

# Lecture 2 - Data and Data Summaries

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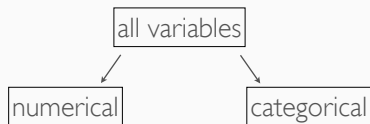
Sta102 / BME 102

January 20, 2016

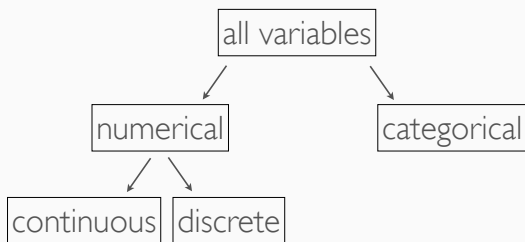
Colin Rundel

**Data**

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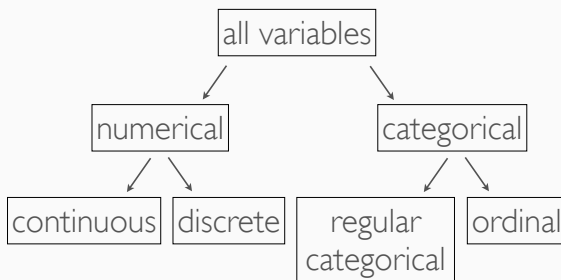


- *Numerical (quantitative)* - takes on a numerical values
  - Ask yourself - is it sensible to add, subtract, or calculate an average of these values?
- *Categorical (qualitative)* - takes on one of a set of distinct categories
  - Ask yourself - are there only certain values (or categories) possible? Even if the categories can be identified with numbers, check if it would be sensible to do arithmetic operations with these values.



- *Continuous* - data that is measured, any numerical (decimal) value
- *Discrete* - data that is counted, only whole non-negative numbers

# Categorical Data



- *Ordinal* - data where the categories have a natural order
- *Regular categorical* - categories do *not* have a natural order

## Example - Class Survey

Students in an introductory statistics course were asked the following questions as part of a class survey:

1. Are you introverted or extraverted?
2. On average, how much sleep do you get per night?
3. When do you go to bed: 8pm-10pm, 10pm-12am, 12am-2am, later than 2am?
4. How many countries have you visited?
5. On a scale of 1 (very little) - 5 (a lot), how much do you dread this semester?

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What type of data is each variable?

## Representing Data - Class Survey

We use a *data matrix (data frame)* to represent responses from this survey.

- Columns represent *variables*
- Rows represent *observations (cases)*

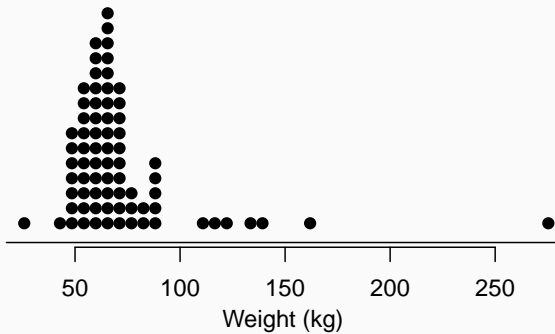
student	gender	intro_extra	sleep	bedtime	countries	dread
1	male	extravert	8	10-12	13	3
2	female	extravert	8	8-10	7	2
3	female	introvert	5	12-2	1	4
4	female	extravert	6.5	12-2	0	2
⋮	⋮	⋮	⋮	⋮	⋮	⋮
86	male	extravert	7	12-2	5	3



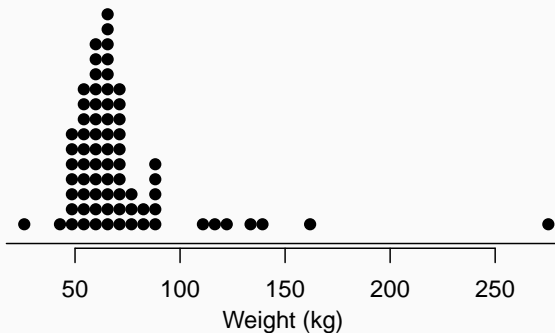
## Numerical data

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# Dot plot

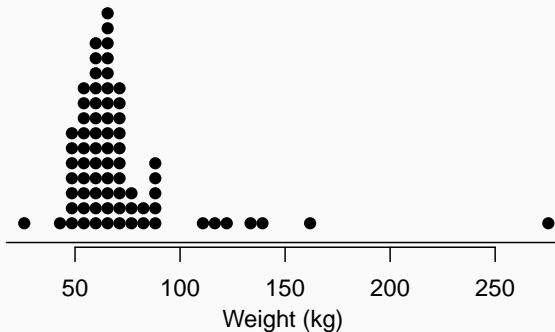


# Dot plot



Useful for visualizing the *distribution* of a numerical variable, especially when individual values are of interest.

