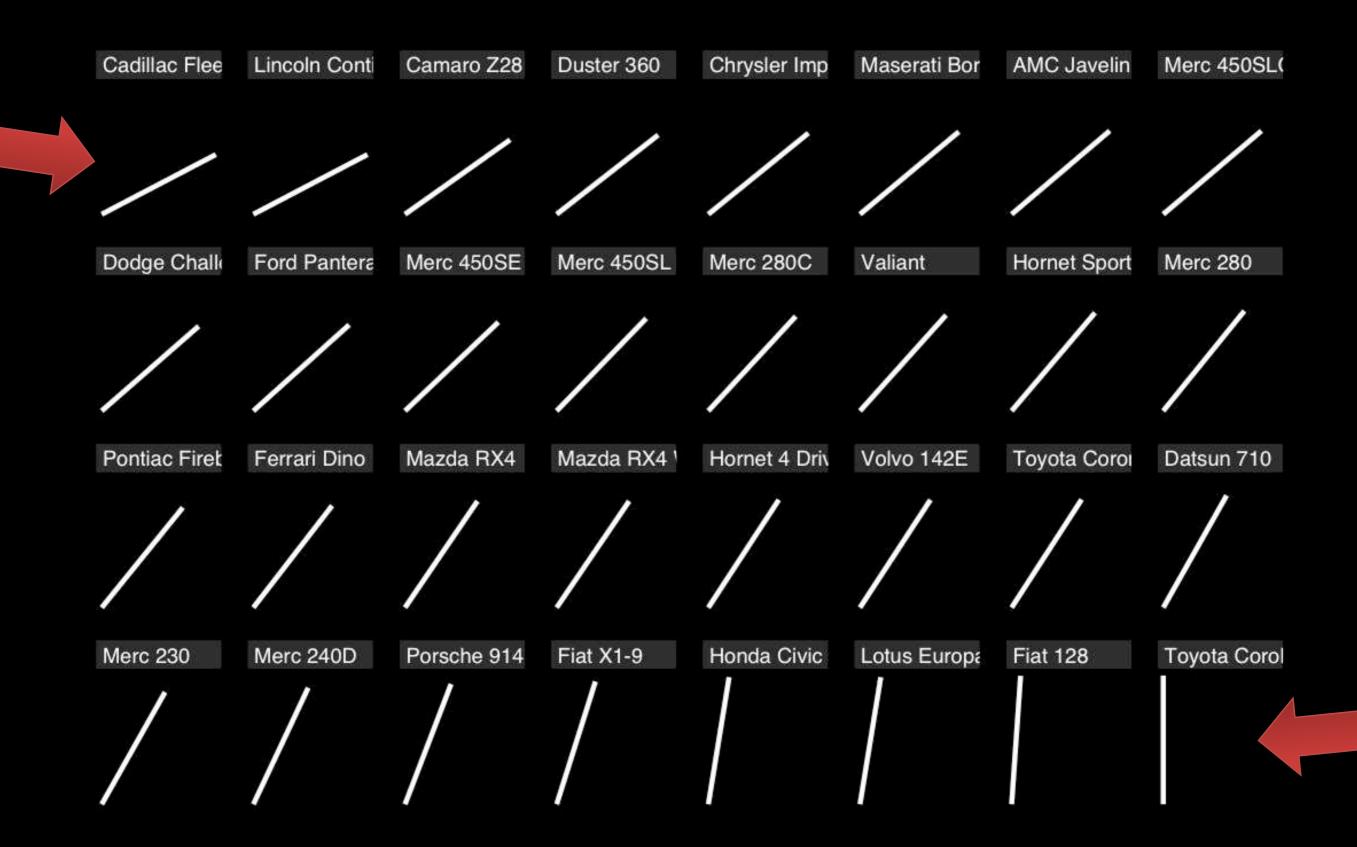


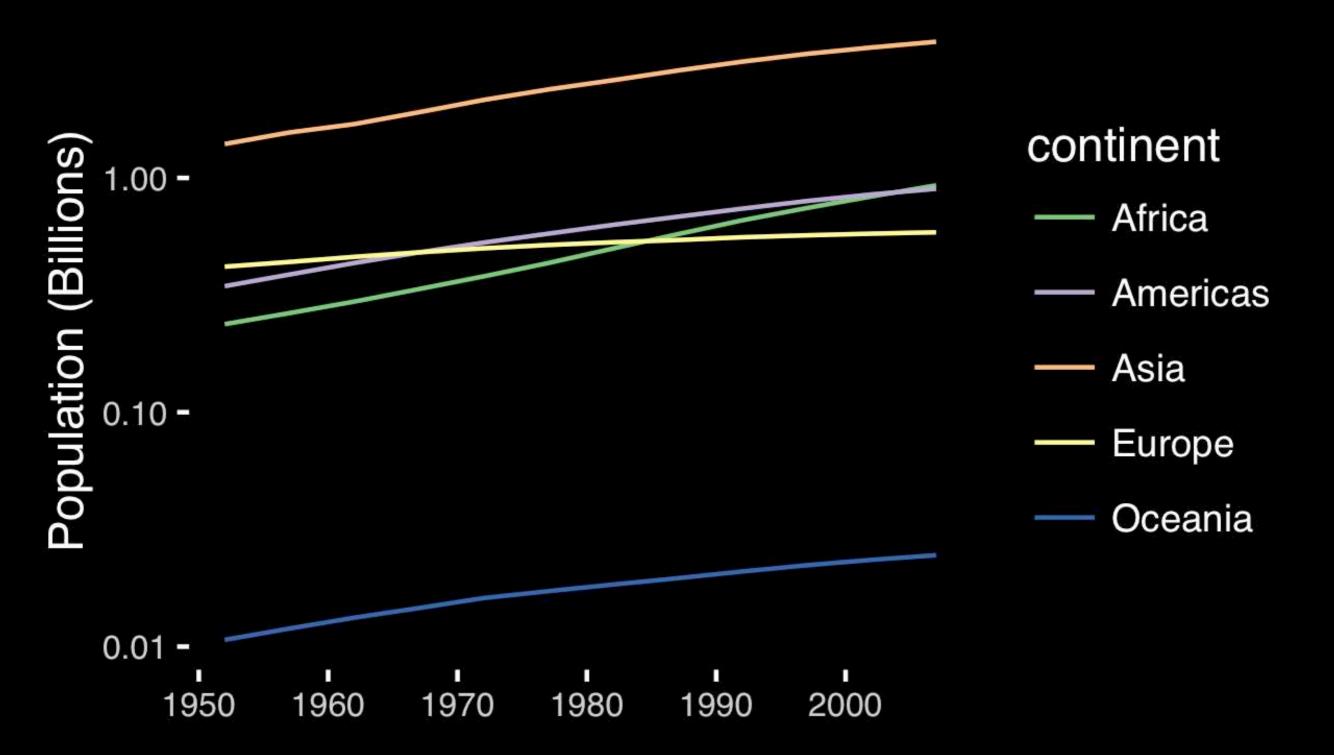


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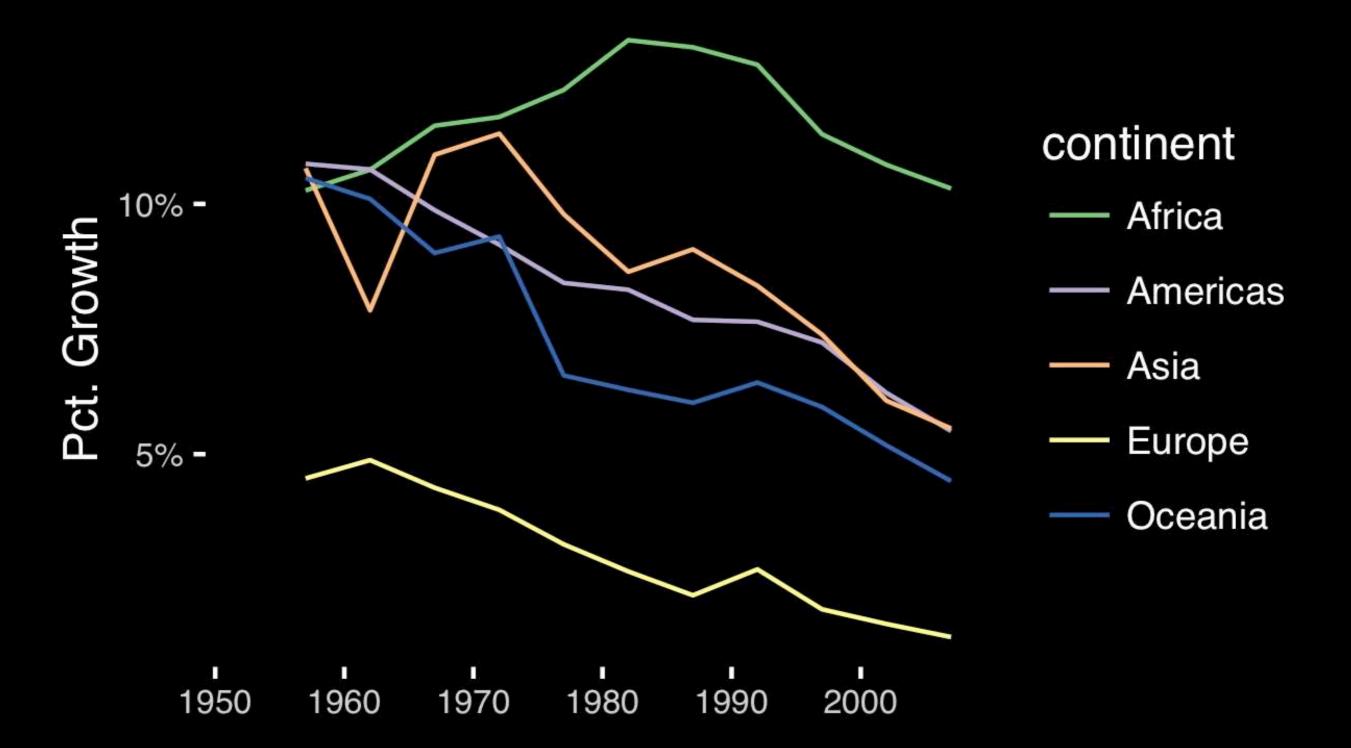
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# If growth (slope) is important, plot it directly.



