



Statistics for Business and Economics

6th Edition

Chapter 3

Describing Data: Numerical



Chapter Goals

After completing this chapter, you should be able to:

- Compute and interpret the **mean, median, and mode** for a set of data
- Find the **range, variance, standard deviation, and coefficient of variation** and know what these values mean
- Apply the **empirical rule** to describe the variation of population values around the mean
- Explain the **weighted mean** and when to use it
- Explain how a **least squares regression line** estimates a linear relationship between two variables



Chapter Topics

- Measures of central tendency, variation, and shape
 - Mean, median, mode, geometric mean
 - Quartiles
 - Range, interquartile range, variance and standard deviation, coefficient of variation
 - Symmetric and skewed distributions
- Population summary measures
 - Mean, variance, and standard deviation
 - The empirical rule and Bienaymé-Chebyshev rule



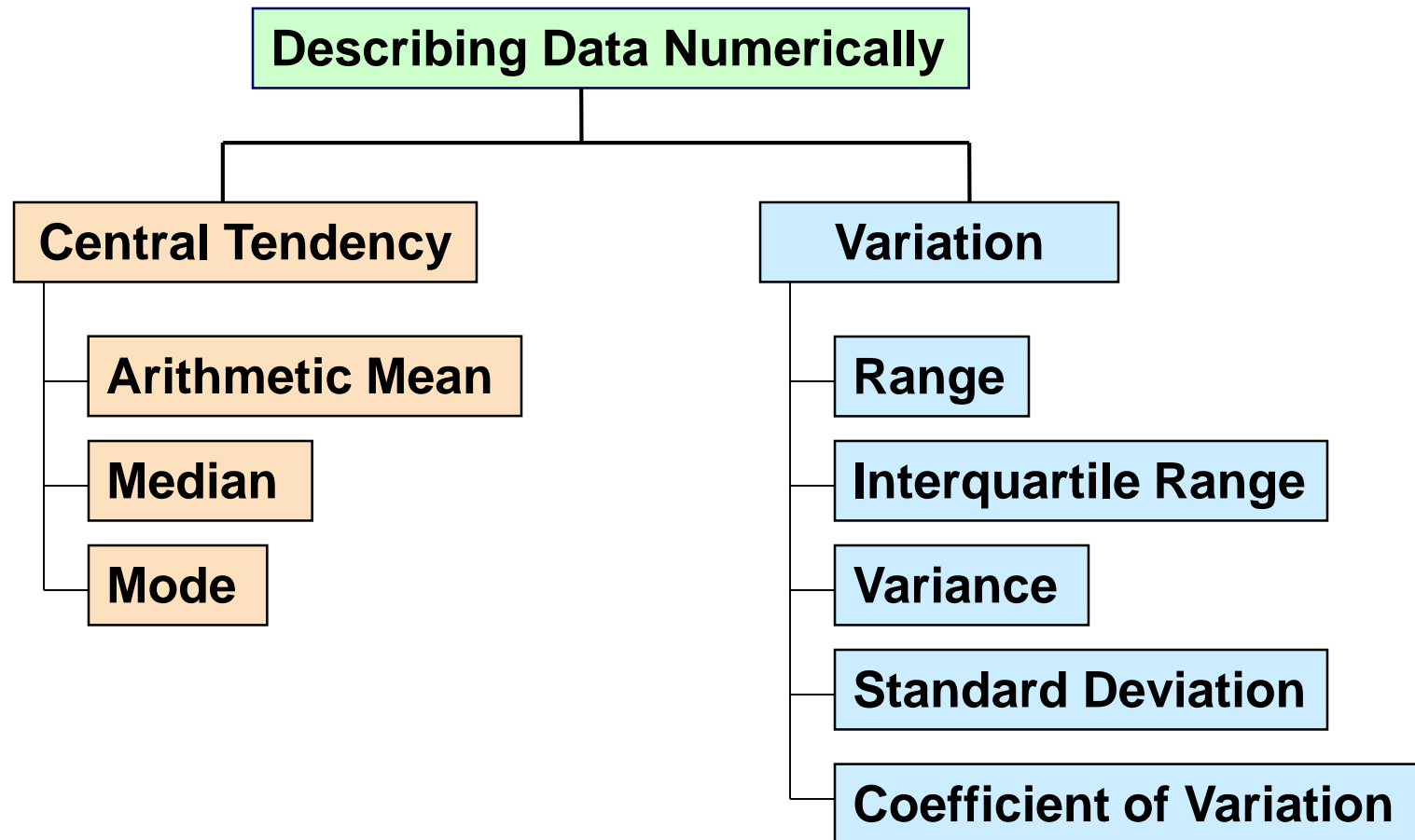
Chapter Topics

(continued)

- Five number summary and box-and-whisker plots
- Covariance and coefficient of correlation
- Pitfalls in numerical descriptive measures and ethical considerations



Describing Data Numerically



Measures of Central Tendency

Overview

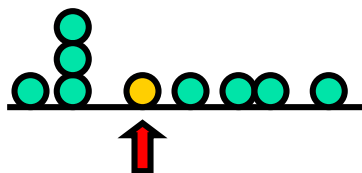
Central Tendency

Mean

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

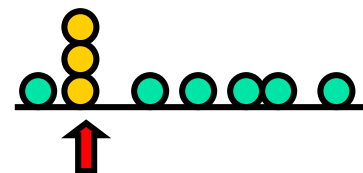
Arithmetic
average

Median



Midpoint of
ranked values

Mode



Most frequently
observed value



Arithmetic Mean

- The arithmetic mean (mean) is the most common measure of central tendency
 - For a population of N values:

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

Population values

Population size

- For a sample of size n :

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

Observed values

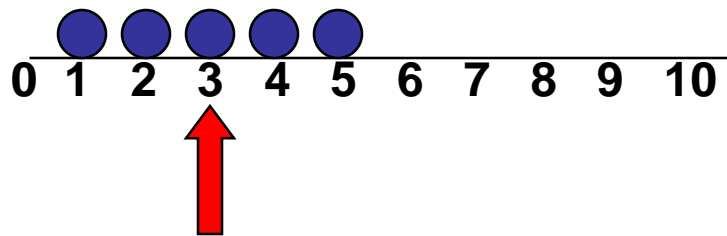
Sample size



Arithmetic Mean

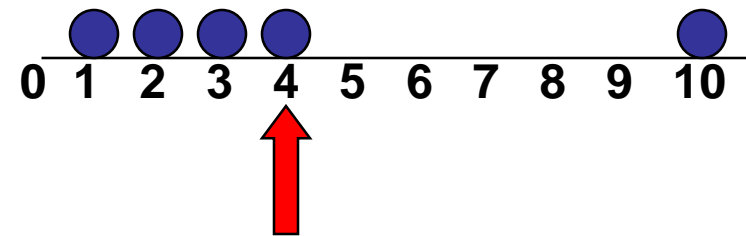
(continued)

- The most common measure of central tendency
- Mean = sum of values divided by the number of values
- Affected by extreme values (outliers)



