

# MATH 4720 / MSSC 5720

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## **Chapter 3**



**Department of Mathematical and Statistical Sciences**

## TOPIC 2 - CHAPTER 3

- **Type of Variables**
- **Frequency distributions**
- **Histograms**
- **Mean, median, variance and standard deviation**
- **Quartiles, interquartile range**
- **Boxplots**
- **Correlation**

# ALL ABOUT VARIABLES

- **Variable:** Any characteristic or quantity to be measured on units in a study
- **Categorical variable:** Places a unit into one of several categories
  - Examples: Gender, race, political party
- **Quantitative variable:** Takes on numerical values for which arithmetic makes sense
  - Examples: SAT score, number of siblings, cost of textbooks
- **Univariate data has one variable.**
- **Bivariate data has two variables.**
- **Multivariate data has three or more variables.**



# TYPES OF VARIABLES

## Examples:

Variable	Numeric		Categorical
	Discrete	Continuous	
Length		X	
Hours Enrolled	X		
Major			X
Zip Code			X



# AUSTRALIAN INSTITUTE OF SPORT DATA

- **Description**

- Data on 102 male and 100 female athletes collected at the Australian Institute of Sport, courtesy of Richard Telford and Ross Cunningham.

- **Source**

- Cook and Weisberg (1994), *An Introduction to Regression Graphics*. John Wiley & Sons, New York.

AIS.mjp

Variable	Description
sex	sex
sport	sport
rcc	red cell count
wcc	white cell count
Hc	Hematocrit
Hg	Hemoglobin
Fe	plasma ferritin concentration
bmi	body mass index, weight/(height)
ssf	sum of skin folds
Bfat	body fat percentage
lbm	lean body mass
Ht	height (cm)
Wt	weight (Kg)

# SUMMARIZING A SINGLE CATEGORICAL VARIABLE



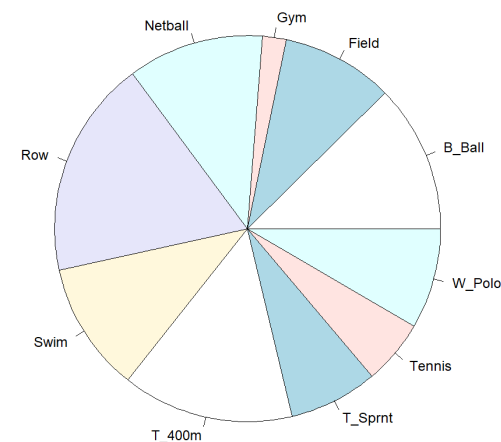
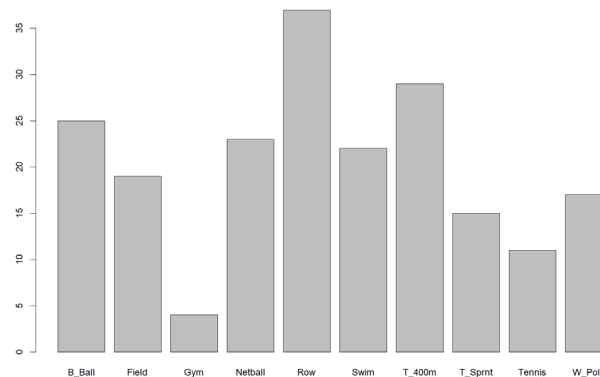
- **Frequency (Count)** - number of times the value occurs in the data
- **Relative frequency (Percent)** - proportion of the data with the value
- **Cumulative Frequency**
- **Cumulative Relative Frequency**
- **ais.csv (D2L/Content/Datasets)**

- **R Code:**

- `library("sn")`
- `data("ais")`
- `tbl <- table(ais$sport)`
- `cumsum(tbl)`
- `prop.table(tbl)`
- `cumsum(prop.table(tbl))`
- `barplot(tbl)`
- `pie(tbl)`

Tally for Discrete Variables: sport

sport	Count	Percent	CumCnt	CumPct
B_Ball	25	12.38	25	12.38
Field	19	9.41	44	21.78
Gym	4	1.98	48	23.76
Netball	23	11.39	71	35.15
Row	37	18.32	108	53.47
Swim	22	10.89	130	64.36
T_400m	29	14.36	159	78.71
T_Sprnt	15	7.43	174	86.14
Tennis	11	5.45	185	91.58
W_Polo	17	8.42	202	100.00
N=	202			



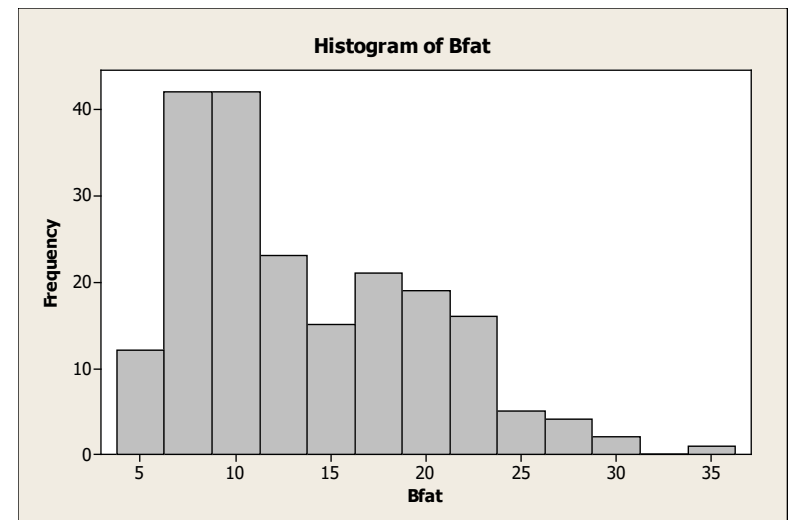
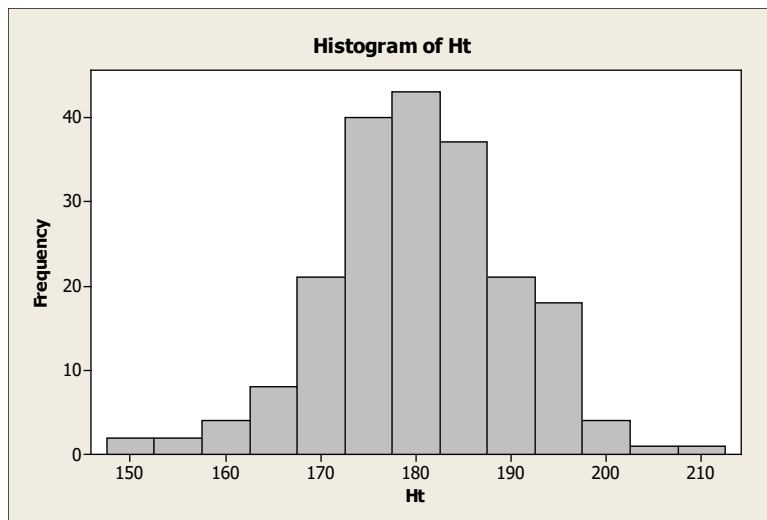
# ANALYZING A SINGLE QUANTITATIVE VARIABLE