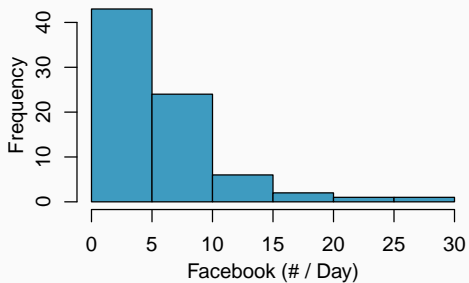
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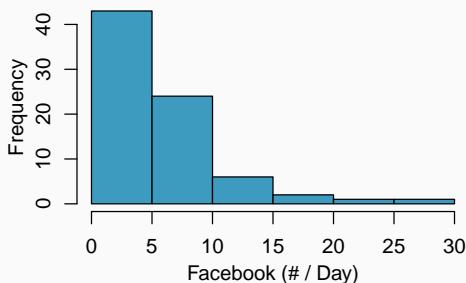
Useful for visualizing the *distribution* of a numerical variable, especially when individual values are of interest.

Do you see anything out of the ordinary?

# Histograms



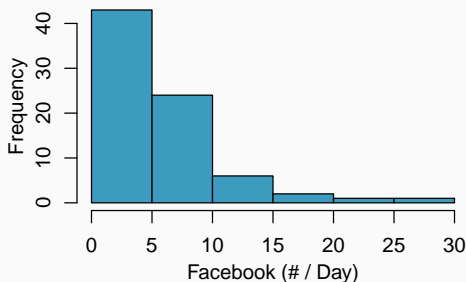
# Histograms



Organizes numeric data into discrete bins, height of each bin is the count or fraction of observations falling in that bin.

Preferable when data is large (but hides finer details)

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Any possible issues? (Hint: is there anything subjective about the plot?)

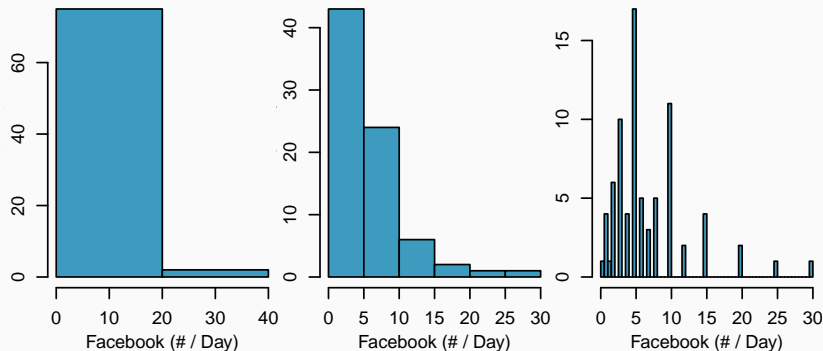
## Histogram bin width

The choice of *bin width* is important and can alter the story the histogram is telling.



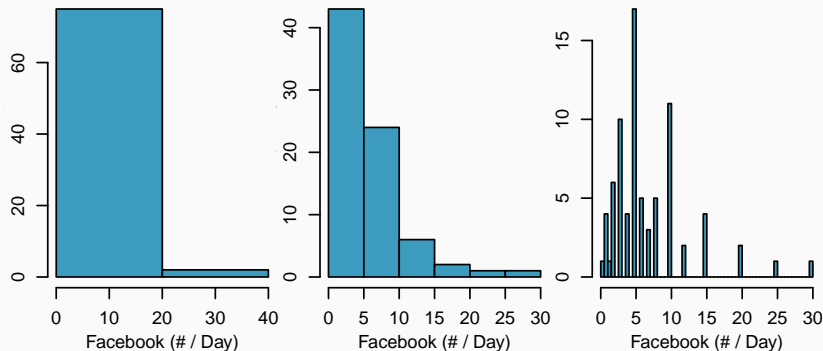
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Which histogram is the most useful? Why?

## Describing Distributions

Dot plots and histograms are approaches for visually approximating the *distribution* of numerical data.

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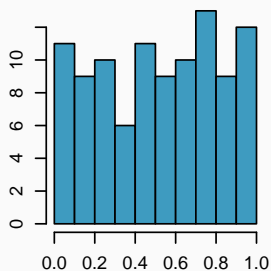
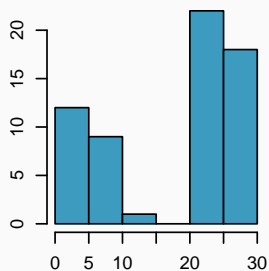
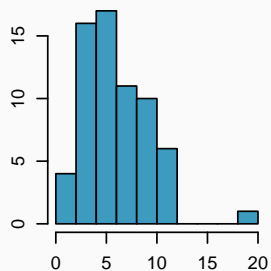
Dot plots and histograms are approaches for visually approximating the *distribution* of numerical data.

