



BIMM 143
Data visualization with R
Lecture 5
Barry Grant
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, vectorization 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**.

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**.

Why visualize at all?

THE HERALD

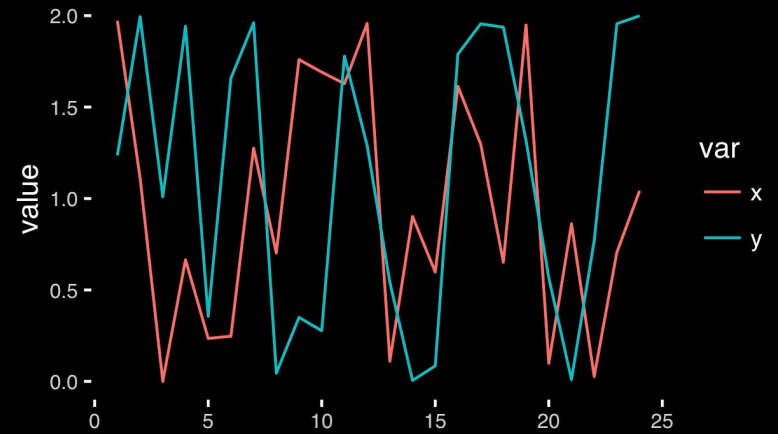
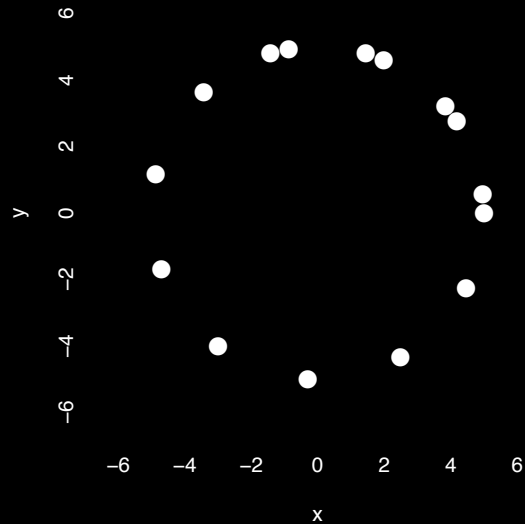
Over-the-Counter

National Market System

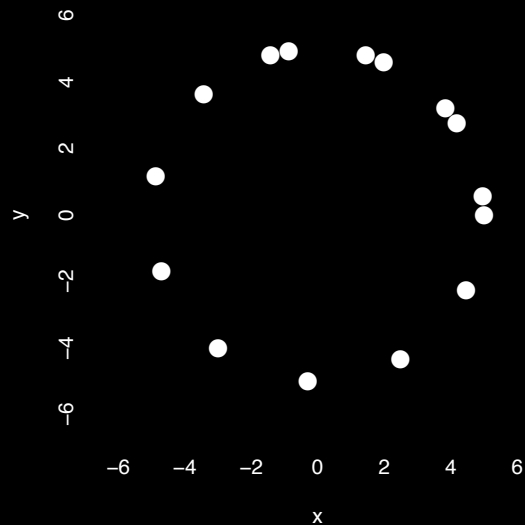
The companies listed below reflect the volume in 2007 of shares on a daily basis, and the rising percentage and change are reflected from the previous day based on trades as reported under the NASD National Market System.

	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



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

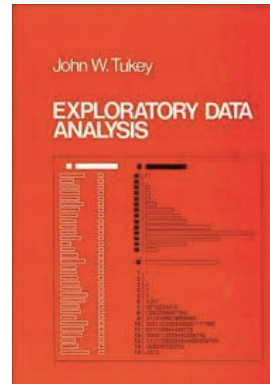


Exploratory Data Analysis

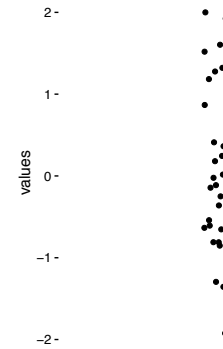
- ALWAYS look at your data!
- If you can't see it, then don't believe it!
- Exploratory Data Analysis (EDA) allows us to:
 1. Visualize distributions and relationships
 2. Detect errors
 3. Assess assumptions for confirmatory analysis
- EDA is the first step of data analysis!

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



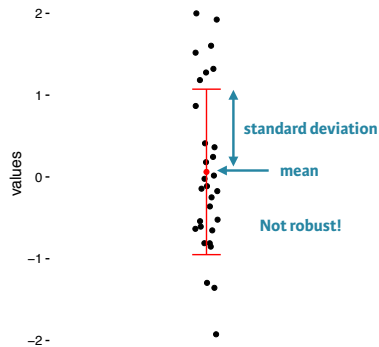
Side-note: How to summarize data?



```
x <- rnorm(1000,0)
```

Side-note: Mean & standard deviation

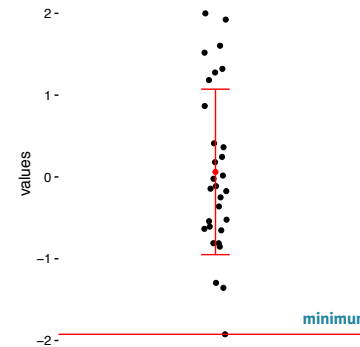
Fine for normally distributed data



```
x <- rnorm(1000,0)  
mean(x)  
sd(x)
```

Side-note: 5 number summary

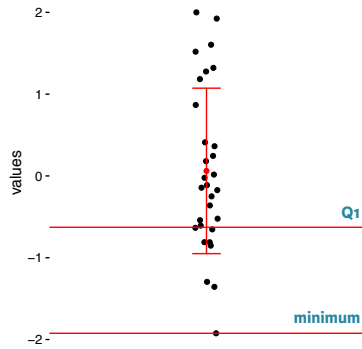
Minimum, Q1, Q2, Q3, and maximum



```
x <- rnorm(1000,0)  
mean(x)  
sd(x)  
summary(x)
```

Side-note: 5 number summary

Minimum, Q1, Q2, Q3, and maximum



```
x <- rnorm(1000,0)
```

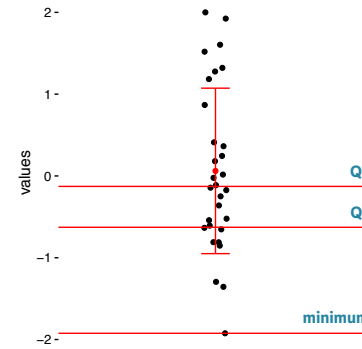
```
mean(x)
```

```
sd(x)
```

```
summary(x)
```

Side-note: 5 number summary

Minimum, Q1, Q2, Q3, and maximum



```
x <- rnorm(1000,0)
```

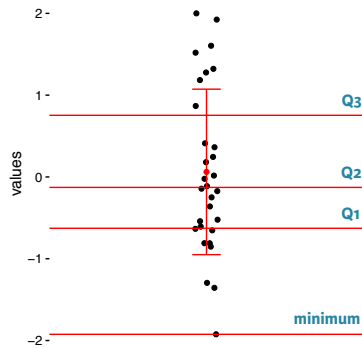
```
mean(x)
```

```
sd(x)
```

```
summary(x)
```

Side-note: 5 number summary

Minimum, Q1, Q2, Q3, and maximum



```
x <- rnorm(1000,0)
```

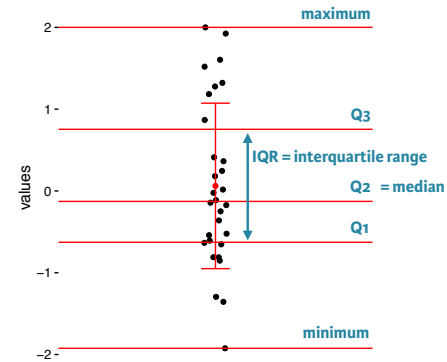
```
mean(x)
```

```
sd(x)
```

```
summary(x)
```

Side-note: 5 number summary

Minimum, Q1, Q2, Q3, and maximum



```
x <- rnorm(1000,0)
```

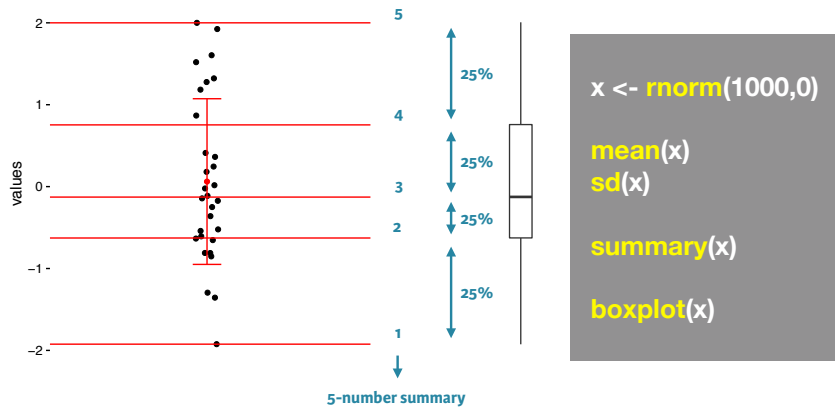
```
mean(x)
```

```
sd(x)
```

```
summary(x)
```

Side-note: boxplot

Graphical form of the 5 number summary!



```
x <- rnorm(1000,0)
```

```
mean(x)
```

```
sd(x)
```

```
summary(x)
```

```
boxplot(x)
```

Also called **box-and-whisker** plots;
See also `hist()`; violin plots etc.