- Consider the ALS data which contains 202 athletes.
- What is a **typical** height of athletes?
- How much **spread** is there in their Body fats?
- **Typical** is generally characterized by the **center** of the data
- **Spread** is generally reported as an interval containing most of the data

# HISTOGRAMS



- **Histogram** - bar graph of binned data where the height of the bar above each bin denotes the frequency (relative frequency) of values in the bin
- Typical concentration?
- Spread?
- Roughly how many athletes are shorter than 180 cm?
- **R Code:**
  - `hist(ais$Ht)`
  - `hist(ais$Bfat, breaks = 16)`







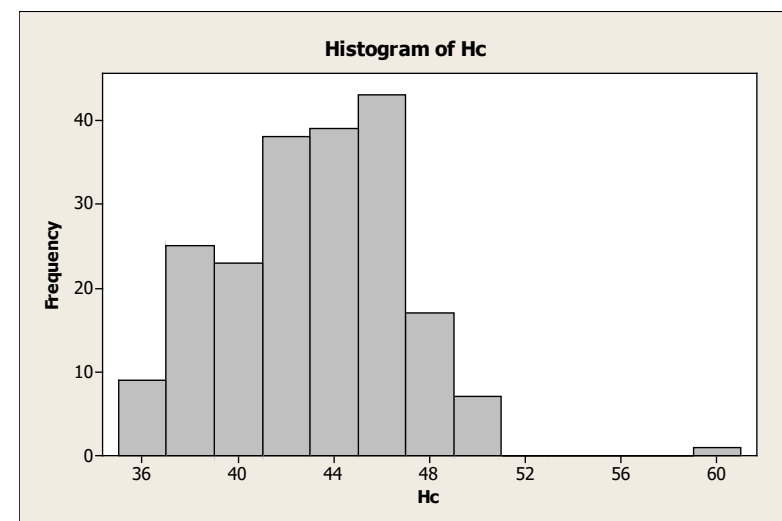
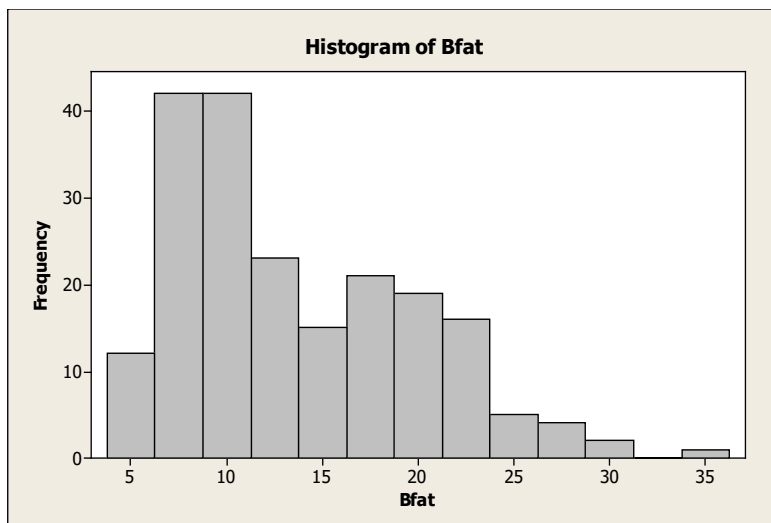
# DESCRIBING THE SHAPE OF QUANTITATIVE DATA

- **Symmetric** data has roughly the same mirror image on each side of a center value.
- **Skewed** data has one side (either **right** or **left**) which is much longer than the other relative to the **mode** (peak value).
- The above definitions are most useful when describing data with a single mode.
- **Multimodal** data has more than one mode.
- Beware of **outliers** when describing shape.
- Shape of the AIS Data?



## DESCRIBING THE SHAPE (CONT...)

- **Bfat: Body Fat**
  - skewed to the right
  - Bimodal
- **Hematocrit (Hc):** Volume percentage (%) of red blood cells in blood
  - Outlier
- **R Code:**
  - `hist(ais$Ht)`
  - `hist(ais$Bfat, breaks = 16)`





# STEM AND LEAF PLOTS

- Separate each value into a **stem** (all but the rightmost digit) and a **leaf** (the rightmost digit)
- Write unique sorted stems in a vertical column
- Add each leaf to the right of its stem in increasing order
- Example from AIS:
- rcc (red cell counts)
  - Female-Row:
    - 4.26 4.63 4.36 3.91 4.51 4.37 4.90 4.46  
3.95 4.46 5.02 4.26 4.46 4.16 4.49 4.21  
4.57 4.87 4.44 4.45 4.41 4.87
  - Male-Row:
    - 4.87 5.04 4.40 4.95 4.78 5.21 5.22 5.18  
5.40 4.92 5.24 5.09 4.83 5.22 4.71
- R Code:
  - `stem(ais[ais$sport == "Row"],$RCC, scale = 2)`

```
Stem-and-leaf of rcc  N  = 37
Leaf Unit = 0.010
```

```
2    39    15
2    40
3    41     6
6    42    166
8    43     67
16   44    01456669
18   45     17
(1)  46     3
18   47     18
16   48    3777
12   49    025
9    50    249
6    51     8
5    52    1224
1    53
1    54     0
```



# HISTOGRAMS VS. STEM AND LEAF PLOTS

- **Stem and leaf plots (typically) display actual data values whereas histograms do not**
- **Stem and leaf plots are more useful for small data sets (less than 100 values)**
- **Histograms can be constructed for larger data sets**

# SUMMARY STATISTICS FOR QUANTITATIVE DATA

- **Measures of center (typical)**

- The **sample median** is the middle observation if the values are arranged in increasing order.
- The **sample mean** of  $n$  observations is the average, the sum of the values divided by  $n$ .