Mean = 3

$$\frac{1 + 2 + 3 + 4 + 5}{5} = \frac{15}{5} = 3$$



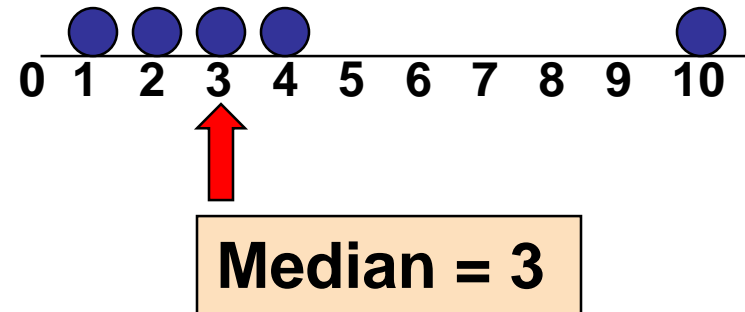
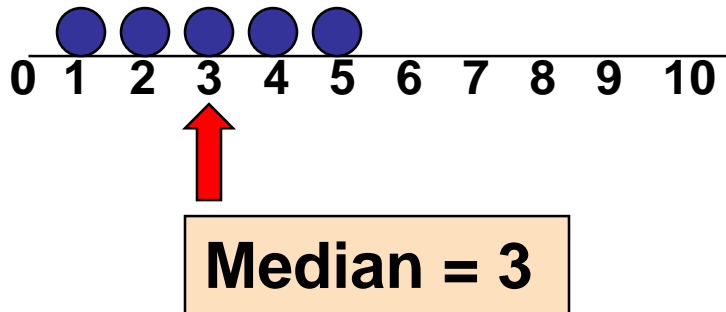
Mean = 4

$$\frac{1 + 2 + 3 + 4 + 10}{5} = \frac{20}{5} = 4$$



Median

- In an ordered list, the median is the “middle” number (50% above, 50% below)



- Not affected by extreme values



Finding the Median

- The location of the median:

$$\text{Median position} = \frac{n+1}{2} \text{ position in the ordered data}$$

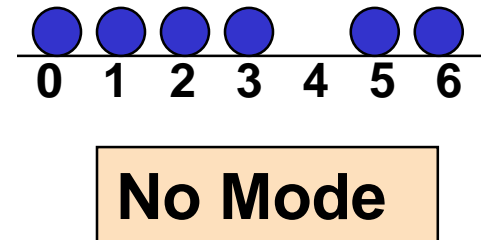
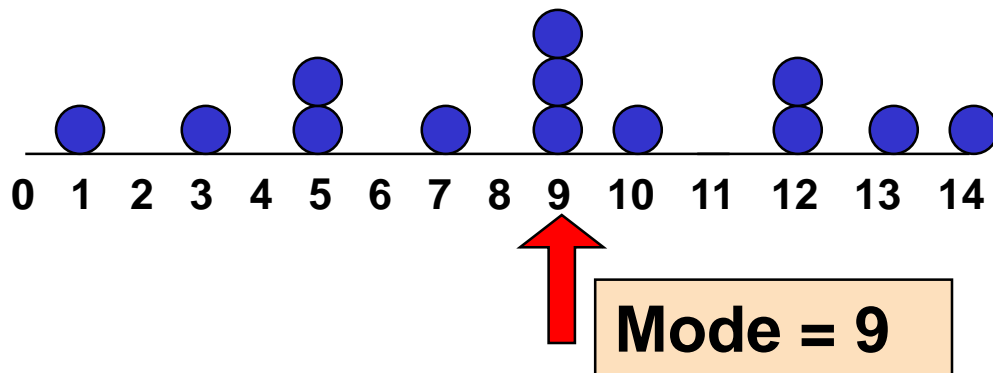
- If the number of values is odd, the median is the middle number
- If the number of values is even, the median is the average of the two middle numbers

- Note that $\frac{n+1}{2}$ is not the *value* of the median, only the *position* of the median in the ranked data



Mode

- A measure of central tendency
- Value that occurs most often
- Not affected by extreme values
- Used for either numerical or categorical data
- There may may be no mode
- There may be several modes



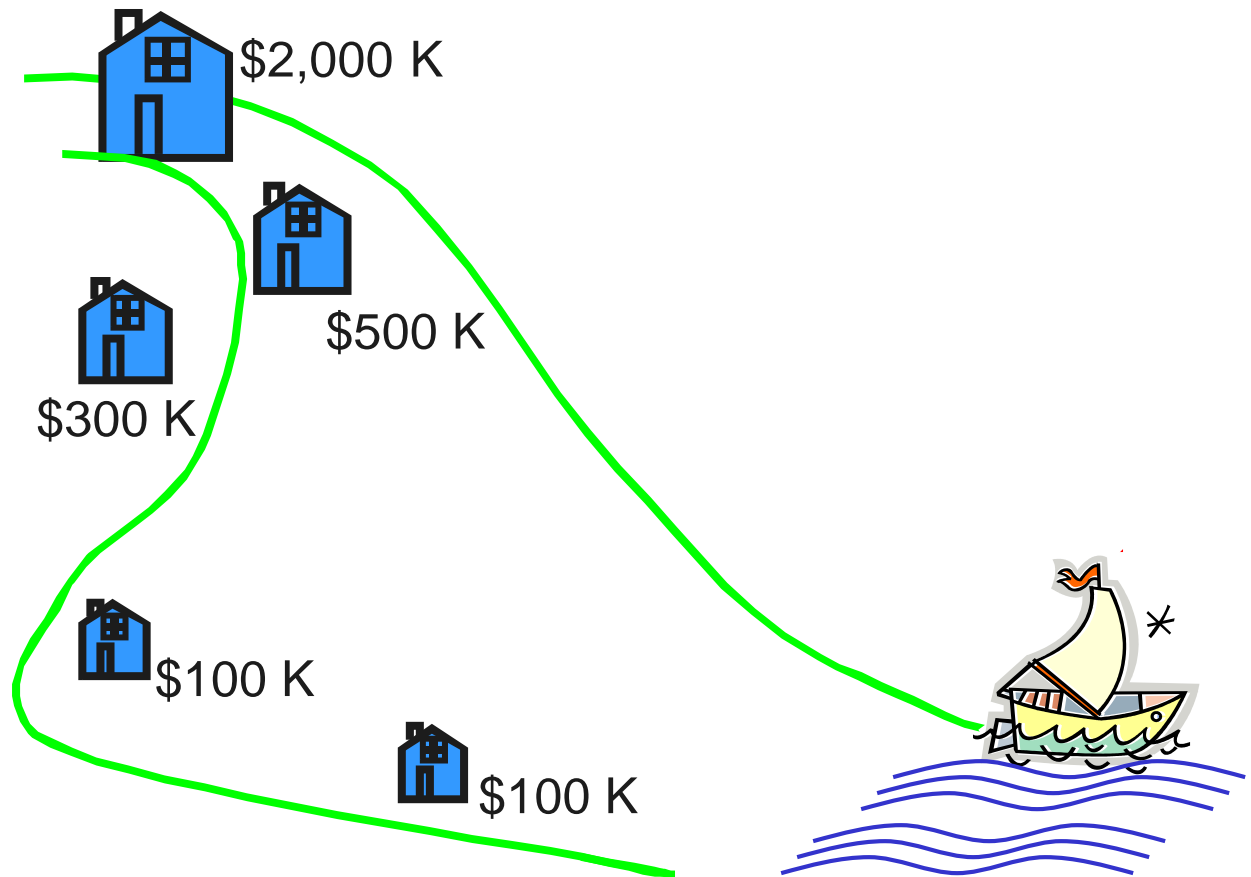


Review Example

- Five houses on a hill by the beach

House Prices:

\$2,000,000
500,000
300,000
100,000
100,000





Review Example: Summary Statistics

House Prices:

\$2,000,000
500,000
300,000
100,000
<u>100,000</u>

Sum 3,000,000

- **Mean:** $(\$3,000,000/5)$
= **\$600,000**
- **Median:** middle value of ranked data
= **\$300,000**
- **Mode:** most frequent value
= **\$100,000**



Which measure of location is the “best”?