We would like to have a standardized way of describing distributions, there are several critical features that we focus on:

- Center
- Spread
- Modality (peaks)
- Skewness (asymmetry)
- Kurtosis (peakedness / tail thickness)

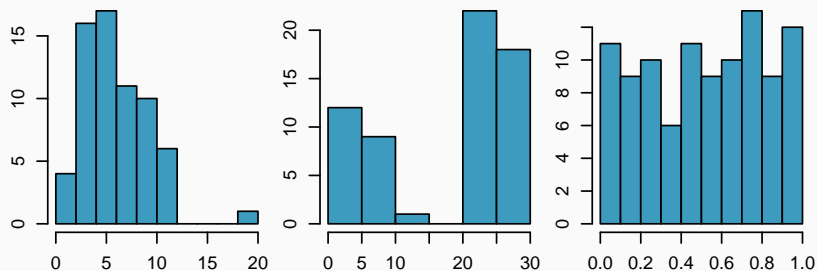
# Modality

This describes the pattern of the peaks in peaks in the histogram - a single prominent peak (*unimodal*), several (*bimodal/multimodal*), or no prominent peaks (*uniform*)?



# Modality

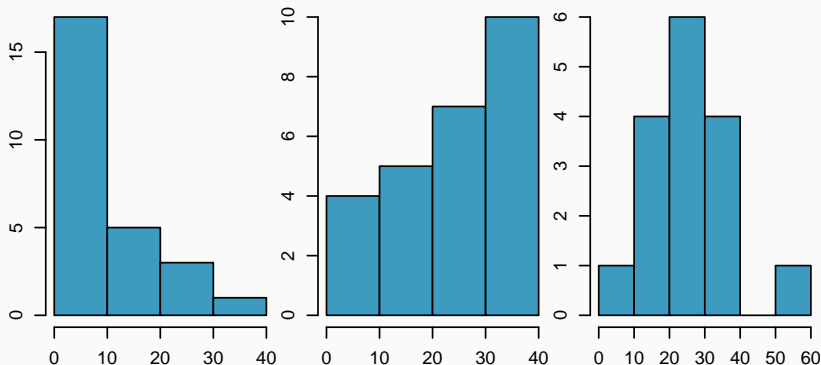
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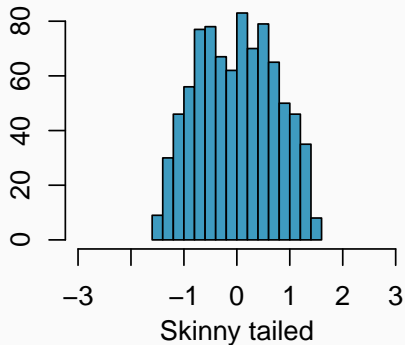
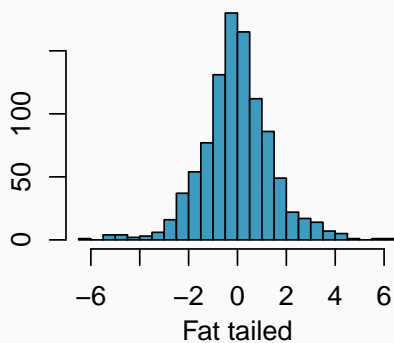


In order to determine modality, it's best to step back and imagine a smooth curve (limp spaghetti) over the histogram.

# Skewness

Histograms are said to be skewed towards the direction with the longer tail. A histogram can be *right skewed*, *left skewed*, or *symmetric*.





Later on we'll talk more about how to better judge this characteristic.



## Thinking about distributions

How are of the following variables likely to be distributed?

1. weights of adult males
2. salaries of a random sample of workers in North Carolina
3. exam scores in Sta 102?
4. birthdays of classmates (day of the month)
5. weights of adults

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5. weights of adults - *Unimodal, Right skewed*

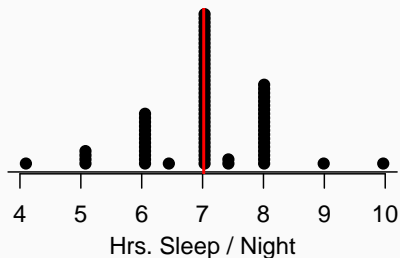
## Guess the center

What would you guess is the average number of hours students sleep per night?



## Guess the center

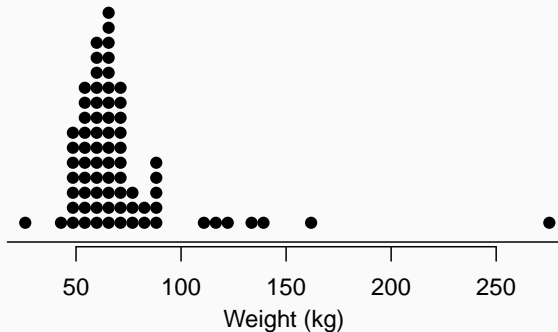
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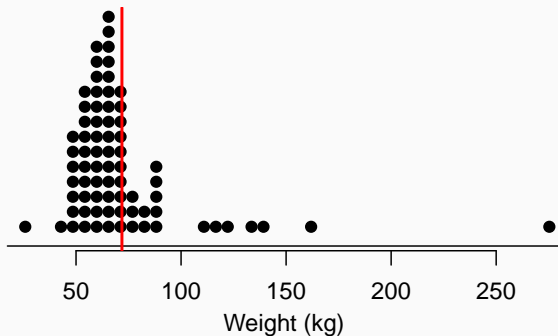
## Guess the center, cont.

