

# Fundamentals of Statistics for Language Sciences LT2206



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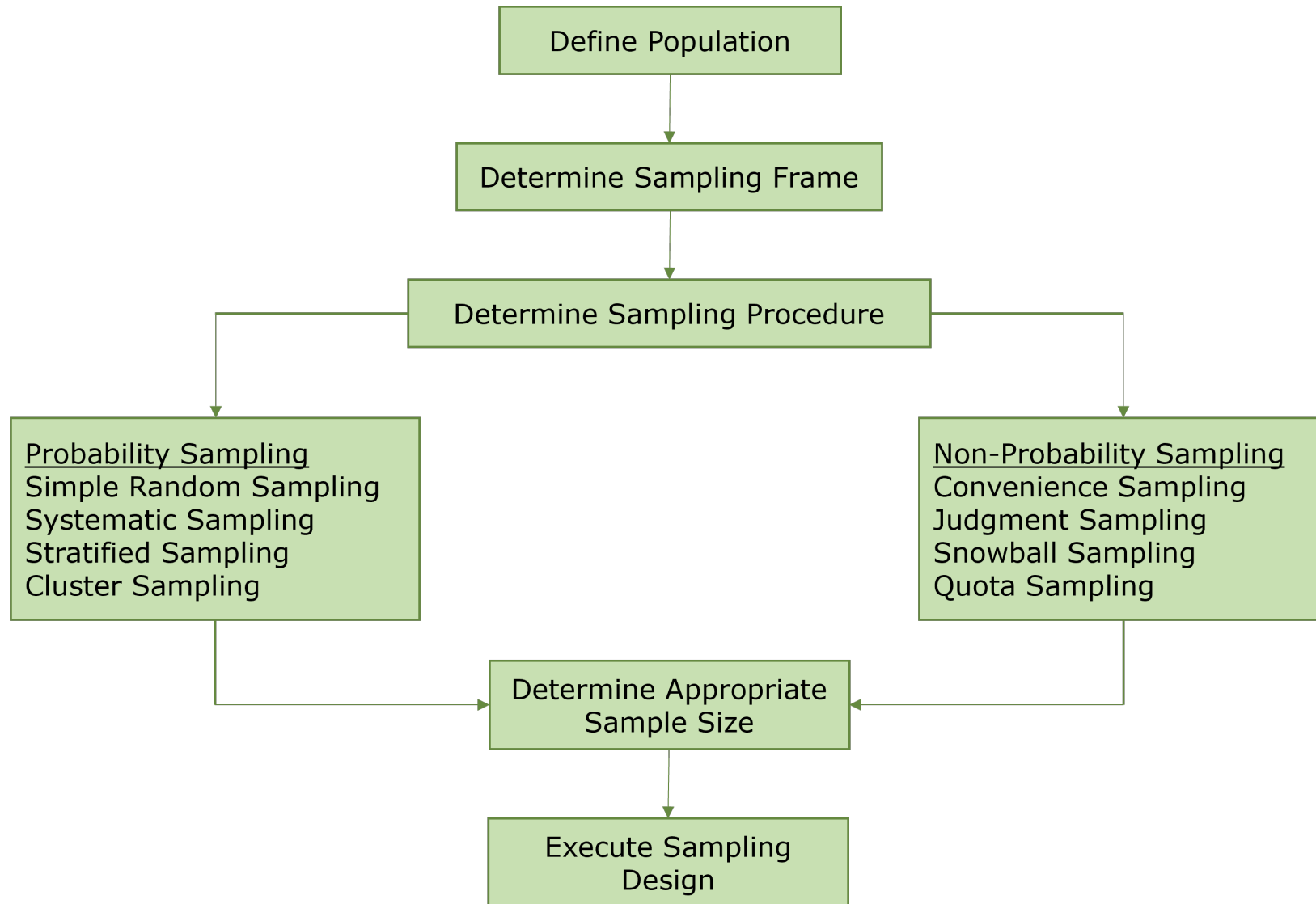
Lecture 3: Descriptive statistics

Slides adapted from Cecilia Earls

# Lecture plan

- Review on sampling
- Descriptive statistics
- Short break (15 mins)
- Hands-on exercises

# Sampling design process



# Descriptive statistics

**Basic goal:** Understanding your data.

- Summary & description
- Look for peculiarities (unusual data values)

Should always be the first step in any data analysis!

# Example: Survivorship on the Titanic

**Goal:** Describe survival patterns for the ill-fated passengers of the Titanic.

The Titanic dataset:  
**n** = 891 passengers

## Variables

- age (in years)
- gender (male, female)
- class (1,2,or 3)
- survived (yes=1, no=0)



# Data types

**Qualitative variables:** categorical; vary in “level”, but lack specific units of measure

- **Nominal:** survived (yes/no), gender (male/female)
- **Ordinal:** passenger class (first, second, or third)

**Quantitative variables:** numerical; vary in magnitude with specific units of measure

- Passenger age (in years)

# Graphical methods

Visually summarize the data to gain an understanding of the composition of the data (in this case, the Titanic passengers)

## **Categorical variables:**

- pie charts
- bar charts

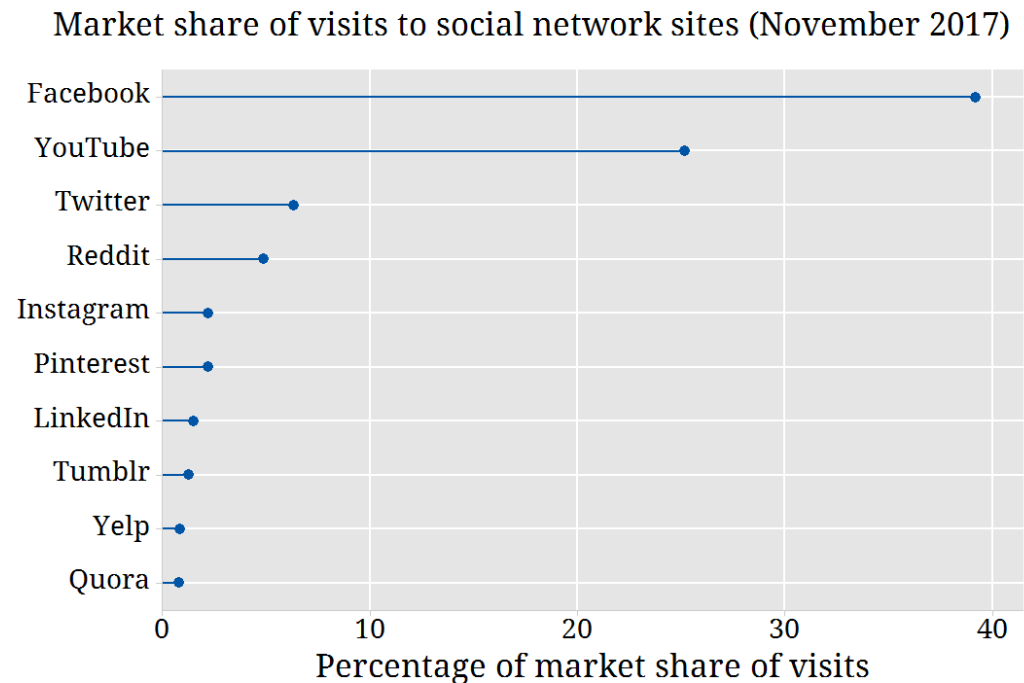
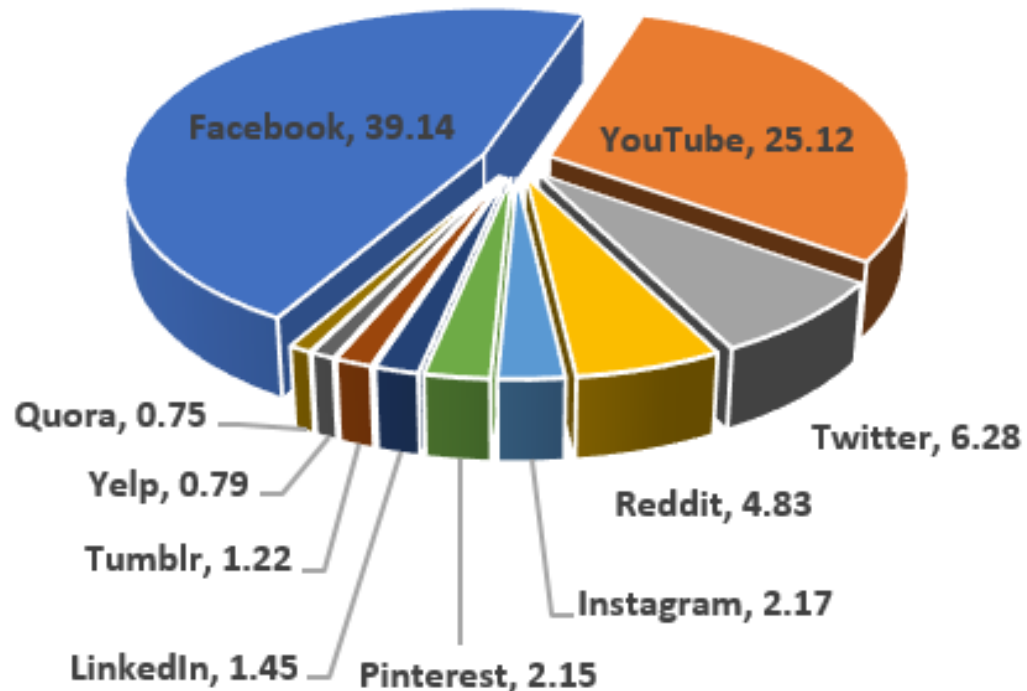
## **Quantitative variables:**