- **Mean** is generally used, unless extreme values (outliers) exist
- Then **median** is often used, since the median is not sensitive to extreme values.
 - **Example:** Median home prices may be reported for a region – less sensitive to outliers

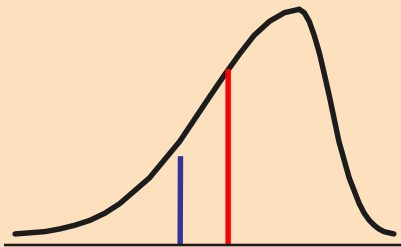


Shape of a Distribution

- Describes how data are distributed
- Measures of **shape**
 - Symmetric or skewed

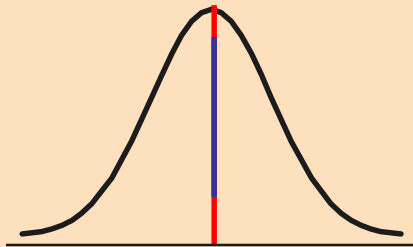
Left-Skewed

Mean < Median



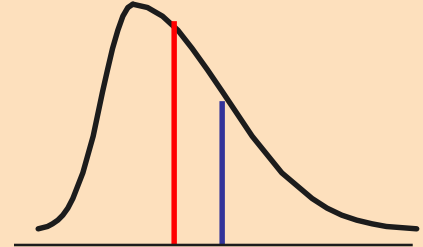
Symmetric

Mean = Median



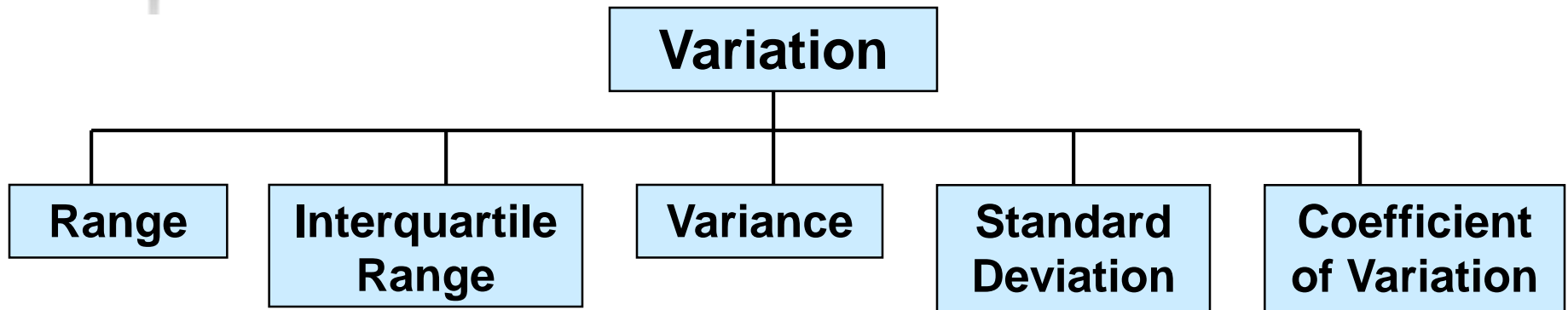
Right-Skewed

Median < Mean

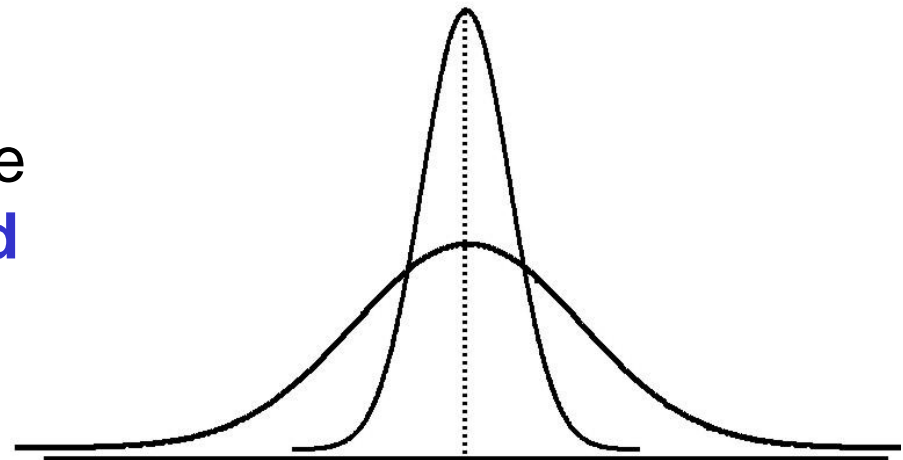




Measures of Variability



- Measures of variation give information on the **spread** or **variability** of the data values.



**Same center,
different variation**

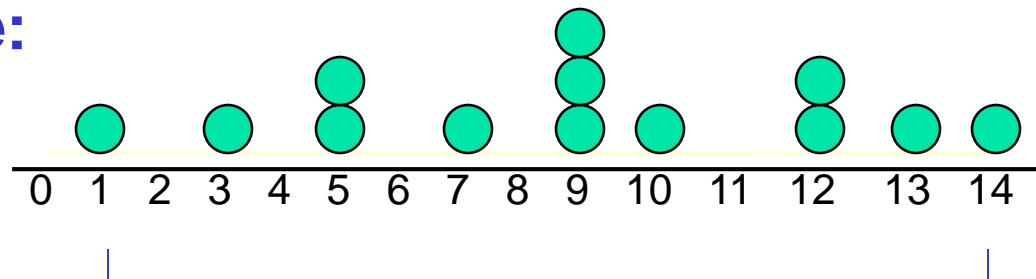


Range

- Simplest measure of variation
- Difference between the largest and the smallest observations:

$$\text{Range} = X_{\text{largest}} - X_{\text{smallest}}$$

Example:

