



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**, **dendograms** 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**.

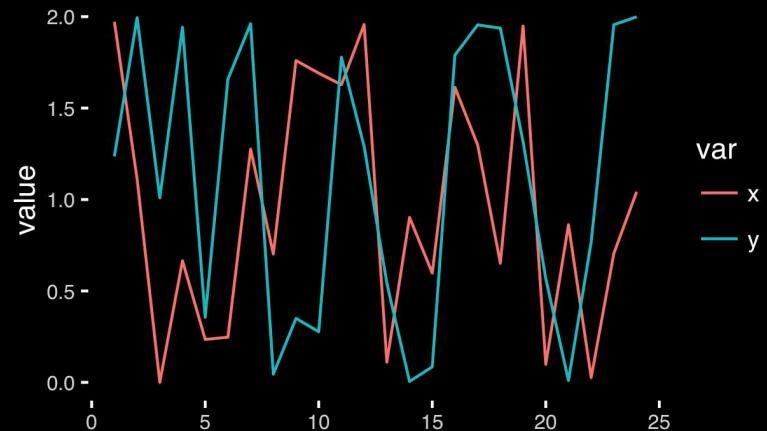
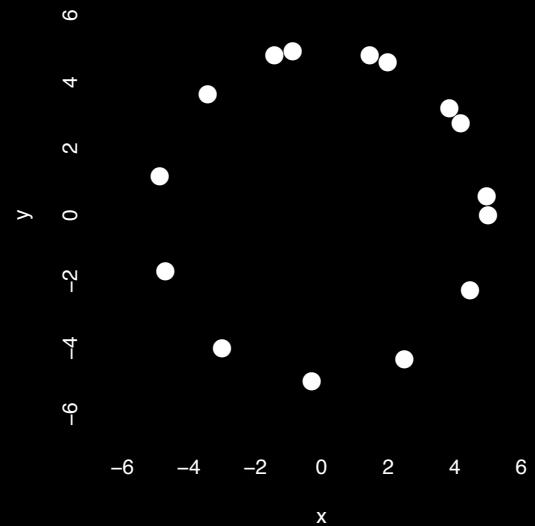
Today's Learning Goals

- 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**, **dendograms** 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**.

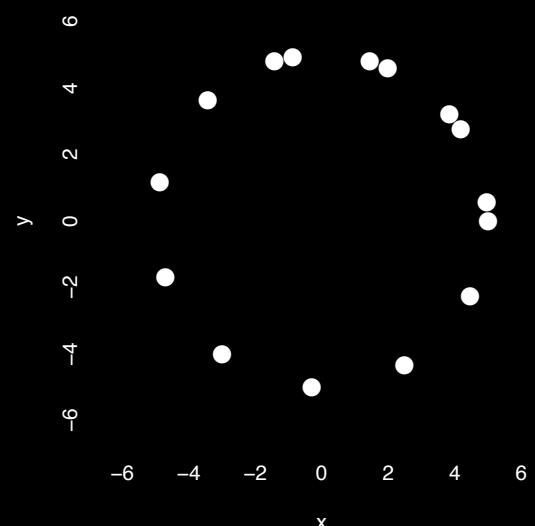
Why visualize at all?

	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/

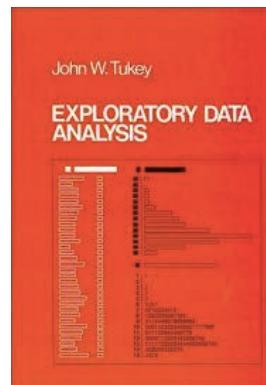


Exploratory Data Analysis

- 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!

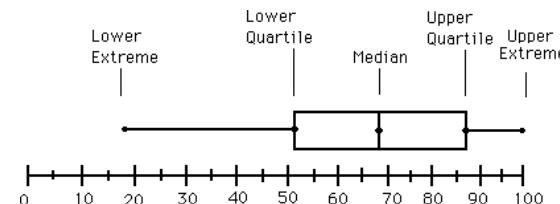
Exploratory Data Analysis 1977

- 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: boxplots

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



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

```
summary(); hist()
```

16

The Trouble with Summary Stats

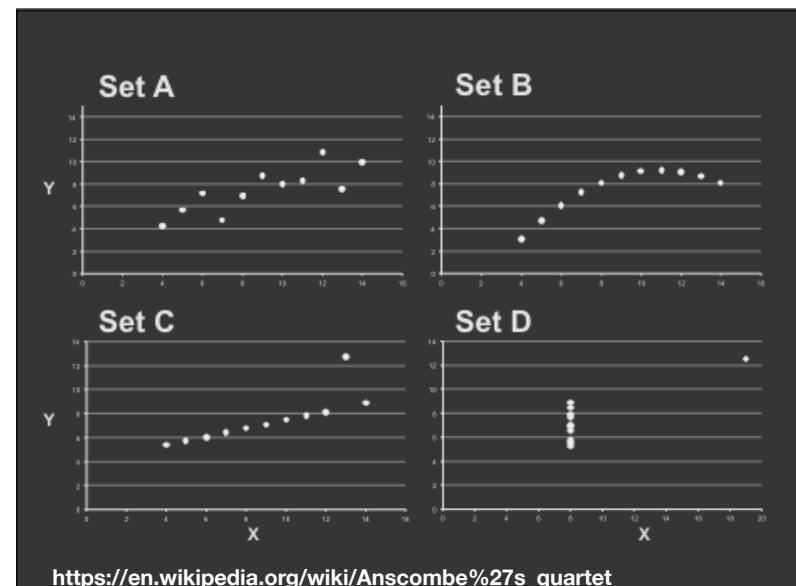
Set A		Set B		Set C		Set D	
X	Y	X	Y	X	Y	X	Y
10	8.04	10	9.14	10	7.46	8	6.58
8	6.95	8	8.14	8	6.77	8	5.76
13	7.58	13	8.74	13	12.74	8	7.71
9	8.81	9	8.77	9	7.11	8	8.84
11	8.33	11	9.26	11	7.81	8	8.47
14	9.96	14	8.1	14	8.84	8	7.04
6	7.24	6	6.13	6	6.08	8	5.25
4	4.26	4	3.1	4	5.39	19	12.5
12	10.84	12	9.11	12	8.15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

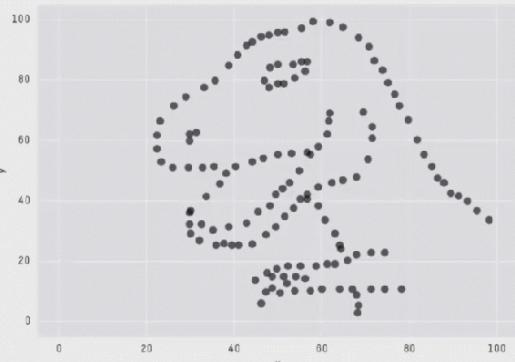
Summary Statistics Linear Regression

$u_x = 9.0 \quad \sigma_x = 3.317 \quad Y = 3 + 0.5 X$
 $u_y = 7.5 \quad \sigma_y = 2.03 \quad R^2 = 0.67$

[Anscombe 73]

Looking at Data





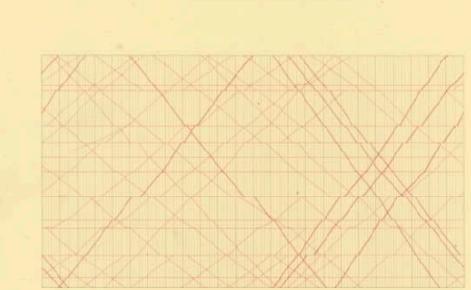
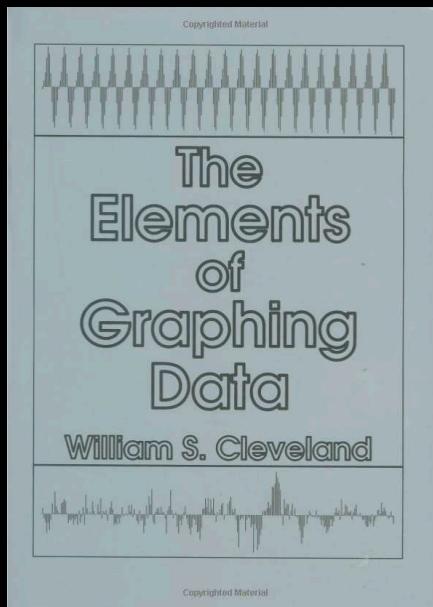
X Mean: 54.2659224
 Y Mean: 47.8313999
 X SD : 16.7649829
 Y SD : 26.9342120
 Corr. : -0.0642526

Key point: You need to visualize your data!