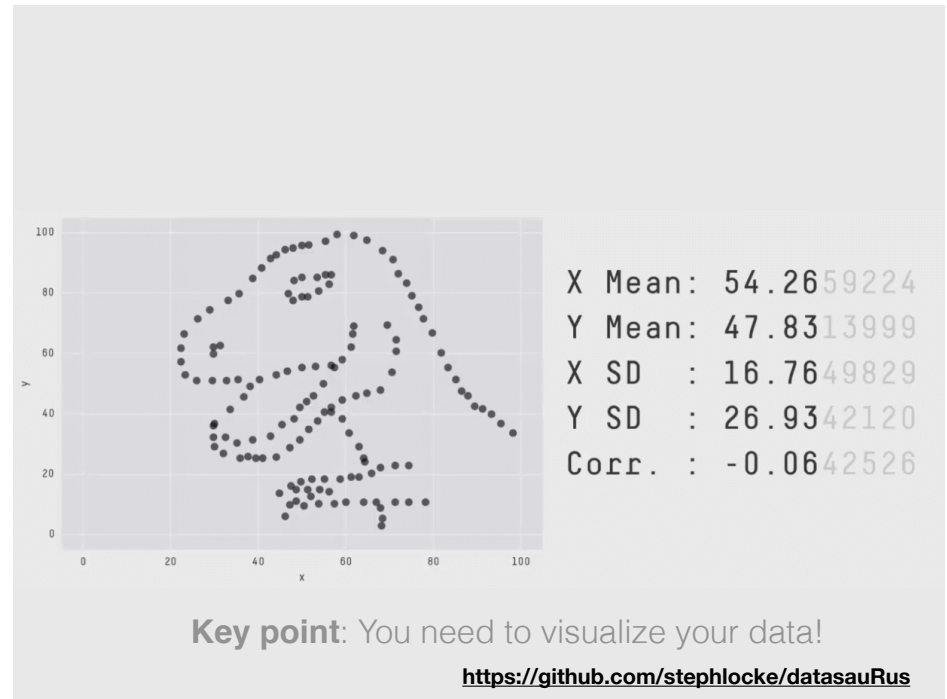
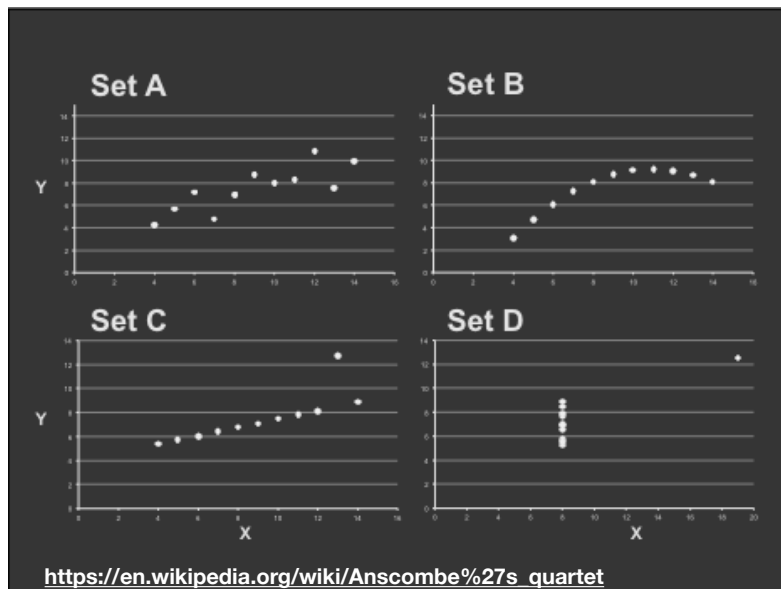
The Trouble with Summary Stats

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

$\mu_x = 9.0$	$\sigma_x = 3.317$	$Y = 3 + 0.5 X$	[Anscombe 73]
$\mu_y = 7.5$	$\sigma_y = 2.03$	$R^2 = 0.67$	

Looking at Data



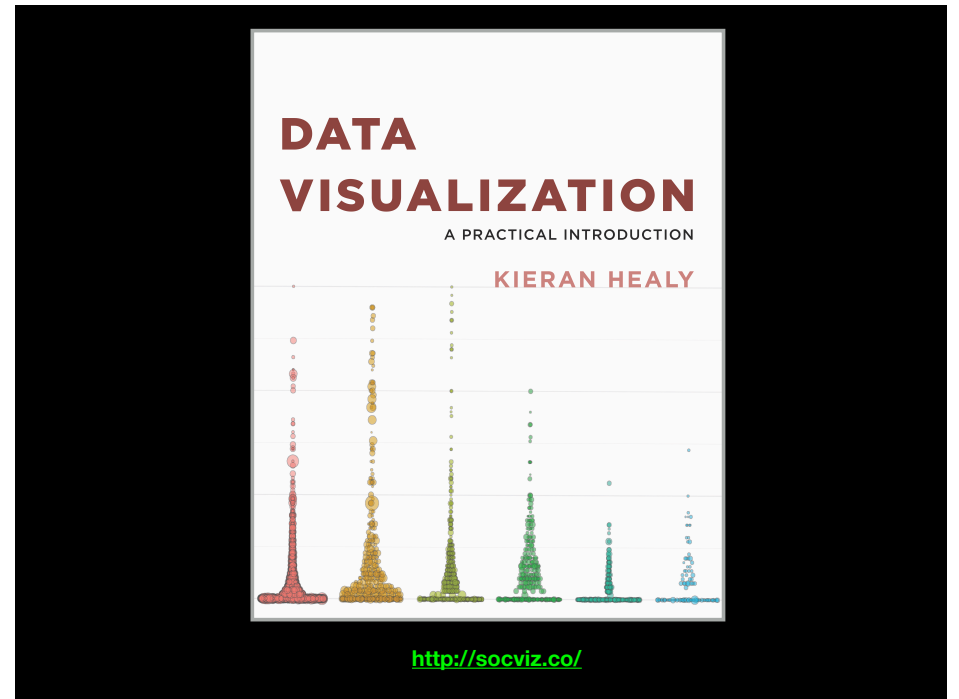
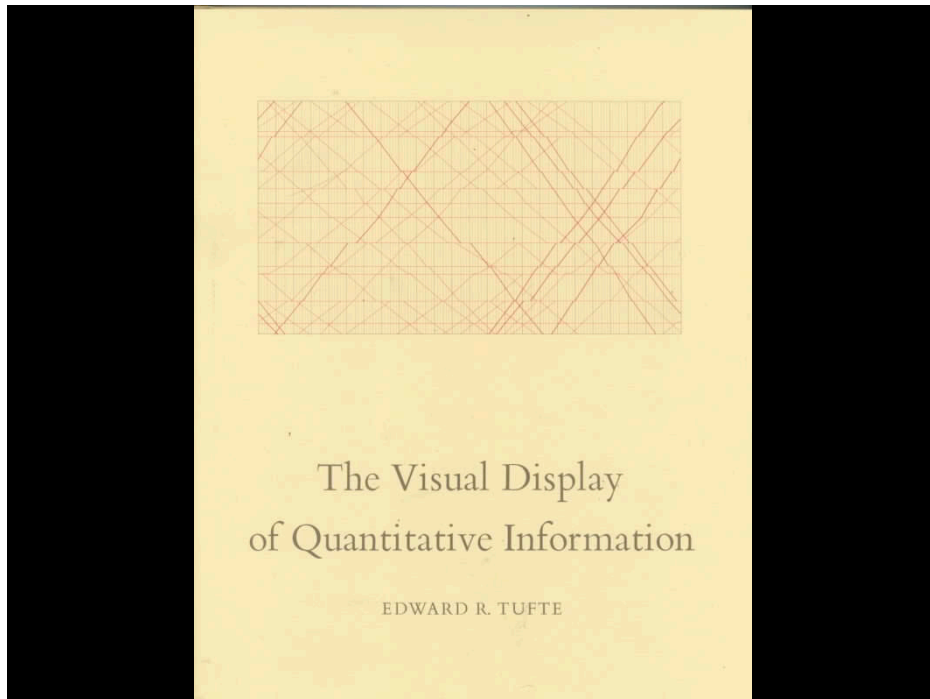
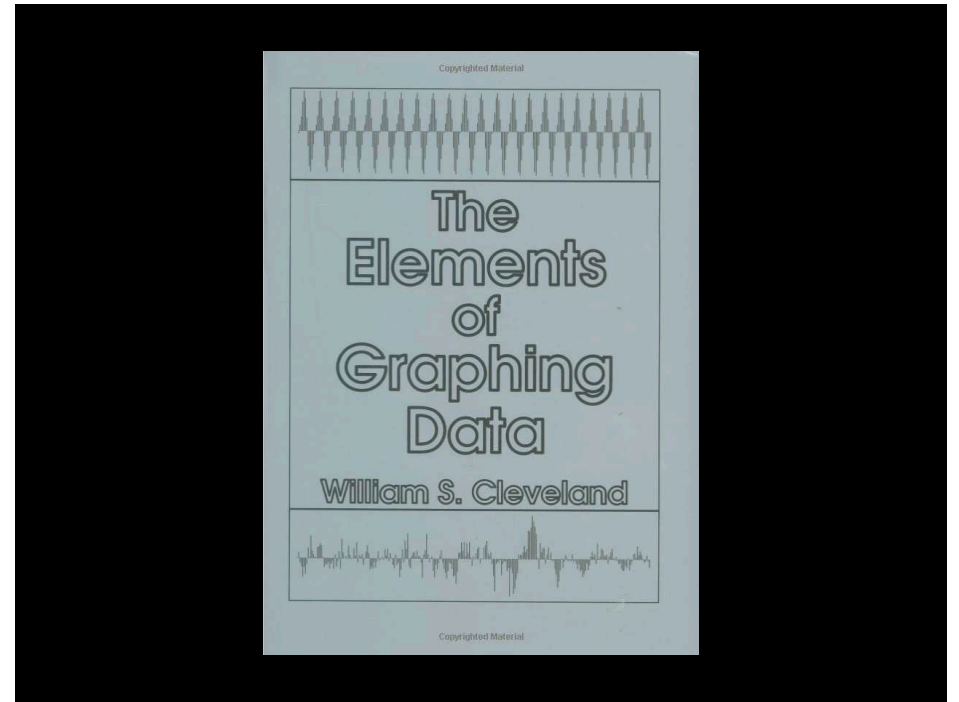
Today's Learning Goals

- 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**.