- histograms
- box plots

# Categorical variables: Pie charts

**Pie charts along with other area-based charts (e.g. donut chart) are not recommended!**

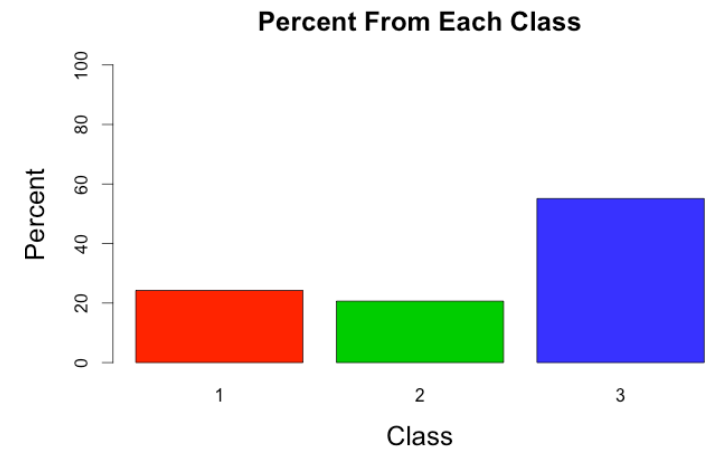
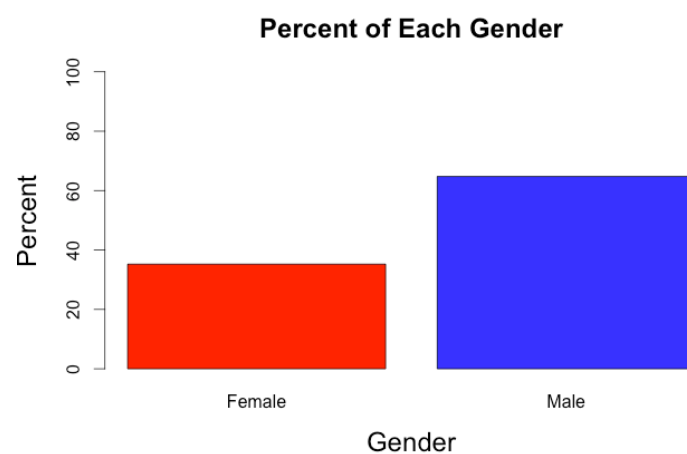
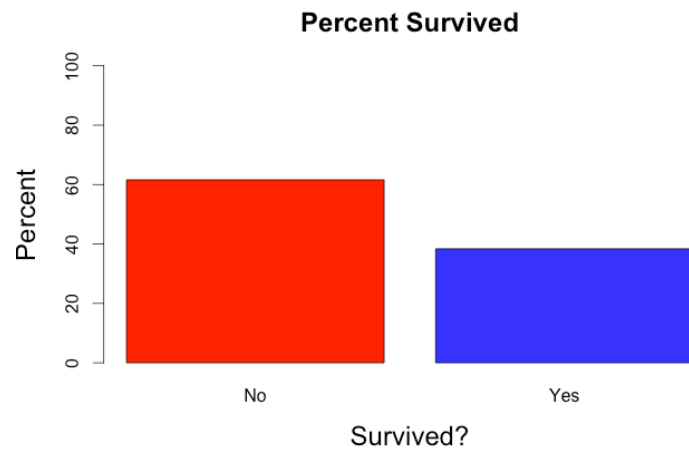
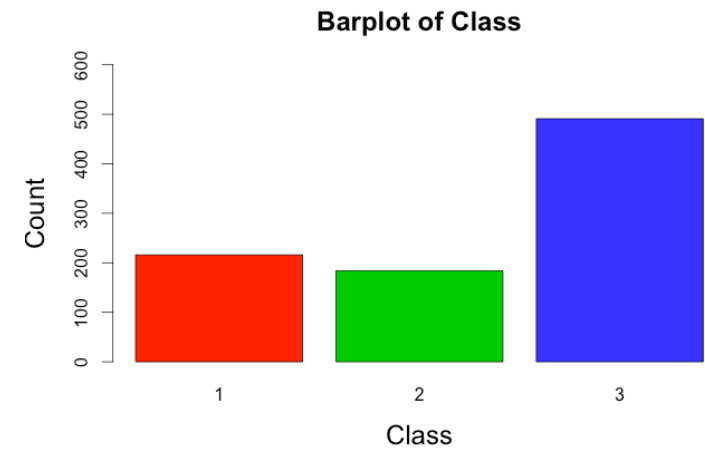
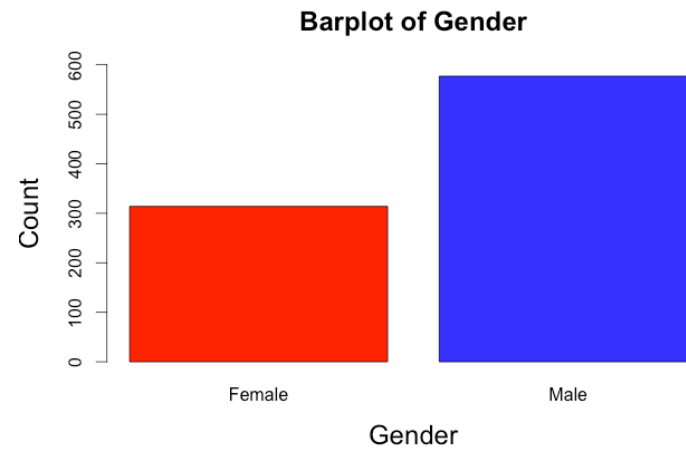
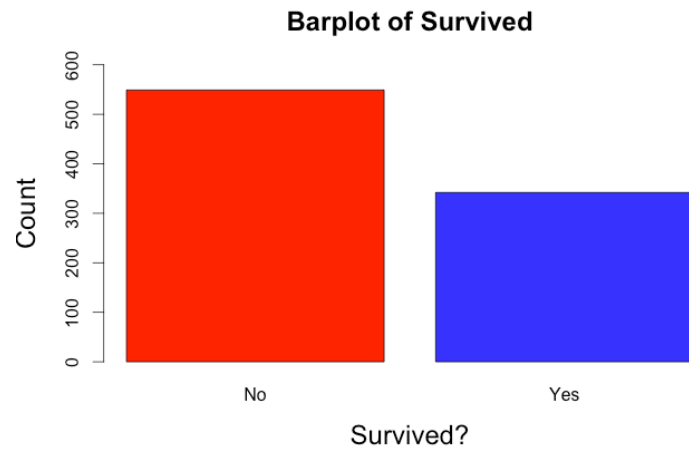
- Difficult to decode the information in the data.
- Completely defeats the purpose of including a chart instead of a table.





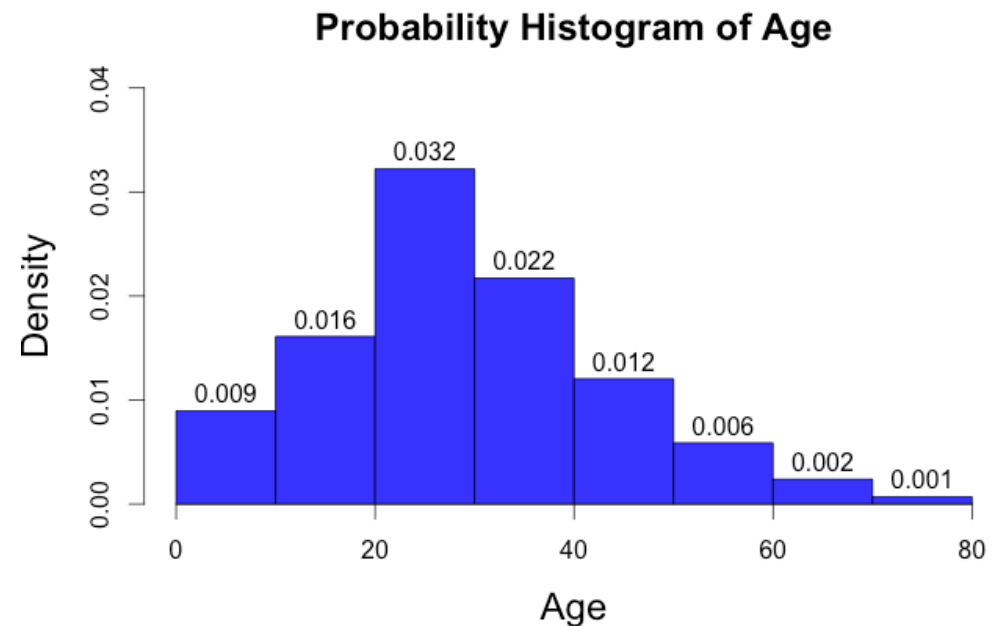
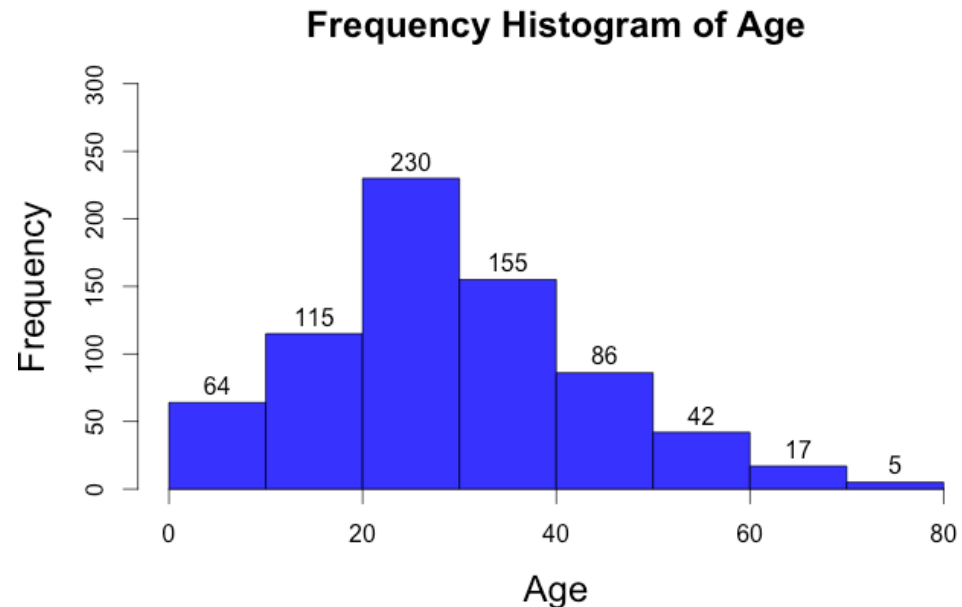
# Categorical variables: Bar charts