<https://github.com/stephlocke/datasauRus>

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, dendograms** 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**.



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

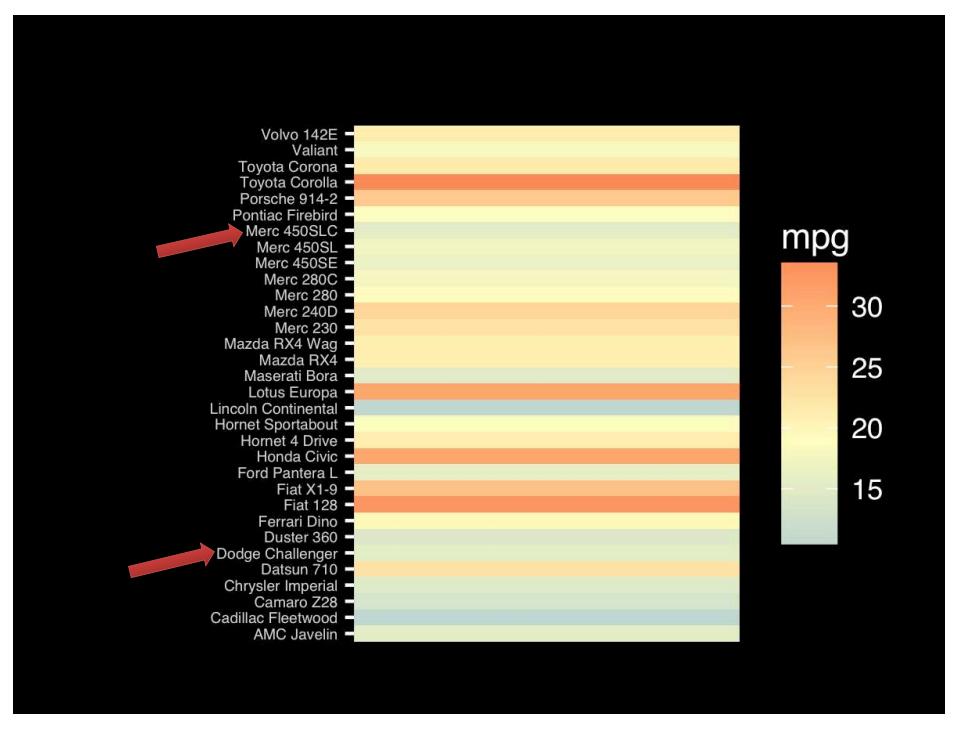
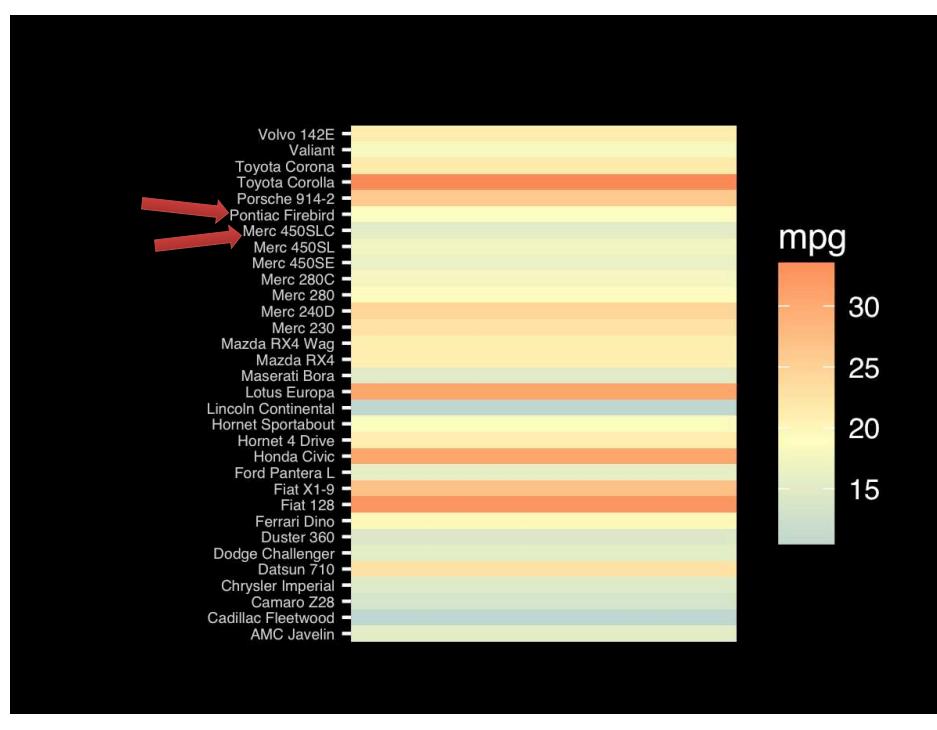
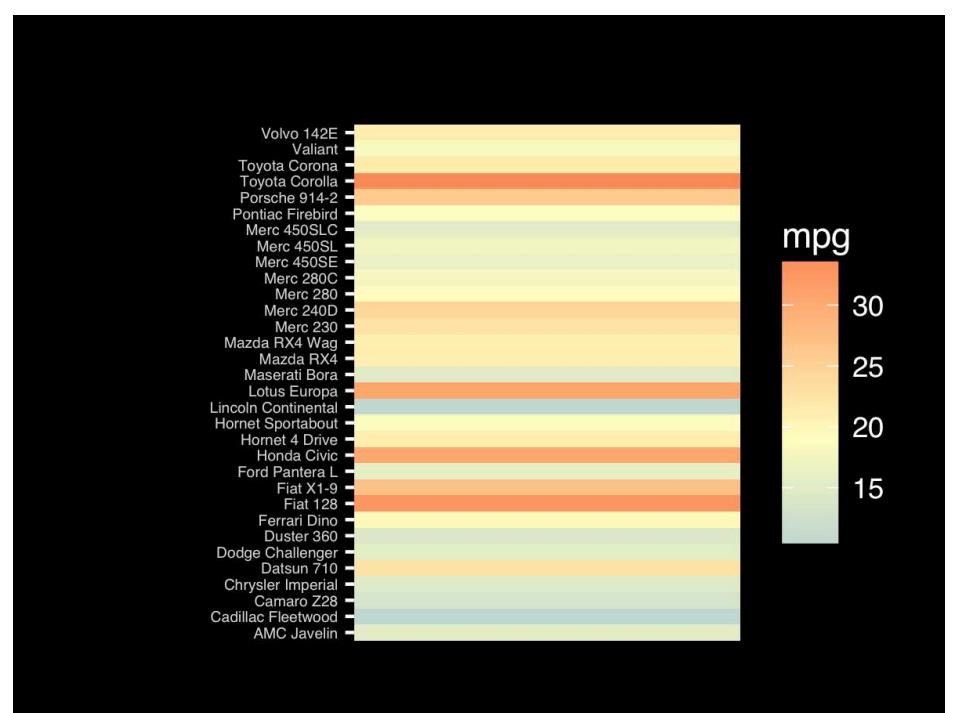
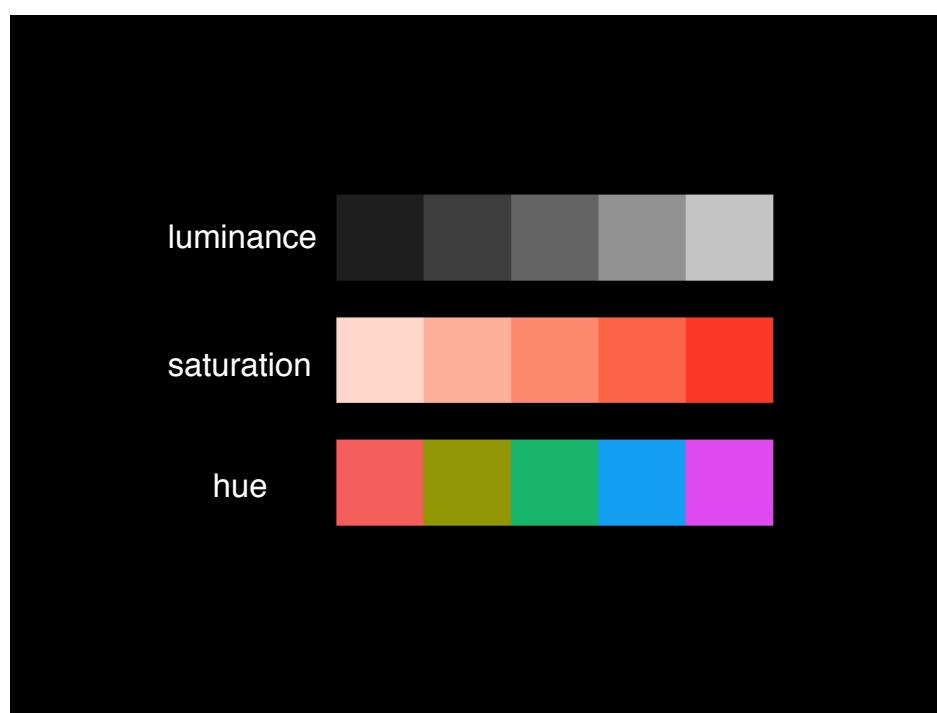
- Position along a common scale
- Position on identical but nonaligned scales
- Length
- Angle or Slope
- Area
- Volume or Density or Color saturation/hue