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

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

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**

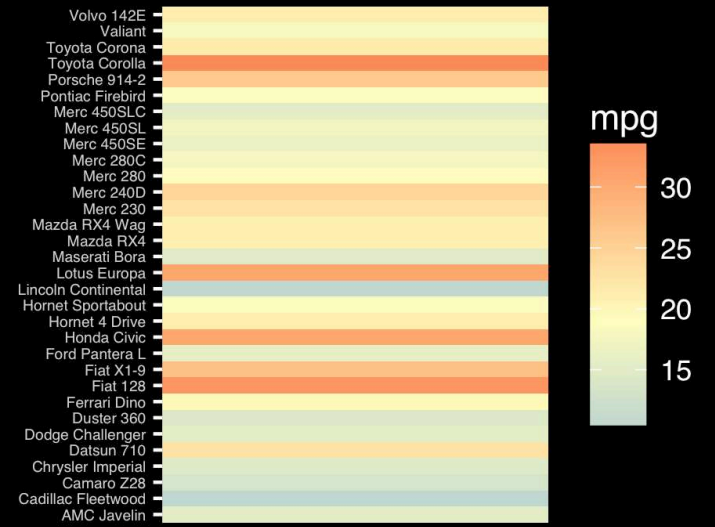
luminance



saturation



hue



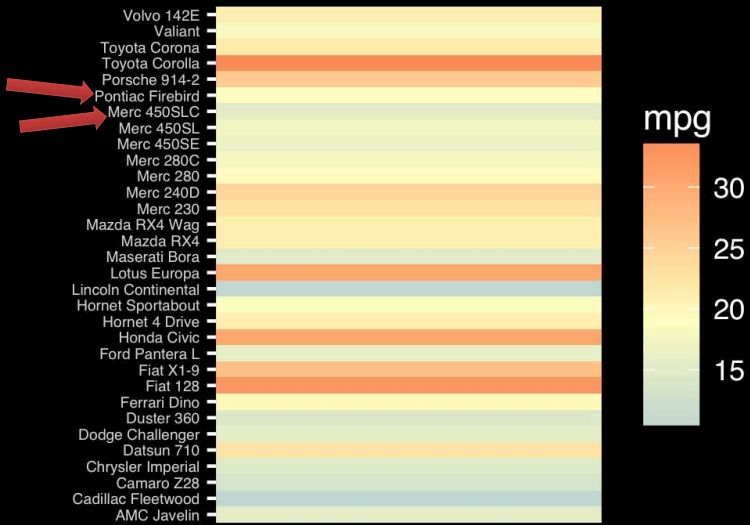
mpg

30

25

20

15



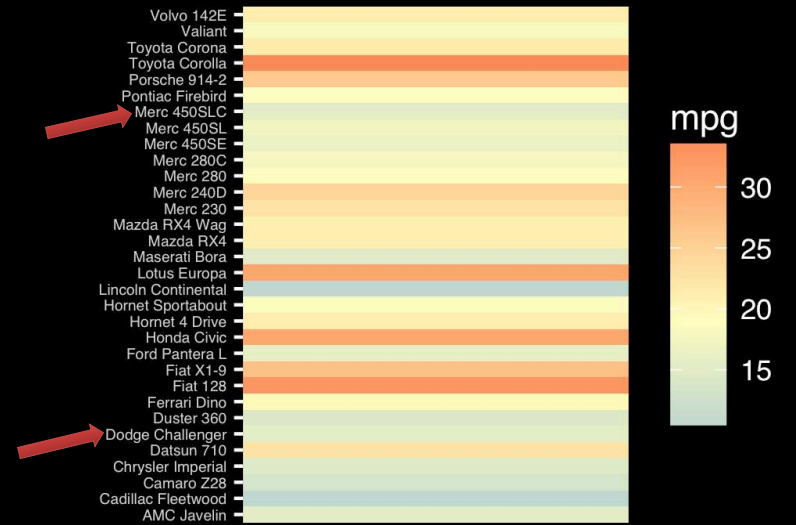
mpg

30

25

20

15



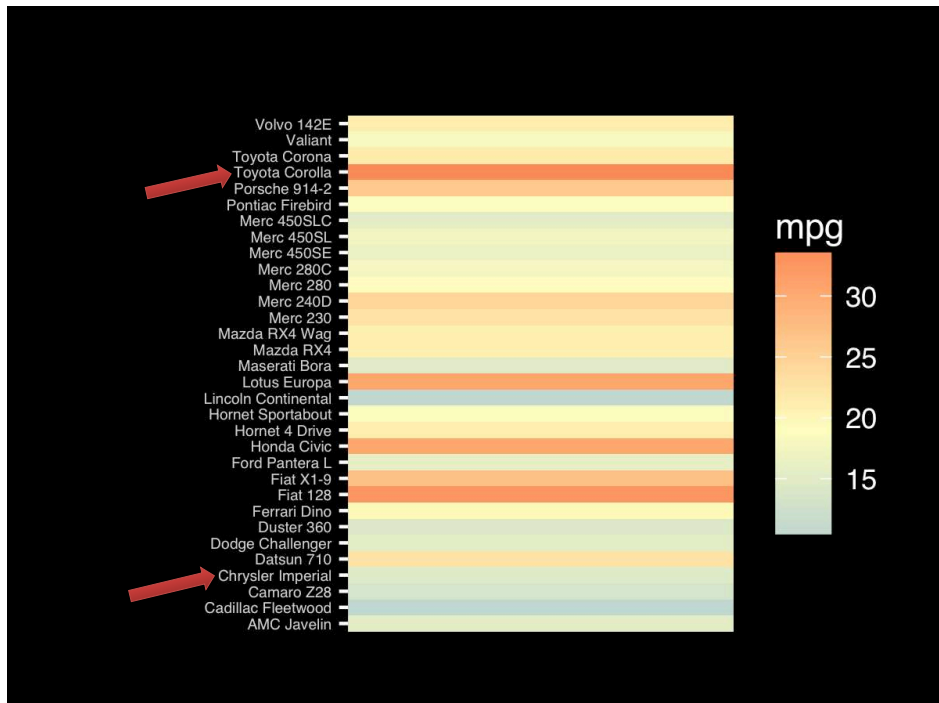
mpg

30

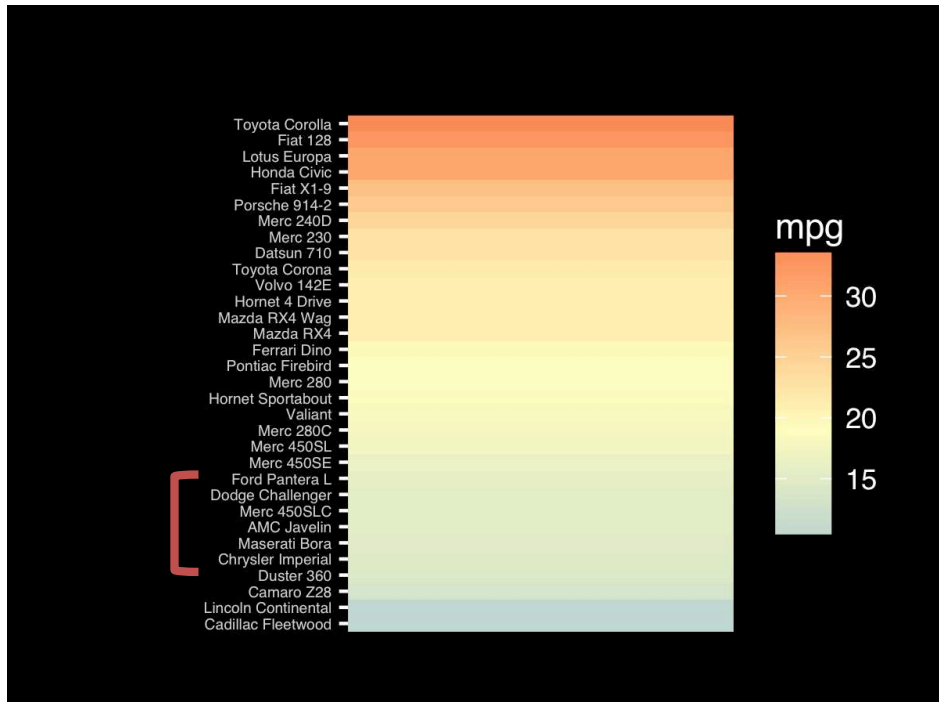
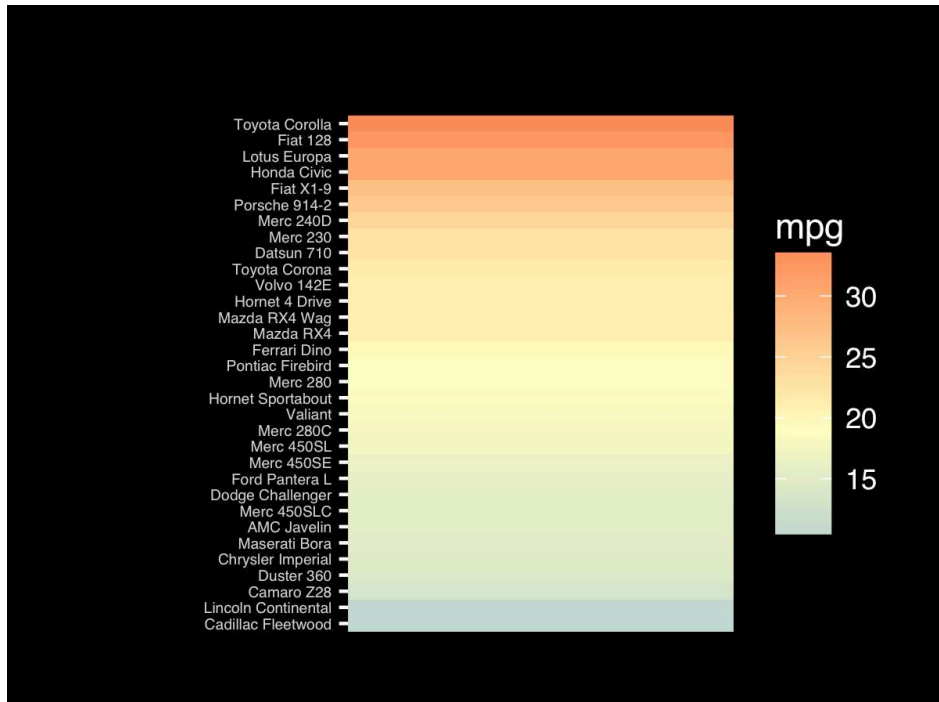
25