Displays the total number or percent of observations falling in each category



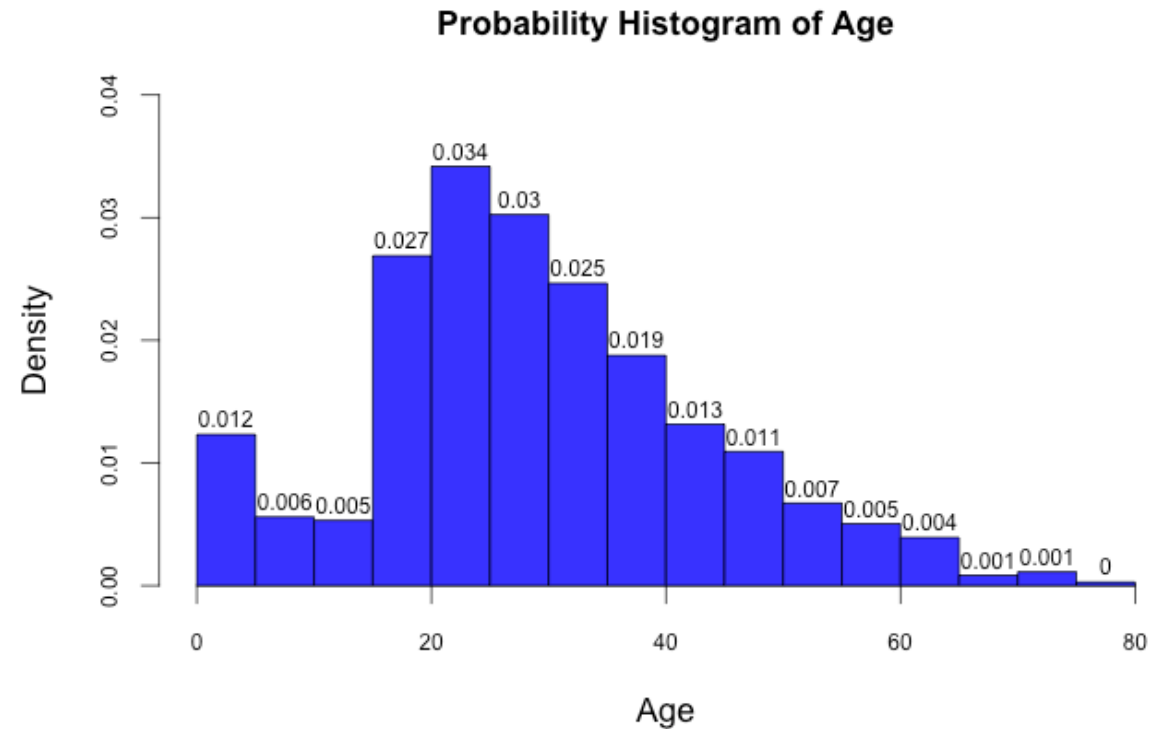
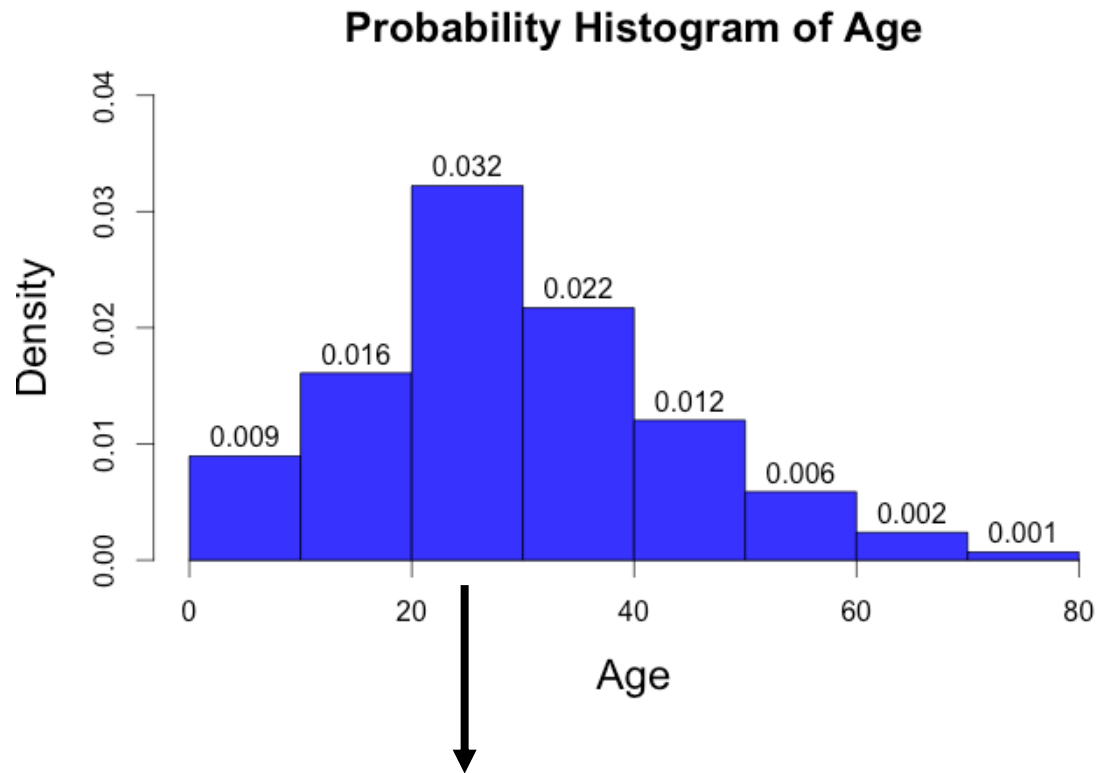
# Quantitative variables: Histograms

- Class-specific **counts** or **relative frequencies** are summarized in a bar-type plot
- Used to summarize the shape of the distribution, assess spread, and look for “extreme” values
- Particularly useful for “large” datasets (>30 observations)



# Probability histograms

**Height of each bar** = Probability **per unit** age for that group.

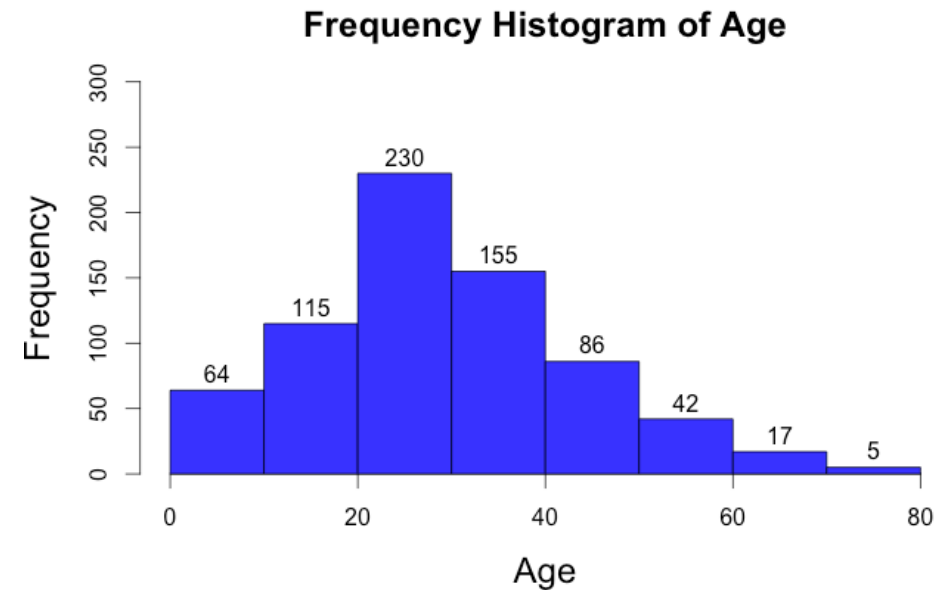
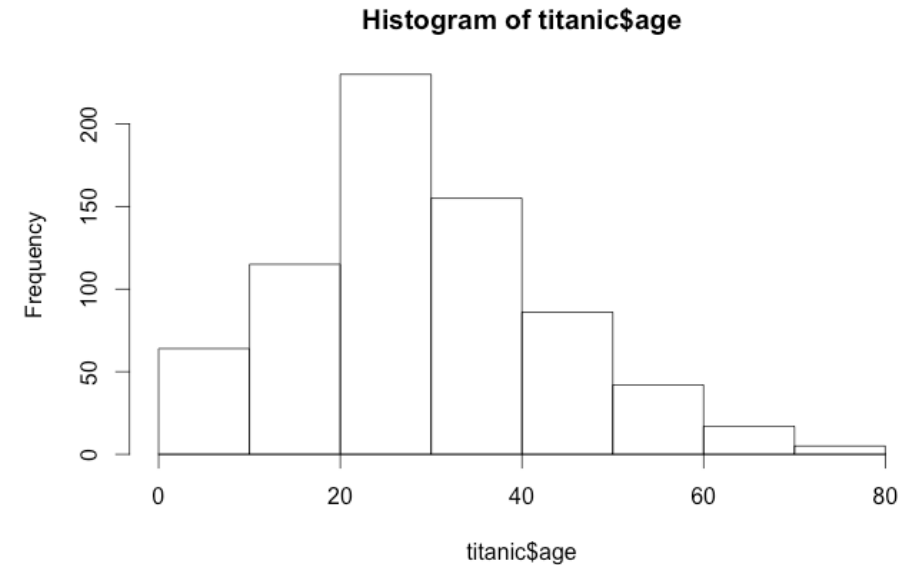


$$P(\text{age between 20 and 30}) = 0.032 \times 10 = 0.32$$

# Histograms in R

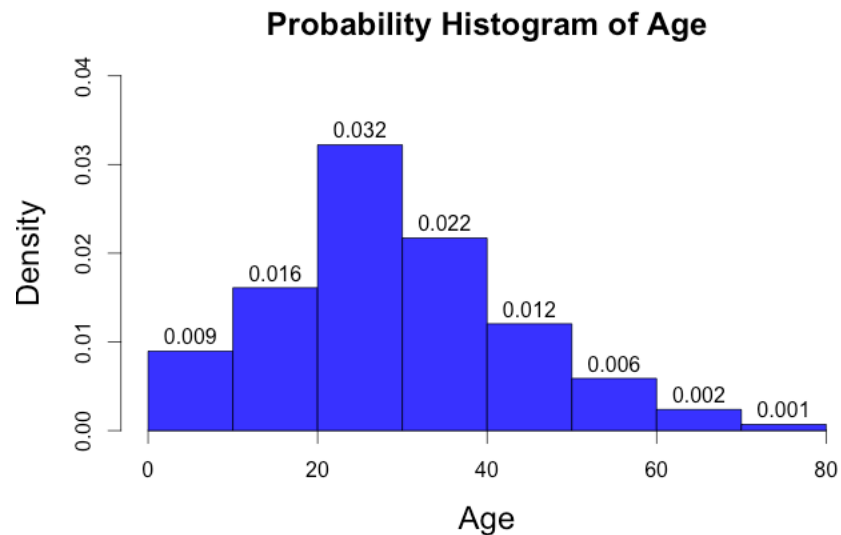
```
titanic = na.omit(titanic)  
hist(titanic$Age)
```

```
hist(titanic$Age,  
     ylim=c(0,300),  
     col='blue',  
     cex.lab=1.5,  
     cex.main=1.5,  
     xlab='Age',  
     main='Frequency  
Histogram of Age',  
     labels=TRUE)
```

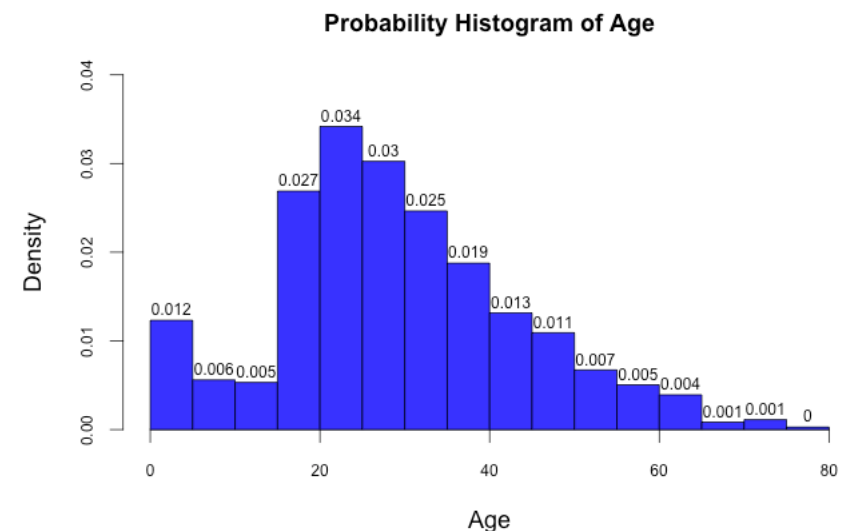


# Histograms in R

```
hist(titanic$Age,  
freq=FALSE,  
ylim=c(0,0.04),  
col='blue',  
cex.lab=1.5,  
cex.main=1.5,  
xlab='Age',  
main='Probability Histogram of Age',  
labels=TRUE)
```