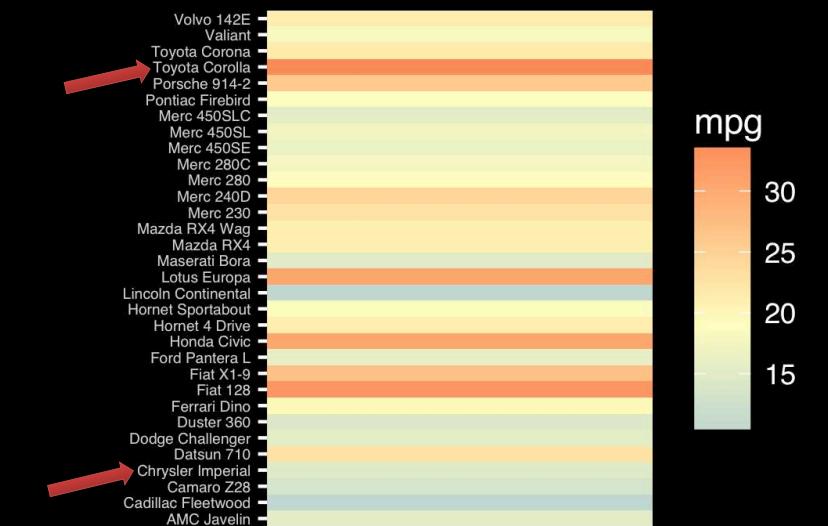
Key Point: The most important measurement should exploit the highest ranked encoding possible

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

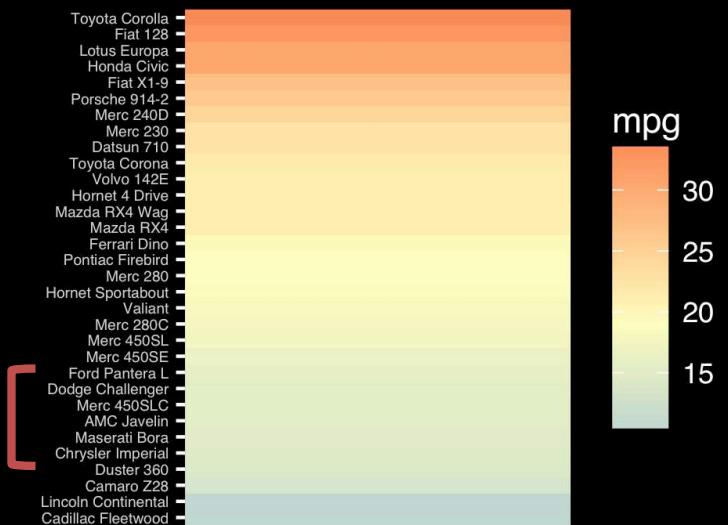
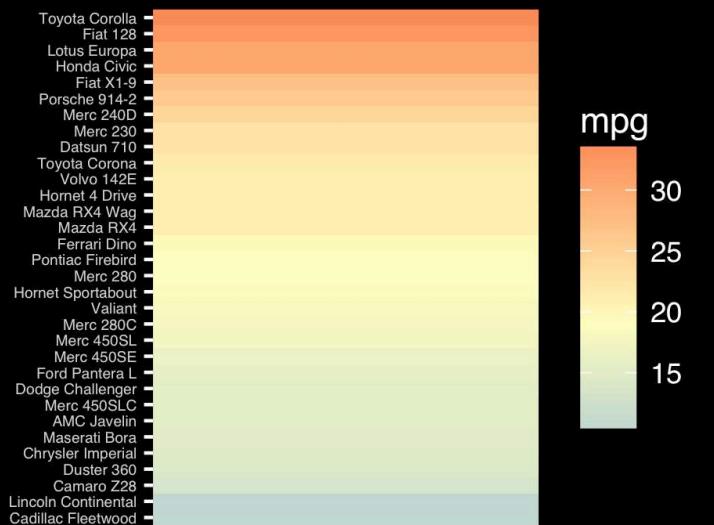
Key Point: The most important measurement should exploit the highest ranked encoding possible

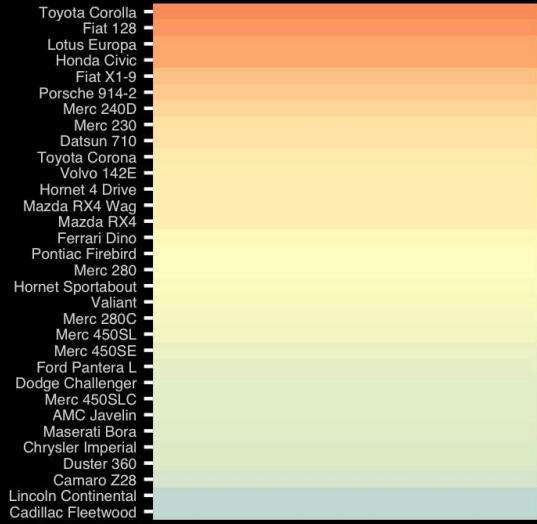
- 
- Position along a common scale
 - Position on identical but nonaligned scales
 - Length
 - Angle or Slope
 - Area
 - Volume or Density or **Color saturation/hue**



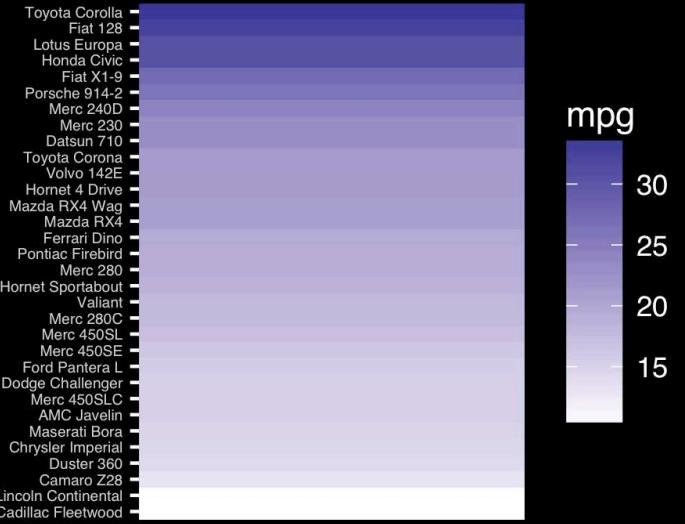


Observation: Alphabetical is almost never the correct ordering of a categorical variable.