What would you guess is the average weight of students?



## Guess the center, cont.

What would you guess is the average weight of students?



- *Sample mean* ( $\bar{x}$ ) - Arithmetic average of values in sample.

$$\bar{x} = \frac{1}{n} (x_1 + x_2 + x_3 + \cdots + x_n) = \frac{1}{n} \sum_{i=1}^n x_i$$

- *Population mean* ( $\mu$ ) - Computed the same way but it is often not possible to calculate  $\mu$  since population data is rarely available.

$$\mu = \frac{1}{N} (x_1 + x_2 + x_3 + \cdots + x_N) = \frac{1}{N} \sum_{i=1}^N x_i$$

- The sample mean is a *sample statistics*, or a *point estimate* of the population mean. This estimate may not be perfect, but if the sample is good (representative of the population) it is usually a good guess.

# Variance

- *Sample Variance* - Average\* deviance from the sample mean.

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

- *Population Variance* - Average deviance from the population mean.

$$\sigma^2 = \frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2$$

- The sample variance is a *sample statistics*, or a *point estimate* of the population variance.
- Variance is a measure of spread, the wider the distribution the larger the variance will be

## Square Deviance?

Why do we use the squared deviation in the calculation of variance?

# Square Deviance?

Why do we use the squared deviation in the calculation of variance?

- *To get rid of negatives so that observations equally distant from the mean are weighed equally*
- *To weigh larger deviations more heavily*

# Standard deviation

- *Sample Standard Deviation*

$$s = \sqrt{s^2} = \sqrt{\frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2}$$

- *Population SD*

$$\sigma = \sqrt{\sigma^2} = \sqrt{\frac{1}{N} \sum_{i=1}^N (x_i - \mu)^2}$$

- Just like Variance, SD is a measure of spread, the wider the distribution the larger the SD will be.
- We often prefer SD to Variance because it has a more natural interpretation. Variance is measured in square units while the SD is in same units as the observed data.

# Diversity vs Variability

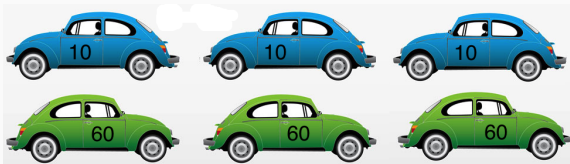
Which group of cars has a more diverse set of colors?





## Diversity vs Variability (cont.)

Which group of cars has a more variable mileage?



## Diversity vs Variability (cont.)

Group 1:

$$\bar{x} = (10 + 20 + 30 + 40 + 50 + 60)/6 = 35$$

$$s^2 = \frac{1}{6-1} ((10-35)^2 + (20-35)^2 + \dots + (60-35)^2) = 350$$

Group 2:

$$\bar{x} = (10 + 10 + 10 + 60 + 60 + 60)/6 = 35$$

$$s^2 = \frac{1}{6-1} ((10-35)^2 + (10-35)^2 + \dots + (60-35)^2) = 750$$

## Median, Quartiles, and IQR

- The *median* is the value that splits the data in half when ordered in ascending order, i.e. *50<sup>th</sup> percentile*.

0, 1, **2**, 3, 4

If there are an even number of observations, then the median is the average of the two values in the middle.

$$0, 1, \underline{2, 3}, 4, 5 \rightarrow \frac{2 + 3}{2} = 2.5$$

- The 25<sup>th</sup> percentile is called the first quartile, *Q1*.
- The 75<sup>th</sup> percentile is called the third quartile, *Q3*.
- The range spanned by the middle 50% of the is the *interquartile range*, or the *IQR*.

The median and IQR are examples of what are known as robust statistics - because they are less affected by skewness and outliers than statistics like mean and SD.

As such:

- for skewed distributions it is more appropriate to use median and IQR to describe the center and spread
- for symmetric distributions it is more appropriate to use the mean and SD to describe the center and spread

## Robust statistics

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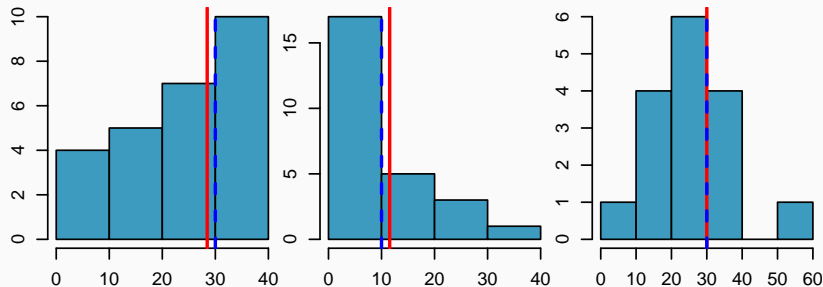
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If you were searching for a house and are price conscious, should you be more interested in the mean or median house price when considering a particular neighborhood?