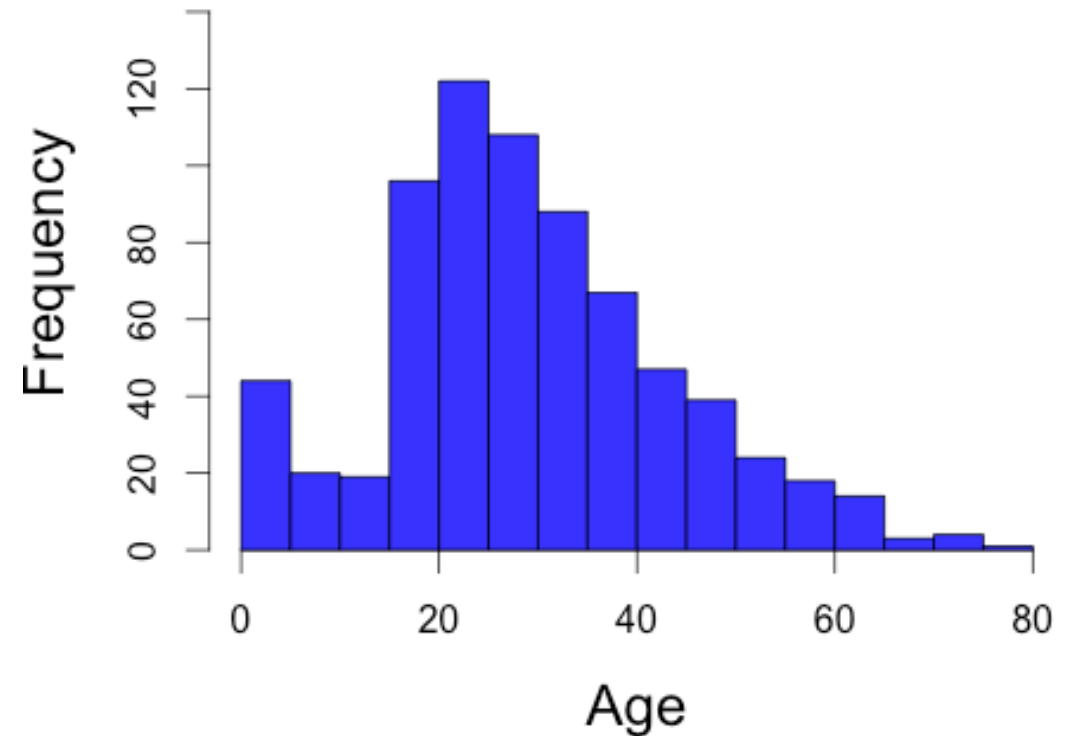
```
hist(titanic$Age,  
freq=FALSE,  
breaks=15,  
ylim=c(0,0.04),  
col='blue',  
cex.lab=1.5,  
cex.main=1.5,  
xlab='Age',  
main='Probability Histogram of Age',  
labels=TRUE)
```



# Common descriptive features

- **Center:** Where is the “middle”?
- **Spread:** How much individual to individual variation exists?
- **Clustering** (number of modes):
  - No bumps: uniform
  - 1 bump: unimodal
  - 2 bumps: bimodal
- **Skewness:** Symmetry? Or is one “tail” longer than the other “tail”?
- **Outliers:** Are there extremes that stand out in the data?

Frequency Histogram of Age



# Unimodal distributions

left-skewed

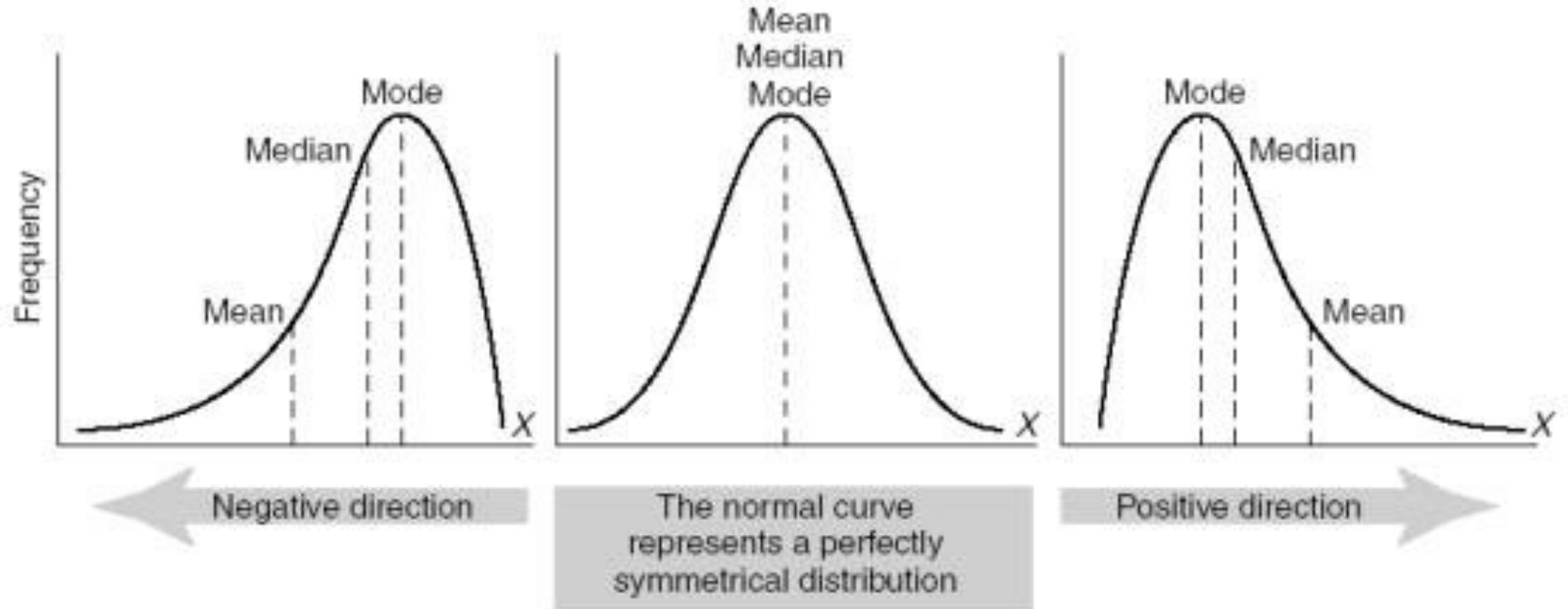
symmetrical

right-skewed

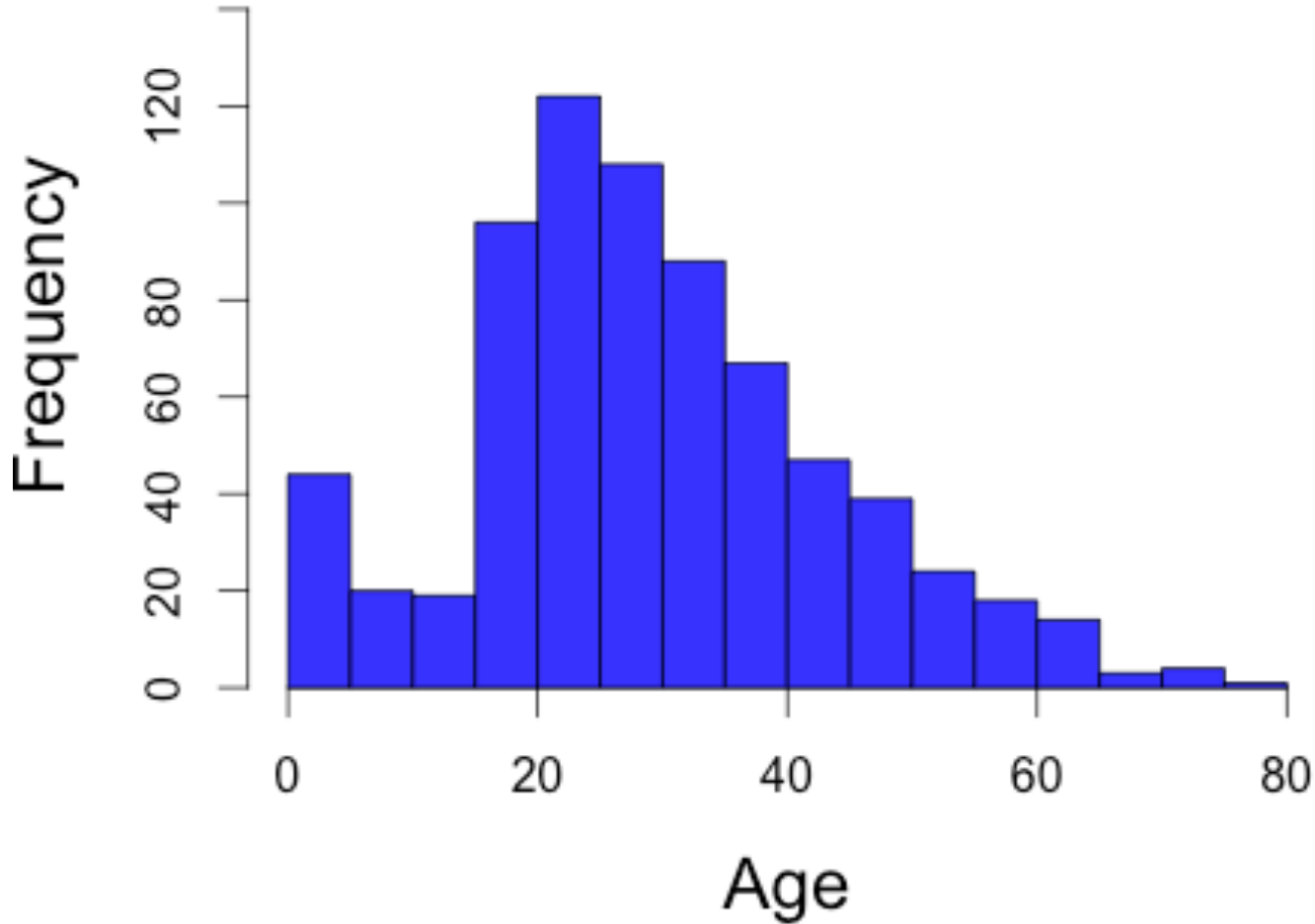
(a) Negatively skewed

(b) Normal (no skew)

(c) Positively skewed



# Frequency Histogram of Age



Slightly skewed to the right  
Mean = 29.70 years  
Median = 28.0 years  
Mode = 24 years



# Numerical methods

To extract meaningful information for purposes of description and comparison; to reduce quantitative data to a few “talking points”.