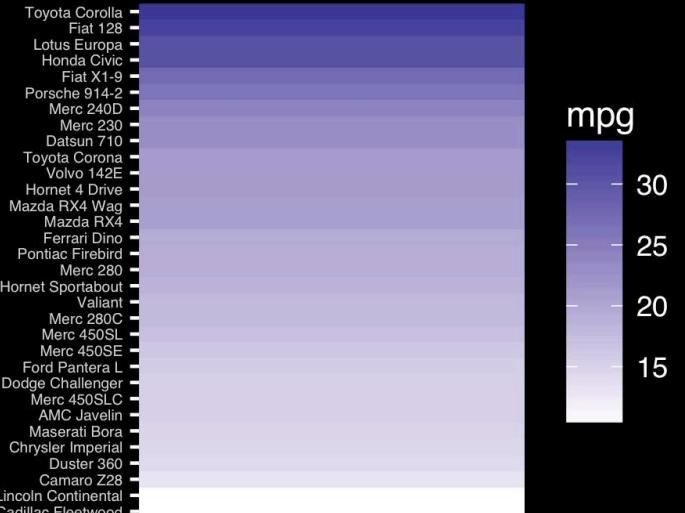


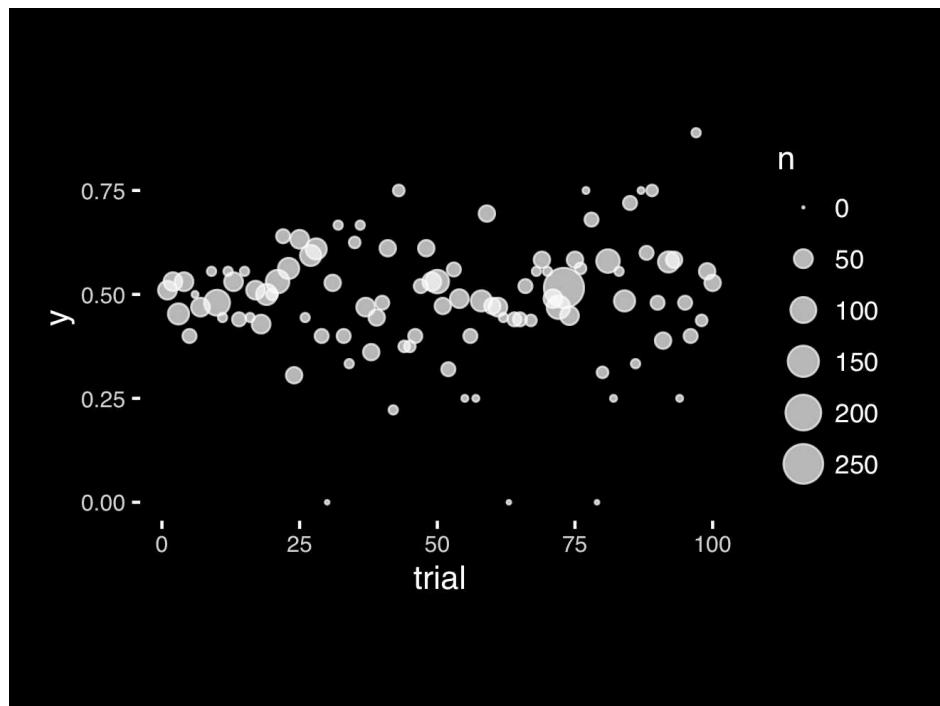
If we did not have the legend would you know which was low or high mpg?



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

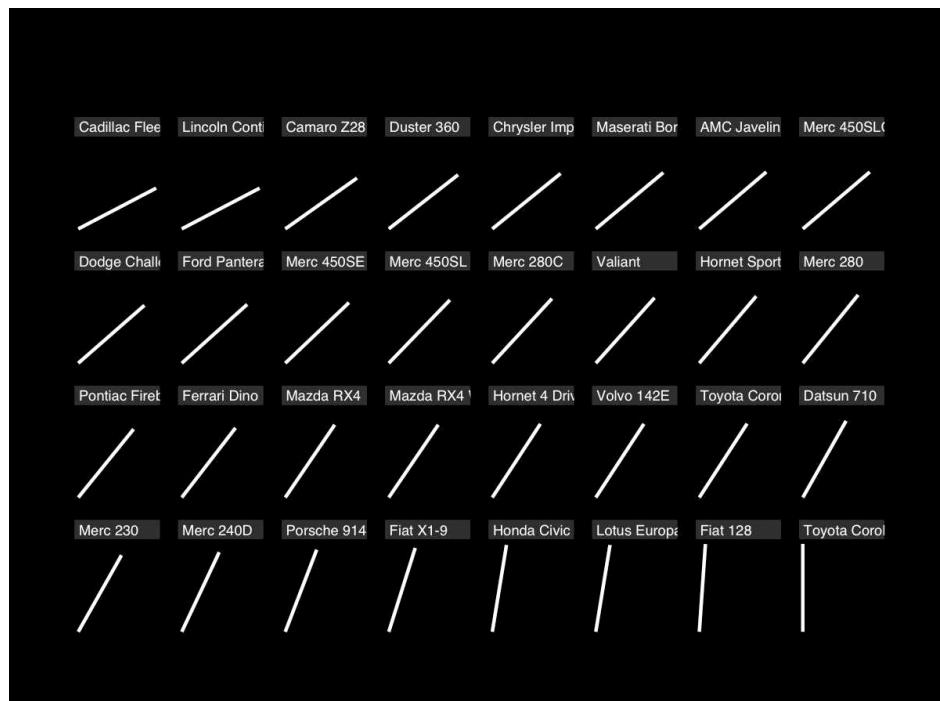
- Position along a common scale
- Position on identical but nonaligned scales
- Length
- Angle or Slope
- **Area**
- Volume or Density or Color saturation/hue

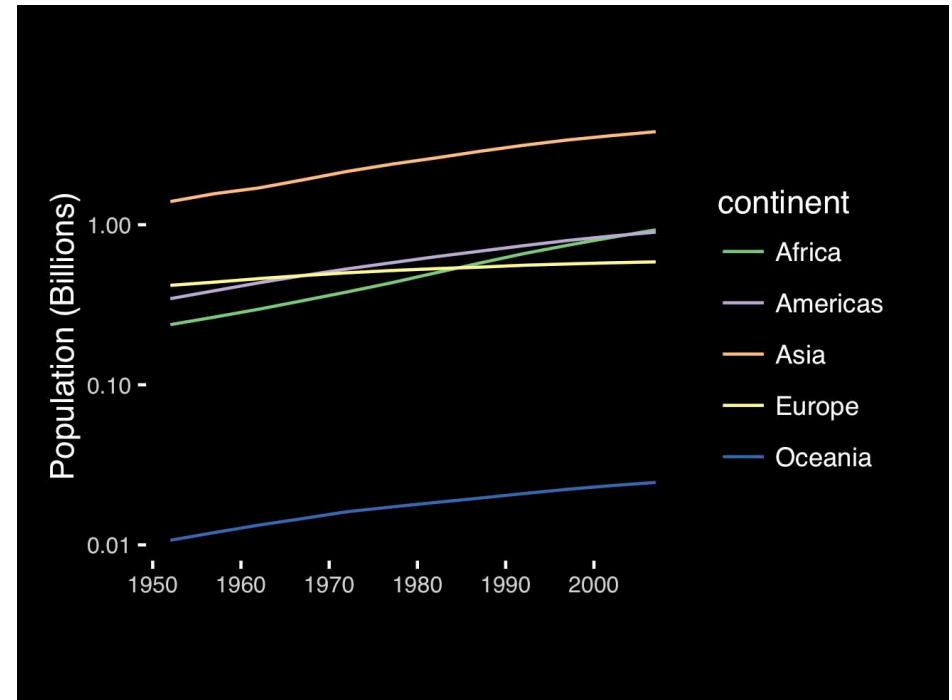
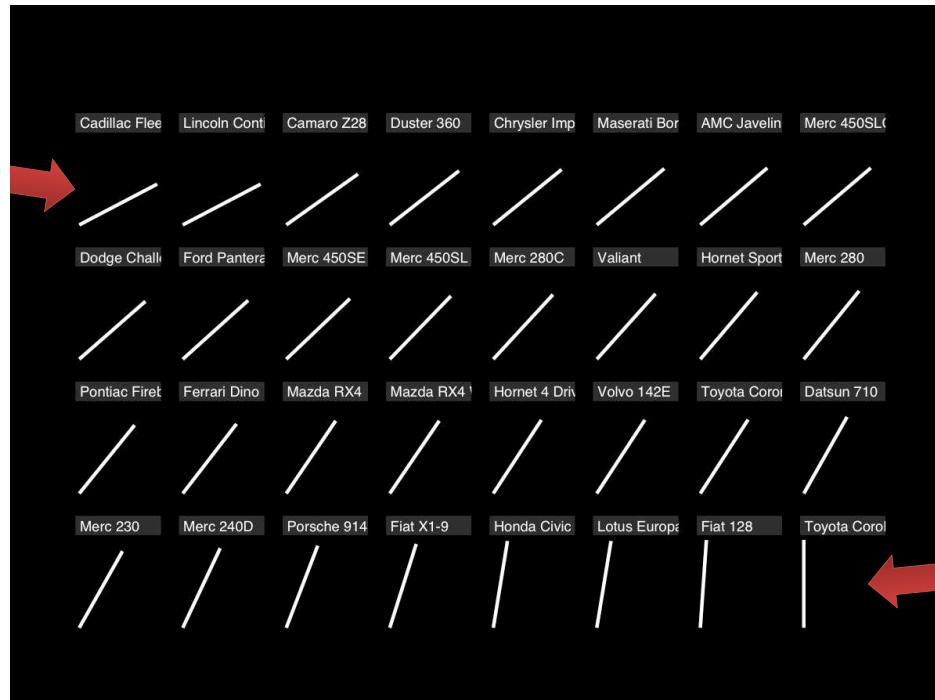