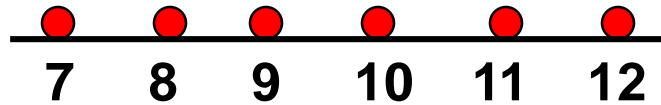


$$\text{Range} = 14 - 1 = 13$$

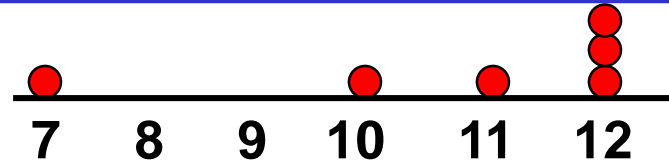


Disadvantages of the Range

- Ignores the way in which data are distributed



$$\text{Range} = 12 - 7 = 5$$



$$\text{Range} = 12 - 7 = 5$$

- Sensitive to outliers

1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 3, 3, 3, 3, 4, 5

$$\text{Range} = 5 - 1 = 4$$

1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 3, 3, 3, 3, 4, 120

$$\text{Range} = 120 - 1 = 119$$



Interquartile Range

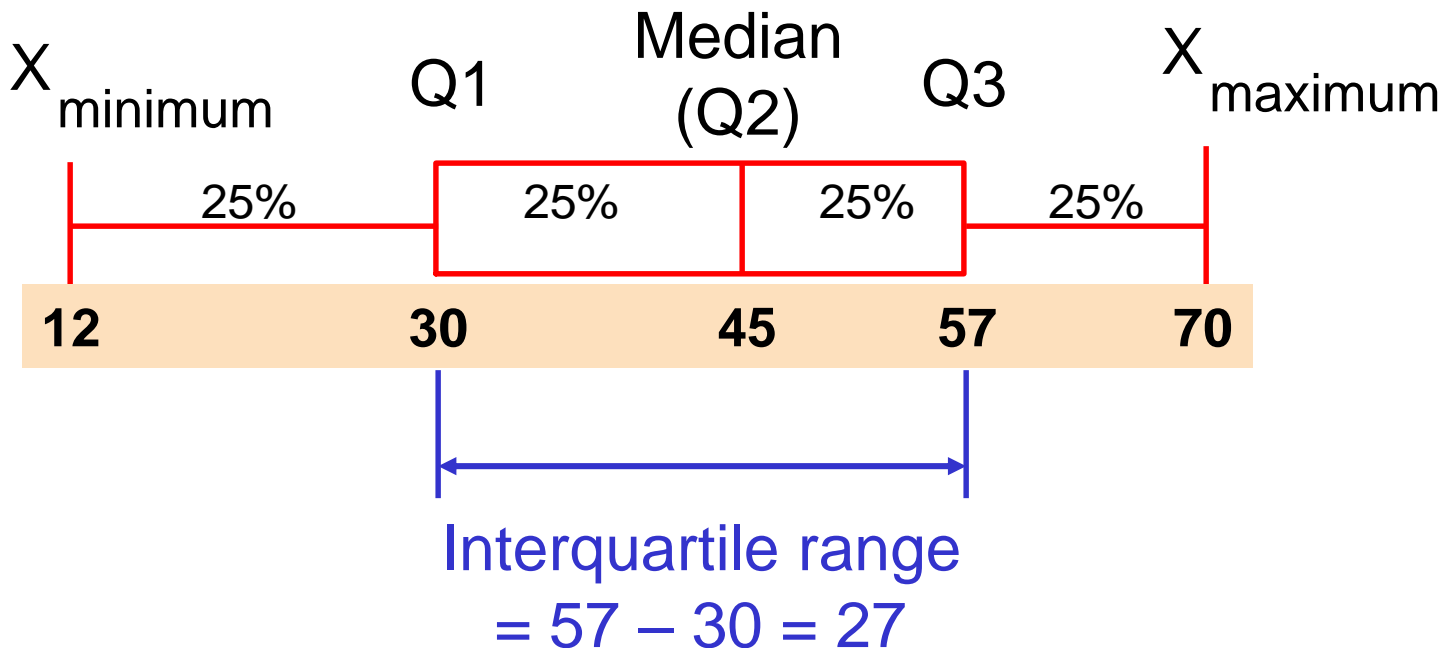
- Can eliminate some outlier problems by using the **interquartile range**
- Eliminate high- and low-valued observations and calculate the range of the middle 50% of the data

- Interquartile range = 3rd quartile – 1st quartile
$$\text{IQR} = Q_3 - Q_1$$



Interquartile Range

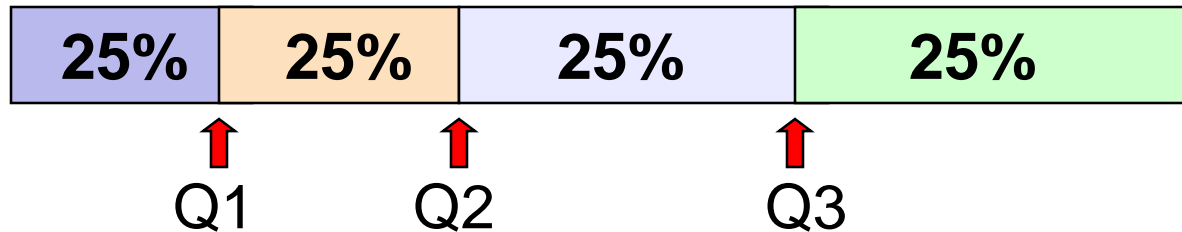
Example:





Quartiles

- Quartiles split the ranked data into 4 segments with an equal number of values per segment



- The first quartile, Q_1 , is the value for which 25% of the observations are smaller and 75% are larger
- Q_2 is the same as the median (50% are smaller, 50% are larger)
- Only 25% of the observations are greater than the third quartile



Quartile Formulas

Find a quartile by determining the value in the appropriate position in the ranked data, where

First quartile position: $Q_1 = 0.25(n+1)$

Second quartile position: $Q_2 = 0.50(n+1)$
(the median position)

Third quartile position: $Q_3 = 0.75(n+1)$

where n is the number of observed values



Quartiles

■ Example: Find the first quartile

Sample Ranked Data: 11 12 13 16 16 17 18 21 22



($n = 9$)

Q_1 = is in the $0.25(9+1) = 2.5$ position of the ranked data
so use the value half way between the 2nd and 3rd values,

so

$$Q_1 = 12.5$$



Population Variance

- Average of squared deviations of values from the mean

- Population variance:

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

Where μ = population mean

N = population size

x_i = i^{th} value of the variable x



Sample Variance

- Average (approximately) of squared deviations of values from the mean

- Sample variance:

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

Where \bar{X} = arithmetic mean

n = sample size

X_i = i^{th} value of the variable X



Population Standard Deviation

- Most commonly used measure of variation
- Shows variation about the mean
- Has the **same units as the original data**

- Population standard deviation:

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



Sample Standard Deviation

- Most commonly used measure of variation
- Shows variation about the mean
- Has the **same units as the original data**

- Sample standard deviation:

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



Calculation Example: Sample Standard Deviation

Sample

Data (x_i) :

10 12 14 15 17 18 18 24

$n = 8$

Mean = $\bar{x} = 16$

$$s = \sqrt{\frac{(10 - \bar{x})^2 + (12 - \bar{x})^2 + (14 - \bar{x})^2 + \cdots + (24 - \bar{x})^2}{n - 1}}$$

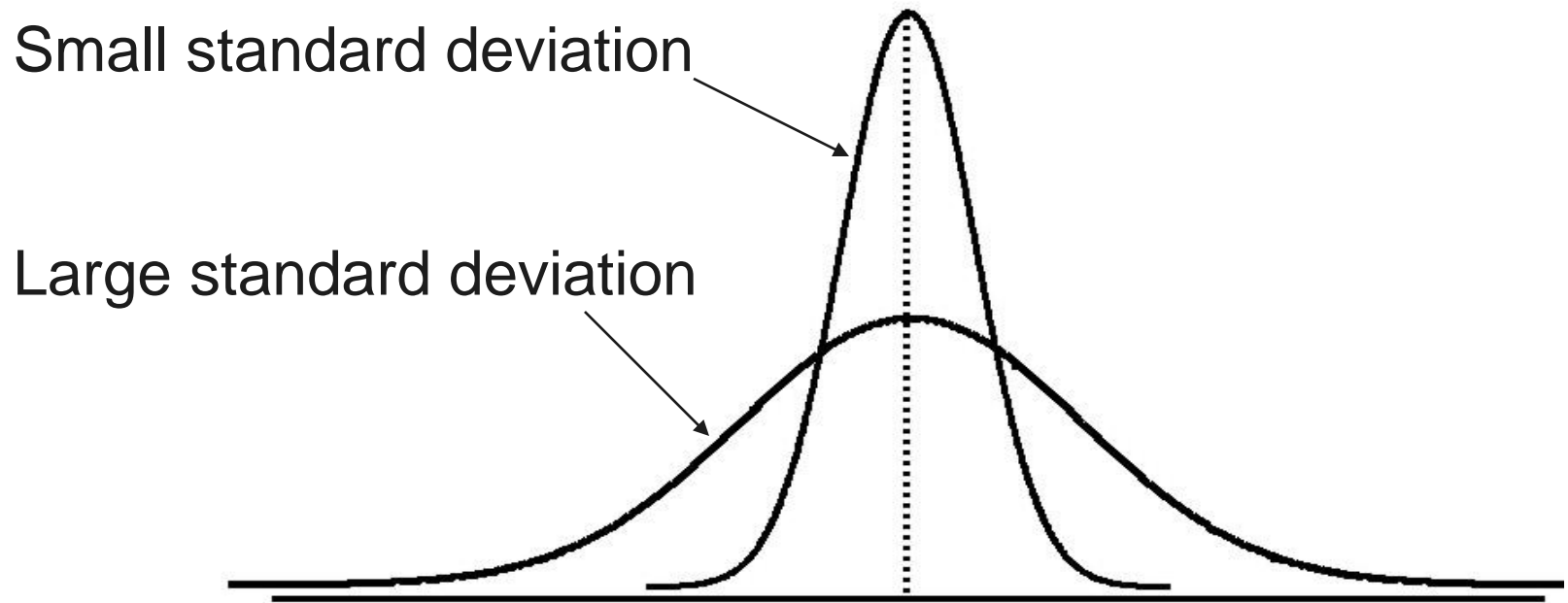
$$= \sqrt{\frac{(10 - 16)^2 + (12 - 16)^2 + (14 - 16)^2 + \cdots + (24 - 16)^2}{8 - 1}}$$