$X_1, \dots, X_n$  represents  $n$  data values

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

# SUMMARY STATISTICS FOR QUANTITATIVE DATA

- **Measures of spread:**

- **Interquartile range**, **IQR = Q3-Q1**, the range of the middle 50% of the data
  - **first quartile (Q1)** is the 25th percentile
  - **third quartile (Q3)** is the 75th percentile
- **sample variance**,  **$s^2$** , is the sum of squared deviations from the sample mean divided by  **$n-1$**

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

- **sample standard deviation**,  **$s$** , is the square root of sample variance. Preferred because it has the same units as the data.



# EXAMPLE ON HOW TO CALCULATE THE VARIANCE

## EXAMPLE 3.9

The time between an electric light stimulus and a bar press to avoid a shock was noted for each of five conditioned rats. Use the given data to compute the sample variance and standard deviation.

Shock avoidance times (seconds): 5, 4, 3, 1, 3

**Solution** The deviations and the squared deviations are shown in Table 3.11. The sample mean  $\bar{y}$  is 3.2.

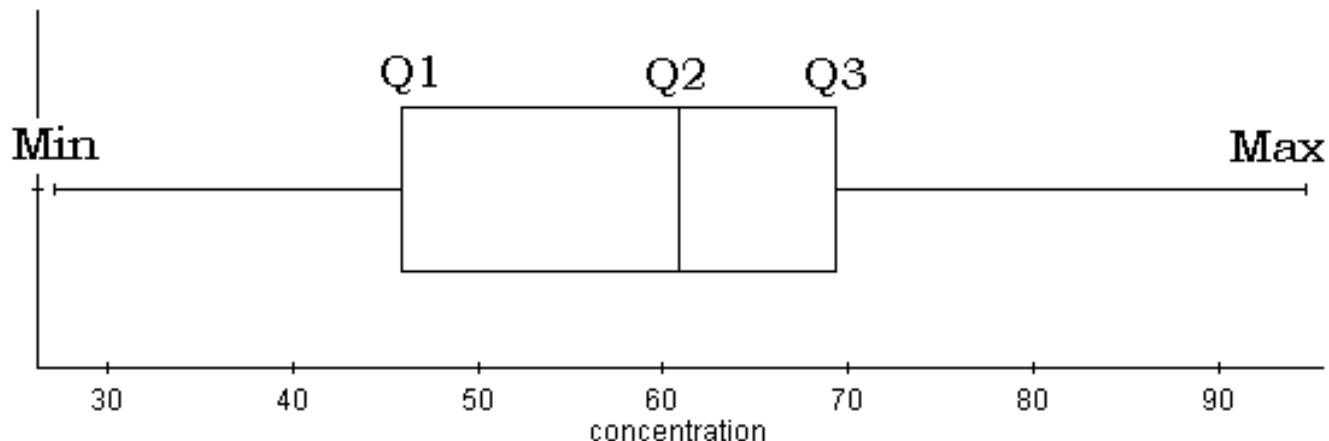
	$y_i$	$y_i - \bar{y}$	$(y_i - \bar{y})^2$
	5	1.8	3.24
	4	.8	.64
	3	-.2	.04
	1	-2.2	4.84
	3	-.2	.04
Totals	16	0	8.80

Using the total of the squared deviations column, we find the sample variance to be

$$s^2 = \frac{\sum_i (y_i - \bar{y})^2}{n - 1} = \frac{8.80}{4} = 2.2$$

# SUMMARY STATISTICS FOR QUANTITATIVE DATA

- **$p$ th percentile** -the value such that  $p \times 100\%$  of values are below it and  $(1-p) \times 100\%$  are above it
  - **first quartile (Q1)** is the 25th percentile
  - **second quartile (Q2)** 50th percentile (median)
  - **third quartile (Q3)** is the 75th percentile
- **5-number summary:** Min, Q1, Q2, Q3, Max
  - **Boxplots:** Stacking boxplots can be very useful for comparing multiple groups