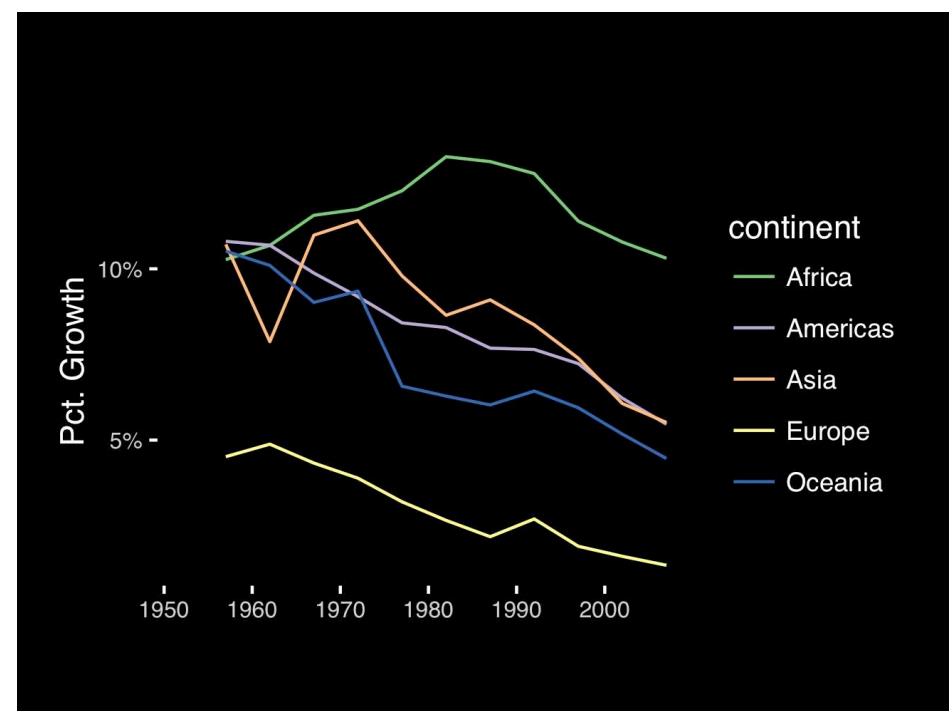
The most important measurement should exploit the highest ranked encoding possible.

- 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?

Among Republicans

Among Democrats

Donald Trump 47%
Hillary Clinton 53%

Among Republicans

Among Democrats

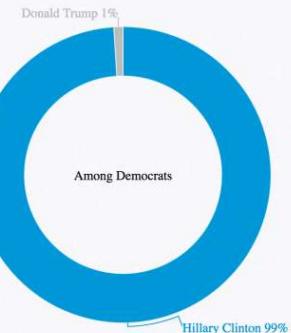
Share

POLITICO

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

Among Republicans

Among Democrats



Share

POLITICO

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.**

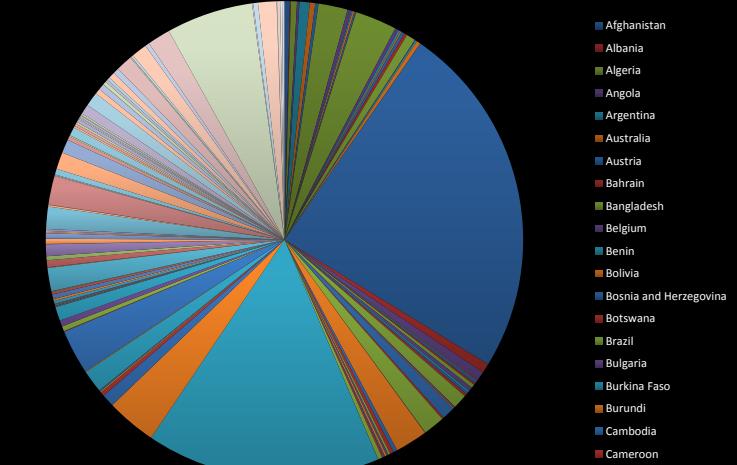
-Edward Tufte, *The Visual Display of Quantitative Information*

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.

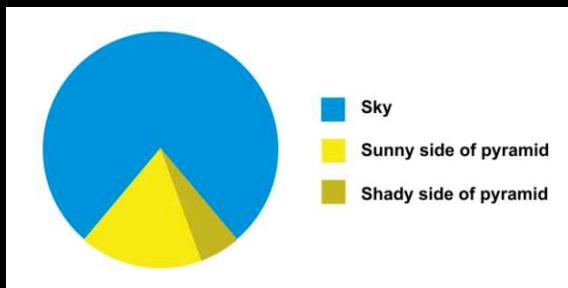
-Edward Tufte, *The Visual Display of Quantitative Information*