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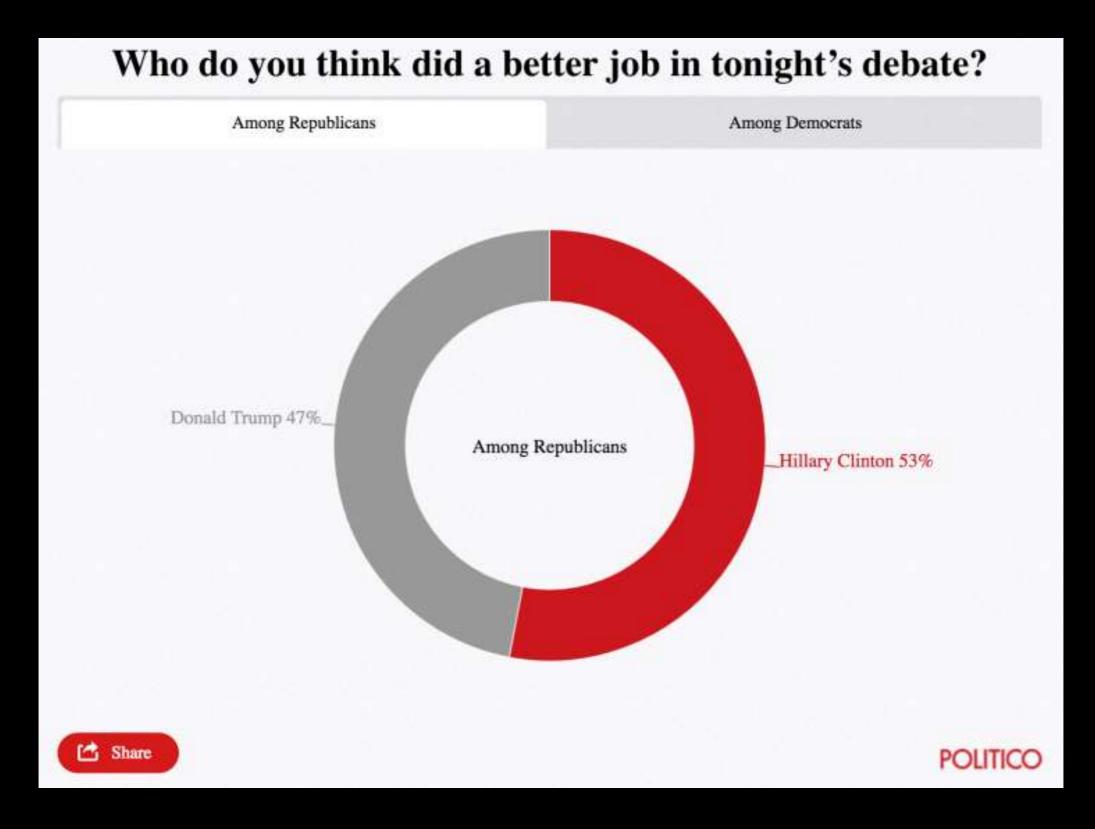
# Observation: Pie charts are <u>ALWAYS</u> a mistake.

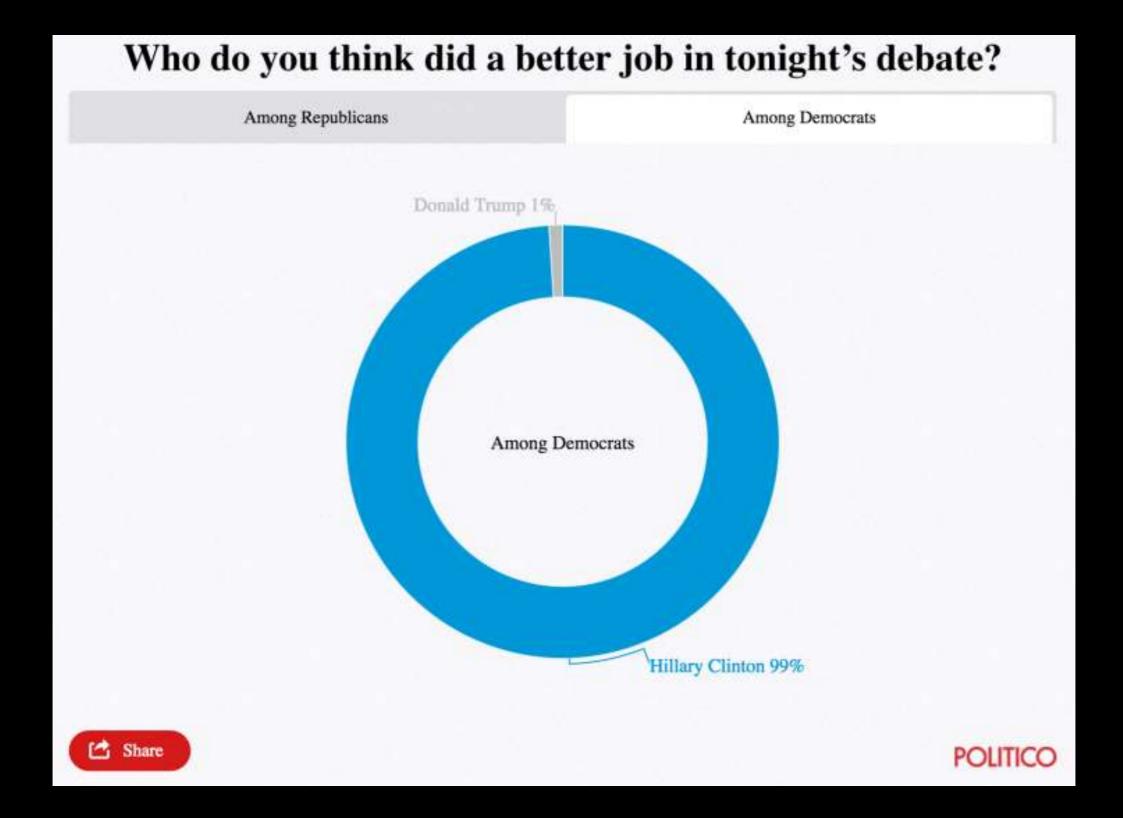
Piecharts are the information visualization equivalent of a roofing hammer to the frontal lobe. They have no place in the world of grownups, and occupy the same semiotic space as short pants, a runny nose, and chocolate smeared on one's face. They are as professional as a pair of assless chaps.

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

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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 compared 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.** 

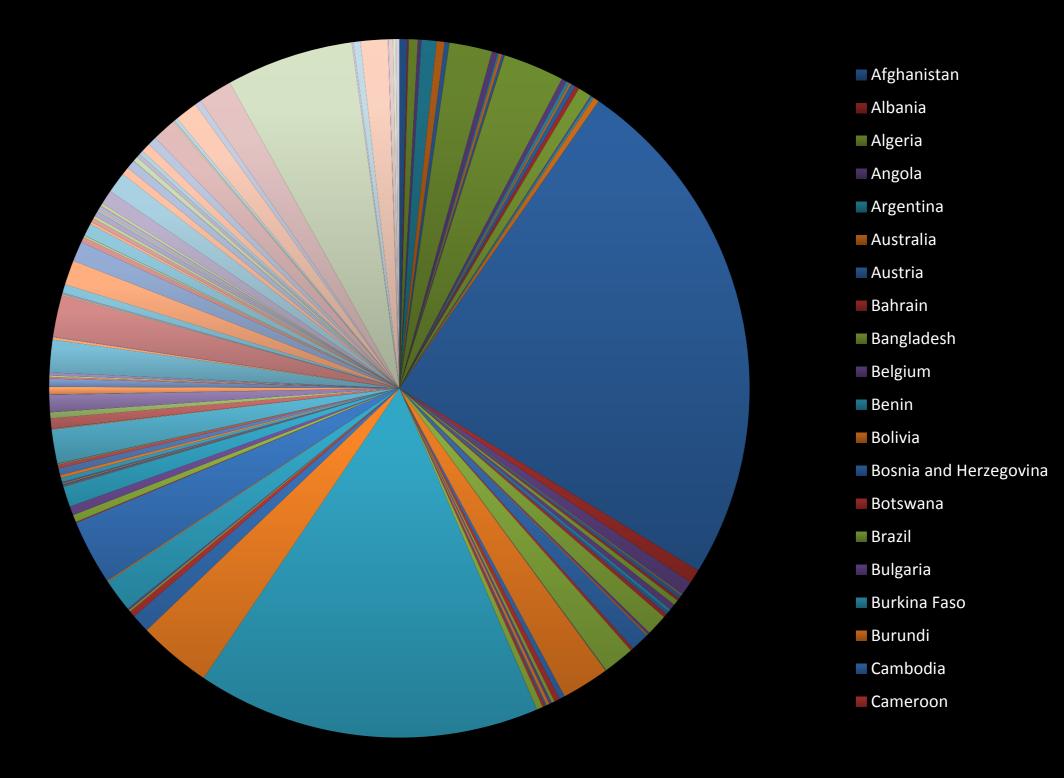
-Edward Tufte, The Visual Display of Quantitative Information

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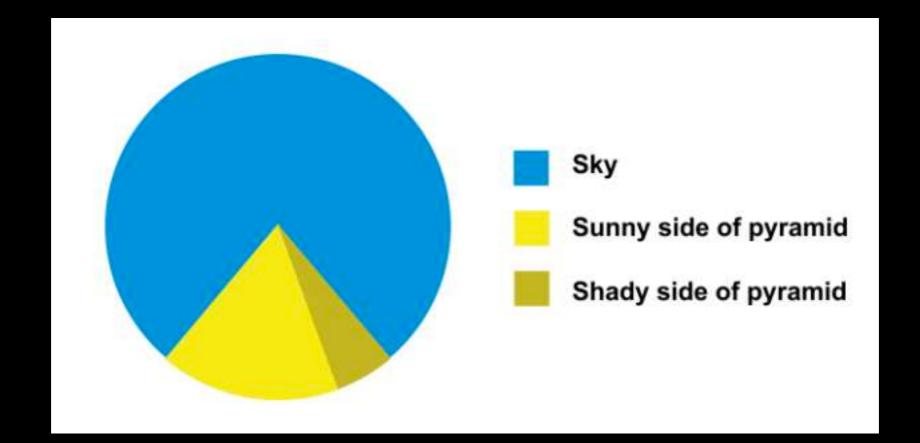
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## Who do you think did a better job in tonight's debate?