# BOX PLOT

## Minimum, $Q_1$ , Median, $Q_3$ , and Maximum of **ALS-weight**

- **R Code:**

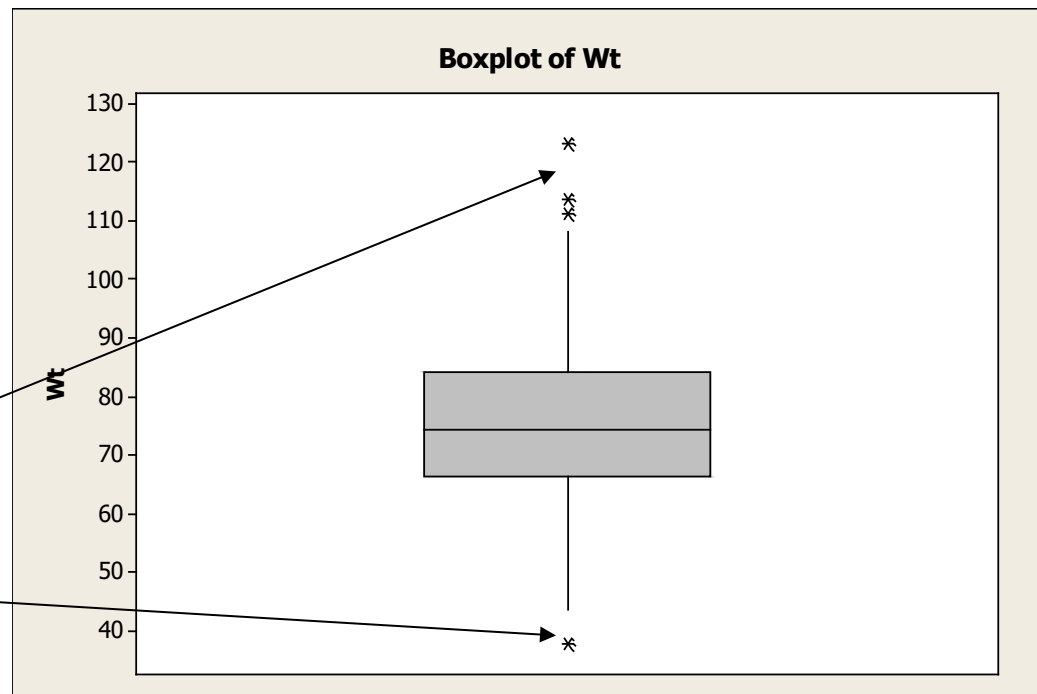
- `boxplot(ais$Wt)`

- These five numbers are called the

- Five Number Summary

- What are these points?

- Outliers



- Interquartile Range (IQR):

- Distance between the first quartile ( $Q_1$ ) and the third quartile ( $Q_3$ ).  $IQR = Q_3 - Q_1$

## BOX PLOT (CONT...)

- Outliers : observations that are unusually far from the bulk of the data.
- What are some possible explanations for outliers?
  - The data point was recorded wrong.
  - The data point wasn't actually a member of the population we were trying to sample.
  - We just happened to get an extreme value in our sample.
- The 1.5 x IQR Criterion for Outliers: Designate an observation a suspected outlier if it falls more than 1.5 x IQR below the first quartile or above the third quartile.



## 1.5\*IQR CRITERION EXAMPLE

- Suppose you had the following data set:  
-2, 15, 3, 7, 10, 21, 1, 5, 12, 8, 1, 35, 10

List data from smallest to largest:

Find  $Q_1$ , Median,  $Q_3$ , Min, and Max:

$$IQR = Q_3 - Q_1 = \underline{\hspace{2cm}}$$

$$1.5 * IQR = \underline{\hspace{2cm}}$$

$$Q_1 - 1.5 * IQR = \underline{\hspace{2cm}} \text{ If less than this number, then outlier}$$

$$Q_3 + 1.5 * IQR = \underline{\hspace{2cm}} \text{ If more than this number, then outlier}$$

Are there any outliers in this data set?



## 1.5\*IQR CRITERION EXAMPLE

- Suppose you had the following data set:

**-2, 15, 3, 7, 10, 21, 1, 5, 12, 8, 1, 35, 10**

**List data from smallest to largest:**

**-2, 1, 1, 3, 5, 7, 8, 10, 10, 12, 15, 21, 35**

**Find  $Q_1$ , Median, and  $Q_3$ :**

$$Q_1 = (1+3)/2 = 2 \quad \text{Median} = 8 \quad Q_3 = (12 + 15)/2 = 13.5$$

$$IQR = Q_3 - Q_1 = 11.5$$

$$1.5*IQR = 17.25$$

$$Q_1 - 1.5*IQR = -15.25 \quad \text{If less than this number, then outlier}$$

$$Q_3 + 1.5*IQR = 30.75 \quad \text{If more than this number, then outlier}$$

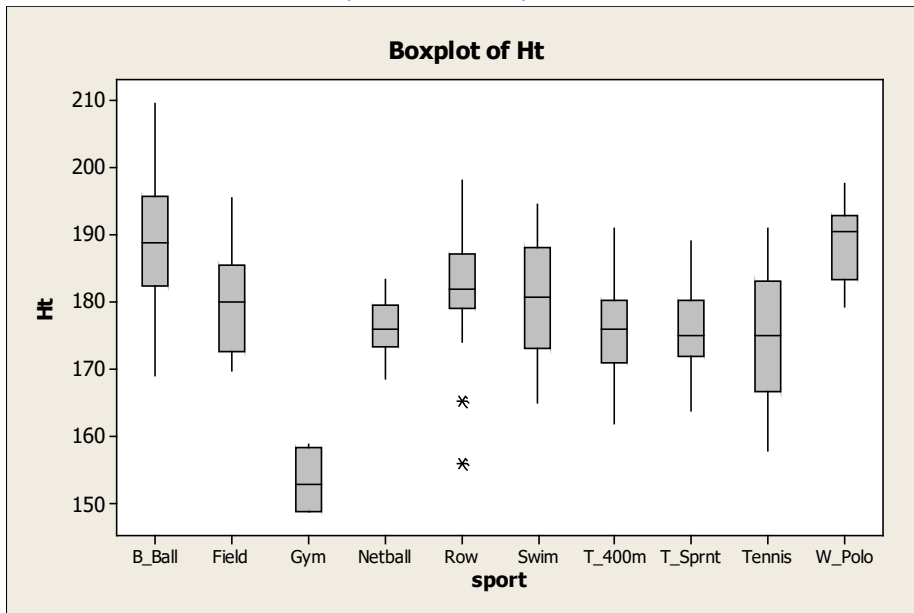
**Are there any outliers in this data set? Yes, 35**