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

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

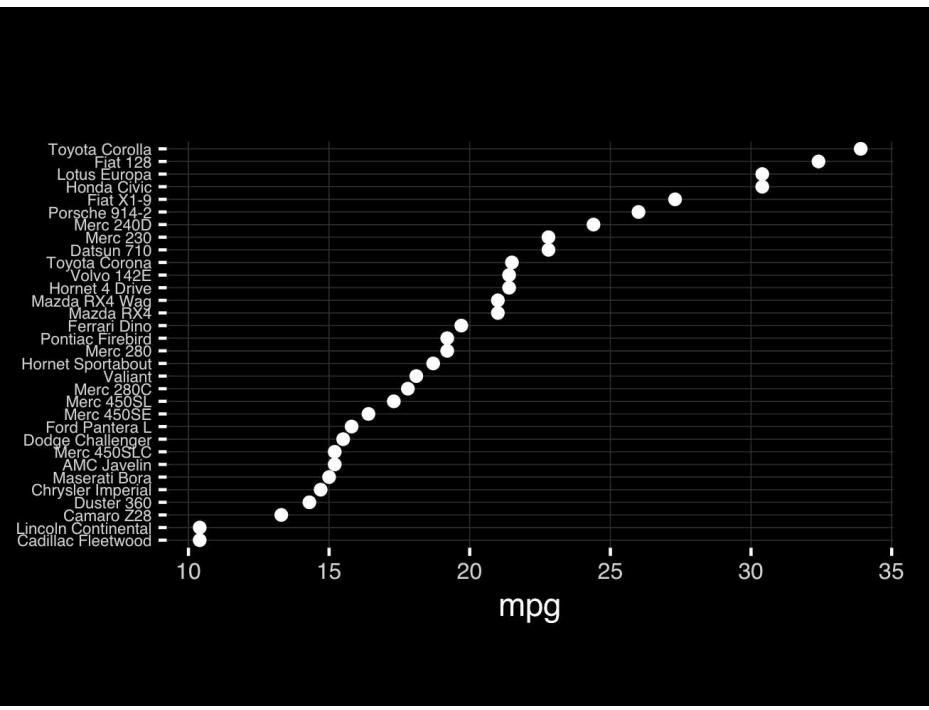
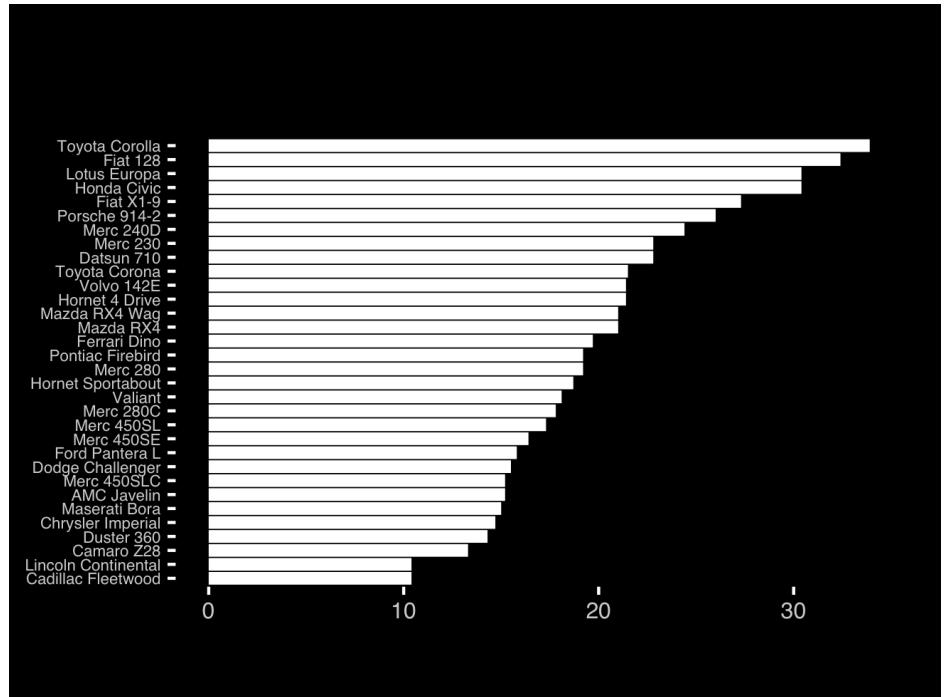
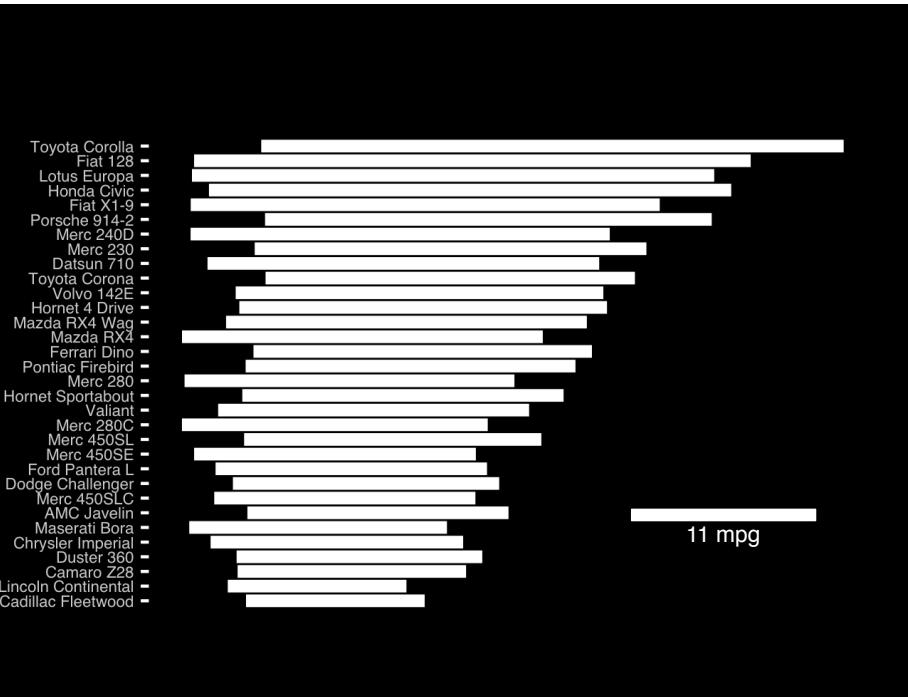


All good pie charts are jokes...



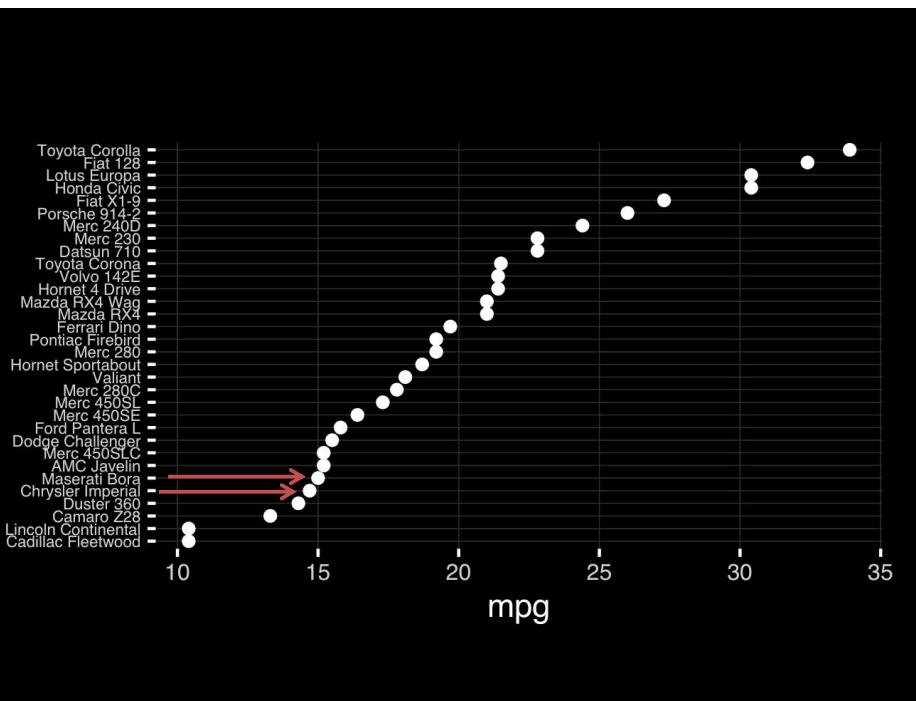
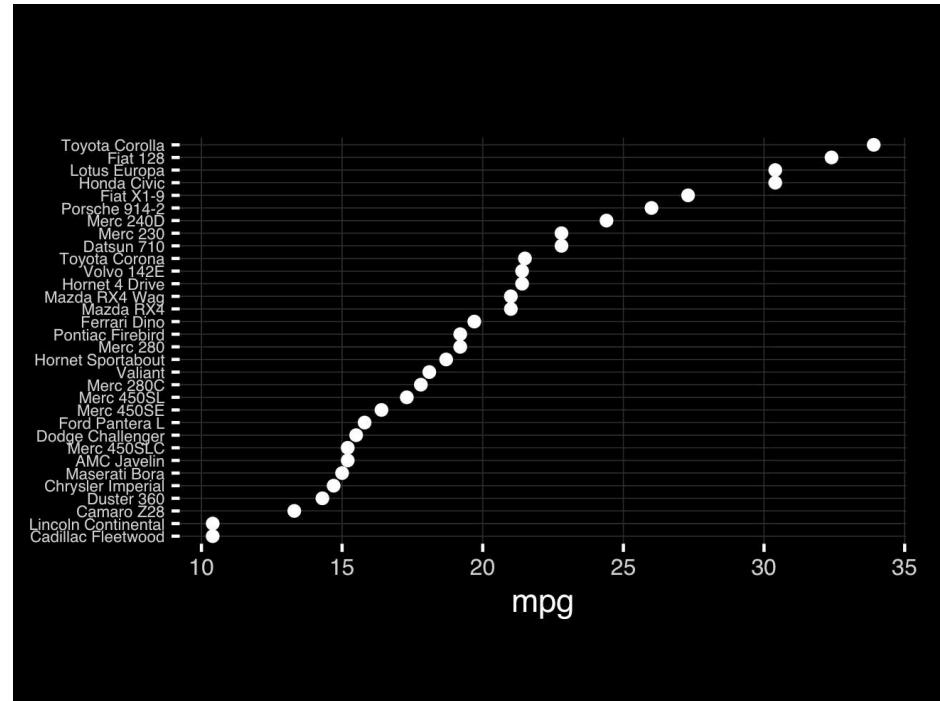
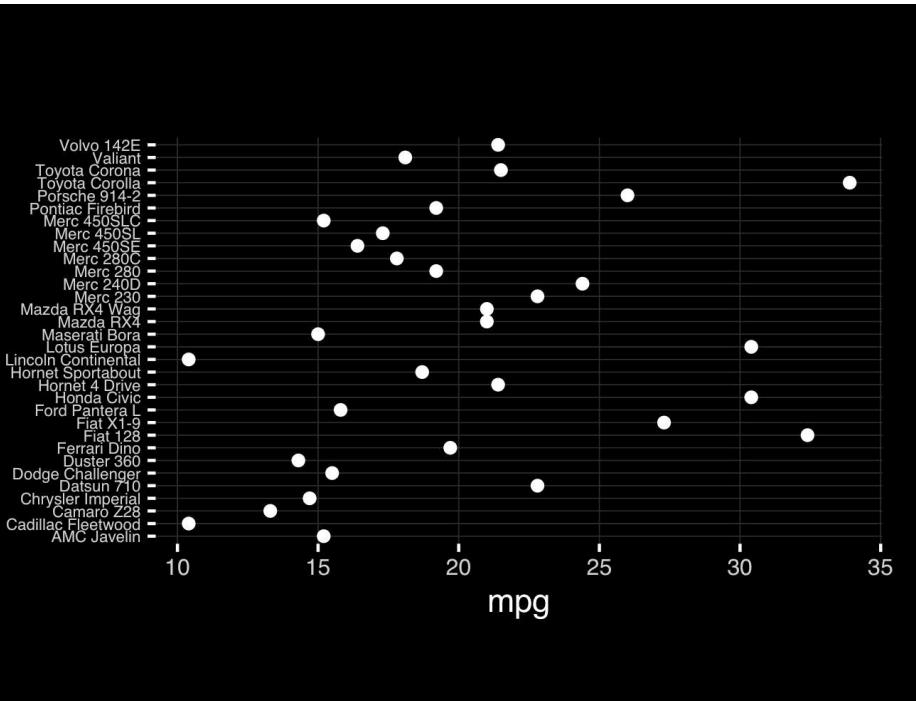
The most important measurement should exploit the highest ranked encoding possible.