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



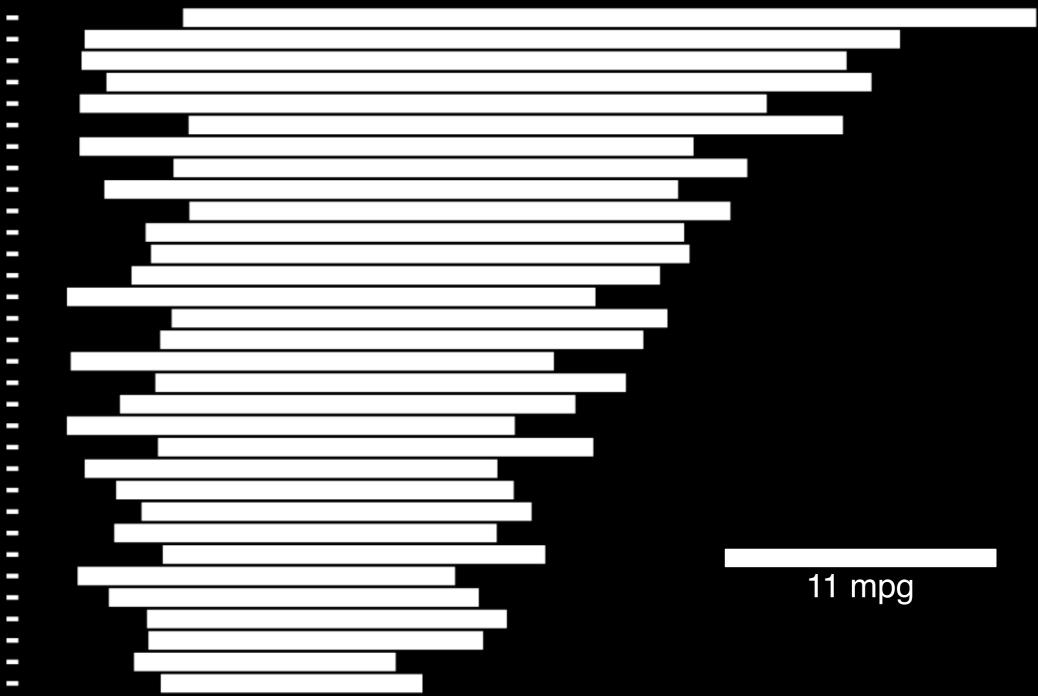
### All good pie charts are jokes...

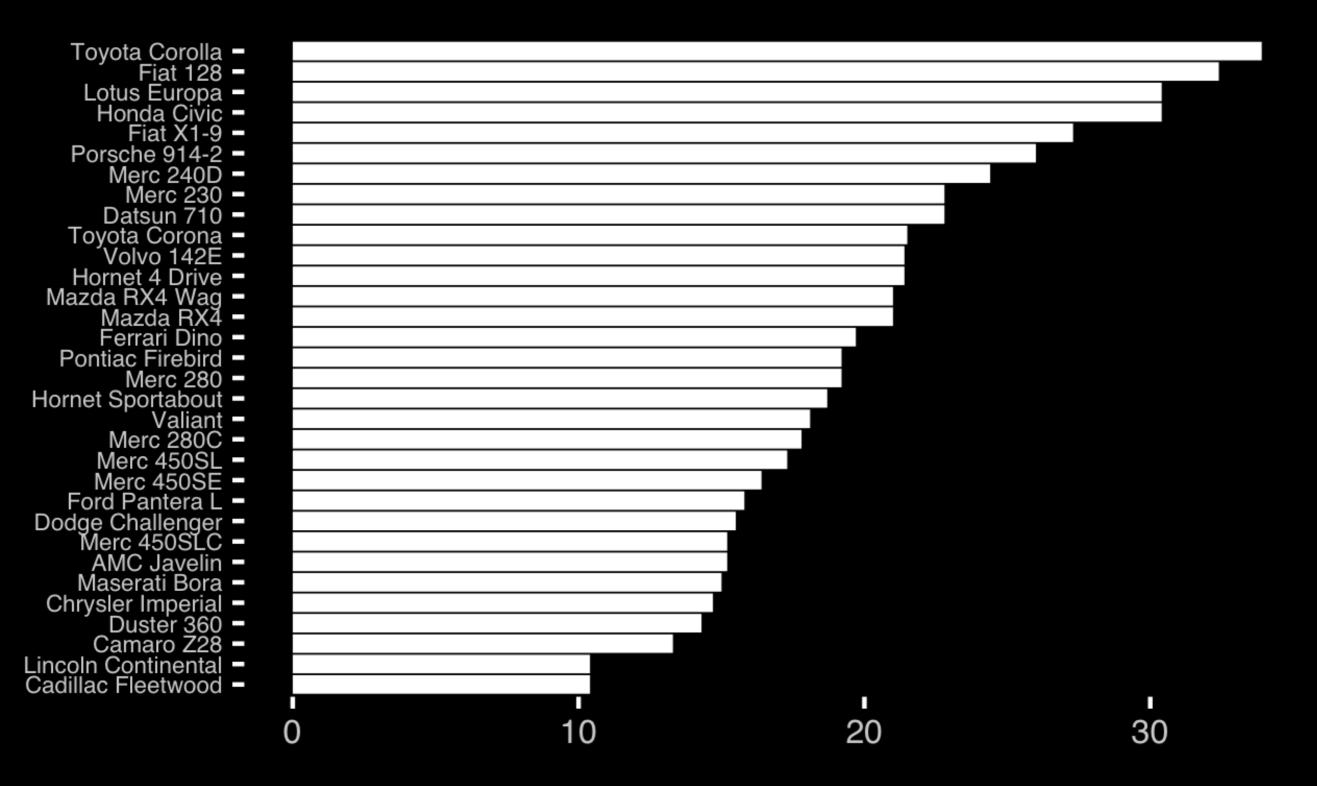


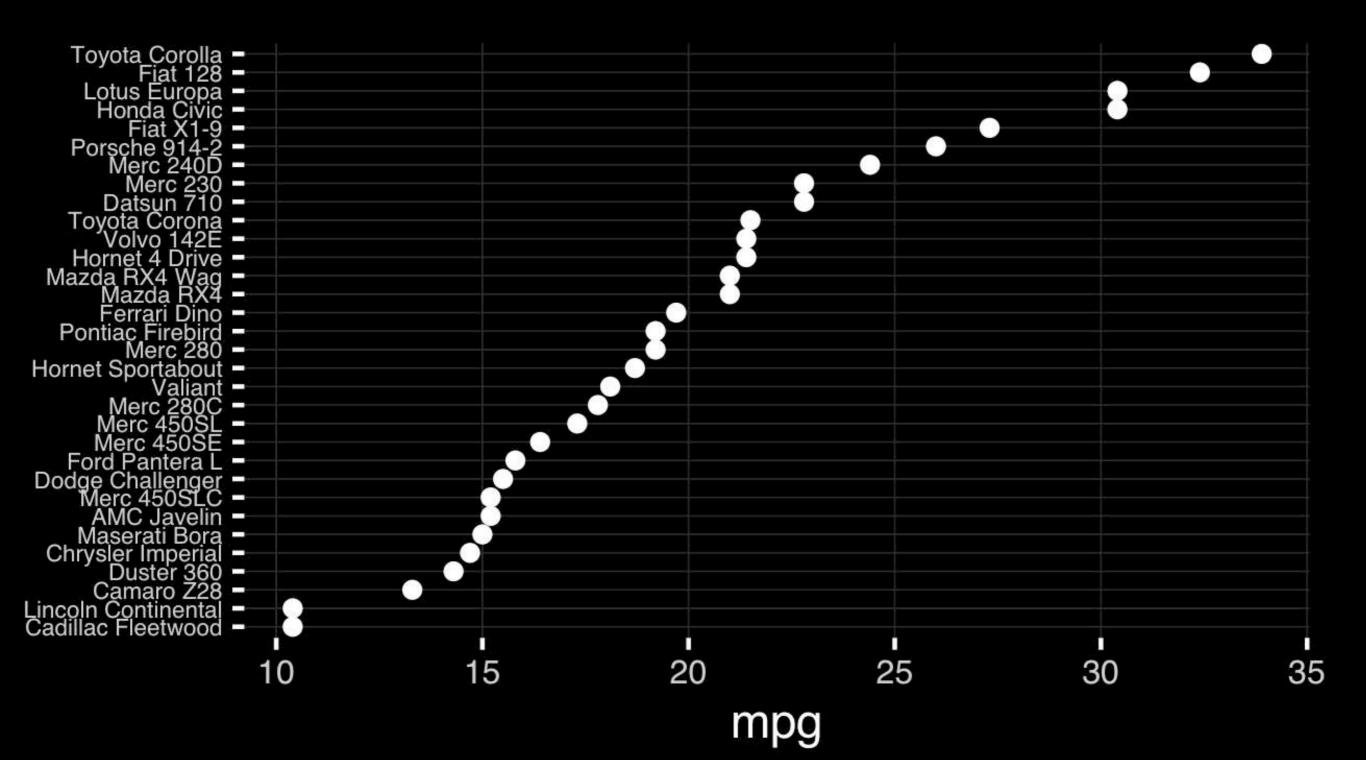
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Toyota Corolla -Fiat 128 -Lotus Europa -Honda Civic -Fiat X1-9 -Porsche 914-2 -Merc 240D -Merc 230 -Datsun 710 -Toyota Corona -Volvo 142E – Hornet 4 Drive – Mazda RX4 Wag -Mazda RX4 -Ferrari Dino -Pontiac Firebird -Merc 280 -Hornet Sportabout -Valiant -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 -