**Statistic:** A numerical summary computed from a **sample** of data on a quantitative variable.

- Population version of these numerical summaries is generally unknown
- One of the goals of the field of statistics is to **estimate** the population numerical summaries

# Numerical measures of central tendency

- **Mean:**
  - Average value
- **Median:**
  - “Midpoint” of the data when values are ordered from smallest to largest (50th percentile)
- **Mode (meaningful with categorical data):**
  - Measurement that occurs most often (most “popular”)
  - Multiple modes are possible

Mean is sensitive to the magnitude of **all data values**, but the median is not. Median may be a more useful statistic for **skewed data**

# Mathematical notation for the sample mean

In general, we can represent a generic sample of  $n$  data points as an indexed list (order irrelevant):

$$x_1, x_2, x_3, \dots, x_n$$

Sample mean = arithmetic average:

$$\bar{x} = \frac{(x_1 + x_2 + \dots + x_n)}{n}$$

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n x_i$$

**Sensitive to the magnitude of each data value.**

# Numerical measures of sample dispersion, spread and variability

- **Range:**
  - maximum data value - minimum data value
- **Rth percentile** (also called **quantiles**):
  - Sort data values from smallest to largest, find the value that has at most R% of measurements below it and at most (100 – R)% above it ( $0 \leq R \leq 100$ ).
- **Quartiles:**
  - 1st, 2nd and 3rd are the 25th, 50th, and 75<sup>th</sup> percentiles, respectively
  - **IQR (interquartile range)** = 3rd quartile – 1st quartile

# Sample variance and standard deviation

Measures dispersion of individual data points about the average

**Sample variance:** expressed in **squared units** of  $x$ .

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

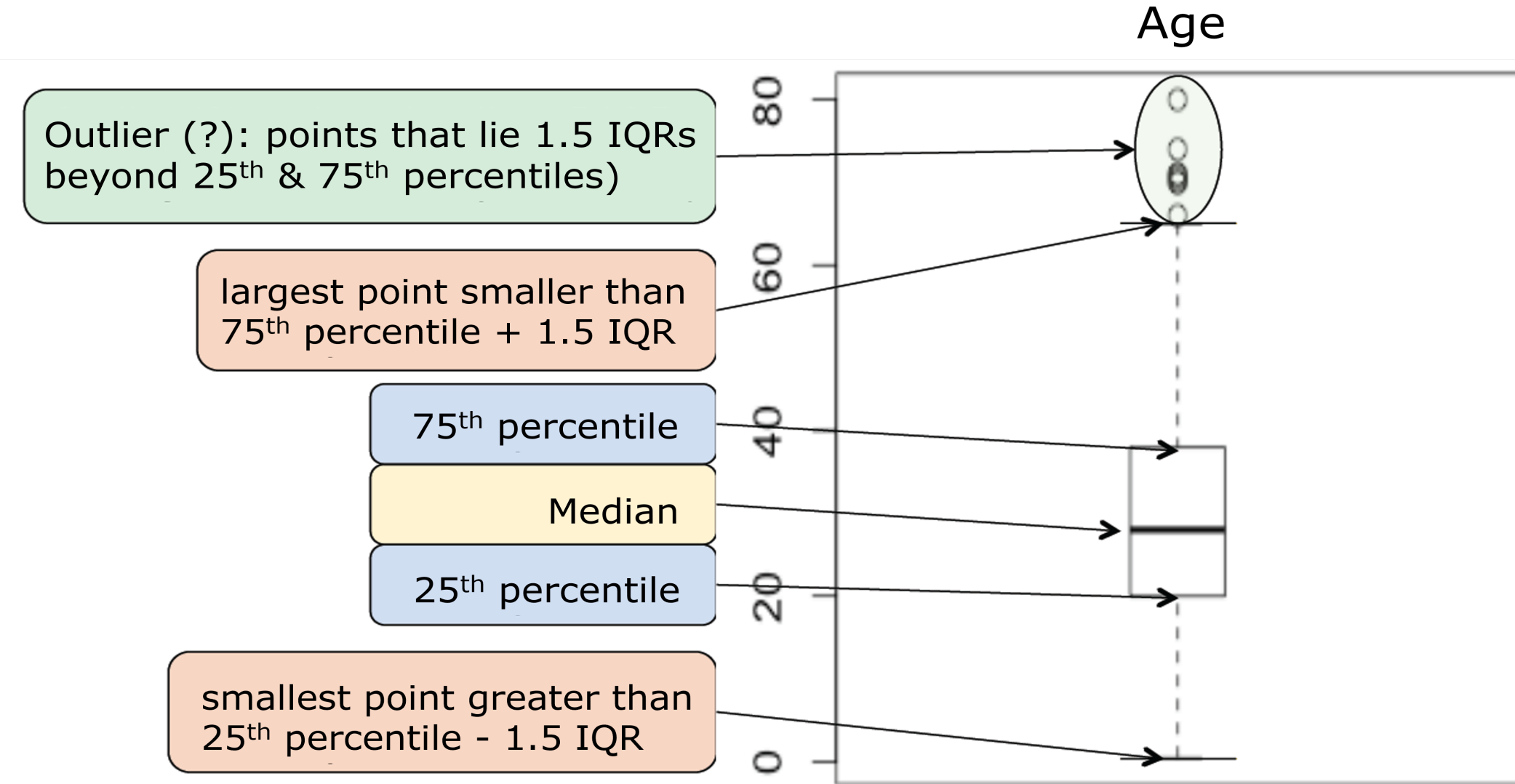
**Sample standard deviation:** expressed in the same units as  $x$ .

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

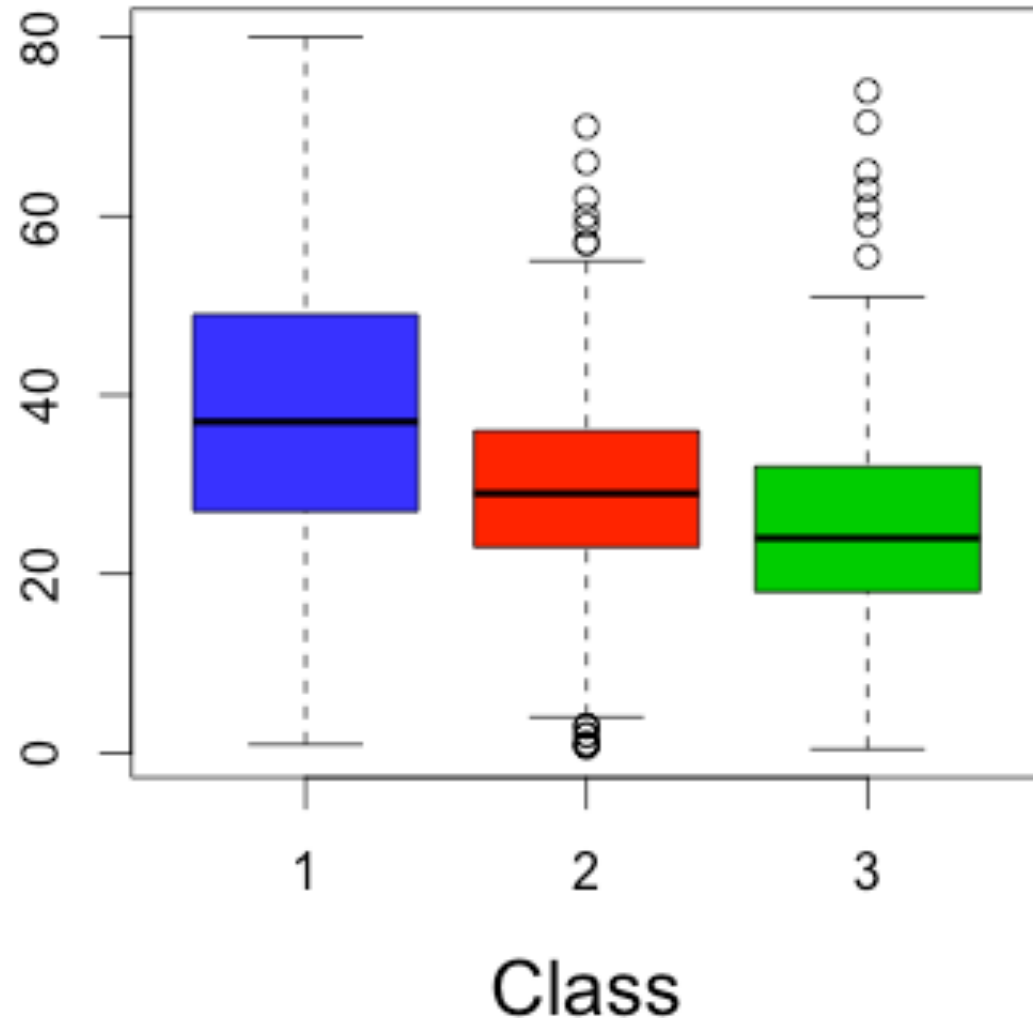
# Sample variance and standard deviation

- Sensitive to the magnitude of each element.
- Used frequently, but may not be very informative about shape in cases where data are highly skewed.
- Both are always non-negative.
- Both equal zero if, and only if, all data values are exactly the same.

# Quantitative variables: Boxplots



## Age by Class



- Median age is noticeably higher in 1st class, but much closer for 2nd and 3rd class.
- Outliers appear to be present in the 2nd and 3rd classes.



# Relationships between two or more variables

## Quantitative vs. Qualitative

- Side-by-side boxplots (used for concrete data)

## • Quantitative vs. Quantitative

- Scatterplots

## • Quantitative-Quantitative-Qualitative

- Coded scatterplots

## • Qualitative vs. Qualitative

- Stacked / unstacked bar charts, contingency tables

# Example: Canadian Prestige Data

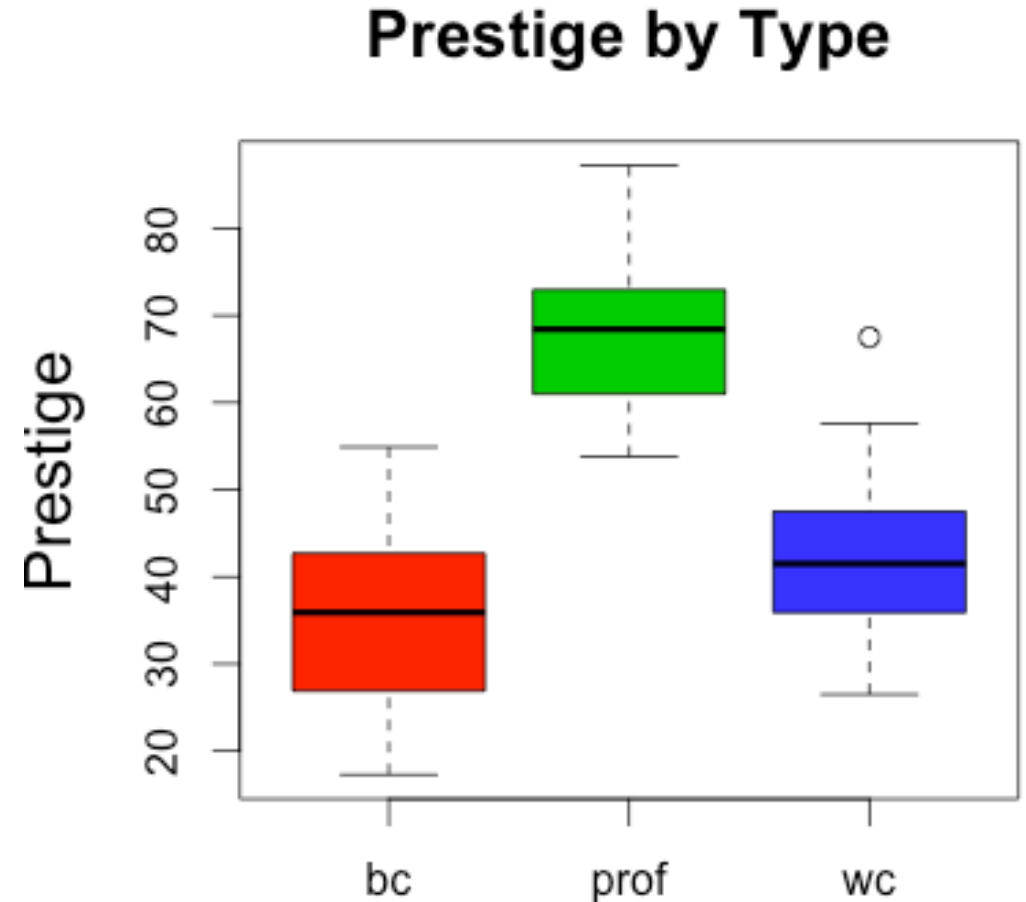
Data obtained on Canadian workers from 98 different occupations in 1971.