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

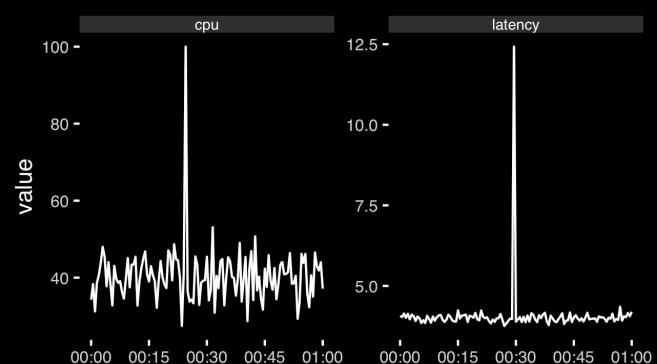


The most important measurement should exploit the highest ranked encoding possible.

- 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**.

Hands-on

Section 1 only please

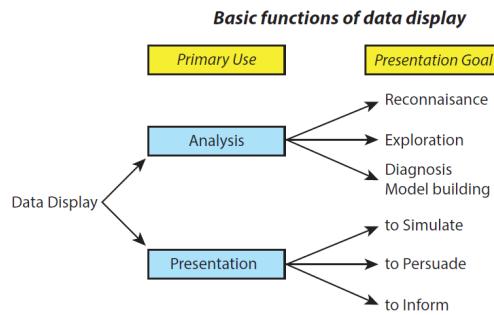
Do it Yourself!

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



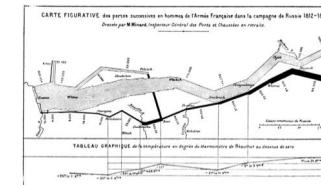
Exploratory (for you!)

Info for others,
publications &
sharing etc.