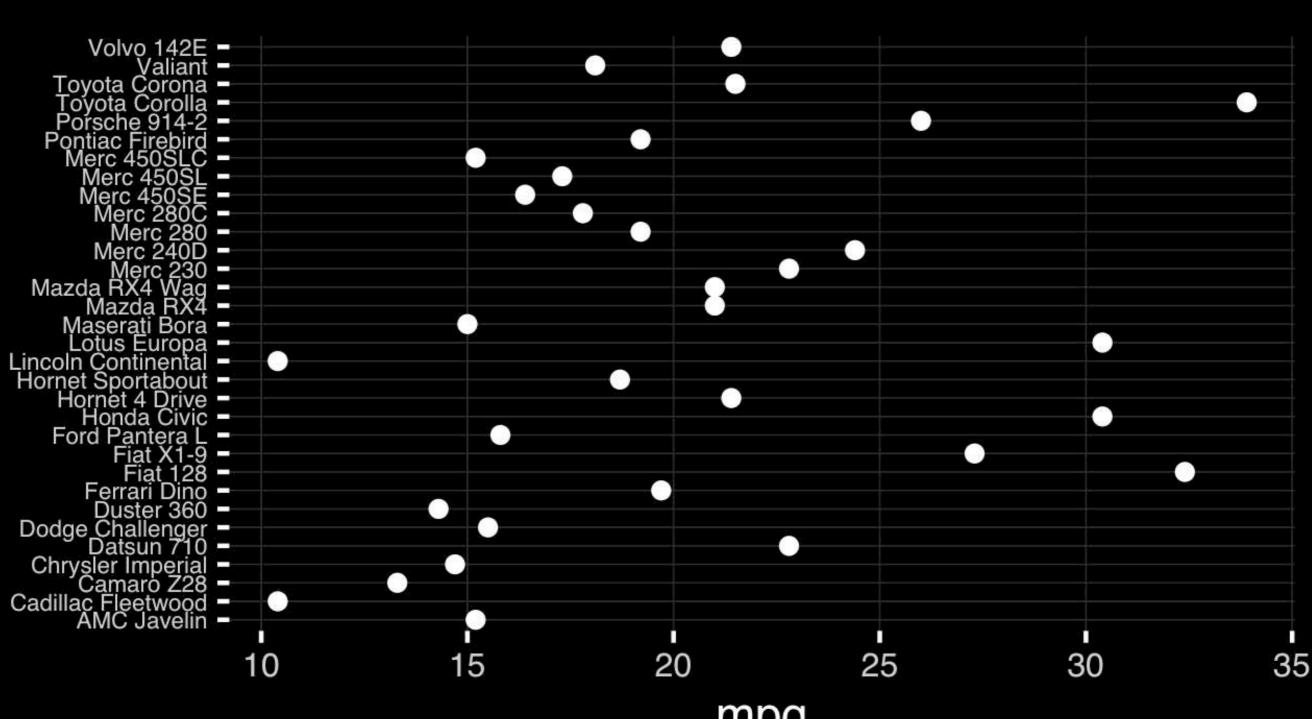




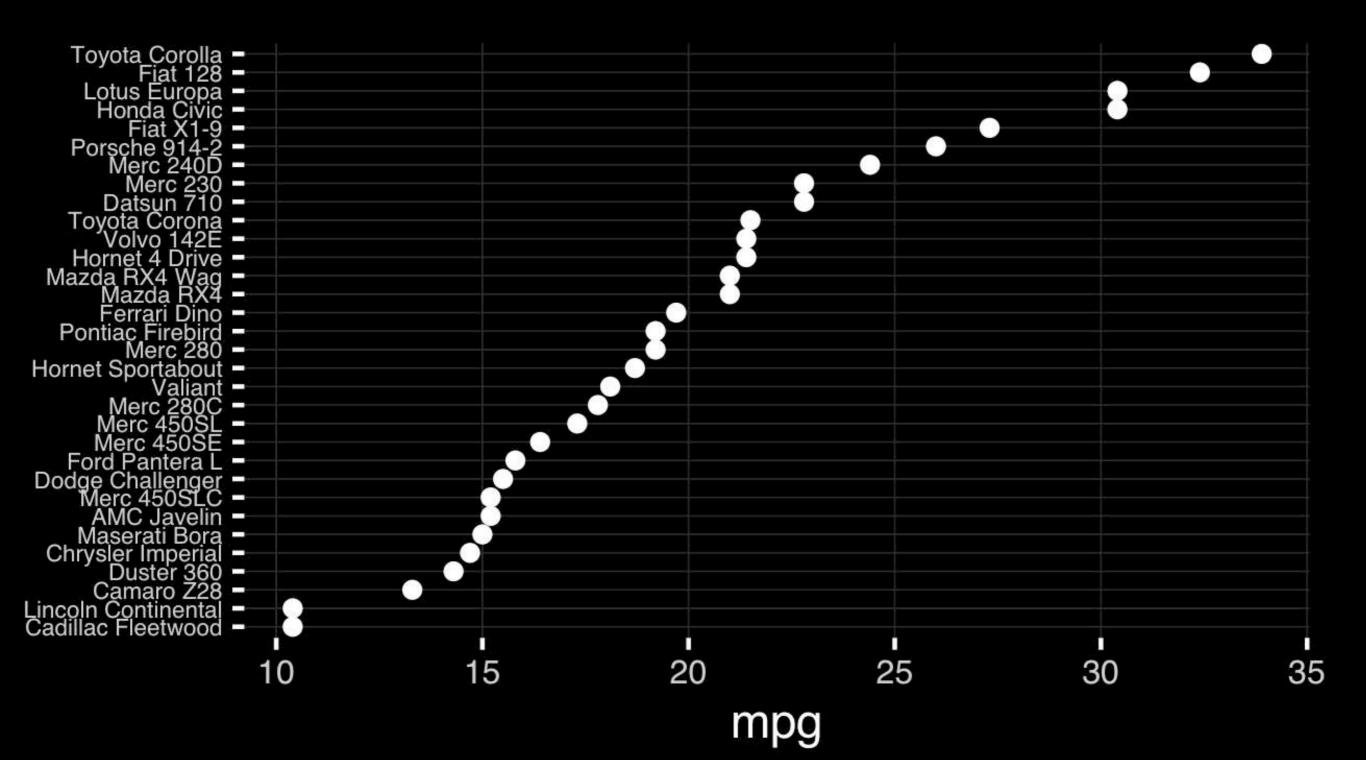
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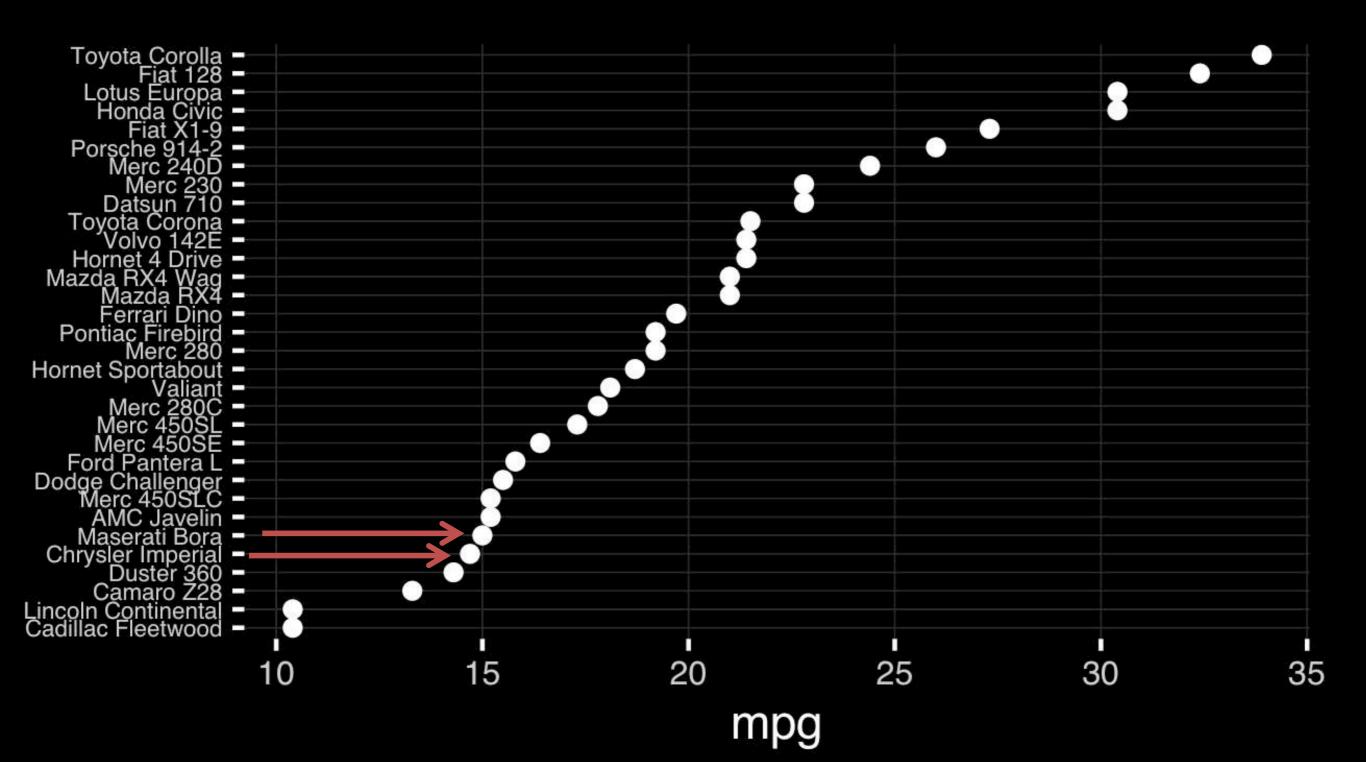
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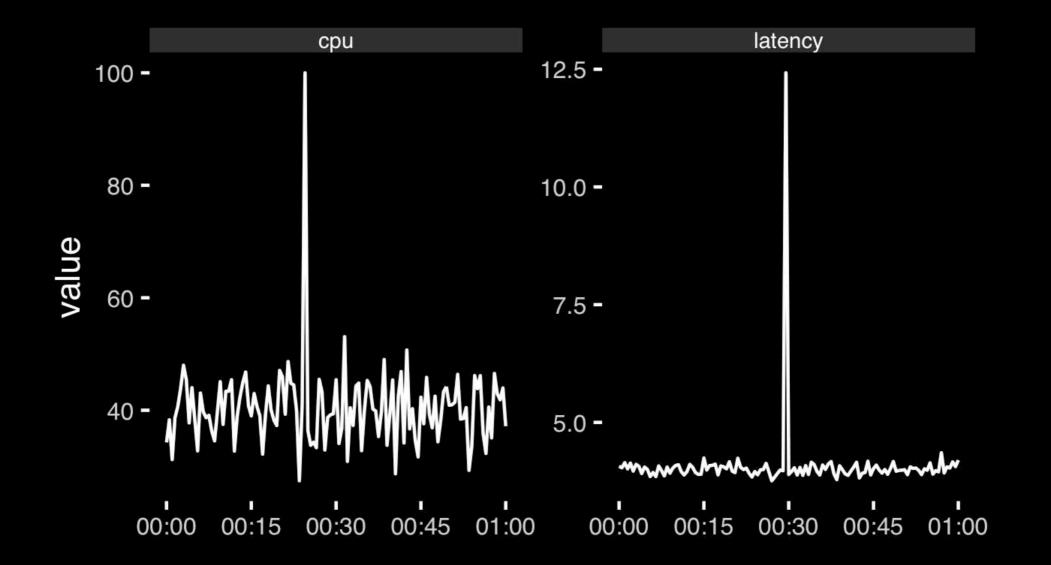