20

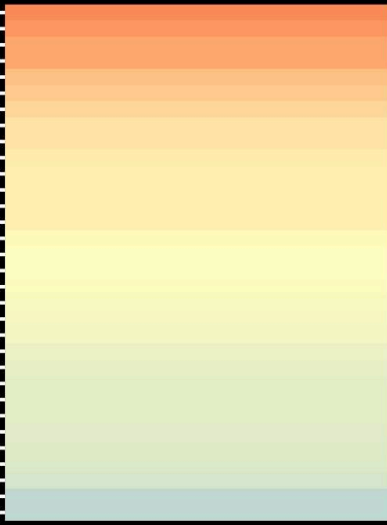
15



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

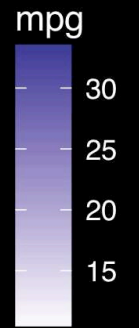
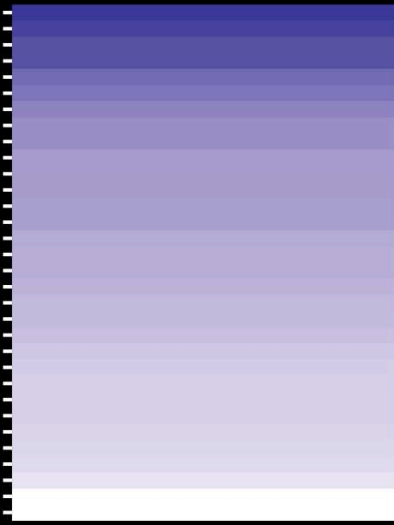


Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
Toyota Corona
Volvo 142E
Hornet 4 Drive
Mazda RX4 Wag
Mazda RX4
Ferrari Dino
Pontiac Firebird
Merc 280
Hornet Sportabout
Valiant
Merc 280C
Merc 450SL
Merc 450SE
Ford Pantera L
Dodge Challenger
Merc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



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

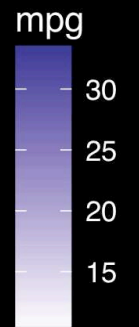
Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
Toyota Corona
Volvo 142E
Hornet 4 Drive
Mazda RX4 Wag
Mazda RX4
Ferrari Dino
Pontiac Firebird
Merc 280
Hornet Sportabout
Valiant
Merc 280C
Merc 450SL
Merc 450SE
Ford Pantera L
Dodge Challenger
Merc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

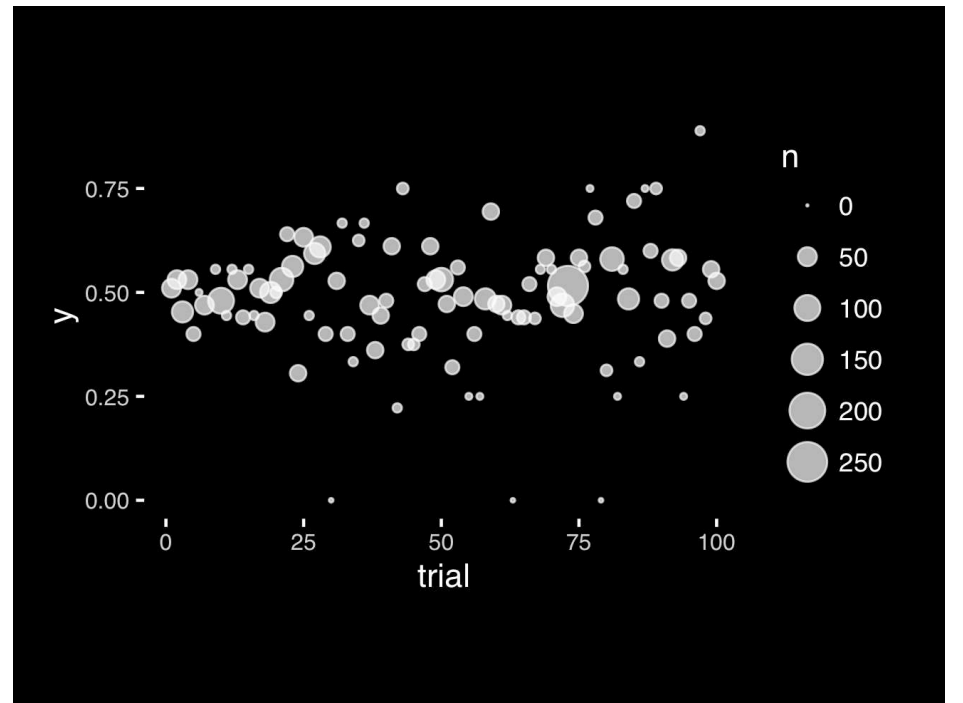


The most important measurement should exploit the highest ranked encoding possible.

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

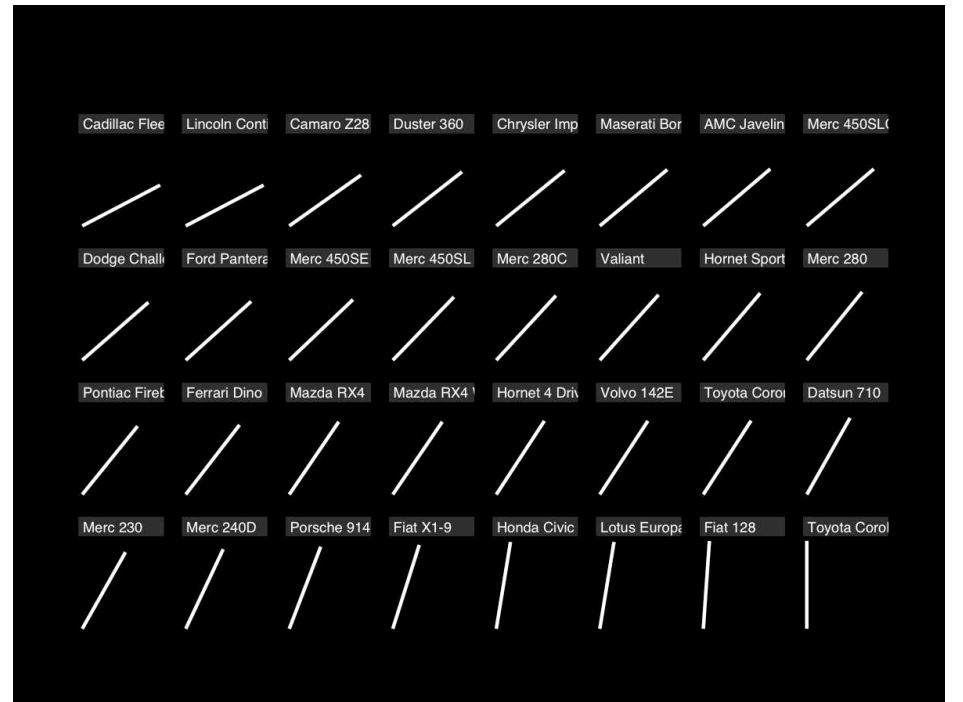
Toyota Corolla
Fiat 128
Lotus Europa
Honda Civic
Fiat X1-9
Porsche 914-2
Merc 240D
Merc 230
Datsun 710
Toyota Corona
Volvo 142E
Hornet 4 Drive
Mazda RX4 Wag
Mazda RX4
Ferrari Dino
Pontiac Firebird
Merc 280
Hornet Sportabout
Valiant
Merc 280C
Merc 450SL
Merc 450SE
Ford Pantera L
Dodge Challenger
Merc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