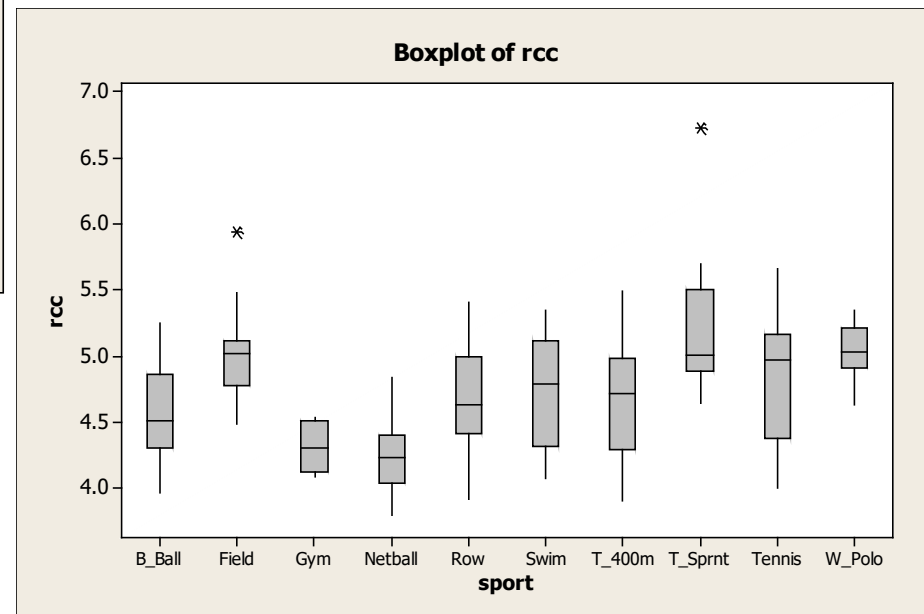
# SIDE-BY-SIDE BOX PLOT

- **R Code:**

➤ `boxplot(Ht ~ sport, data=ais)`



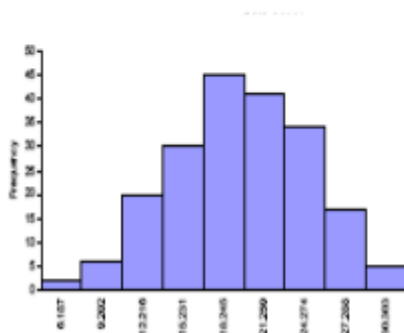
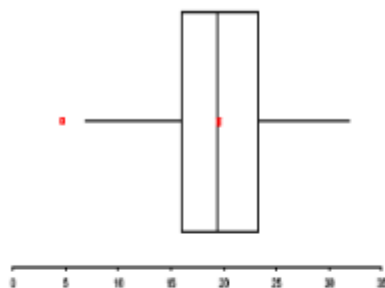
➤ `boxplot(RCC ~ sport, data=ais)`



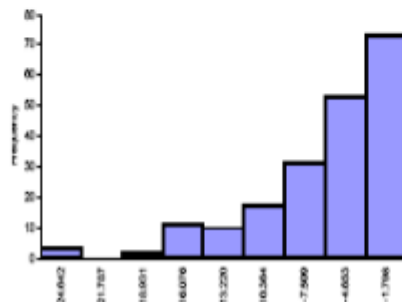
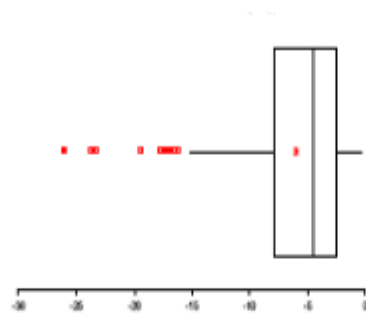


# COMPARING HISTOGRAMS AND CORRESPONDING BOXPLOTS

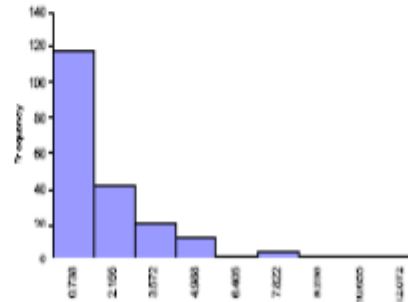
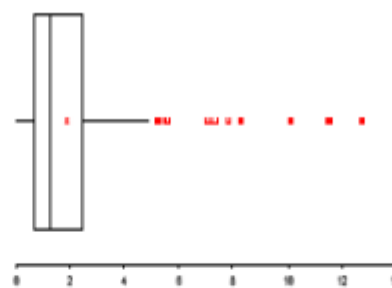
Symmetric



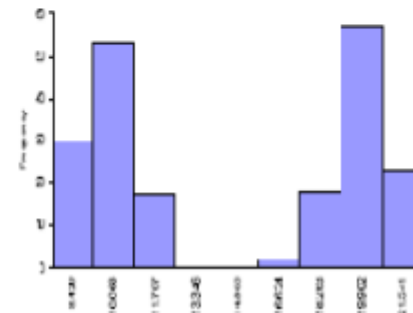
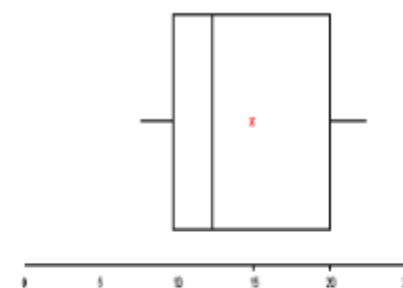
Left Skewed