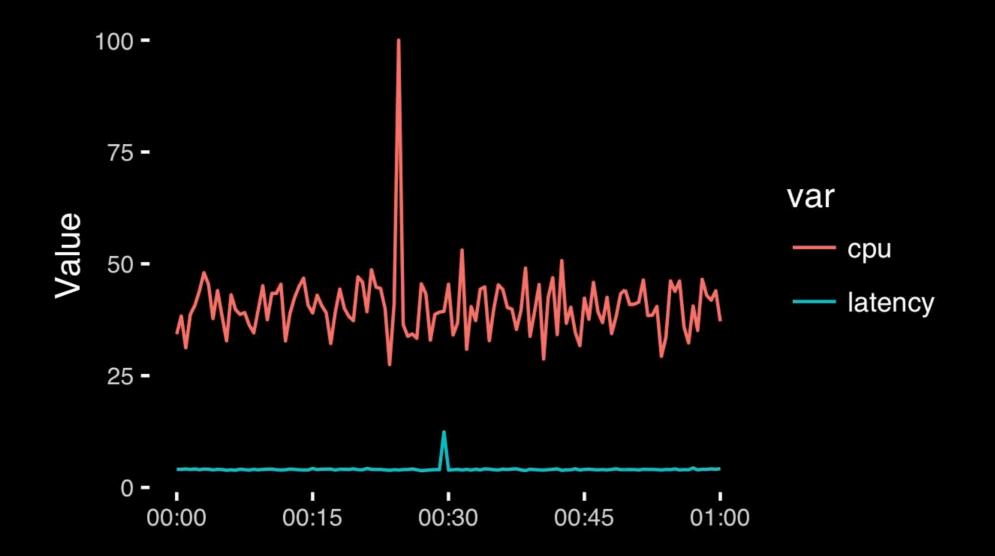
mpg

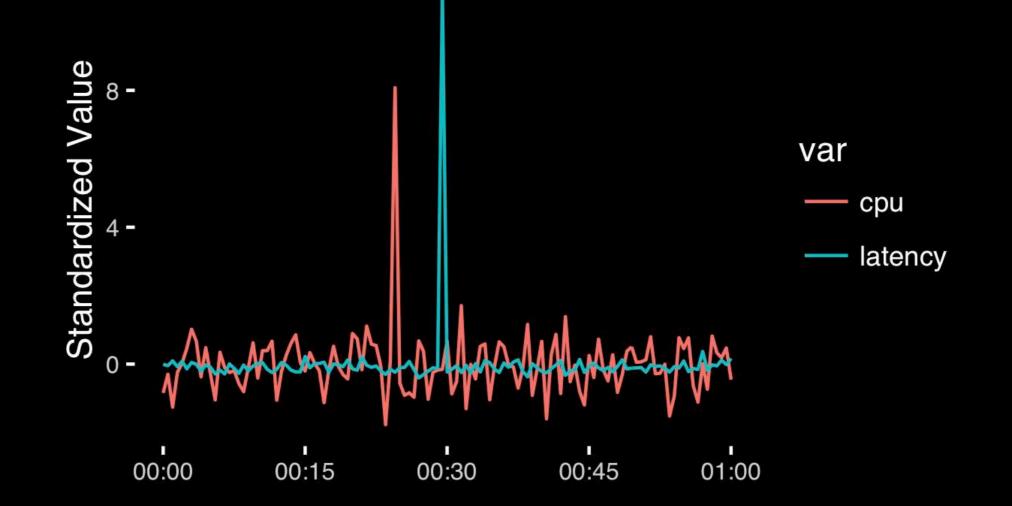




# Observation: Comparison is trivial on a common scale.





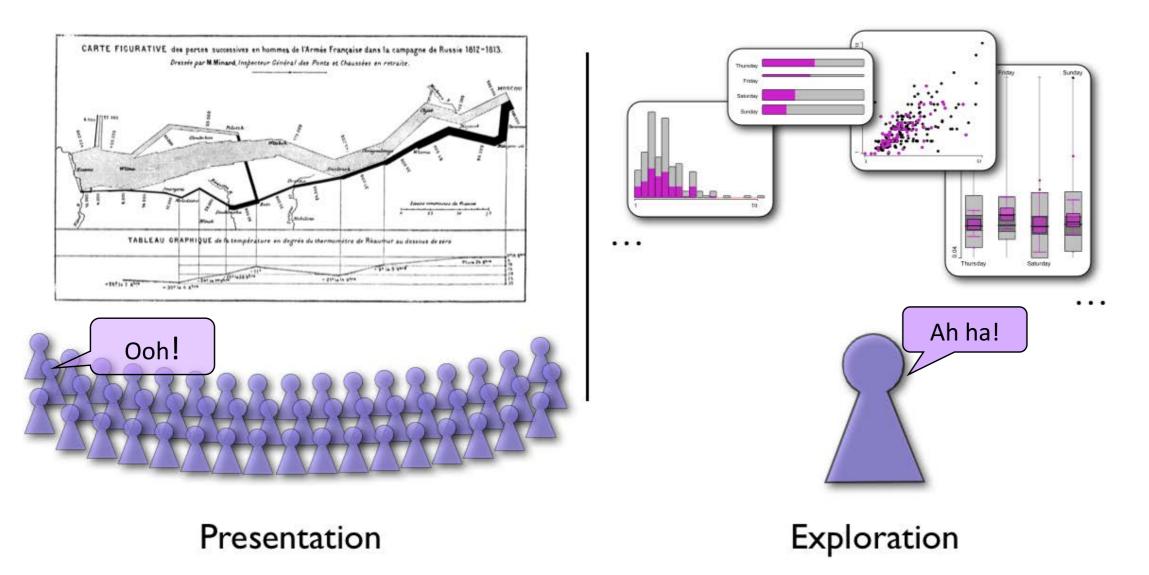


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### Different graphs for different purposes

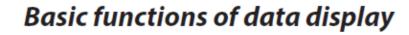
**Exploratory graphs**: many images for a narrow audience (you!) **Presentation graphs**: single image for a large audience

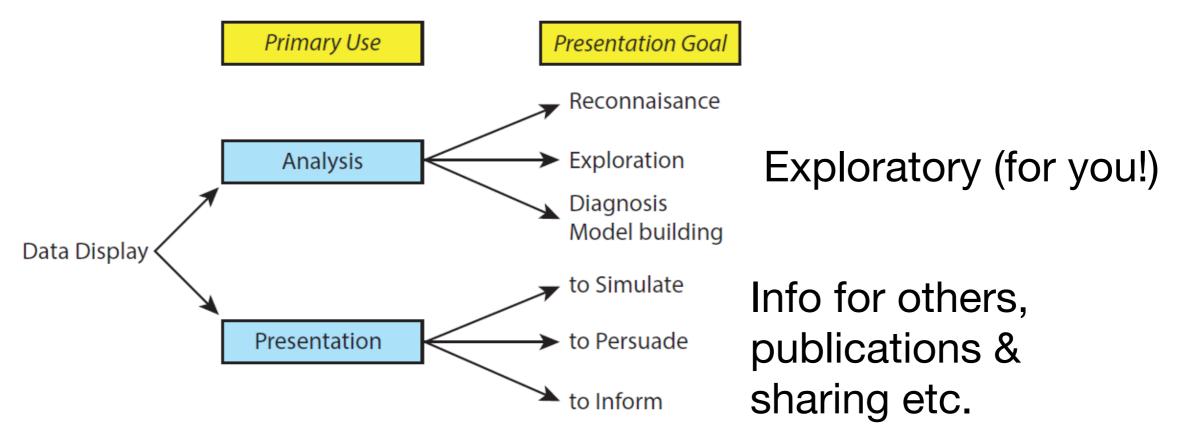


### Roles of graphics in data analysis

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

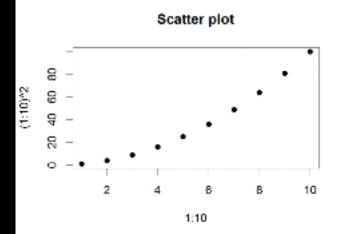
Analysis graphs: design to see patterns, trends, aid the process of data description, interpretation **Presentation graphs**: design to attract attention, make a point, illustrate a conclusion



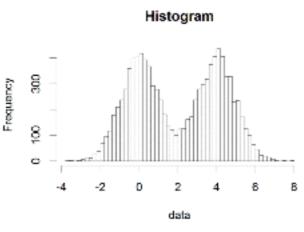


# Core R Graph Types

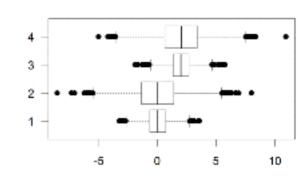
raises



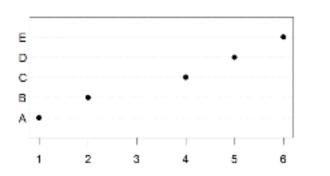
Bar Chart





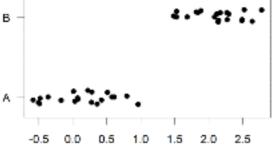


Dot Chart

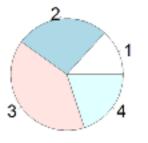


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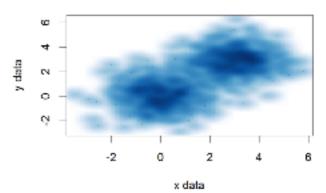
Stripchart