# Mean vs. median

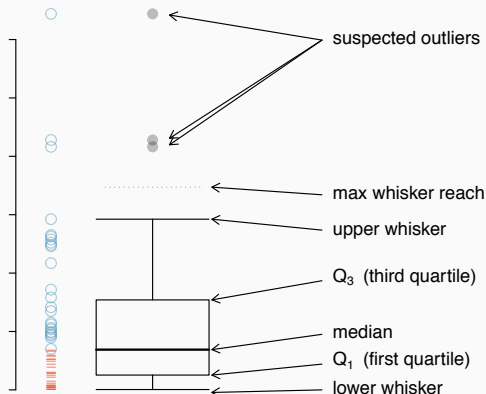
- If the distribution is symmetric, center is the mean
  - Symmetric: mean = median
- If the distribution is skewed or has outliers center is the median
  - Right-skewed: mean  $>$  median
  - Left-skewed: mean  $<$  median



red solid - mean, black dashed - median

# Box plot

A *box plot* visualizes the median, the quartiles, and suspected outliers.



## Box plot - Example

Resting Pulses:

(62, 64, 68, 70, 70, 74, 74, 76, 76, 78, 78, 80)

Steps:

1. Calculate median, Q1, Q3, IQR, min, and max
2. Calculate upper and lower fences ( $Q1 - 1.5 \text{ IQR}$ ,  $Q3 + 1.5 \text{ IQR}$ )
3. Find the location of the upper and lower whiskers
4. Consider data points outside whiskers as potential outliers



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Steps:

1. Calculate median, Q1, Q3, IQR, min, and max

- $med = 74$
- $Q1 = 69$
- $Q2 = 77$
- $IQR = 77 - 69 = 8$
- $min = 62$
- $max = 80$

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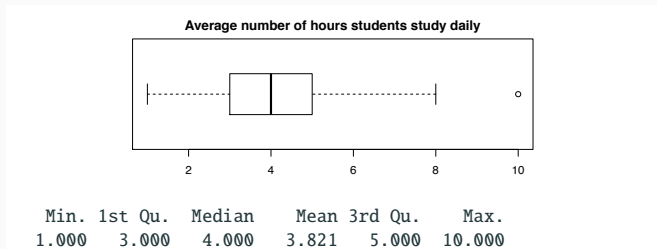
- $F_L = Q1 - 1.5 IQR = 69 - 12 = 57$
- $F_U = Q3 + 1.5 IQR = 77 + 12 = 89$

3. Find the location of the upper and lower whiskers

- $W_L = 62$

## Reading a boxplot

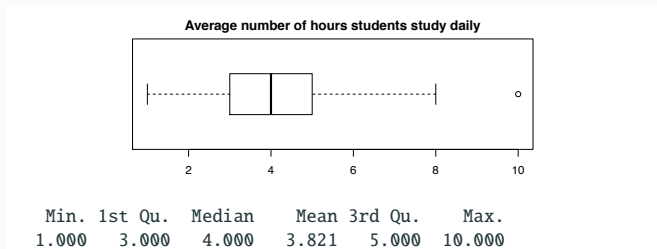
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- a) There are no students who don't study at all.
- b) The IQR is 2 hours.
- c) 75% of the students study  $\geq 5$  hours daily, on average.
- d) 25% of the students study  $\leq 3$  hours, on average.

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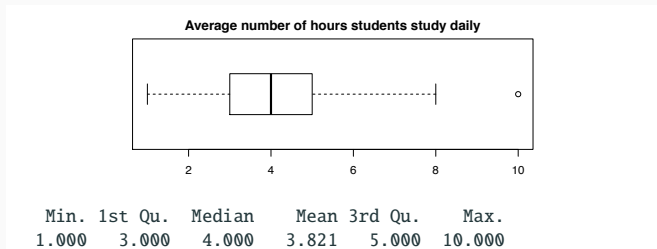
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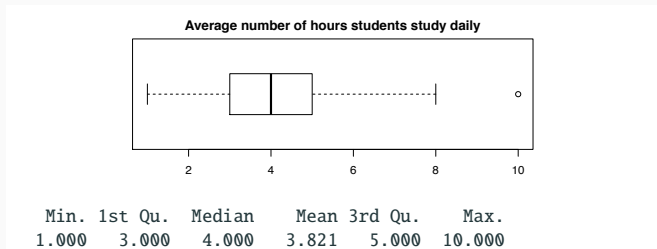
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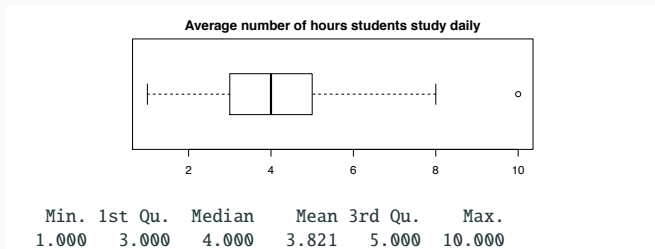
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# Categorical data

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## Tables and Contingency tables

For a single categorical variable we can always summarize it by showing the absolute # of counts for each category.