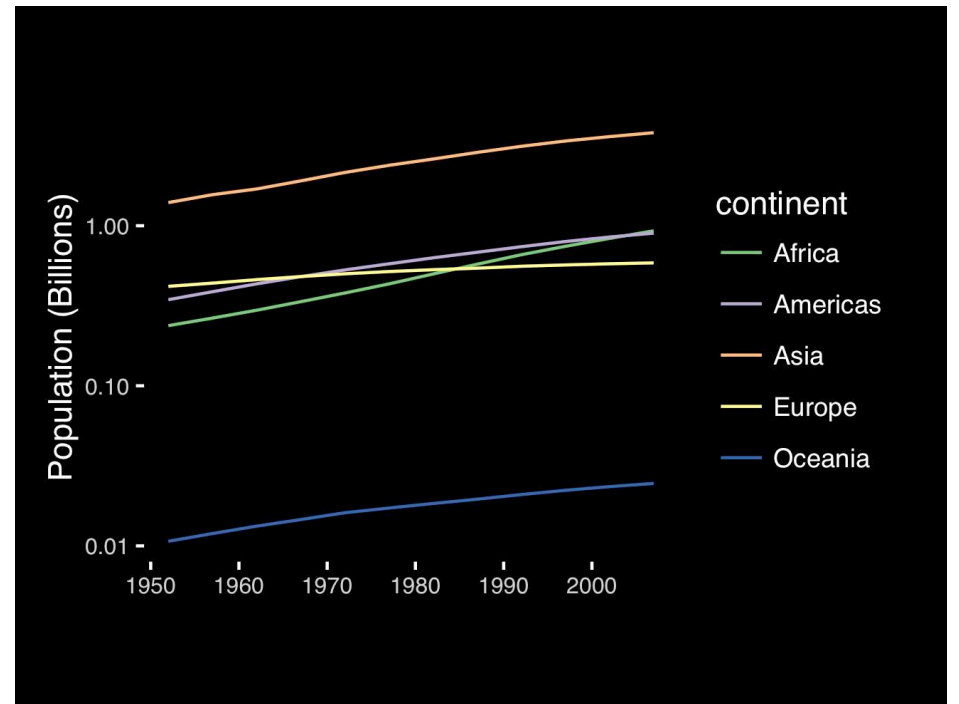
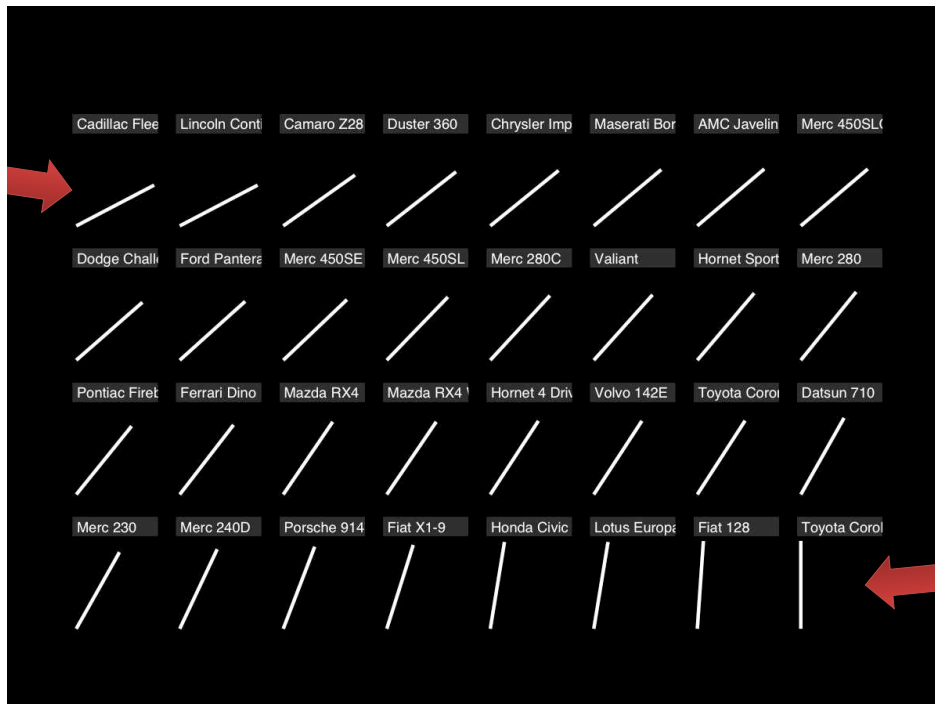




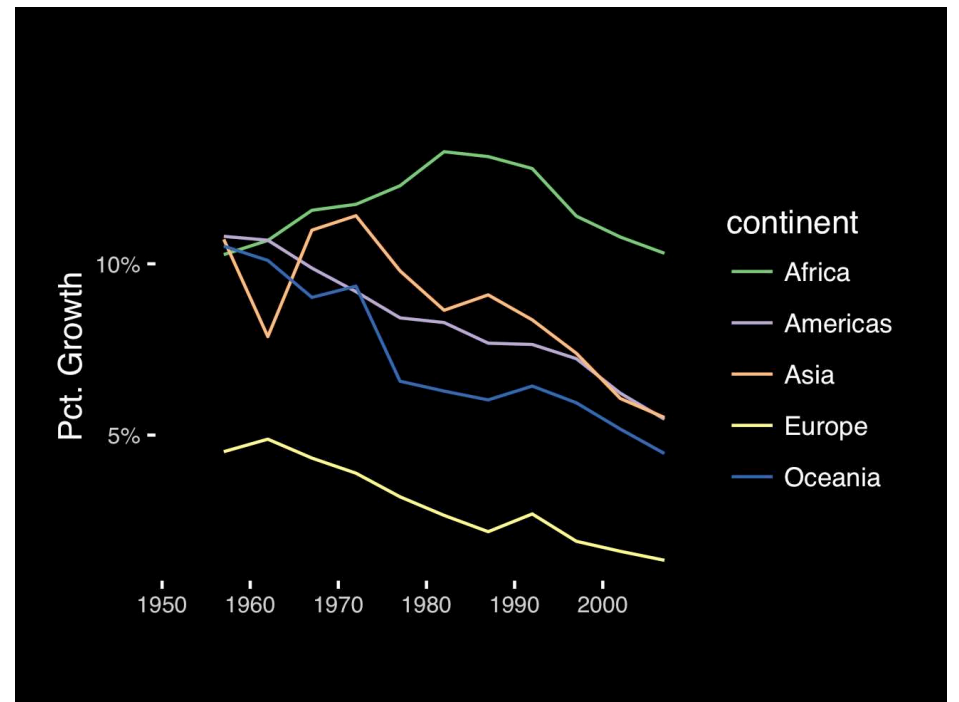
The most important measurement should exploit the highest ranked encoding possible.

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





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



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

Observation: Pie charts are ALWAYS a mistake.

Apart from MPAs :-)

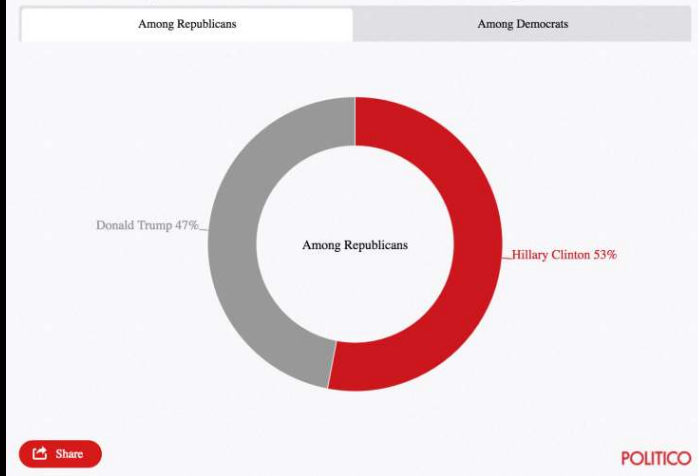
Piecharts are the information visualization equivalent of a roofing hammer to the frontal lobe. They have no place in the world of grownups, and occupy the same semiotic space as short pants, a runny nose, and chocolate smeared on one's face. They are as professional as a pair of assless chaps.

<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.**

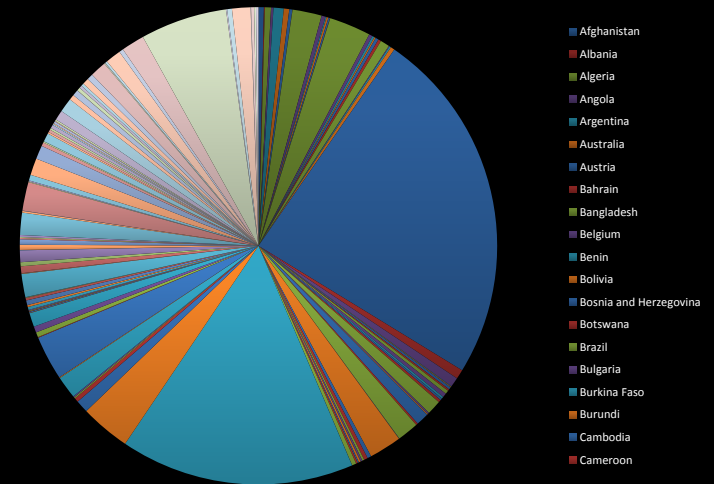
-Edward Tufte, The Visual Display of Quantitative Information

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.

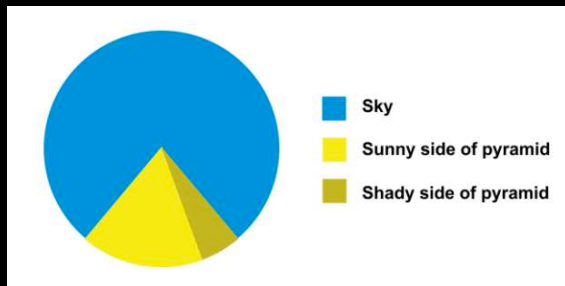
-Edward Tufte, The Visual Display of Quantitative Information

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

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

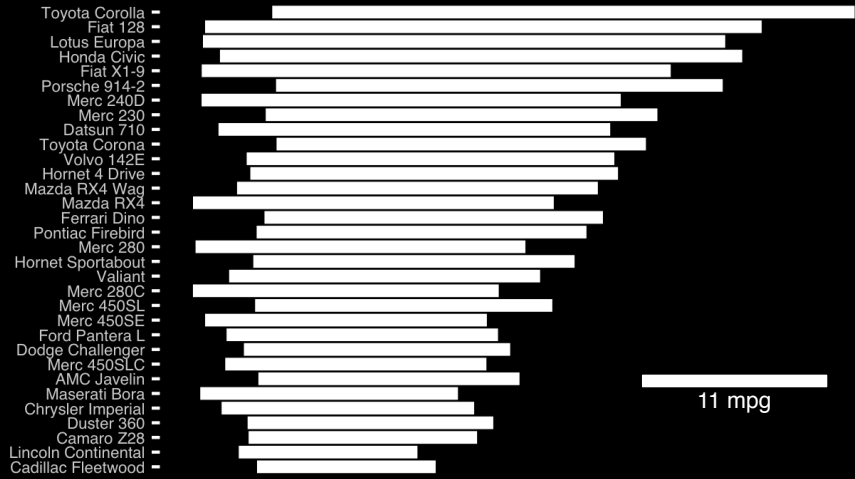


All good pie charts are jokes...