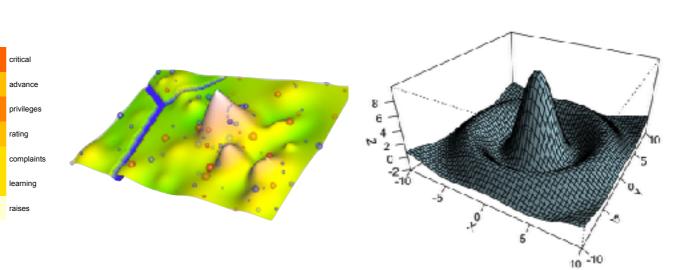




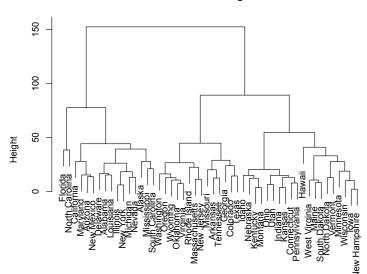


Smooth Scatter

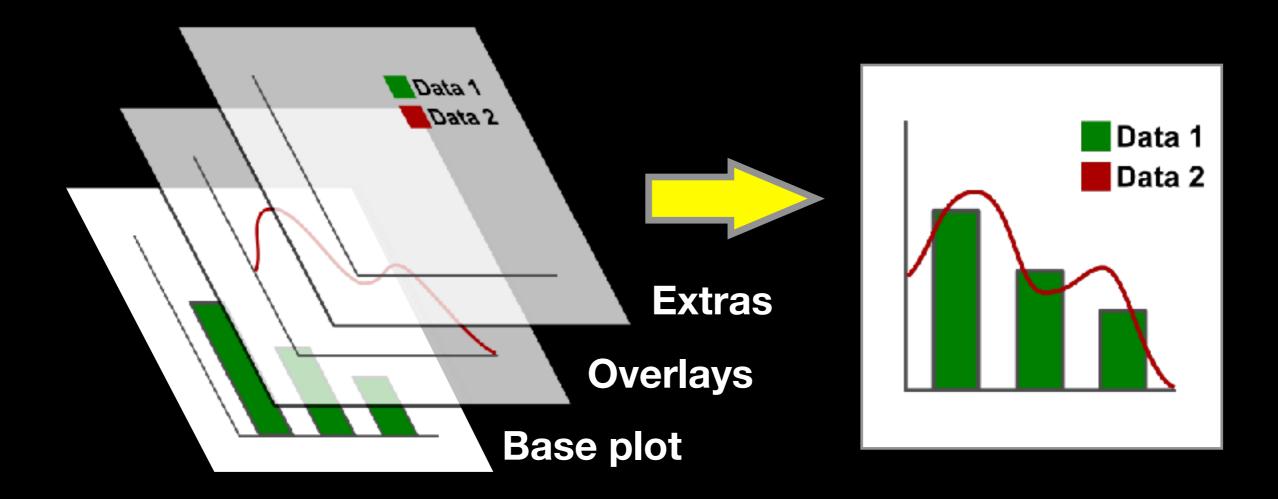




Cluster Dendrogram

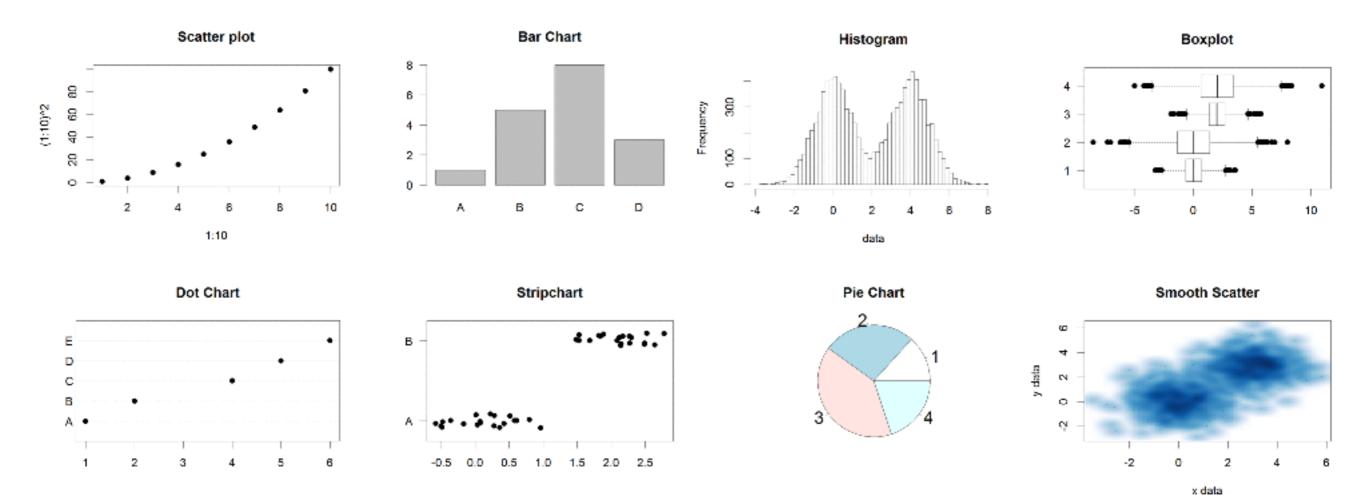


## The R Painters Model



### 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

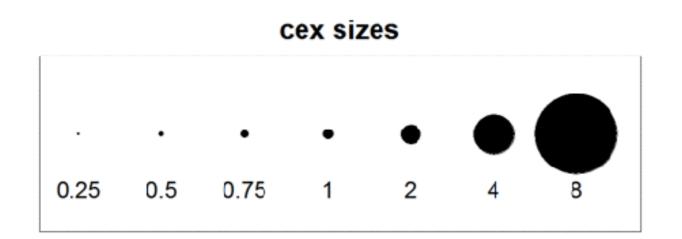
## **Common Options**

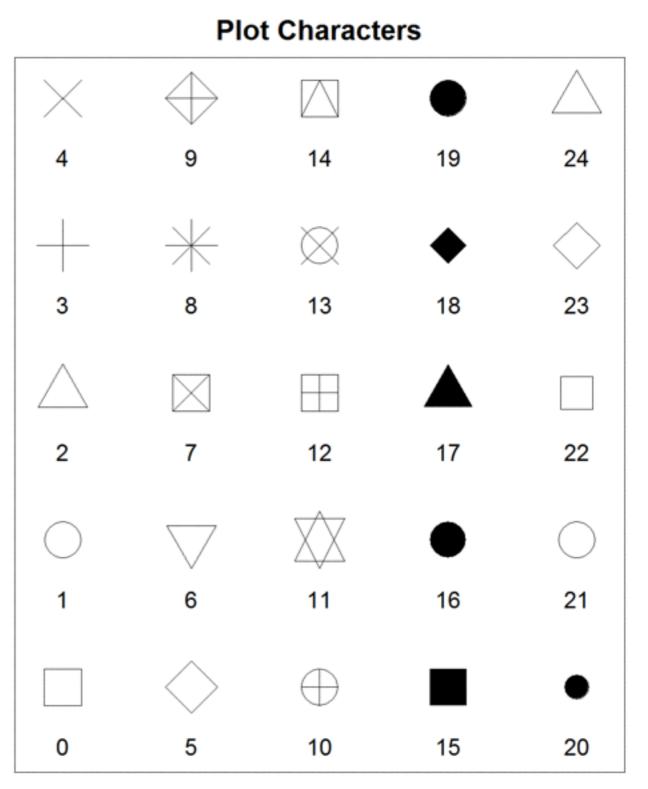
- Axis scales
  - xlim c(min, max)
  - ylim c(min, max)
- Axis labels
  - xlab(text)
  - ylab(text)

- 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

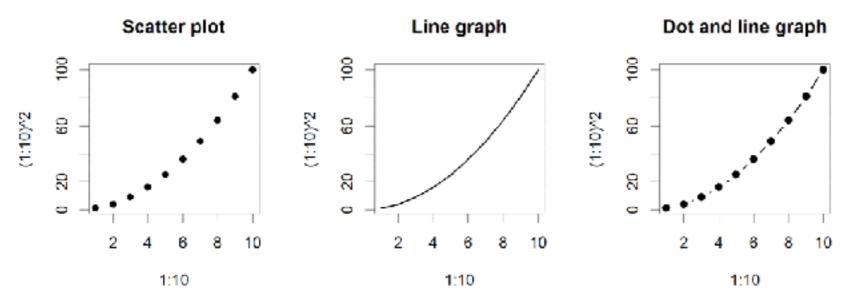
#### **Plot Characters**





#### Plot 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)

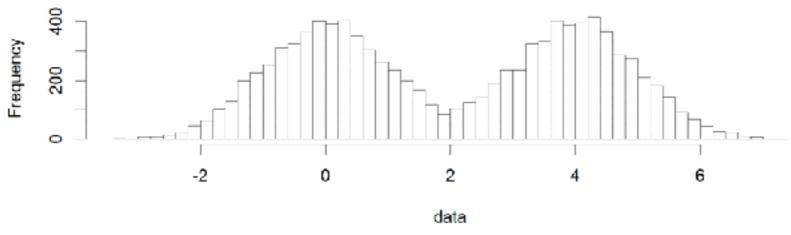


- Input: Vector (single) or Matrix (stack or group)
- Output: Bar centre positions
- Options:
  - names.arg Bar labels (if not from data)
  - horiz=TRUE Plot horizontally
  - beside=TRUE Plot multiple series as a group not stacked

barplot(VADeaths, beside = TRUE)

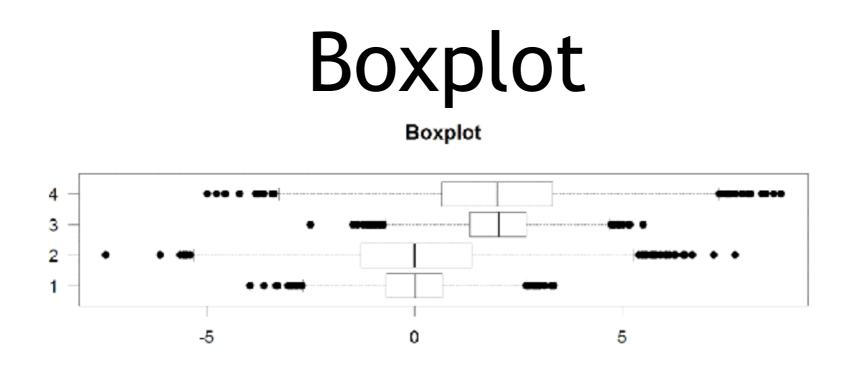
## Hist (histograms)