Right Skewed



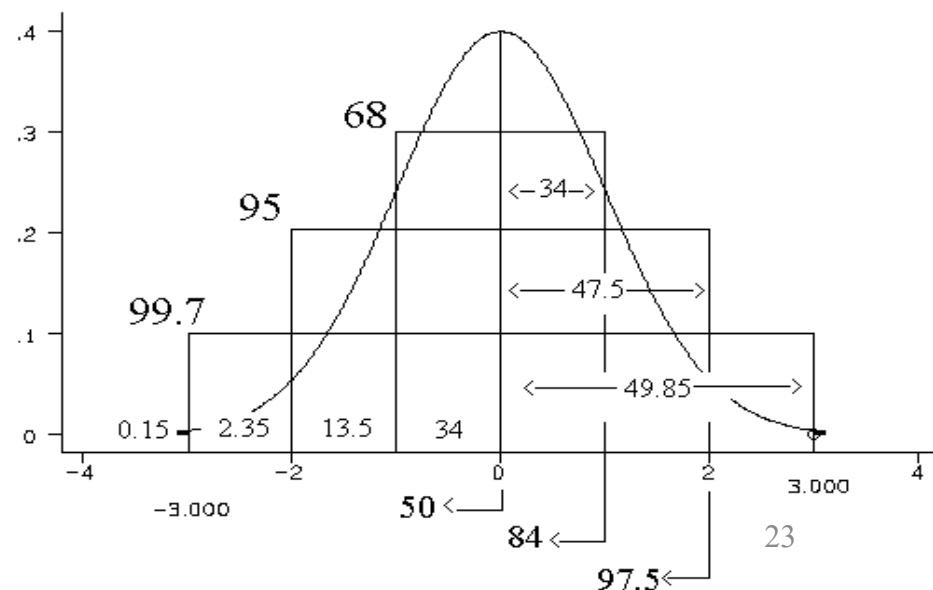
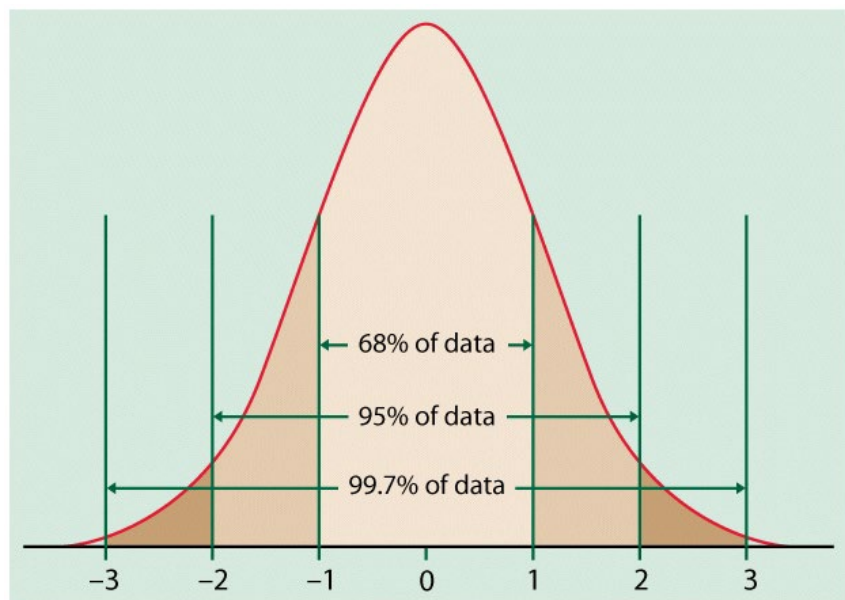
Bimodal





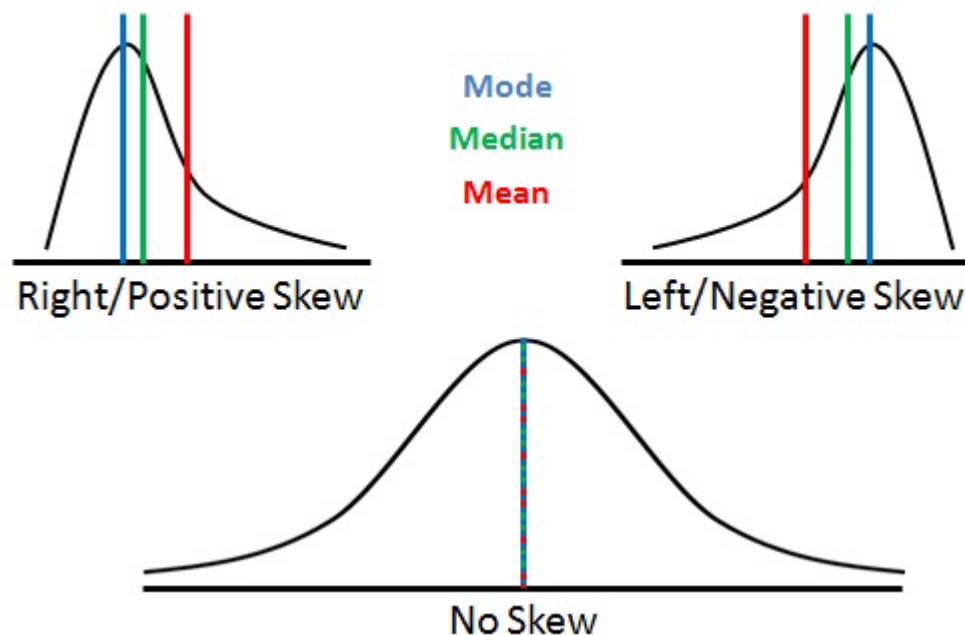
# EMPIRICAL RULE (THE 68-95-99.7 RULE)

- If the distribution is mound-shaped, then
  - Approximately 68% of the data falls within one standard deviation of the mean
  - Approximately 95% of the data falls within two standard deviations of the mean
    - Approximate value of  $s = \frac{\text{range}}{4}$
  - Approximately 99.7% of the data falls within three standard deviations of the mean



# COMPARING MEASURES OF CENTER AND SPREAD

- The **sample mean** and the **sample standard deviation** are good measures of center and spread, respectively, for **symmetric** data
- If the data set is **skewed** or has **outliers**, the **sample median** and the **interquartile range** are more commonly used
- Mean versus median



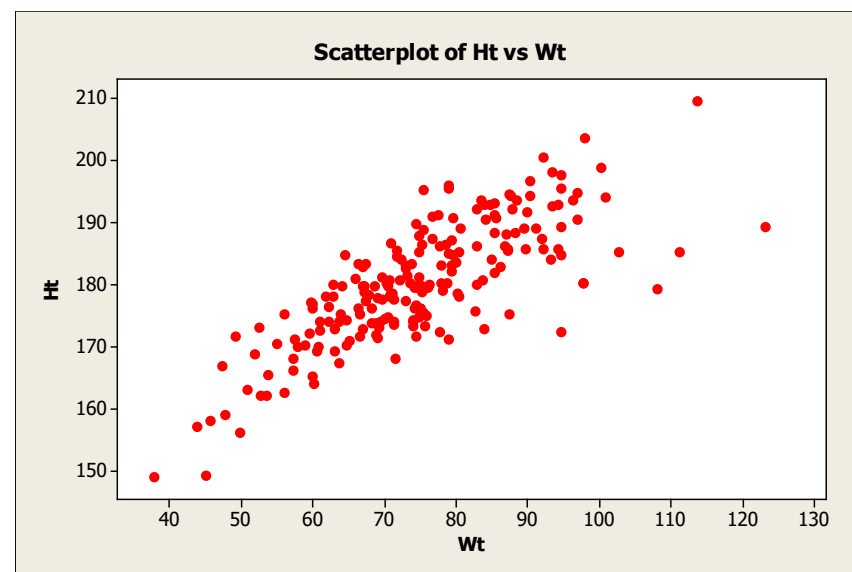


# RELATIONSHIPS BETWEEN 2 NUMERIC VARIABLES

- Depending on the situation, one of the variables is the explanatory variable and the other is the response variable.
- There is not always an explanatory-response relationship.
- Examples:
  - Height and Weight
  - Income and Age
  - SAT scores on math exam and on verbal exam
  - Amount of time spent studying for an exam and exam score