- **Education:** Average education of subjects working in a given occupation in 1971 (years after grade 4)
- **Income:** Average income (1971 Canadian dollars)
- **Women:** % of women in a given occupation
- **Prestige:** Occupational prestige score
- **Occupation class:** Blue Collar (bc), Professional, Managerial, and Technical (prof), White Collar (wc)

# Quantitative vs. Qualitative

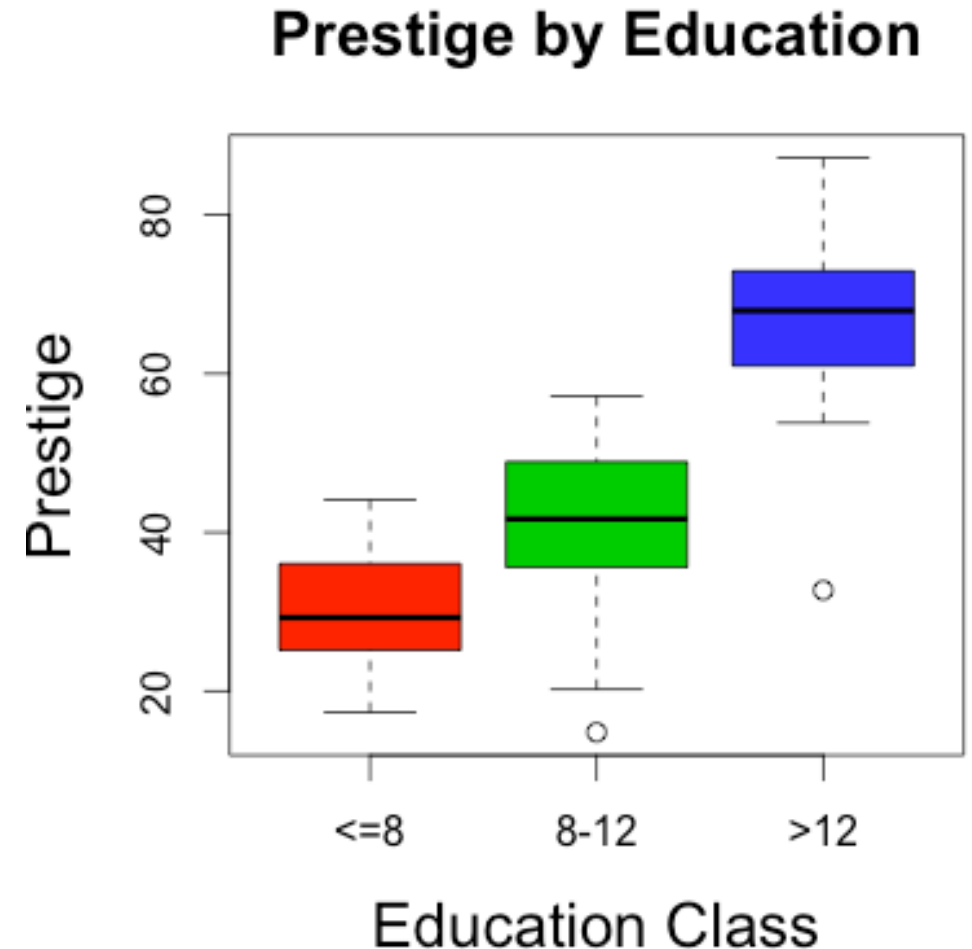
Side-by-side boxplots allow us to quickly compare medians, variability and general shape of a quantitative variable for different levels of a qualitative variable.

- Professionals enjoy higher prestige ratings than blue or white collar workers.
- White collar workers may have slightly higher prestige than blue collar workers.
- Variability in prestige looks similar across occupation classes



Boxplots of prestige vs. education can be constructed by **categorizing education**.

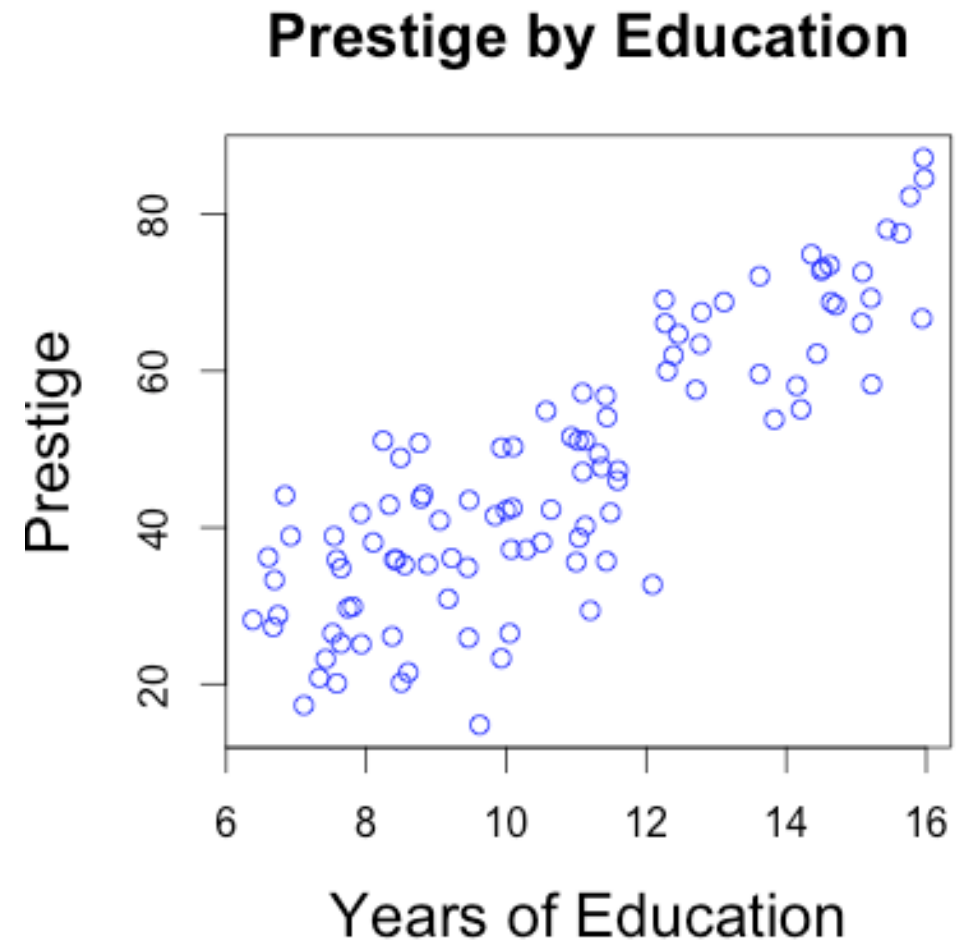
- Prestige increases with education.
- The greatest difference is between those who have education beyond high school and those who do not.

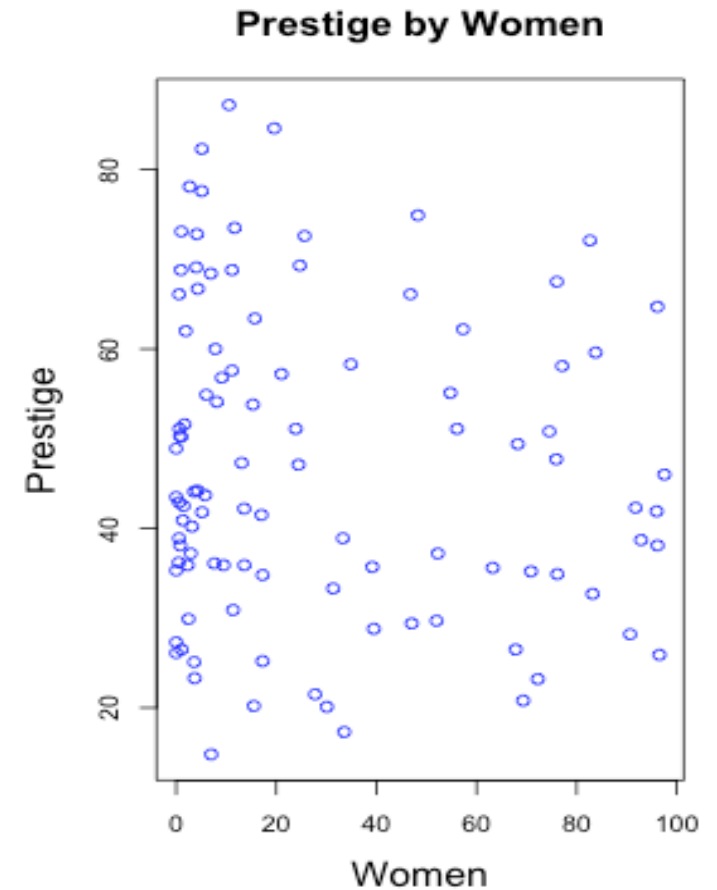
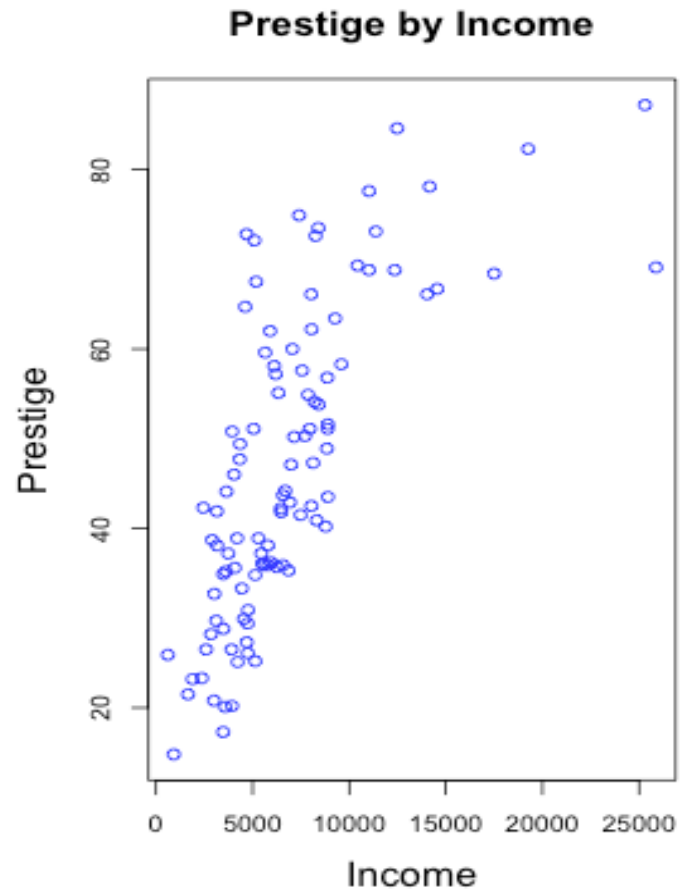
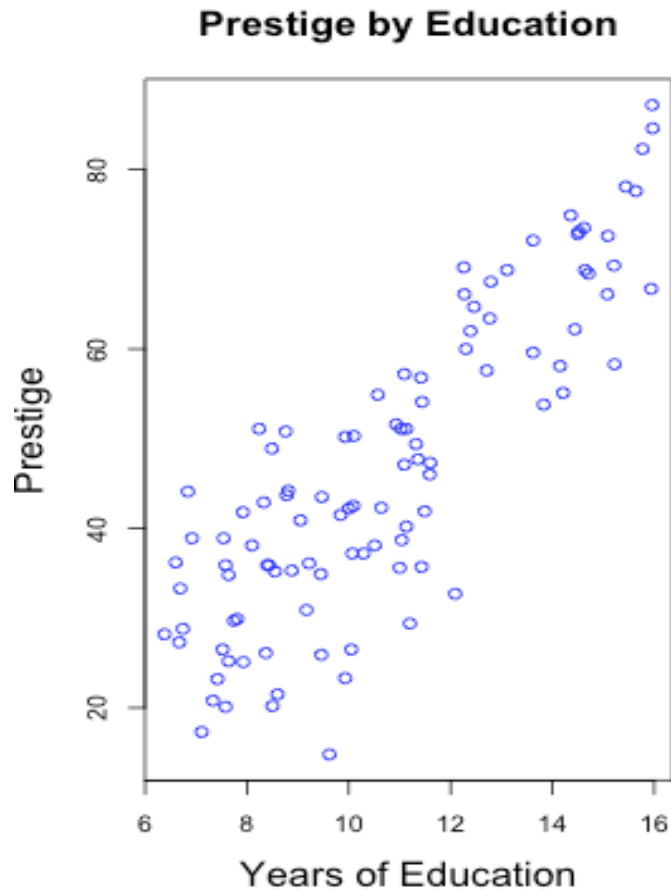


# Quantitative vs. Quantitative

Scatterplots are the best way to visualize the relationship between two quantitative variables.