$$= \sqrt{\frac{126}{7}} = \boxed{4.2426} \rightarrow$$

A measure of the “average”
scatter around the mean



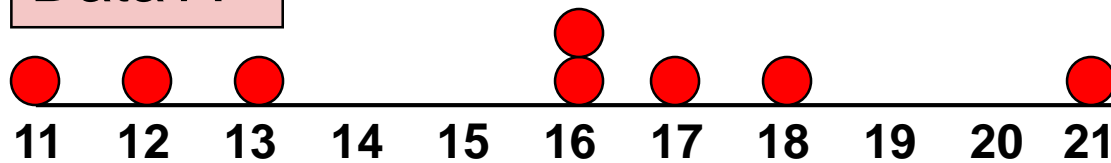
Measuring variation





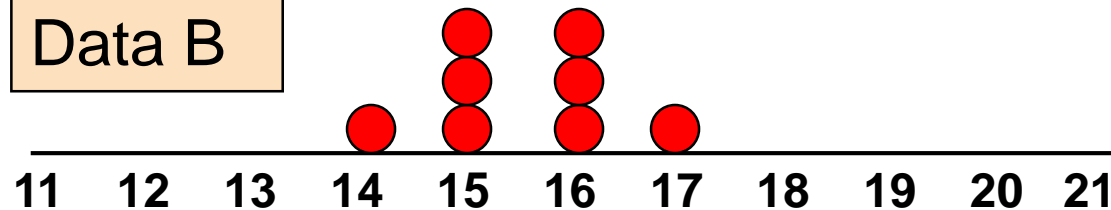
Comparing Standard Deviations

Data A



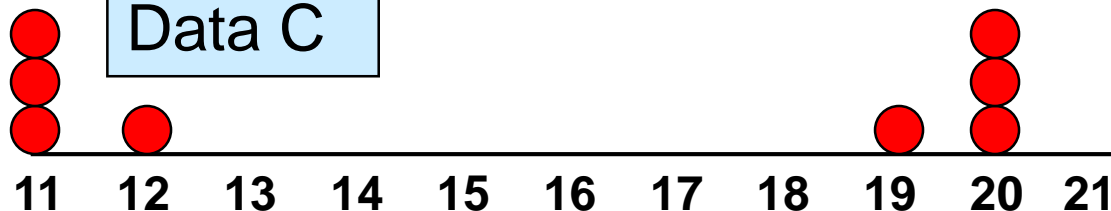
Mean = 15.5
 $s = 3.338$

Data B



Mean = 15.5
 $s = 0.926$

Data C



Mean = 15.5
 $s = 4.570$



Advantages of Variance and Standard Deviation

- Each value in the data set is used in the calculation
- Values far from the mean are given extra weight
(because deviations from the mean are squared)



Chebyshev's Theorem

- For any population with mean μ and standard deviation σ , and $k > 1$, the percentage of observations that fall within the interval

$$[\mu + k\sigma]$$

Is *at least*

$$100[1 - (1/k^2)]\%$$



Chebyshev's Theorem

(continued)

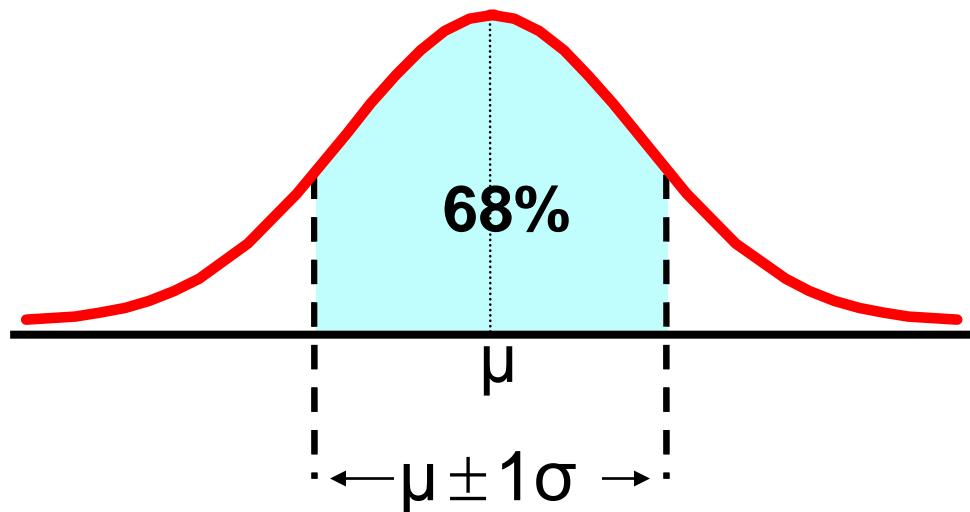
- Regardless of how the data are distributed, at least $(1 - 1/k^2)$ of the values will fall within k standard deviations of the mean (for $k > 1$)
- Examples:

At least	within
$(1 - 1/1^2) = 0\%$	$k=1 \quad (\mu \pm 1\sigma)$
$(1 - 1/2^2) = 75\%$	$k=2 \quad (\mu \pm 2\sigma)$
$(1 - 1/3^2) = 89\%$	$k=3 \quad (\mu \pm 3\sigma)$



The Empirical Rule

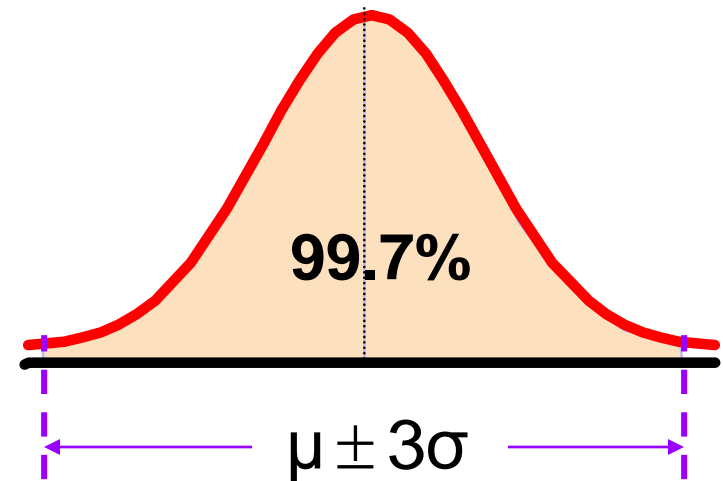
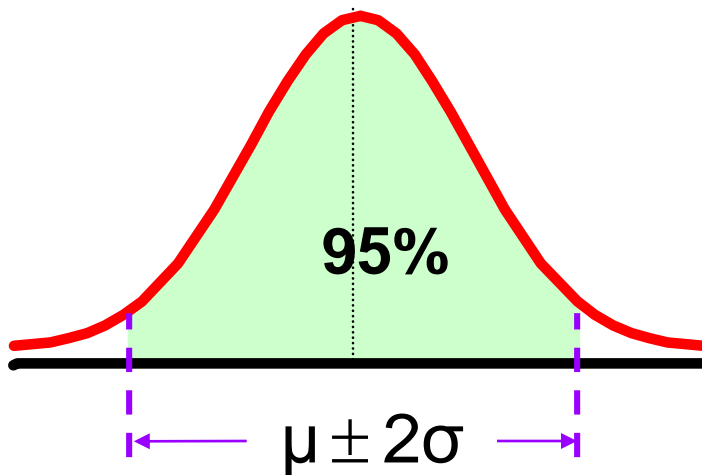
- If the data distribution is bell-shaped, then the interval:
- $\mu \pm 1\sigma$ contains about 68% of the values in the population or the sample