Belief in a higher power:

No	Somewhat	Yes
22	23	36

Gender:

Female	Male
57	25

Similarly we often report the counts as a relative proportion of the total.

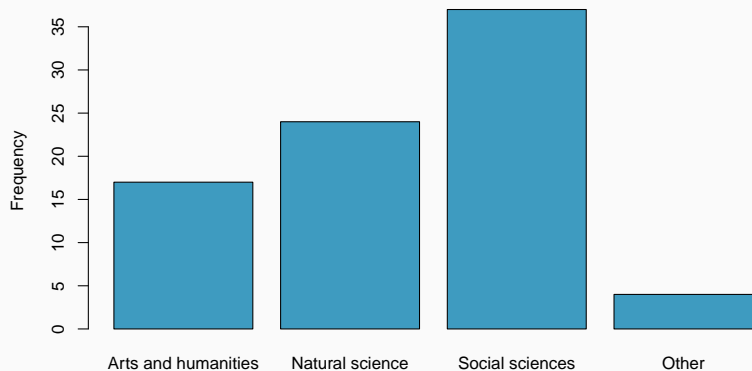
Belief in a higher power:

No	Somewhat	Yes
0.272	0.284	0.444

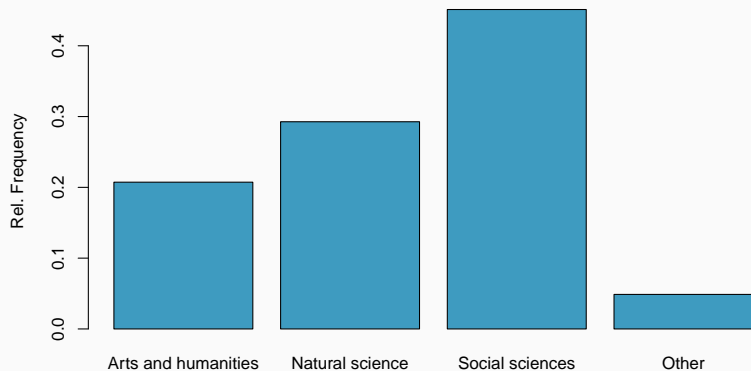
Gender:

Female	Male
0.695	0.305

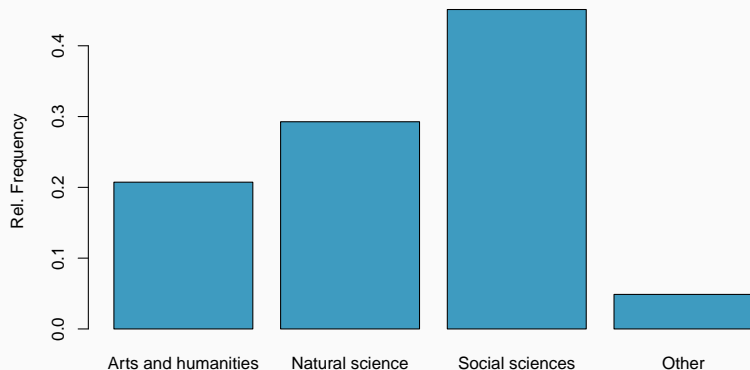
## Barplots - Absolute (Counts) vs Relative (Proportions)



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Why is this different from a histogram?

# Bivariate Relationship Plots

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# Bivariate Relationships

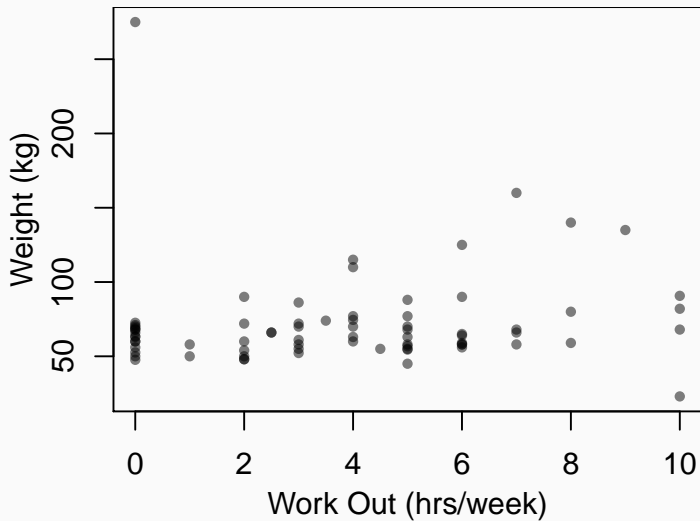
We are often interested in exploring the relationship between two different variables. The appropriate visualization will depend on the type of variables being plotted and their intrinsic relationship.