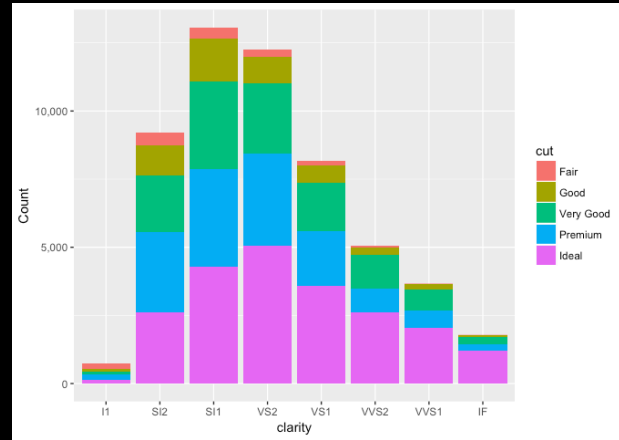


The most important measurement should exploit the highest ranked encoding possible.

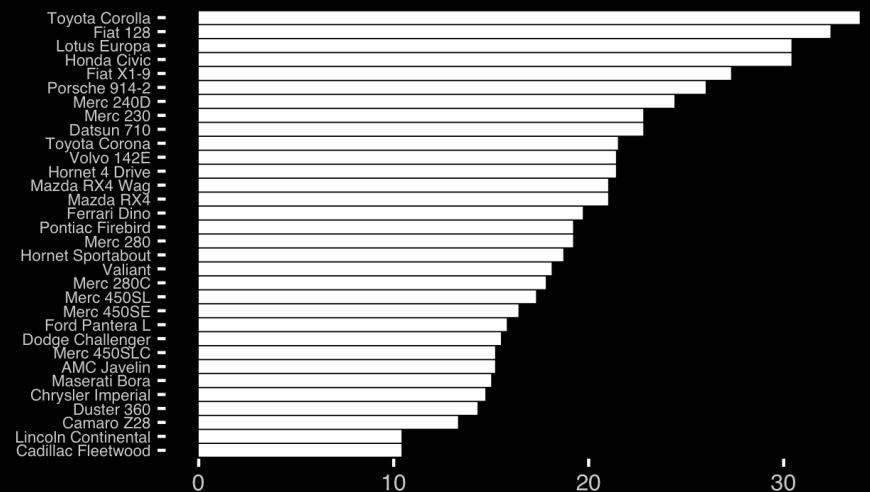
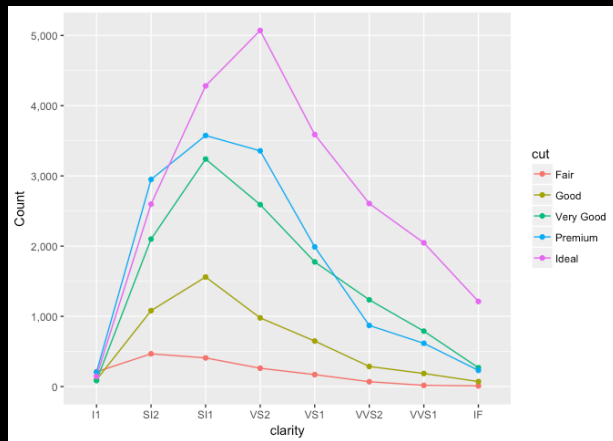
- Position along a common scale
- **Position on identical but nonaligned scales**
- Length
- Angle or Slope
- Area
- Volume or Density or Color saturation/hue

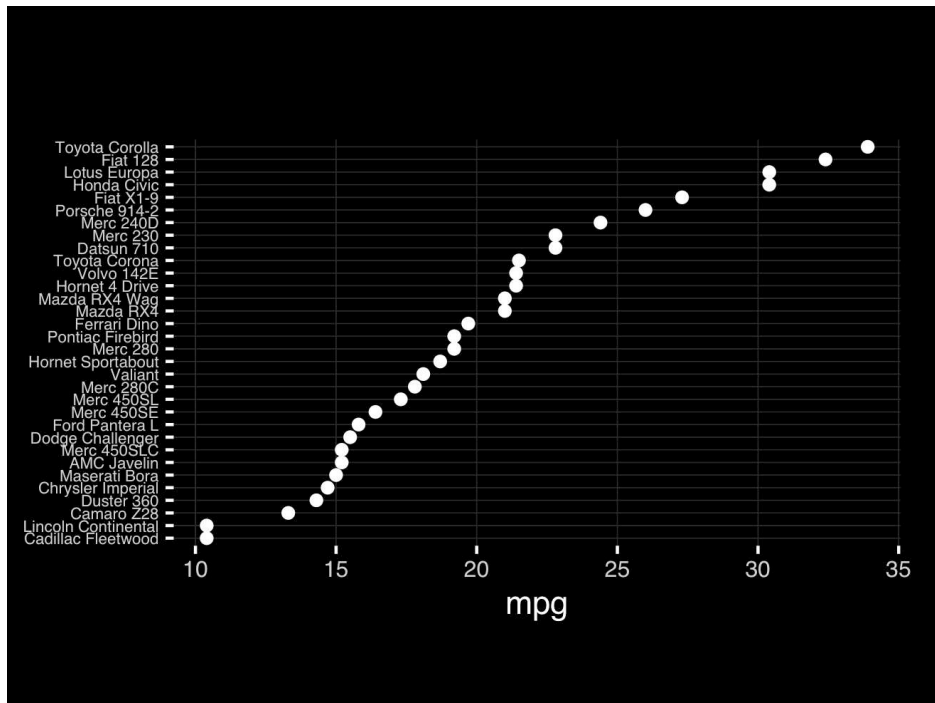


Stacked anything is nearly always a mistake



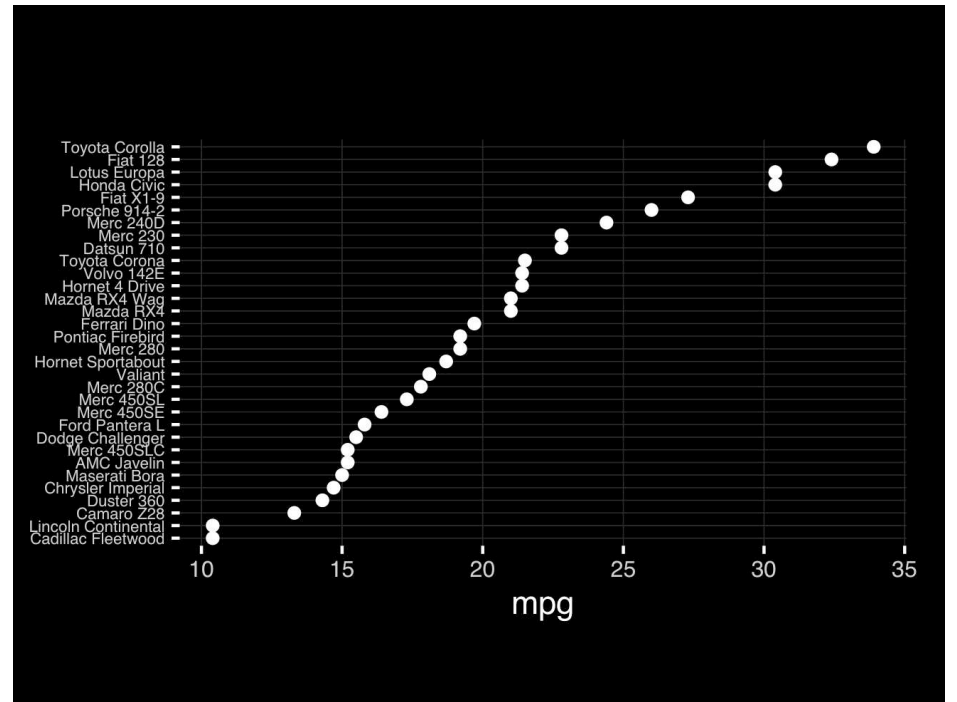
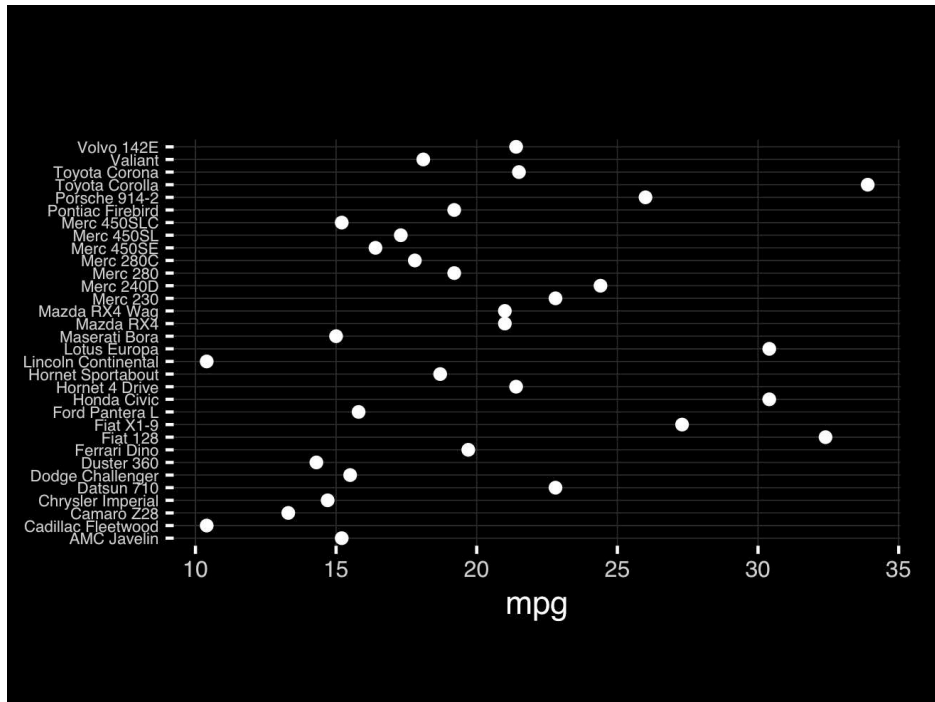
This is much better...