- **R Code:**

- `plot(Ht ~ Wt, data=ais)`





# RELATIONSHIPS BETWEEN 2 NUMERIC VARIABLES

- **Scatterplots**

- Look for overall pattern and any striking deviations from that pattern.
- Look for outliers, values falling outside the overall pattern of the relationship
- You can describe the overall pattern of a scatterplot by the form, direction, and strength of the relationship.
  - Form: Linear or clusters
  - Direction
    - Two variables are positively associated when above-average values of one tend to accompany above-average values of the other and likewise below-average values also tend to occur together.
    - Two variables are negatively associated when above-average values of one variable accompany below-average values of the other variable, and vice-versa.
  - Strength-how close the points lie to a line

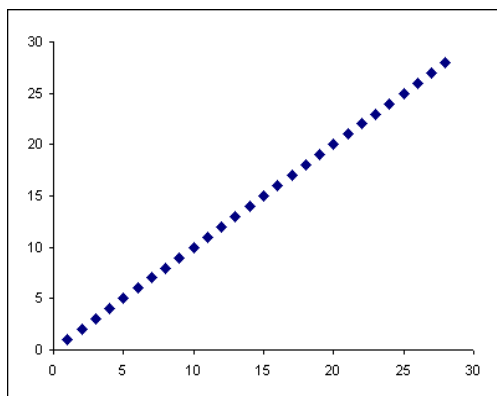


# RELATIONSHIPS BETWEEN 2 NUMERIC VARIABLES

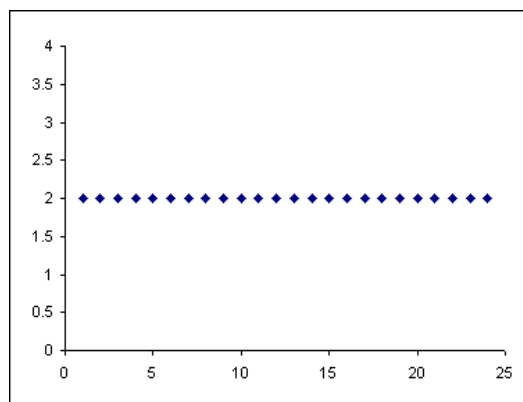
$$\begin{aligned} r &= \frac{1}{n-1} \sum_{i=1}^n \left( \frac{x_i - \bar{x}}{s_x} \right) \left( \frac{y_i - \bar{y}}{s_y} \right) \\ &= \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}} \end{aligned}$$

- Examples of extreme cases

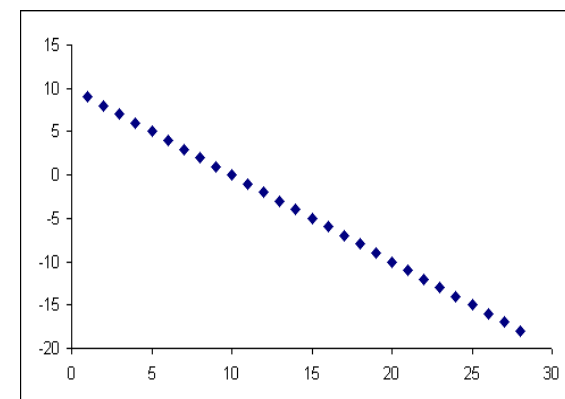
$r = 1$



$r = 0$



$r = -1$





# EXAMPLE FOR CORRELATION

## EXAMPLE 3.16

For the data in Table 3.17, compute the value of the correlation coefficient.

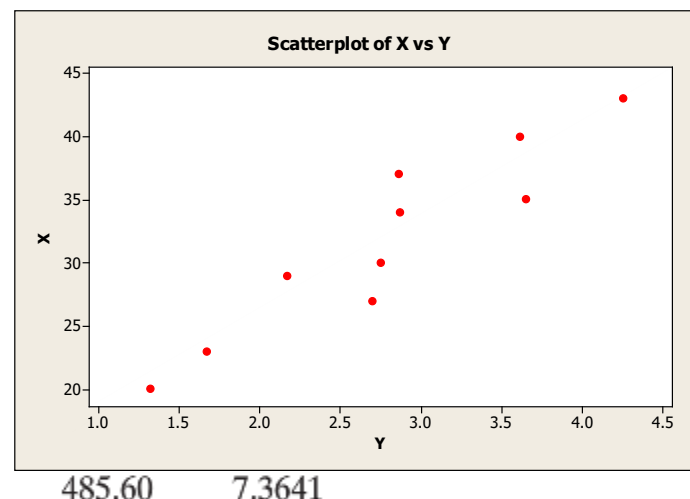
$$\bar{x} = 31.80 \text{ and } \bar{y} = 2.785$$

$$x - \bar{x} = 20 - 31.8 = -11.8, \quad y - \bar{y} = 1.32 - 2.785 = -1.465,$$

$$(x - \bar{x})(y - \bar{y}) = (-11.8)(-1.465) = 17.287,$$

$$(x - \bar{x})^2 = (-11.8)^2 = 139.24, \quad (y - \bar{y})^2 = (-1.465)^2 = 2.14623$$

	$x$	$y$	$x - \bar{x}$	$y - \bar{y}$	$(x - \bar{x})(y - \bar{y})$	$(x - \bar{x})^2$	$(y - \bar{y})^2$
	20	1.32	-11.8	-1.465	17.287		
	23	1.67	-8.8	-1.115	9.812		
	29	2.17	-2.8	-0.615	1.722		
	27	2.70	-4.8	-0.085	0.408		
	30	2.75	-1.8	-0.035	0.063		
	34	2.87	2.2	0.085	0.187		
	35	3.65	3.2	0.865	2.768		
	37	2.86	5.2	0.075	0.390		
	40	3.61	8.2	0.825	6.765		
	43	4.25	11.2	1.465	16.408		
Total	318	27.85	0	0	55.810		
Mean	31.80	2.785					