The Empirical Rule

- $\mu \pm 2\sigma$ contains about 95% of the values in the population or the sample
- $\mu \pm 3\sigma$ contains about 99.7% of the values in the population or the sample





Coefficient of Variation

- Measures **relative variation**
- Always in percentage (%)
- Shows **variation relative to mean**
- Can be used to compare two or more sets of data measured in different units

$$CV = \left(\frac{s}{\bar{x}} \right) \cdot 100\%$$



Comparing Coefficient of Variation

■ Stock A:

- Average price last year = \$50
- Standard deviation = \$5

$$CV_A = \left(\frac{s}{\bar{x}} \right) \cdot 100\% = \frac{\$5}{\$50} \cdot 100\% = 10\%$$

■ Stock B:

- Average price last year = \$100
- Standard deviation = \$5

$$CV_B = \left(\frac{s}{\bar{x}} \right) \cdot 100\% = \frac{\$5}{\$100} \cdot 100\% = 5\%$$

Both stocks have the same standard deviation, but stock B is less variable relative to its price

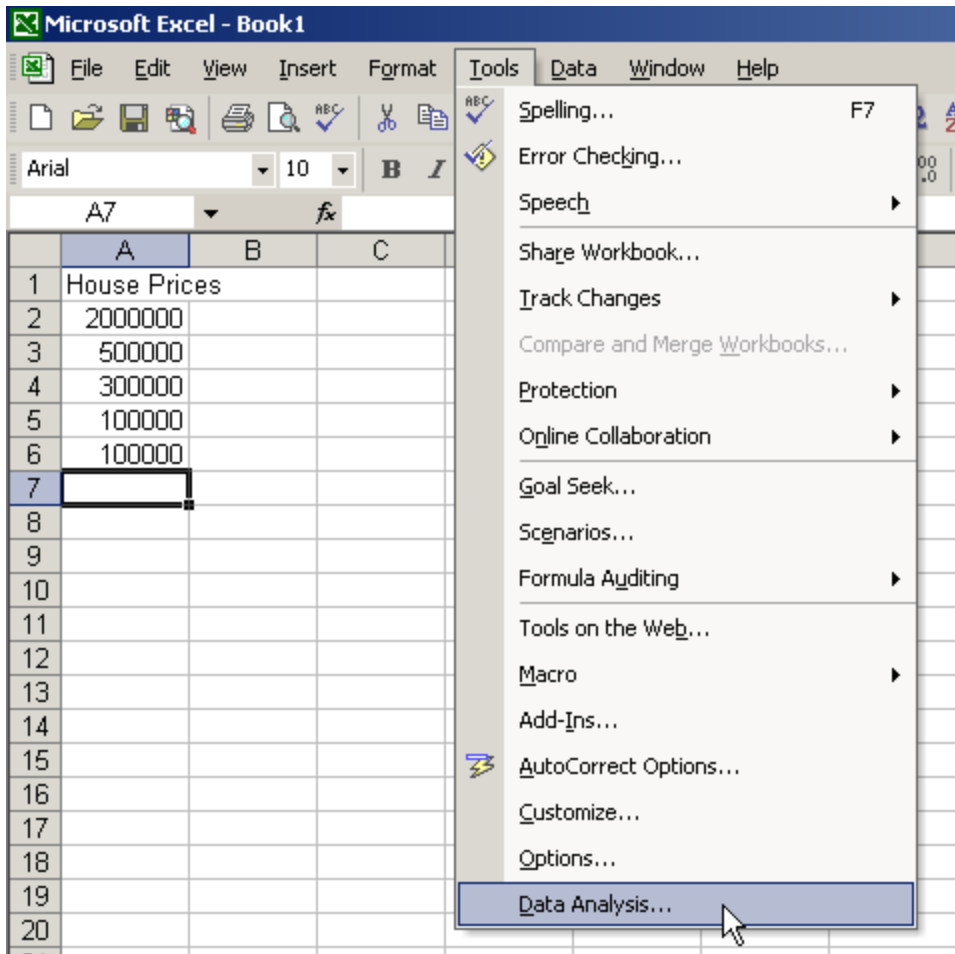


Using Microsoft Excel

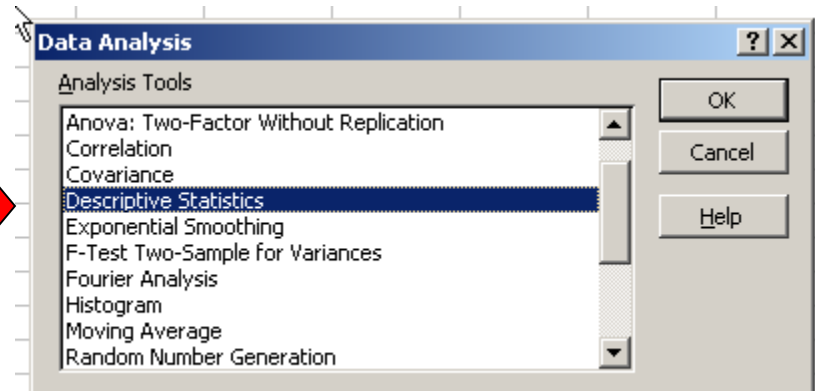
- Descriptive Statistics can be obtained from Microsoft® Excel
 - Use menu choice:
tools / data analysis / descriptive statistics
 - Enter details in dialog box