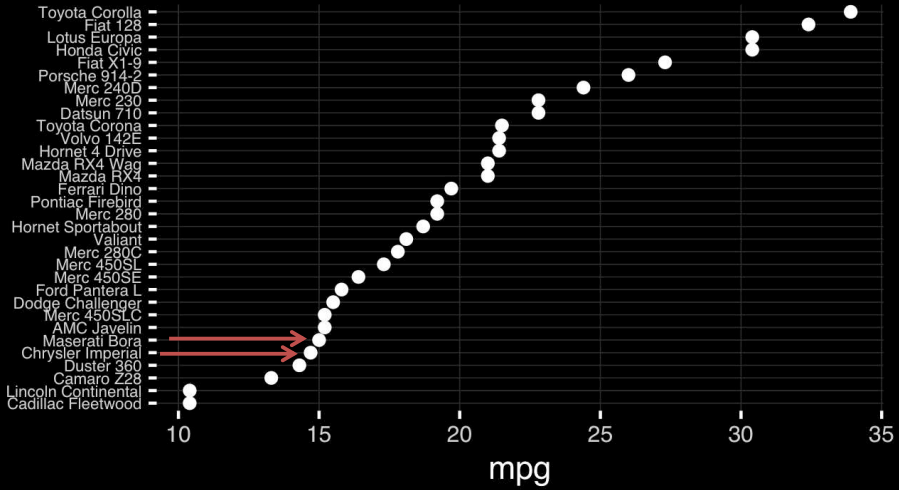




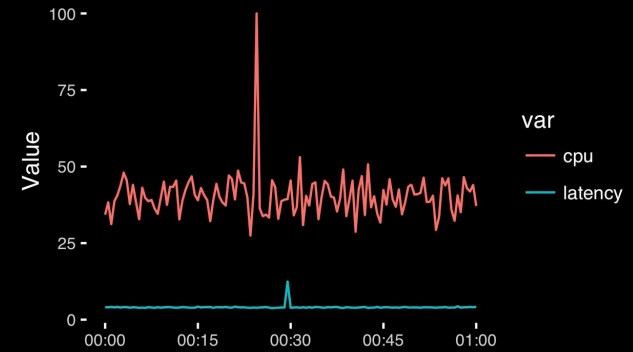
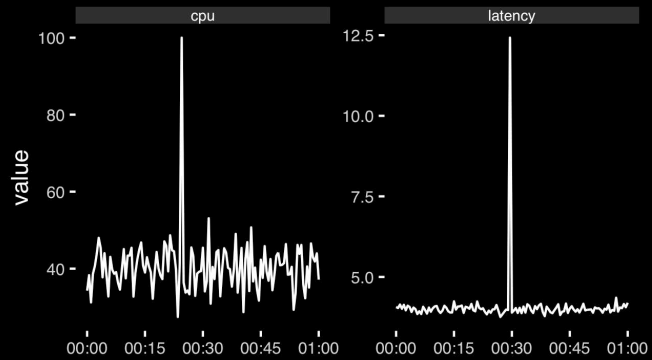
The most important measurement should exploit the highest ranked encoding possible.

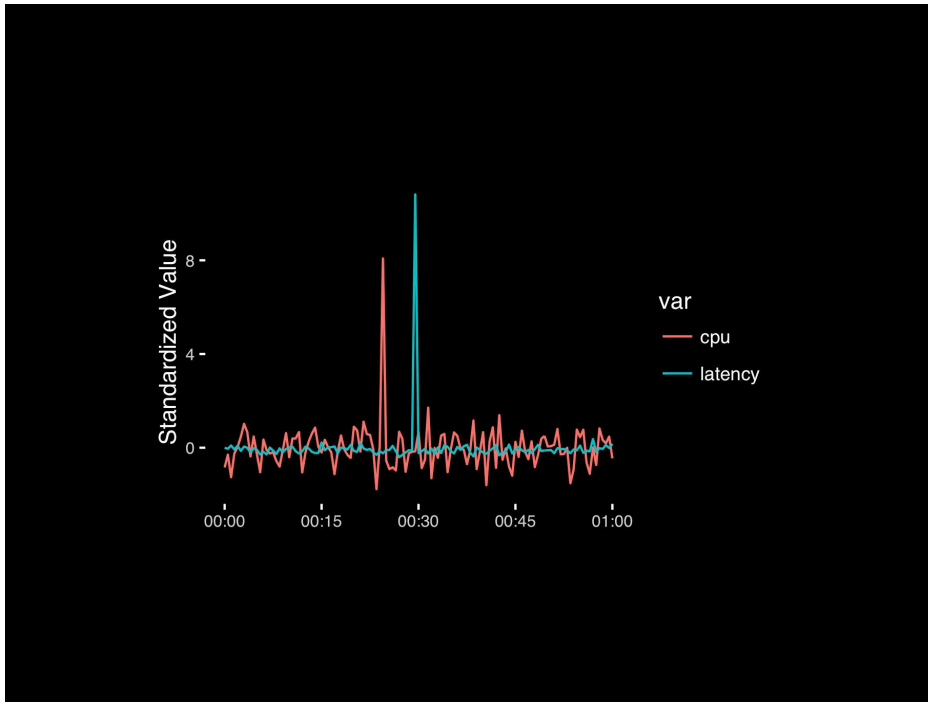
- Position along a common scale
- Position on identical but nonaligned scales
- Length
- Angle or Slope
- Area
- Volume or Density or Color saturation/hue





Observation: Comparison is trivial on a common scale.





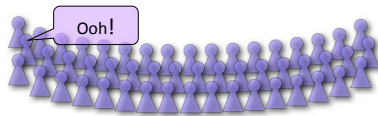
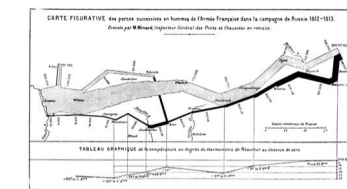
Today's Learning Goals

- 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**.

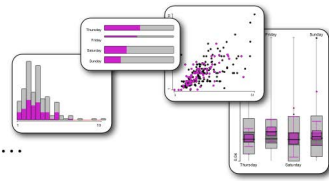
Different graphs for different purposes

Exploratory graphs: many images for a narrow audience (you!)

Presentation graphs: single image for a large audience

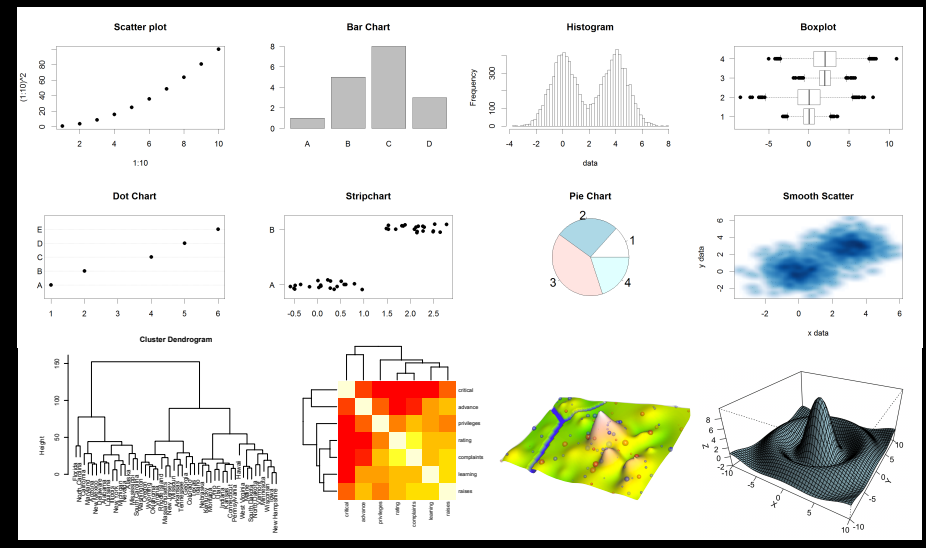


Presentation

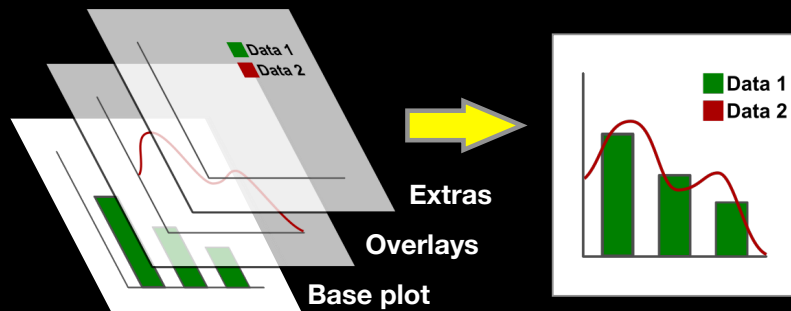


Exploration

Core R Graph Types



The R Painters Model

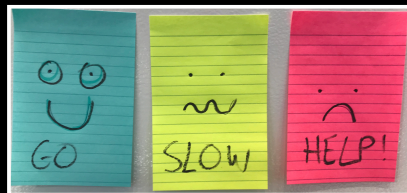


Side-Note: "Red and green should never be seen"

Do it Yourself!

Hands-on Section 1 only please

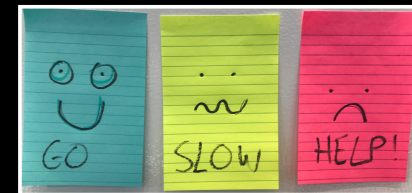
- Create a new **RStudio Project** for this class,
- **Download** the example data files and move them to your project directory,
- Focus on **Sections 1A & 1B** in the **handout**.