- **x-axis**
  - Predictor
  - Explanatory variable
  - Independent variable
- **y-axis**
  - Response
  - Dependent variable



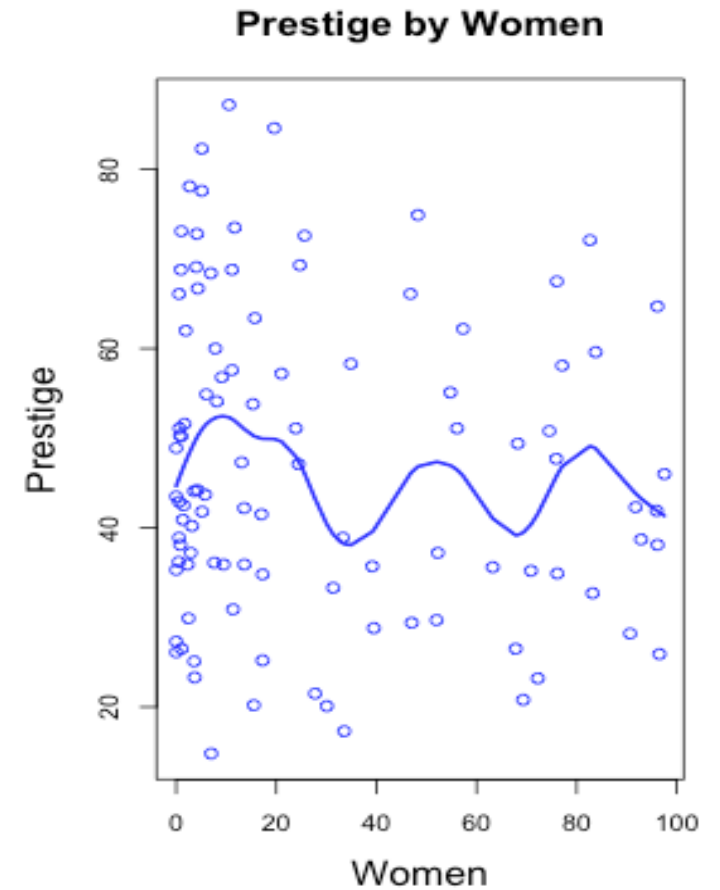
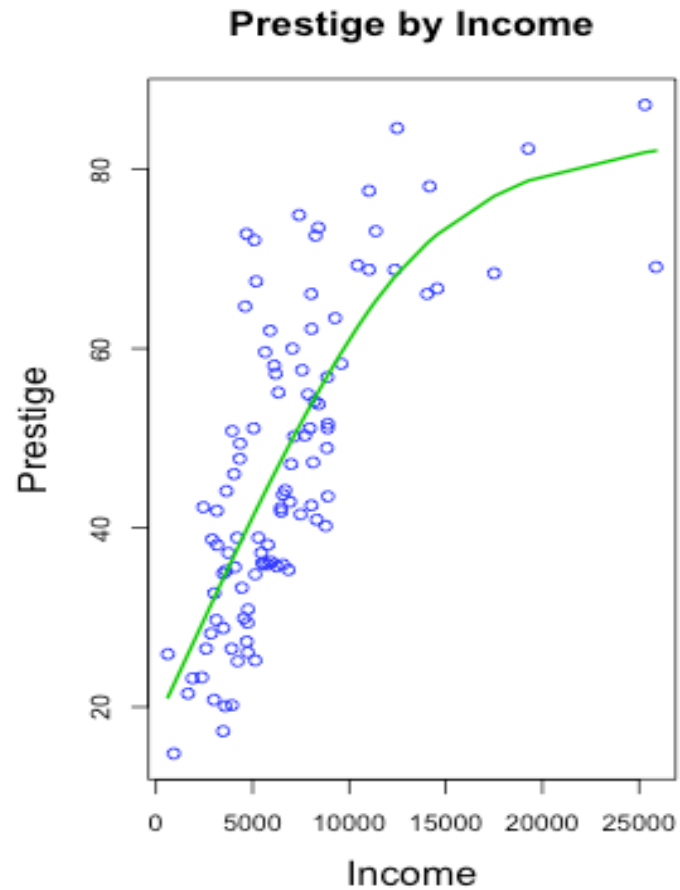
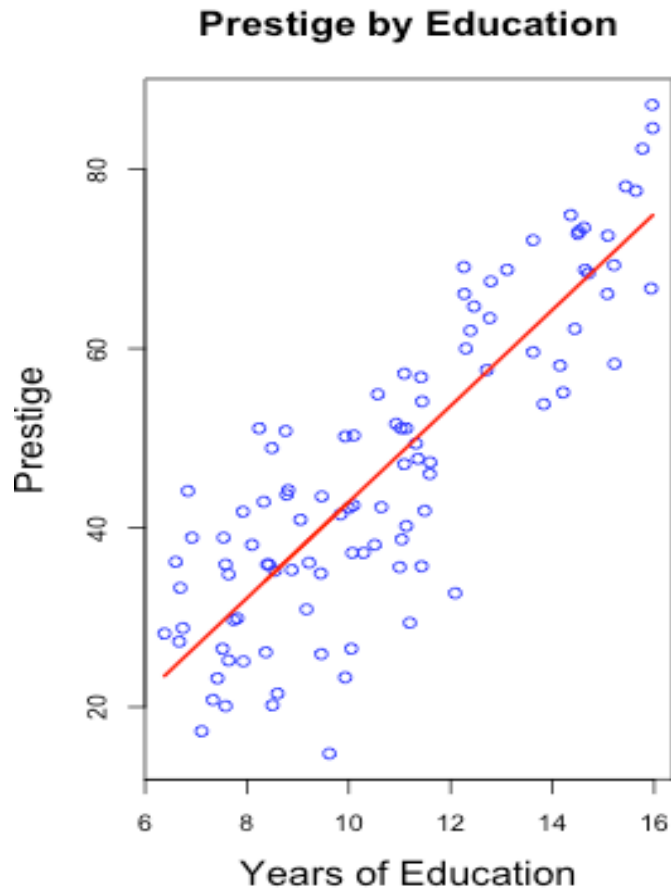


**Direction:** Does “Y” increase or decrease with “X”?

**Trend:** Linear?

**Strength of association:** Amount and width of scatter?

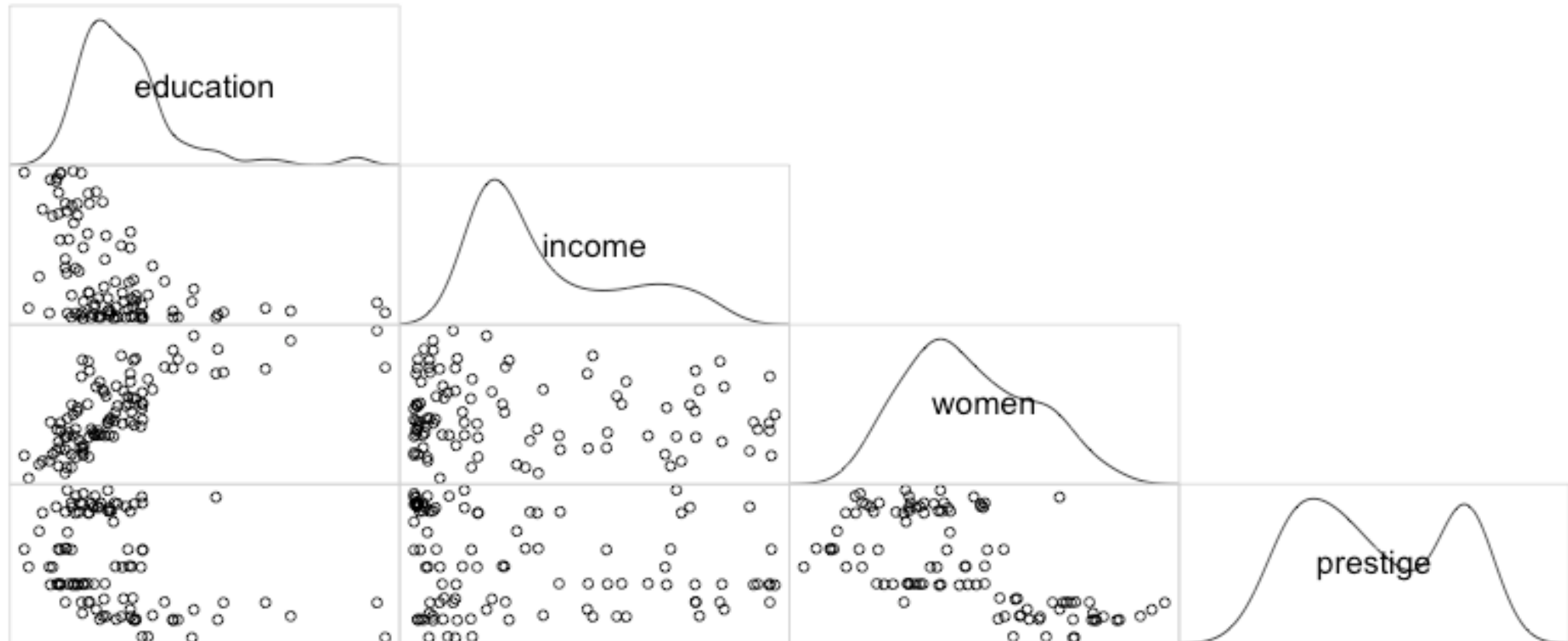
**Outliers:** Any unusual or extreme observations?



Smooth curves can be fit to scatterplots to help visualize trends. The most common “smooth” curve is a line (**linear regression**).

Scatterplot matrices can be used to look at all pairwise relationships between quantitative variables.