Using Excel



- Use menu choice:
tools / data analysis /
descriptive statistics

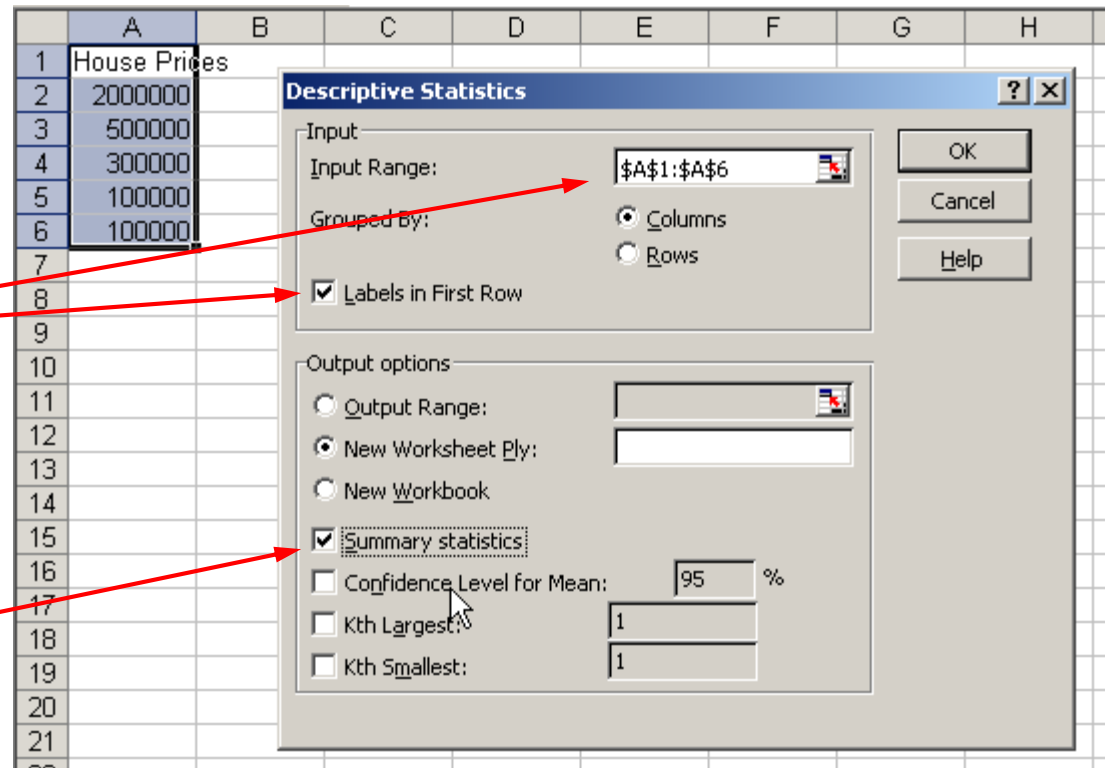




Using Excel

(continued)

- Enter dialog box details
- Check box for summary statistics
- Click OK





Excel output

Microsoft Excel
descriptive statistics output,
using the house price data:

House Prices:

\$2,000,000
500,000
300,000
100,000
100,000

	A	B
1	<i>House Prices</i>	
2		
3	Mean	600000
4	Standard Error	357770.8764
5	Median	300000
6	Mode	100000
7	Standard Deviation	800000
8	Sample Variance	6.4E+11
9	Kurtosis	4.130126953
10	Skewness	2.006835938
11	Range	1900000
12	Minimum	100000
13	Maximum	2000000
14	Sum	3000000
15	Count	5
16		
17		



Weighted Mean

- The **weighted mean** of a set of data is

$$\bar{x} = \frac{\sum_{i=1}^n w_i x_i}{\sum w} = \frac{w_1 x_1 + w_2 x_2 + \cdots + w_n x_n}{\sum w_i}$$

- Where w_i is the weight of the i^{th} observation
- Use when data is already grouped into n classes, with w_i values in the i^{th} class



Approximations for Grouped Data

Suppose a data set contains values m_1, m_2, \dots, m_k , occurring with frequencies f_1, f_2, \dots, f_k

- For a **population** of N observations the mean is

$$\mu = \frac{\sum_{i=1}^K f_i m_i}{N}$$

where $N = \sum_{i=1}^K f_i$

- For a **sample** of n observations, the mean is

$$\bar{x} = \frac{\sum_{i=1}^K f_i m_i}{n}$$

where $n = \sum_{i=1}^K f_i$



Approximations for Grouped Data

Suppose a data set contains values m_1, m_2, \dots, m_k , occurring with frequencies f_1, f_2, \dots, f_k

- For a **population** of N observations the variance is

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

- For a **sample** of n observations, the variance is

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



The Sample Covariance

- The covariance measures the strength of the linear relationship between **two variables**
- The **population covariance**:

$$\text{Cov}(x, y) = \sigma_{xy} = \frac{\sum_{i=1}^N (x_i - \mu_x)(y_i - \mu_y)}{N}$$

- The **sample covariance**:

$$\text{Cov}(x, y) = s_{xy} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{n-1}$$

- Only concerned with the strength of the relationship
- No causal effect is implied



Interpreting Covariance

■ **Covariance** between two variables:

$\text{Cov}(x,y) > 0 \rightarrow$ x and y tend to move in the **same** direction

$\text{Cov}(x,y) < 0 \rightarrow$ x and y tend to move in **opposite** directions

$\text{Cov}(x,y) = 0 \rightarrow$ x and y are independent



Coefficient of Correlation

- Measures the relative strength of the linear relationship between two variables
- Population correlation coefficient:

$$\rho = \frac{\text{Cov}(x, y)}{\sigma_x \sigma_y}$$

- Sample correlation coefficient:

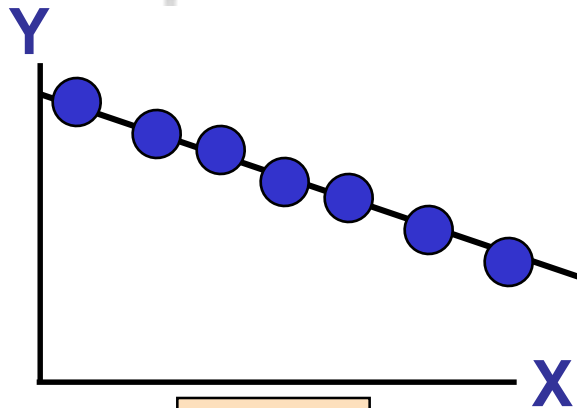
$$r = \frac{\text{Cov}(x, y)}{s_x s_y}$$



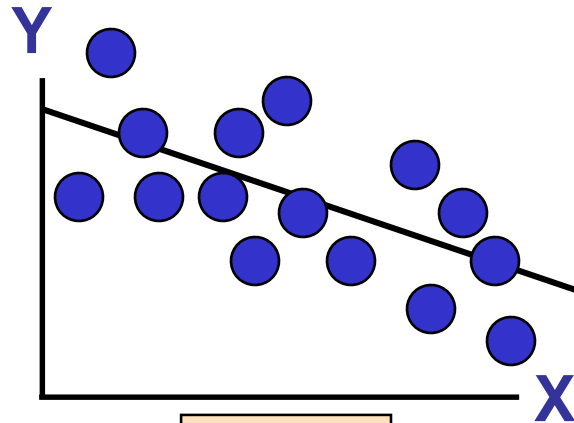
Features of Correlation Coefficient, r

- Unit free
- Ranges between -1 and 1
- The closer to -1 , the stronger the negative linear relationship
- The closer to 1 , the stronger the positive linear relationship
- The closer to 0 , the weaker any positive linear relationship

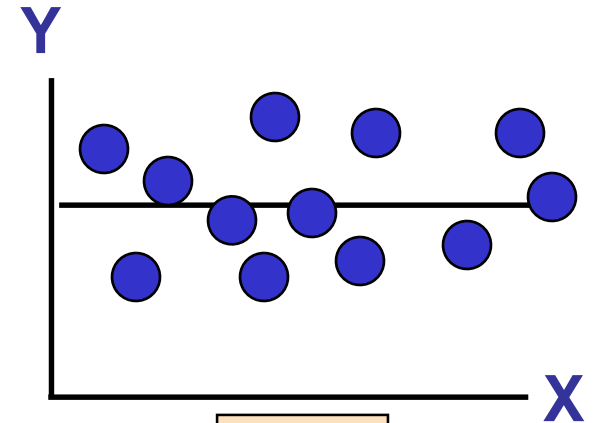
Scatter Plots of Data with Various Correlation Coefficients