Almost always we will want:

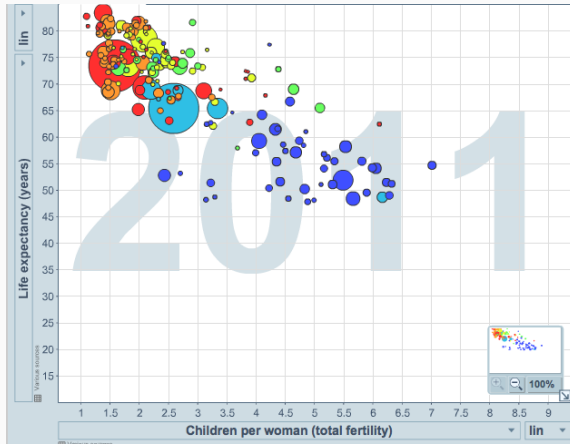
- X-axis - the independent variable
- Y-axis - the dependent variable

To decide which is which, think about cause and effect - the dependent variable (X) is “causing” the change in independent variable (Y).

# Scatterplots



# Multivariate Scatterplots



<http://www.gapminder.org/world>



## Contingency tables

If we are interested in looking at a relationship between two categorical variables we construct a contingency table (cross tabulation).

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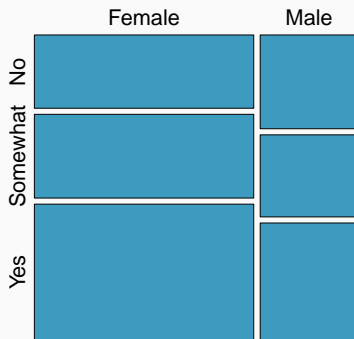
Female	Male
57	25

	Female	Male
No	14	8
Somewhat	16	7
Yes	26	10

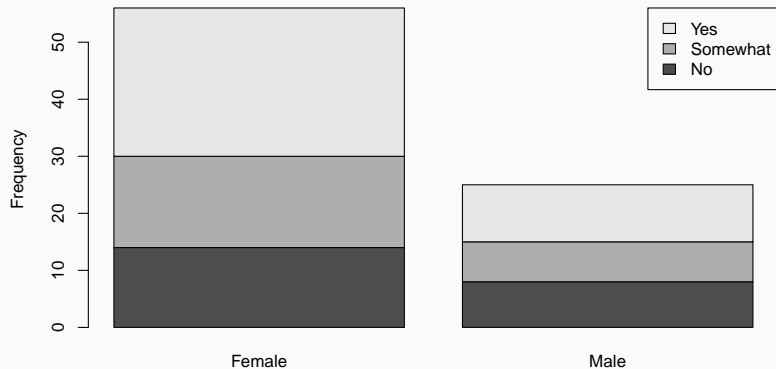
## Mosaic plots

Is there a relationship between major and relationship status?

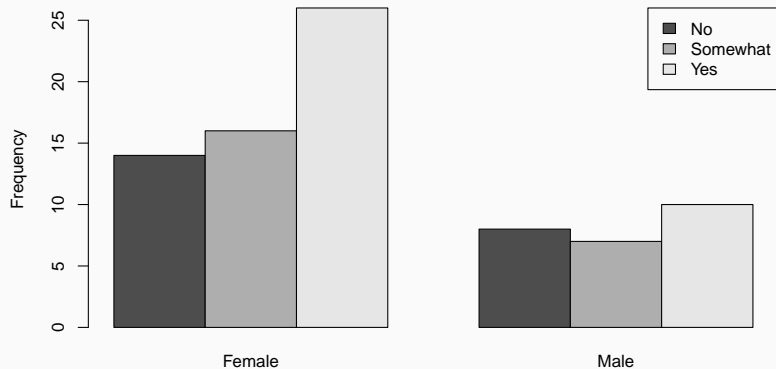
	Female	Male
No	14	8
Somewhat	16	7
Yes	26	10



# Bivariate Barplots - Stacked vs Juxtaposed

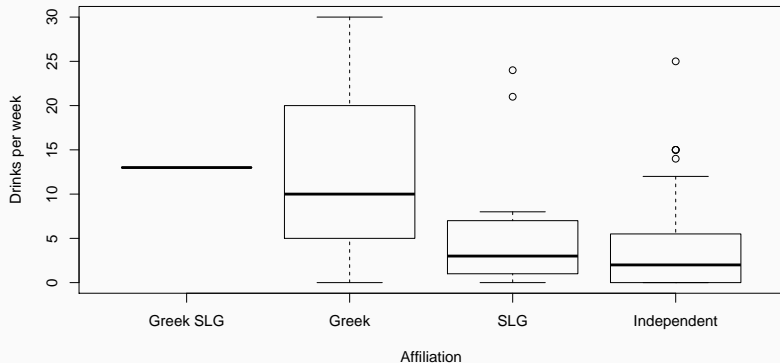


## Bivariate Barplots - Stacked vs Juxtaposed



## Side-by-side box plots

How does number of drinks consumed per week vary by affiliation?