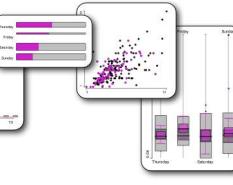
Different graphs for different purposes

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

Presentation graphs: single image for a large audience



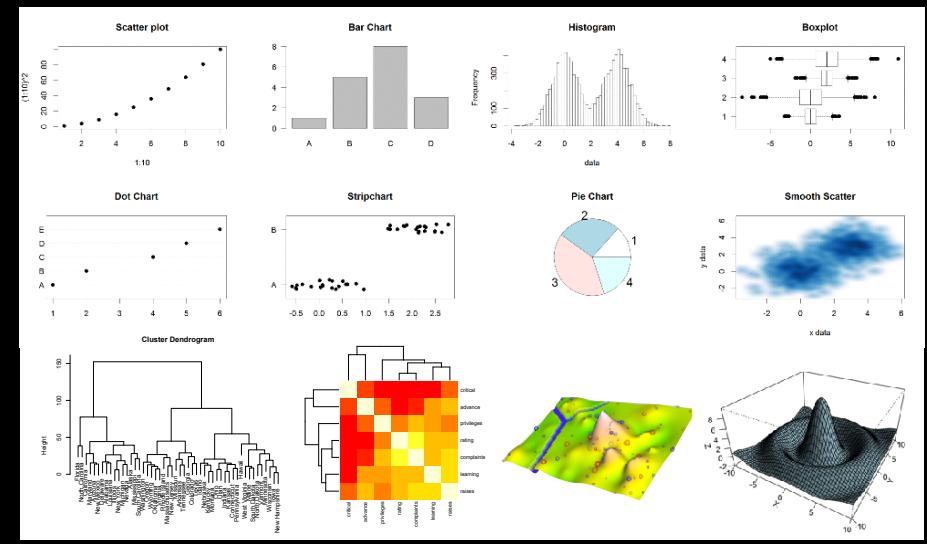
Presentation



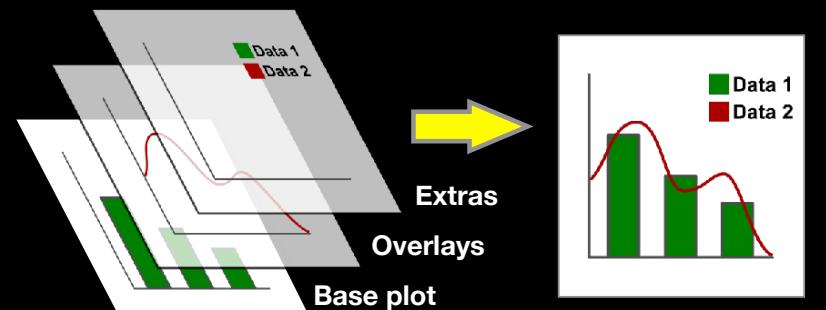
Exploration

17

Core R Graph Types

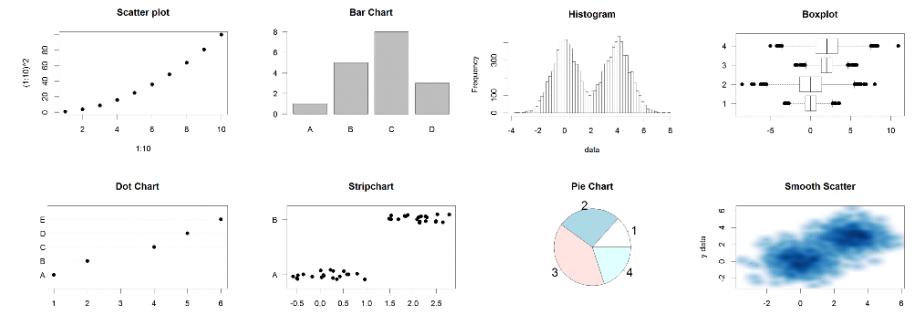


The R Painters Model



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

Core Graph Types



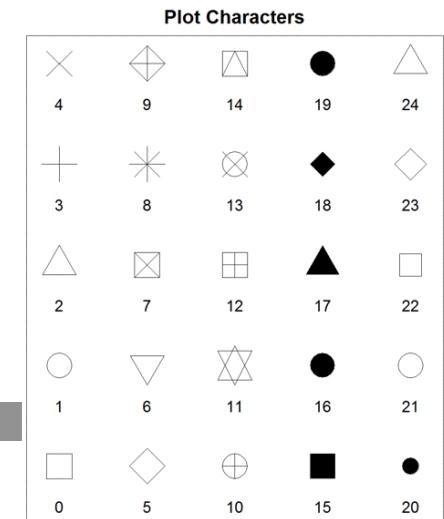
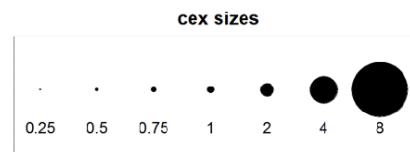
- Local options to change a specific plot
- Global options to affect all graphs

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

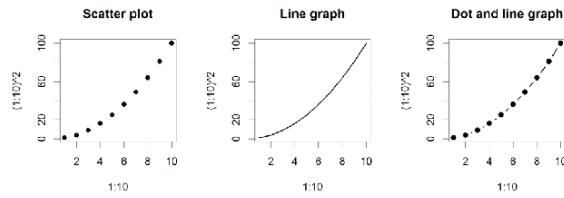
Plot Characters



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

Plot Type Specific Options

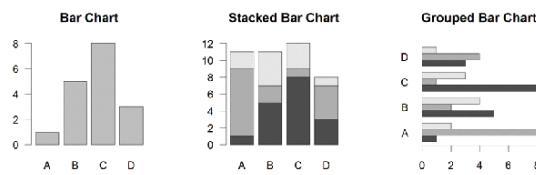
Plot (scatterplots and line graphs)



- 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 )
```

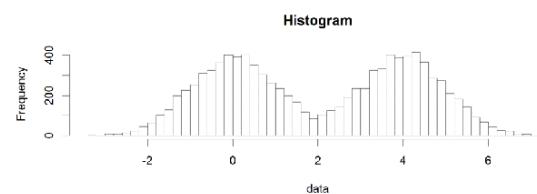
Barplot (bar graphs)



- 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)
```

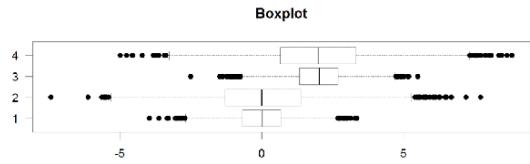
Hist (histograms)



- 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
```

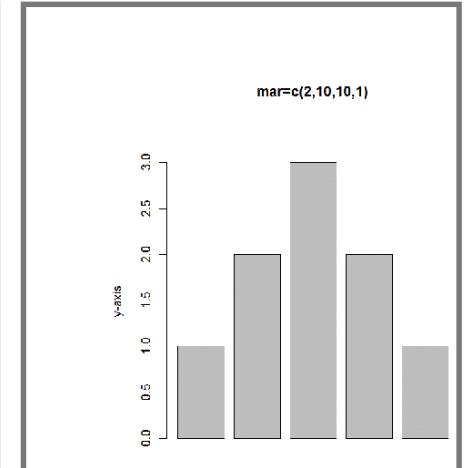
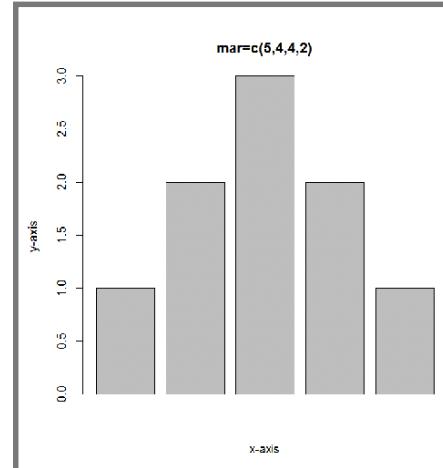
Par examples

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

Par options

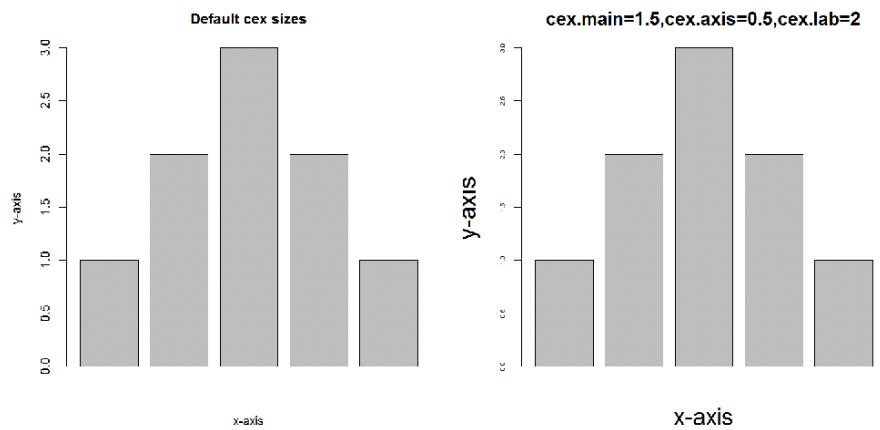
- Margins
 - mai (set margins in inches)
 - mar (set margins in number of lines)
 - mex (set lines per inch)
 - 4 element vector (bottom, left, top, right)
- Warning
 - Error in plot.new() : figure margins too large

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



Par options

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



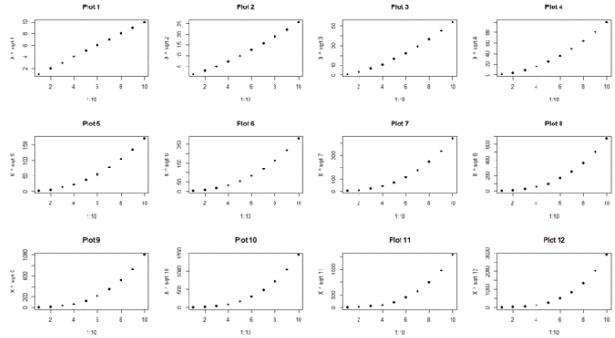
```
par( cex.main=1.5, cex.axis=0.5, cex.lab=2 )
```

Do it Yourself!

Par options

- Multi-panel

– `par(mfrow(rows,cols))`



`par(mfrow(3 , 4))`

Using Color

Hands-on Section 2 only please

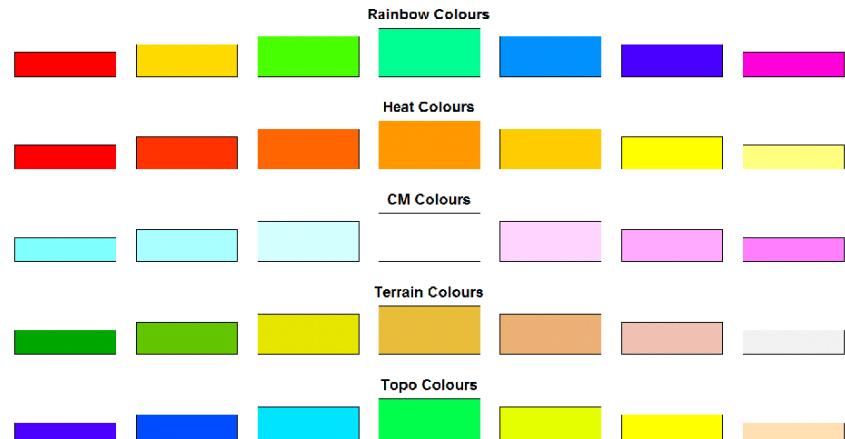
Specifying colors

- 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( 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**](#)