A form of  $r$  that is somewhat more direct in its calculation is given by

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2 \sum_{i=1}^n (y_i - \bar{y})^2}} = \frac{55.810}{\sqrt{(485.6)(7.3641)}} = .933$$



# RELATIONSHIPS BETWEEN 2 NUMERIC VARIABLES

- **Correlation** or  **$r$**  : measures the direction and strength of the linear relationship between two numeric variables
  - General Properties
    - It must be between -1 and 1, or  $(-1 \leq r \leq 1)$ .
    - If  $r$  is negative, the relationship is negative.
    - If  $r = -1$ , there is a perfect negative linear relationship (extreme case).
    - If  $r$  is positive, the relationship is positive.
    - If  $r = 1$ , there is a perfect positive linear relationship (extreme case).
    - If  $r$  is 0, there is no **linear** relationship.
    - $r$  measures the strength of the **linear** relationship.
    - If explanatory and response are switched,  $r$  remains the same.
    - $r$  has no units of measurement associated with it
    - Scale changes do not affect  $r$
- **Correlation Applet**



Correlation  $r = 0$



Correlation  $r = -0.3$



Correlation  $r = 0.5$



Correlation  $r = -0.7$



Correlation  $r = 0.9$



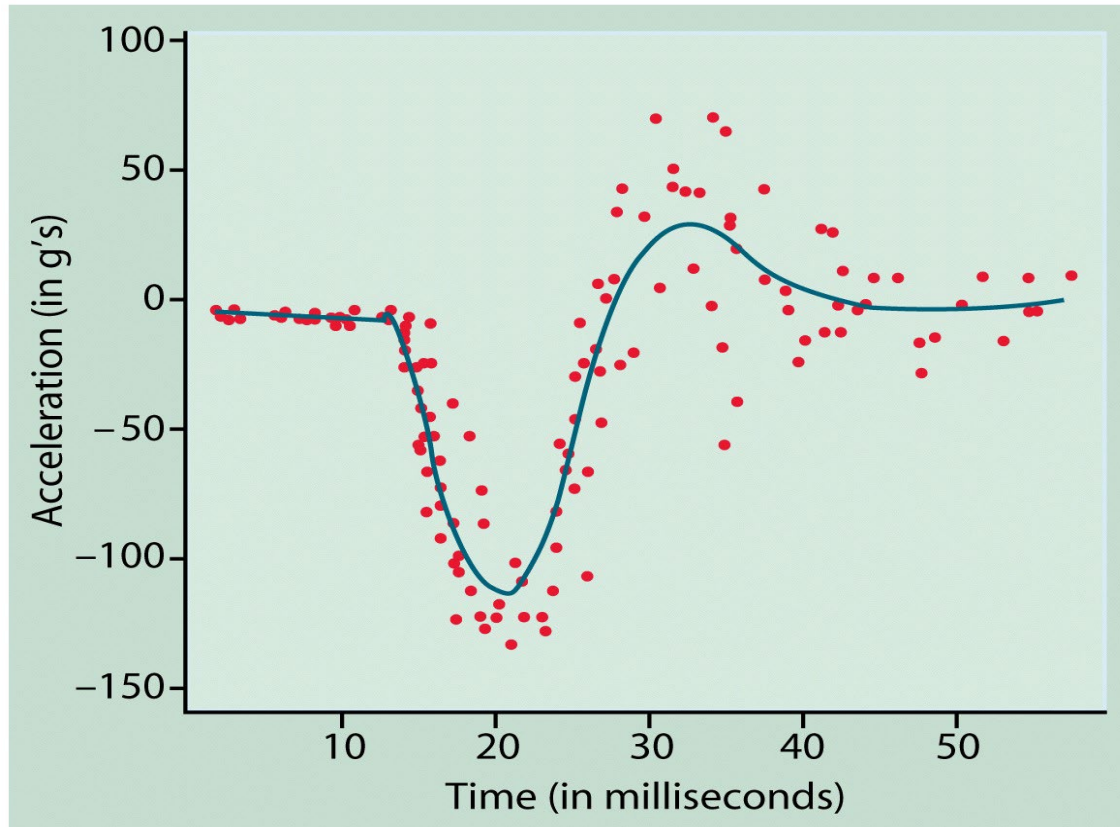
Correlation  $r = -0.99$

# RELATIONSHIPS BETWEEN 2 NUMERIC VARIABLES



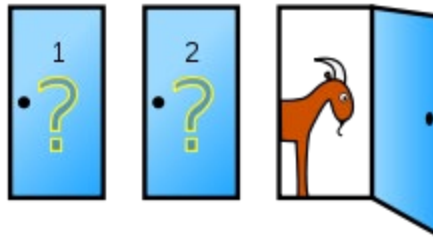
**It is possible for there to be a strong relationship between two variables and still have  $r \approx 0$ .**

**EX.**



# LET'S MAKE A DEAL

- Let's Make a Deal (Monty Hall problem)
  - [http://en.wikipedia.org/wiki/Monty\\_Hall\\_problem](http://en.wikipedia.org/wiki/Monty_Hall_problem)



- This is motivation to study probability.
- Should you switch or should you stay with your original choice?



# BIRTHDAY PARADOX

- What's the chances that two people in our class have the same birthday?

- **R Code:**

- ```
p <- function(n) {  
  p <- NA  
  for (i in 1:length(n)) {  
    p[i] <- prod(365:(365 - (n[i] - 1))) / 365^n[i]  
  }  
  return(p)  
}
```
- ```
plot(n,p(n), col="blue", type="l", lwd=2, xlab= "Number of people")
```
- ```
points(n,1-p(n), col="red", lwd=2, type="l")
```
- ```
abline(v=23, lty=2)
```
- ```
abline(h=0.5, lty=2)
```
- ```
legend("right",c("Probability of a pair","Probability of no maching pair"),  
      lty=1, lwd=2, col=c("red","blue"), cex=2)
```