Do it Yourself!

Hands-on Section 1 only please

- Create a new **RStudio Project** for this class,
- **Download** the example data files and move them to your project directory,
- Focus on **Sections 1A & 1B** in the **handout**.



Do it Yourself!

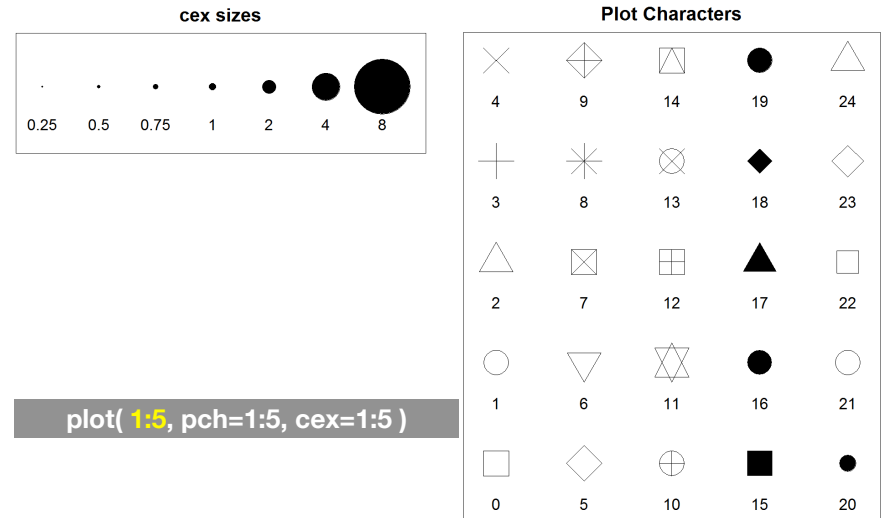
Hands-on Section 2 Notes

- Focus on **Sections 2A & 2B** in the lab **handout**.
- Try **Section 2C** if you have time.
- See notes on the following slides...

Common Options

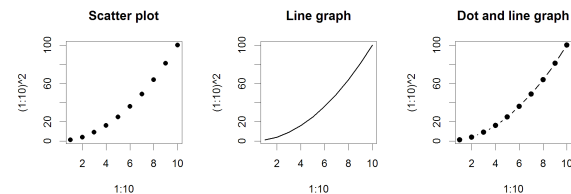
- **Axis scales**
 - `xlim c(min,max)`
 - `ylim c(min,max)`
 - **Axis labels**
 - `xlab(text)`
 - `ylab(text)`
 - **Plot titles**
 - `main(text)`
 - `sub(text)`
 - **Plot characters**
 - `pch(number)`
 - `cex(number)`
- Local options to change a specific plot
 - Global options to affect all graphs

Plot Characters



Plot Type Specific Options

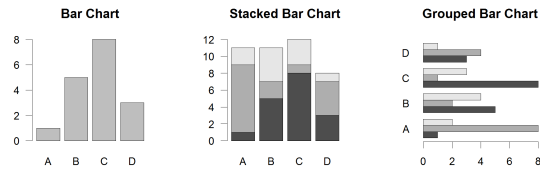
Plot (scatterplots and line graphs)



- **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)`

Section 2B: Barplot (a.k.a. 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)
```

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
```

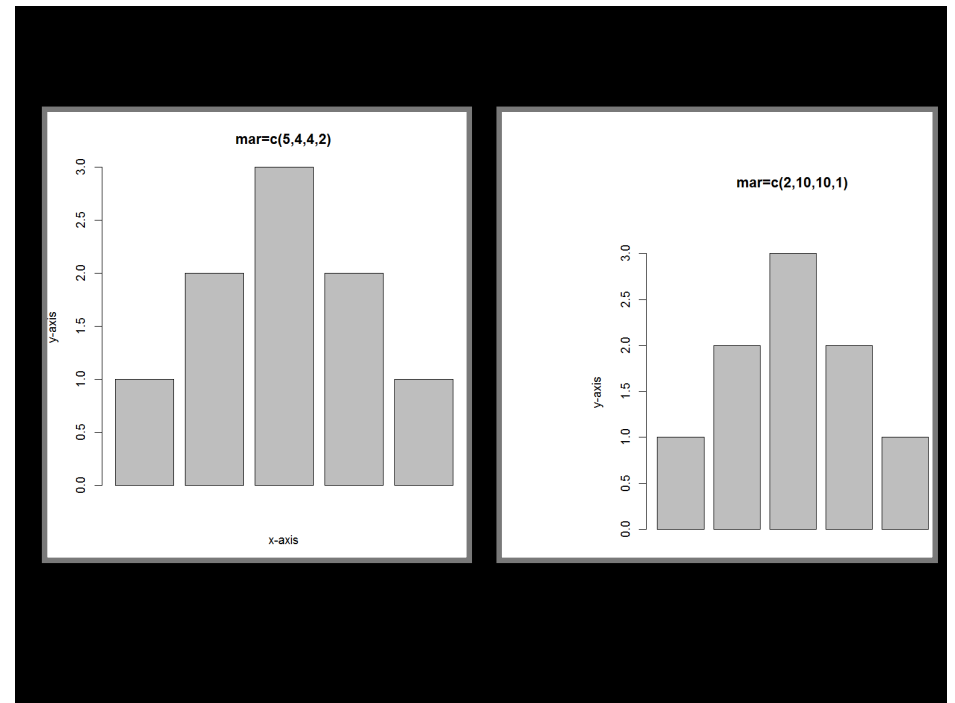
Par examples

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

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) )
```



Par options

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

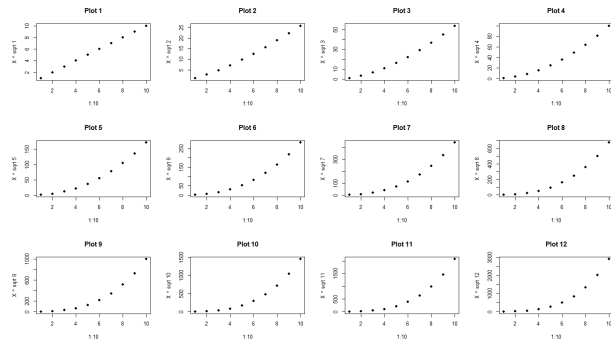


```
par( cex.main=1.5, cex.axis=0.5, cex.lab=2 )
```


Par options

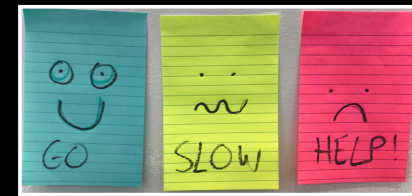
- Multi-panel

- `par(mfrow=c(rows, cols))`



`par(mfrow=c(3, 4))`

Using Color



Hands-on Section 3 only please

Specifying colors

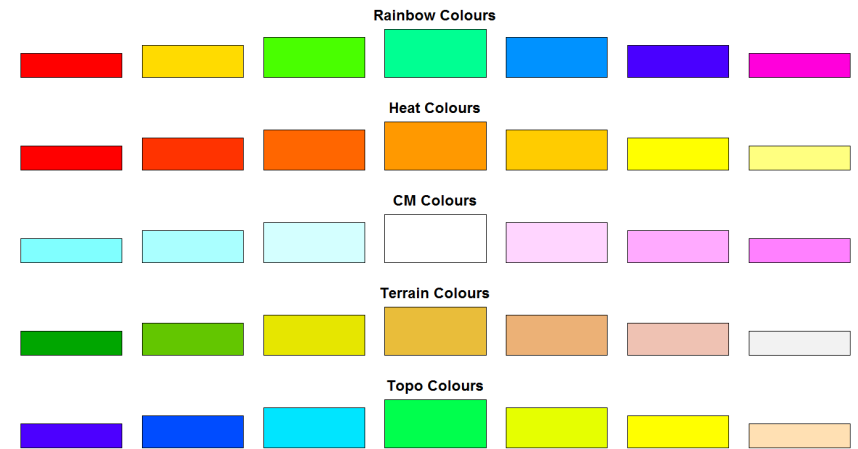
- Color by number
 - `col=c(1, 2, 3)`
 - Will give black, red, green etc.
- Hexadecimal strings
 - `#FF0000` (red)
 - `#0000FF` (blue)
 - `#CC00CC` (purple)
- Controlled names
 - `col=c("red", "green")` etc.
 - see `colors()`
- Also RGB values, HCL values, etc.

Built in color schemes

- Functions to generate colors
- Pass in number of colors you want
- E.G. the functions:

- `rainbow()`
- `heat.colors()`
- `cm.colors()`
- `terrain.colors()`
- `topo.colors()`

```
rainbow(7)
```



```
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 **palette()** function
 - `palette()`
 - `palette(brewer.pal(9, "Set1"))`

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

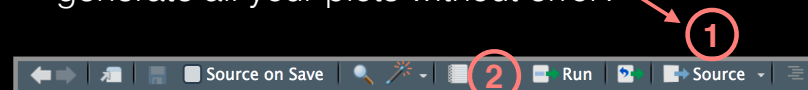
Dynamic use of color

- Coloring by density
 - Pass data and palette to `densCols()`
 - Vector of colors returned
- See **Lab Supplement** (online):
 - Plotting with color in R

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

Make a lab report!

- Open your previous **class05** RStudio **project** (and your saved **R script**)
- Can you **source** your **class05.R** file to re-generate all your plots without error?



- If so you can now generate a nice **HTML report** of your work to date...

[Take 2-3 minutes]

Homework!

New **DataCamp** Assignments

- **RStudio IDE (Pt 1)**
- **Intermediate R**
 - Conditionals and Control Flow
 - Functions
 - Loops

Muddy Point Assessment Form Link

Useful new website: <https://www.data-to-viz.com/>