Histogram



- 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 )



- 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

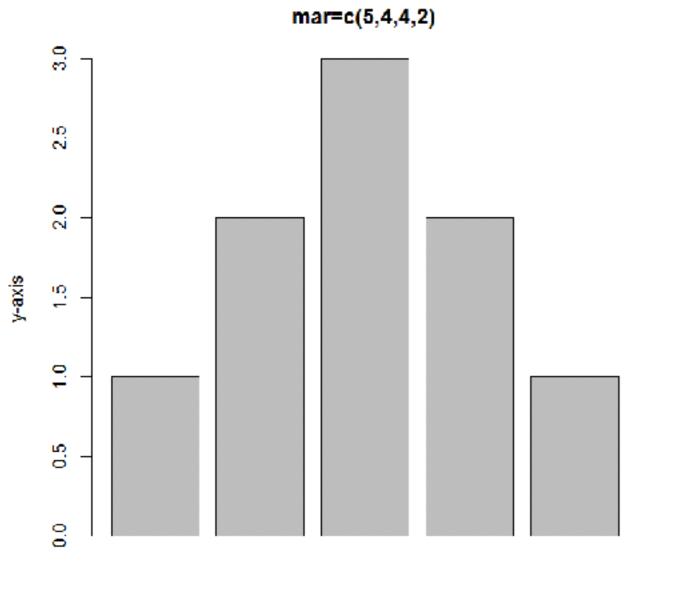
- The par function controls global parameters affecting all plots in the current plot area
- Changes affect all subsequent plots
- Many par options can also be passed to individual plots

#### Par examples

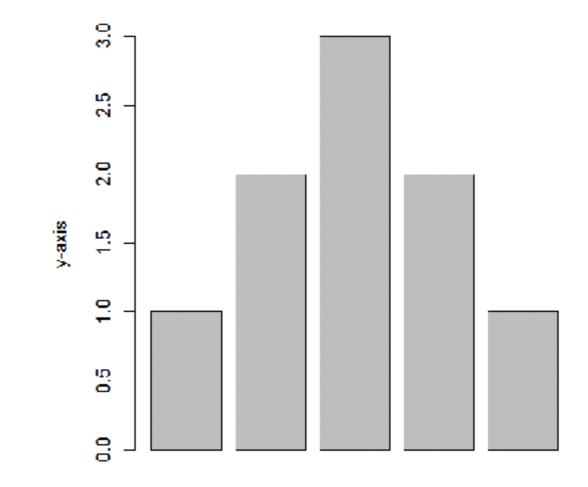
- Reading current value
   par() \$cex
- Setting a value
   -par(cex=1.5) -> old.par
- Restoring a value
  - -par(old.par)
  - -dev.off()

## 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



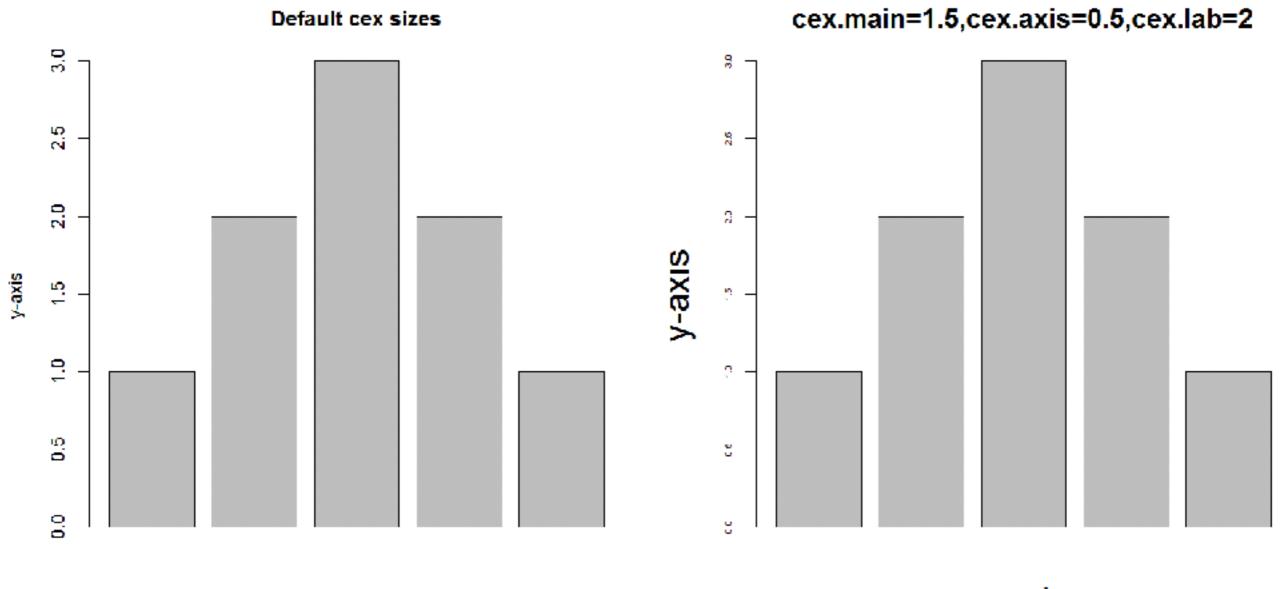
mar=c(2,10,10,1)



x-axis

### Par options

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

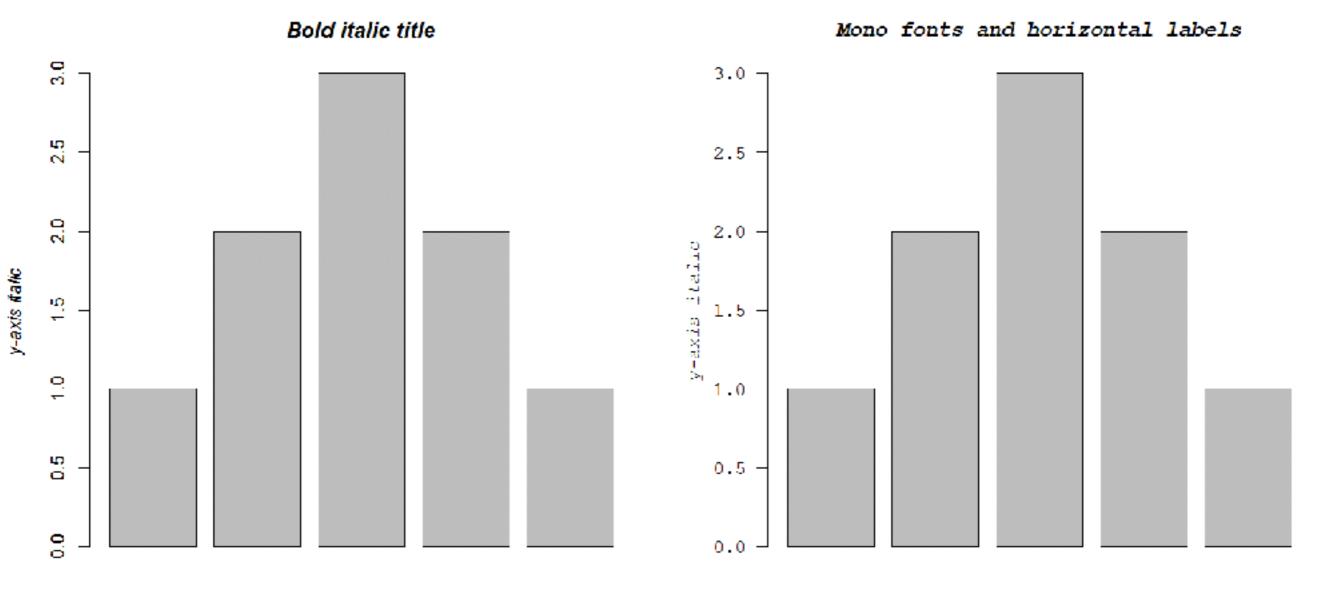


x-axis

x-axis

## Par options

- Font style
  - font (font.axis, font.main, font.sub, font.lab)
    - 1 = Plain text
    - 2 = Bold text
    - 3 = Italic text
    - 4 = Bold italic text
  - las (label orientation)
    - 0 = Parallel to axis
    - 1 = Horizontal
    - 2 = Perpendicular
    - 3 = Vertical