# Summary

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## Visualization Summary

- Single numeric - dot plot, *box plot*, *histogram*
- Single categorical - bar plot (or a *table*)
- Two numeric - *scatter plot*
- Two categorical - *mosaic plot*, stacked or side-by-side bar plot
- Numeric and categorical - *side-by-side box plot*



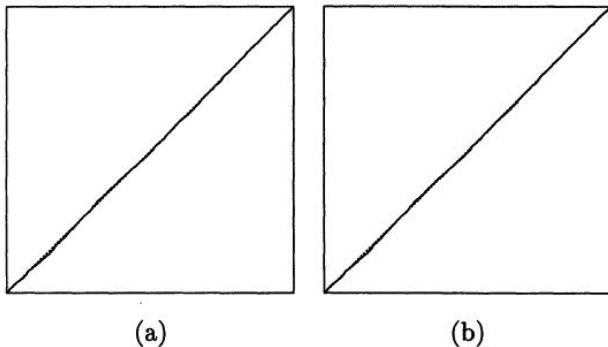
# Visualization Summary

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## Tufte's Principles:

1. Above all else show data.
2. Maximize the data-ink ratio.
3. Erase non-data-ink.
4. Erase redundant data-ink.
5. Revise and edit

## Karl Broman's Top Ten Worst Graphs

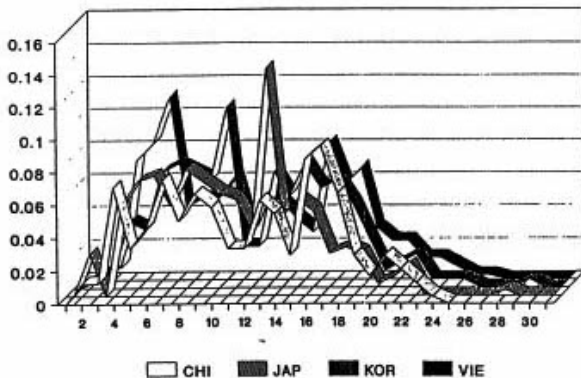


*Figure 1. SRQ Plots of  $T_i/T_n$  (Vertical Axes) Against  $i/n$  (Horizontal Axes) for the Gibbs Sampler (a) and an Alternating Gibbs/Independence Sampler (b) for the Pump Failure Data Based on Runs of Length 5,000. Lines through the origin with unit slope are shown dashed; axis ranges are from 0 to 1 for all axes.*

# Karl Broman's Top Ten Worst Graphs

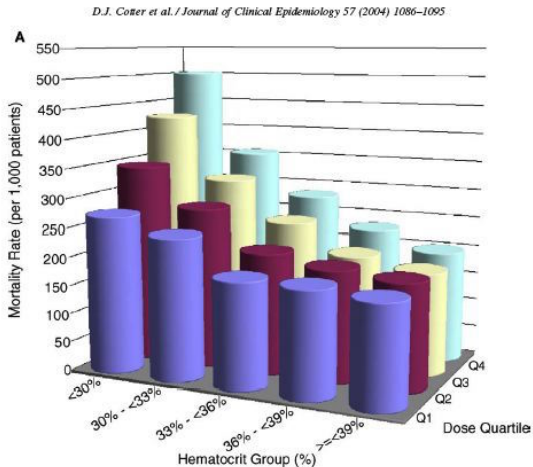
**B**

**BINNED FREQUENCY DATA - D10S28**  
CHINESE, JAPANESE, KOREAN, VIETNAMESE



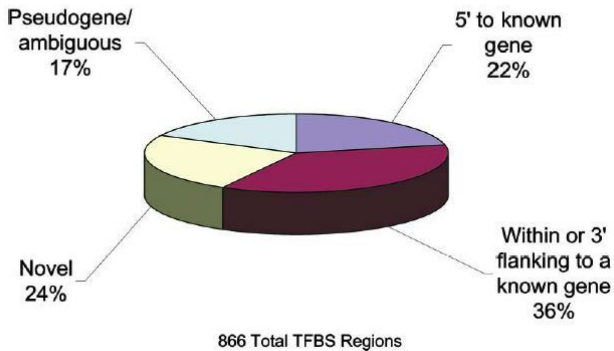
[https://www.biostat.wisc.edu/~kbroman/topten\\_worstgraphs/](https://www.biostat.wisc.edu/~kbroman/topten_worstgraphs/)

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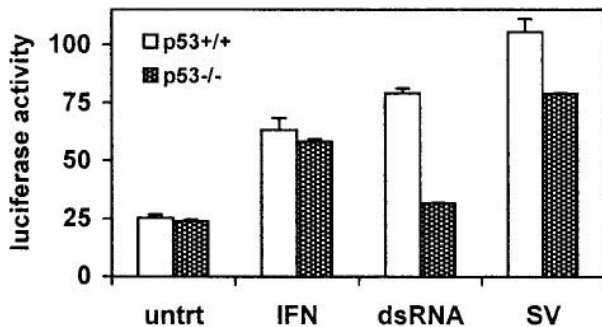
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## Distribution of All TFBS Regions



[https://www.biostat.wisc.edu/~kbroman/topten\\_worstgraphs/](https://www.biostat.wisc.edu/~kbroman/topten_worstgraphs/)

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