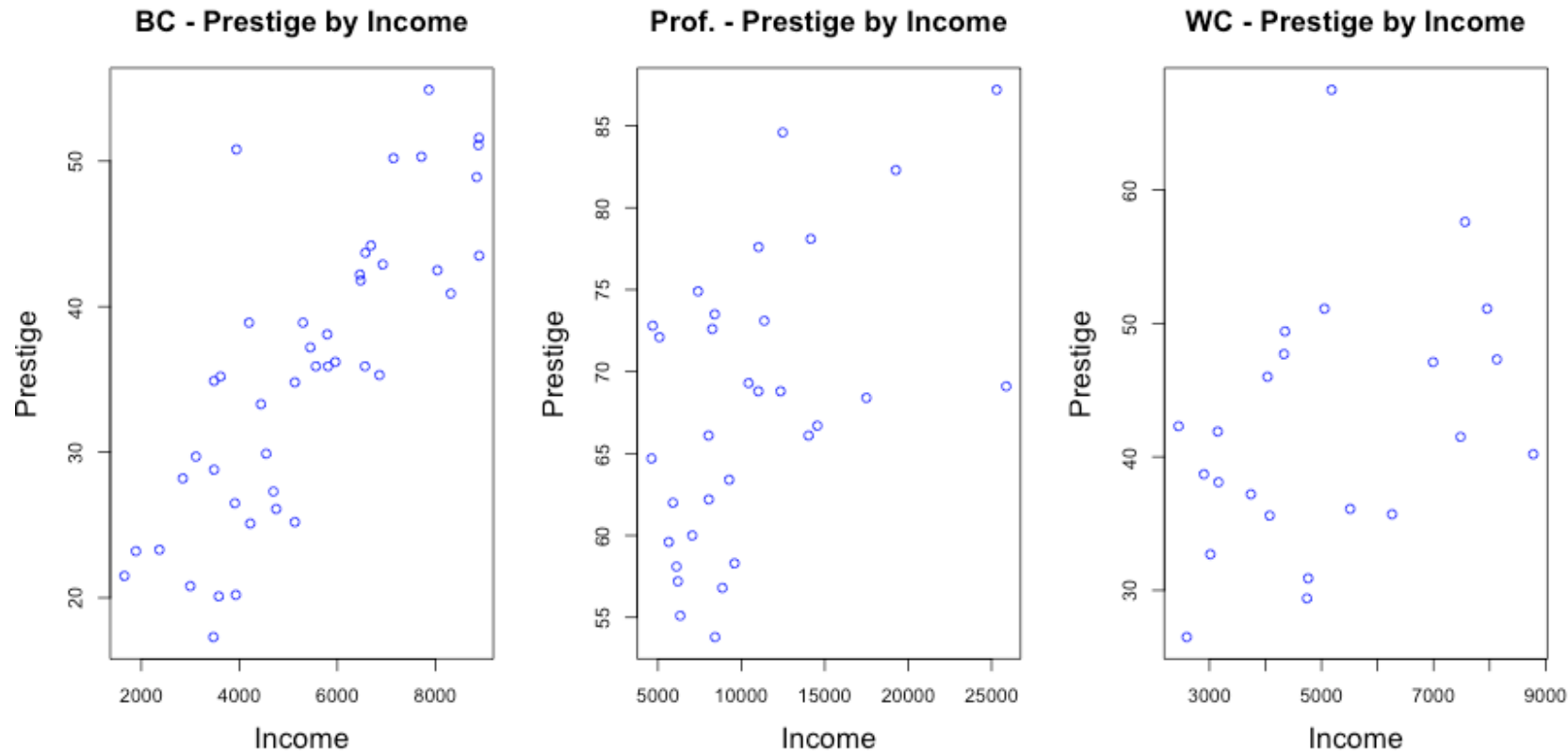
### Scatterplot Matrix





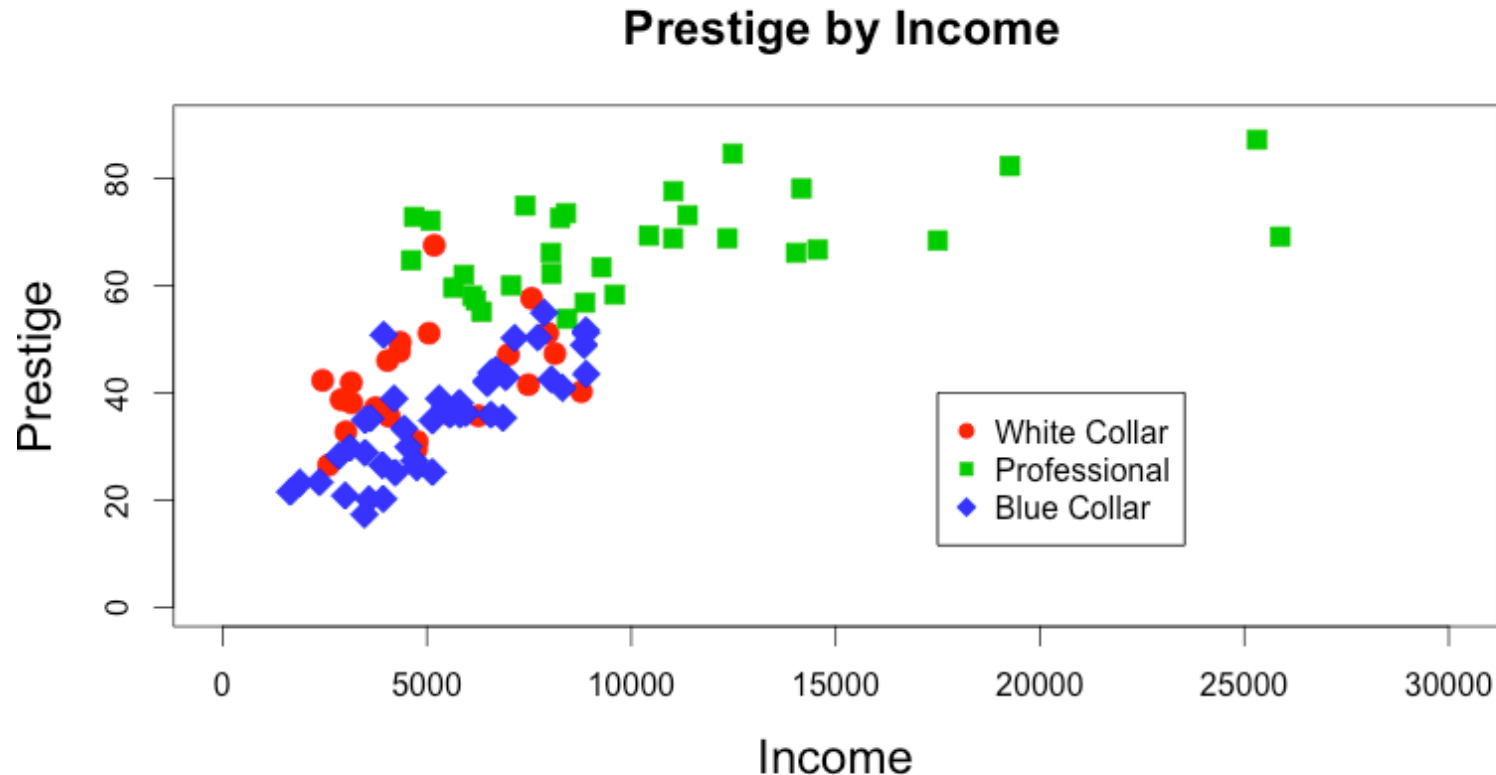
# Quantitative-Quantitative-Qualitative

**Question:** Does the relationship between prestige and income depend on the type of profession?



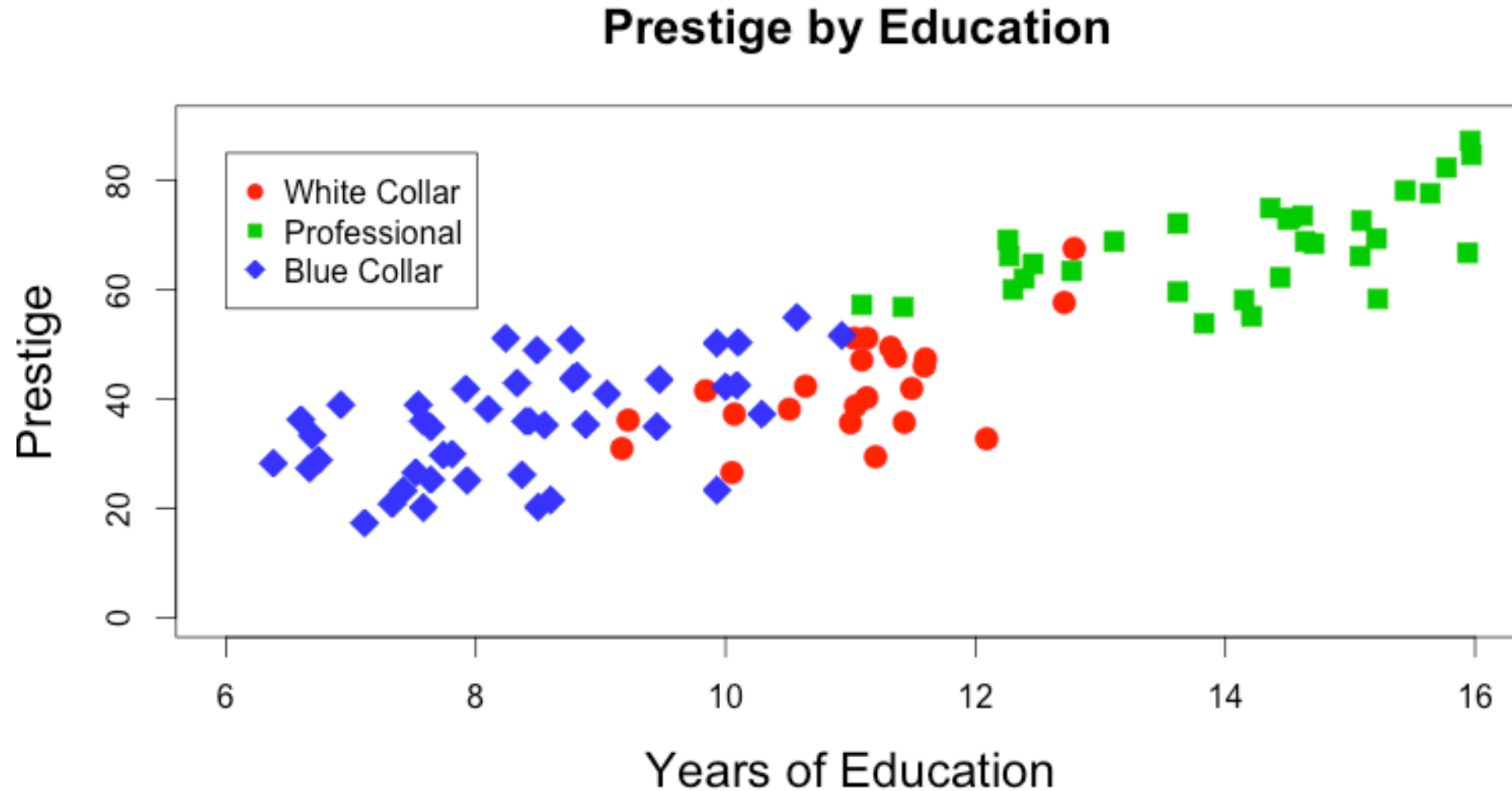
**Not easy to tell from this set of plots!**

**Coded scatterplot:** Label the income & prestige pairs by type of occupation.



- The relationship between prestige and income is similar for blue collar and white collar workers.
- Range of prestige and income values is different for professionals and the relationship is “flatter”.

Relationship between prestige and education also appears to depend on occupation type.



# To do

- Finish Lab 3
- Read: This lecture: Textbook Ch3
- Next lecture: Textbook Ch4