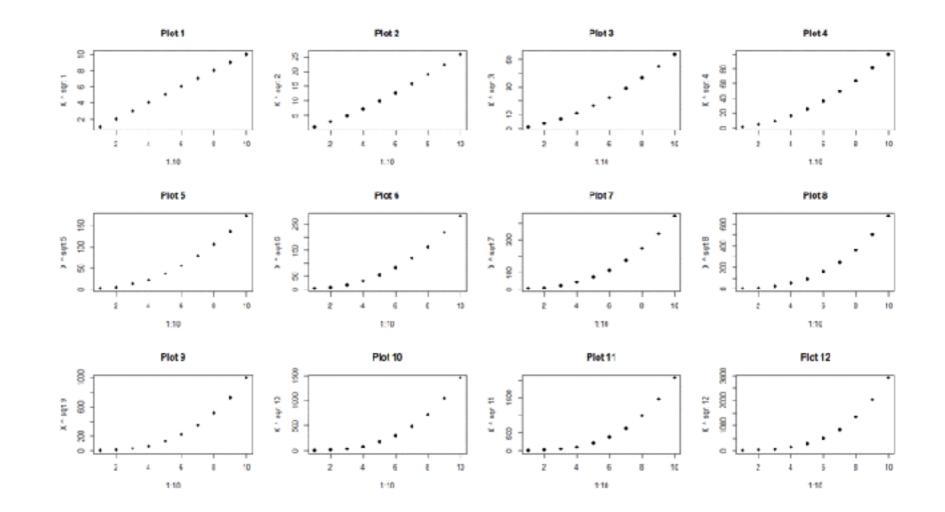


x-axis italic

x-axis italic

#### Par options

- Multi-panel
  - mfrow (rows, cols)
  - Not supported by some packages



#### Exercise 1

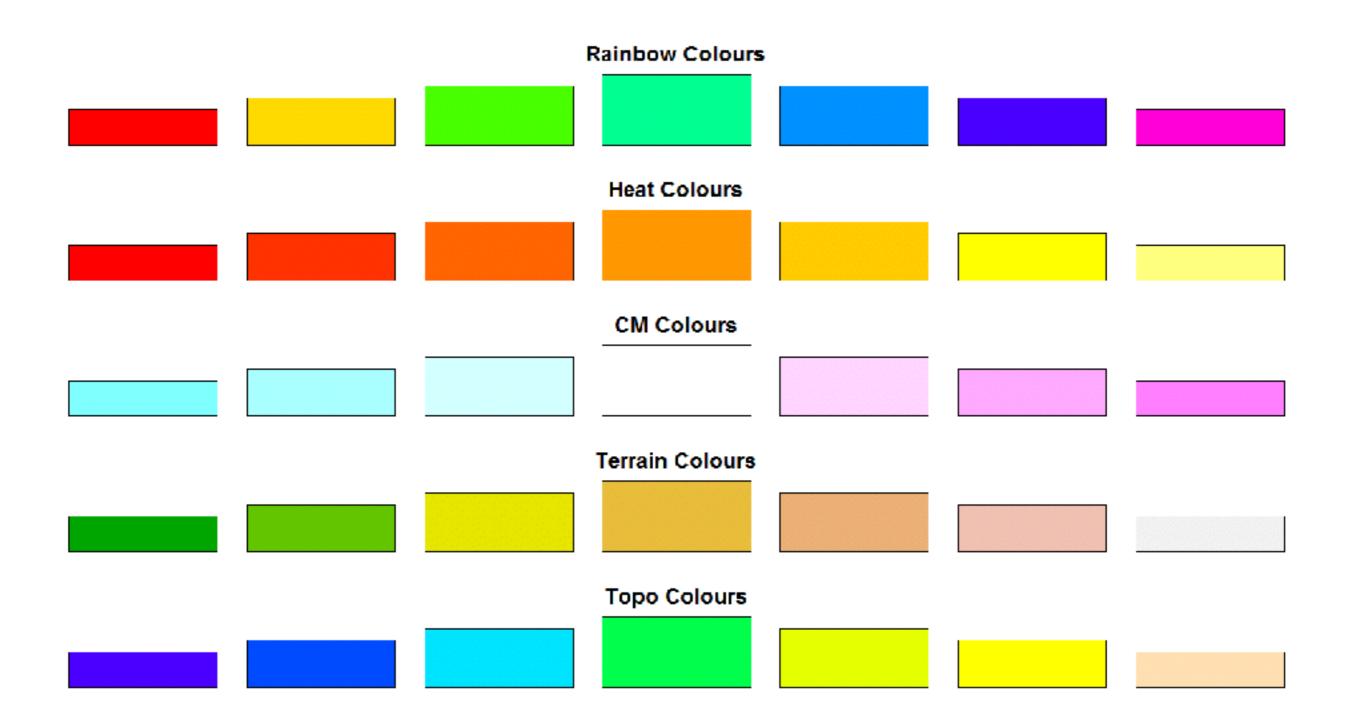
#### Using Color

# 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



## Color Packages

- Color Brewer
  - Set of pre-defined, optimized palettes
  - library(RColorBrewer)
  - -brewer.pal(no 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)

#### Color Packages

#### Colorspace

- library(colorspace)
- choose.palette()

Choose Color Palette	
File	
The nature of your data Diverging	-
-Default color schemes	
-Palette description: Hue, Chroma, Luminance, P	ower
IN	
H2	128
a	
C2 ]	0
n — — — — — — — — — — — — — — — — — — —	
L2	95
м	0.7
P2	13
Correct all colors to valid RGB color model val	ues
Number of colors in palette	1
n	70
Show example	
Pict type 👻 🛽	Reverse colors
Desaturation     Color blindness:     e deutan     protan     thtan	
OK	Cancel

# Applying Color to Plots

- Vector of colors passed to the col parameter
- Vector of factors used to divide the data

   Colors taken from pallete
  - Can read or set using pallete function
    - palette()
    - palette(brewer.pal(9,"Set1")
    - Ordered by levels of factor vector

## 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

## **Color Mapping Function**

map.colors <- function(value,range,palette) {</pre>

```
proportion <- (value-range[1])/(range[2]-range[1])
index <- round((length(palette)-1)*proportion)+1</pre>
```

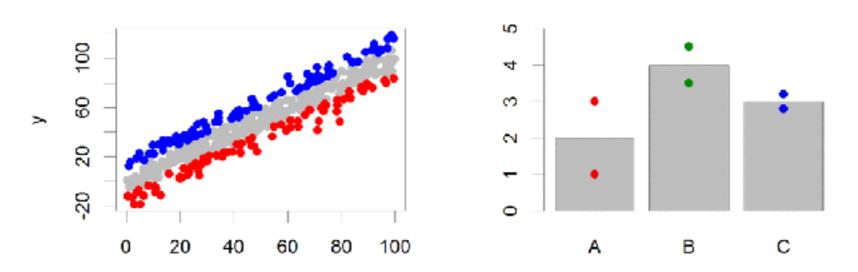
```
return(palette[index])
```

}

#### Exercise 2

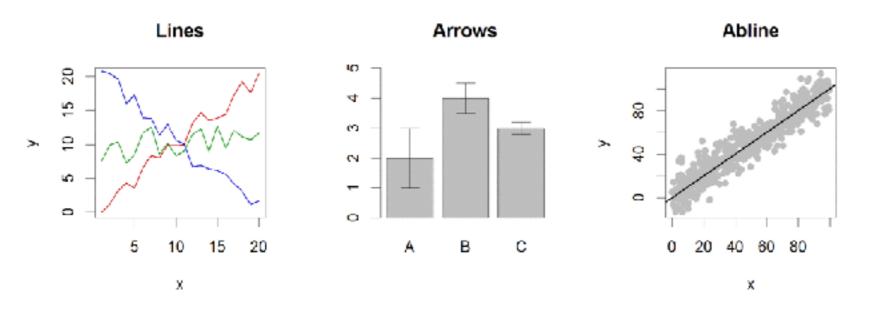
Plot Overlays Exercise 3

#### Points



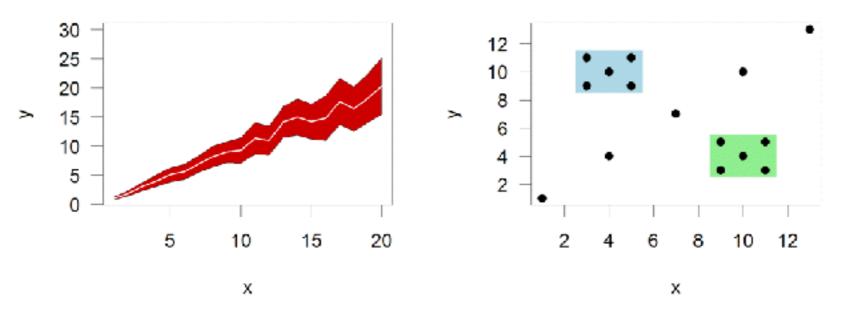
- Input: 2 Vectors (x and y positions)
- Options:
  - -pch
  - cex

## Lines / Arrows / Abline



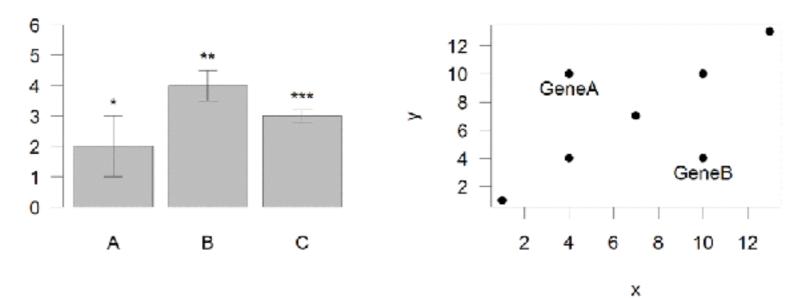
- Input:
  - Lines 2 vectors (x and y)
  - Arrows 4 vectors (x0,x1,y0,y1)
  - Abline Intercept and slope (or correlation object)
- Options:
  - \_ lwd
  - angle (arrows)

#### Polygon (shaded areas)

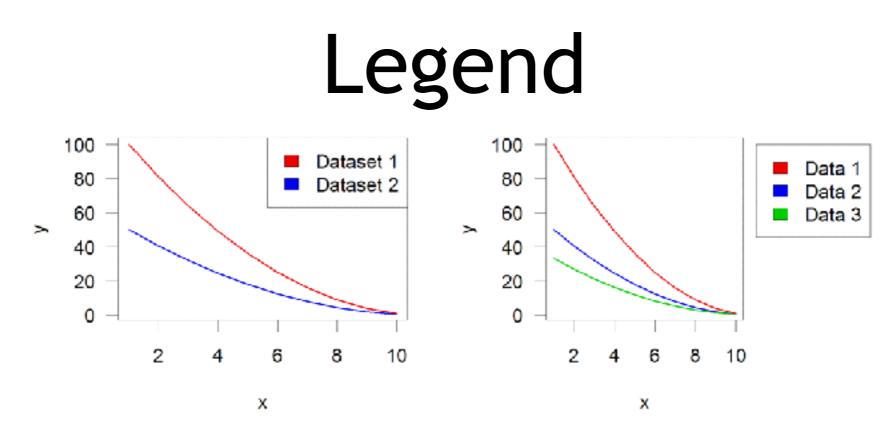


- Input:
  - 2 vectors (x and y) for bounding region
- Options:
  - col

#### Text (in plot text)



- Input:
  - Text, x, y
- Options:
  - adj (x and y offsets)
  - pos (auto offset 1=below,2=left,3=above, 4=right)



- Input:
  - Position (x,y or "topright", "bottomleft" etc)
  - Text labels
- Options:
  - fill (colours for shaded boxes)
  - xpd=NA (draw outside plot area)

#### Exercise 3