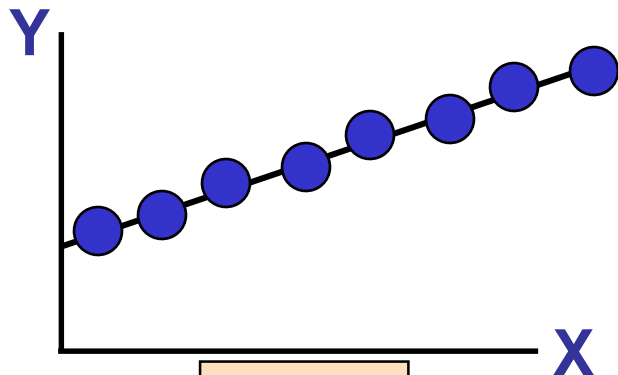
$r = -1$



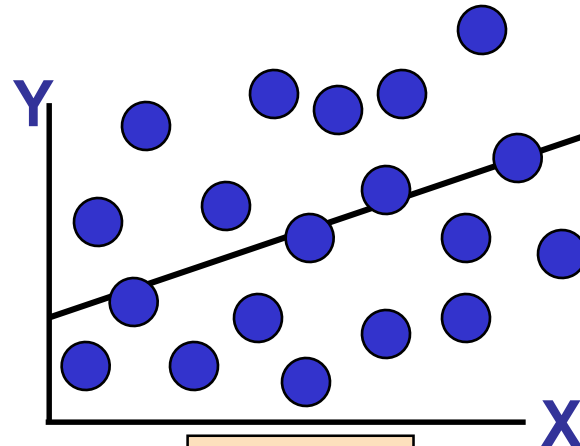
$r = -.6$



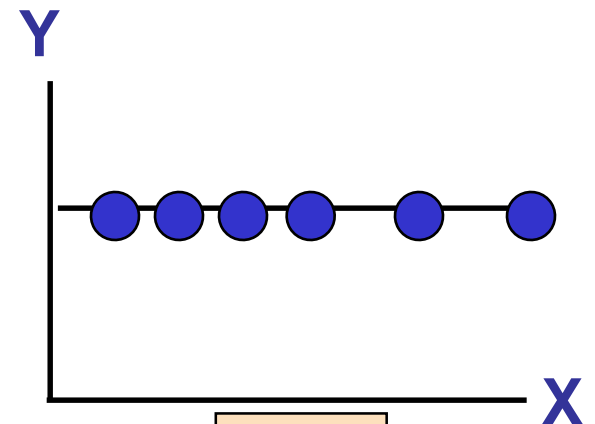
$r = 0$



$r = +1$

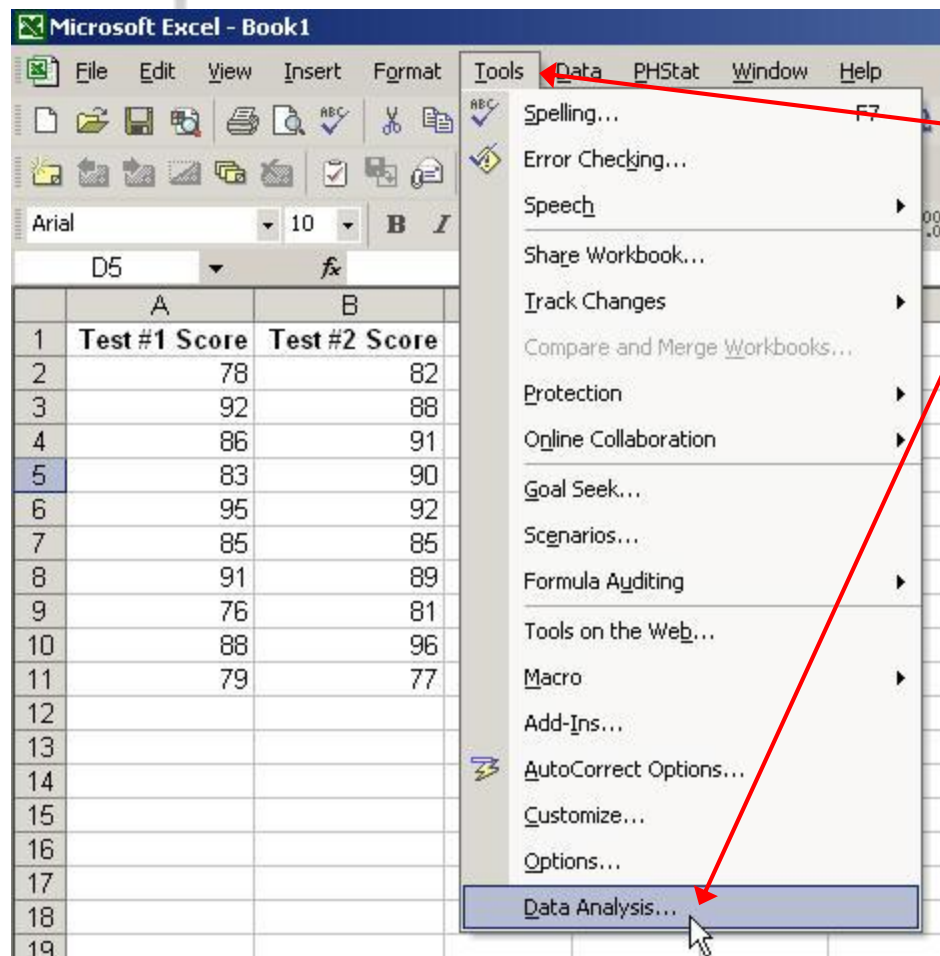


$r = +.3$

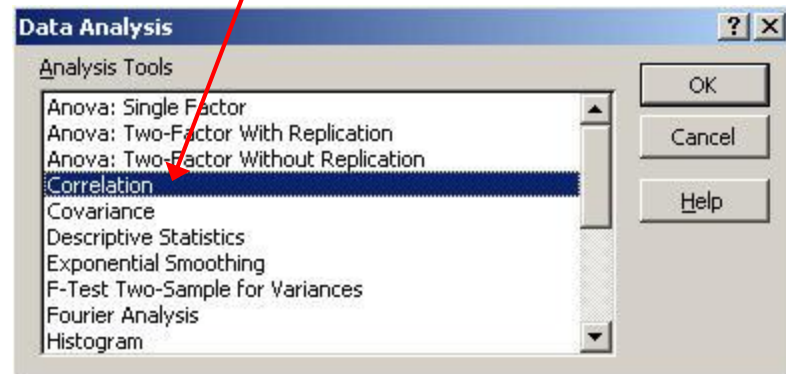


$r = 0$

Using Excel to Find the Correlation Coefficient



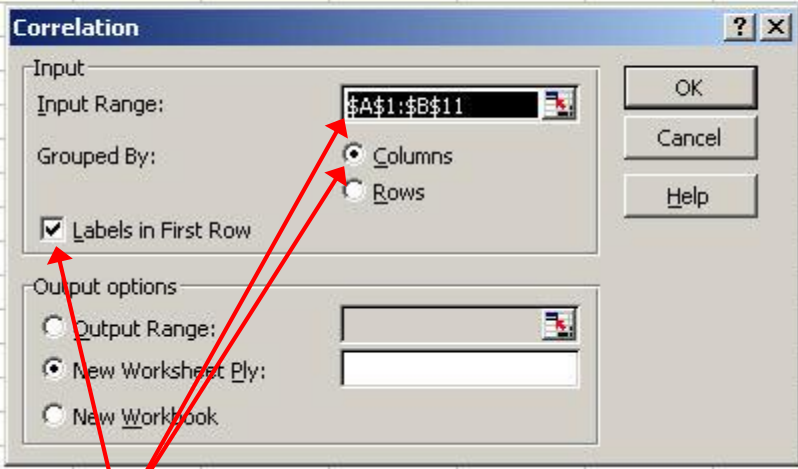
- Select **Tools/Data Analysis**
- Choose **Correlation** from the selection menu
- Click OK . . .



Using Excel to Find the Correlation Coefficient

(continued)

	A	B	C	D	E	F	G	H	I
1	Test #1 Score	Test #2 Score							
2	78	82							
3	92	88							
4	86	91							
5	83	90							
6	95	92							
7	85	85							
8	91	89							
9	76	81							
10	88	96							
11	79	77							
12									
13									
14									
15									
16									
17									
18									



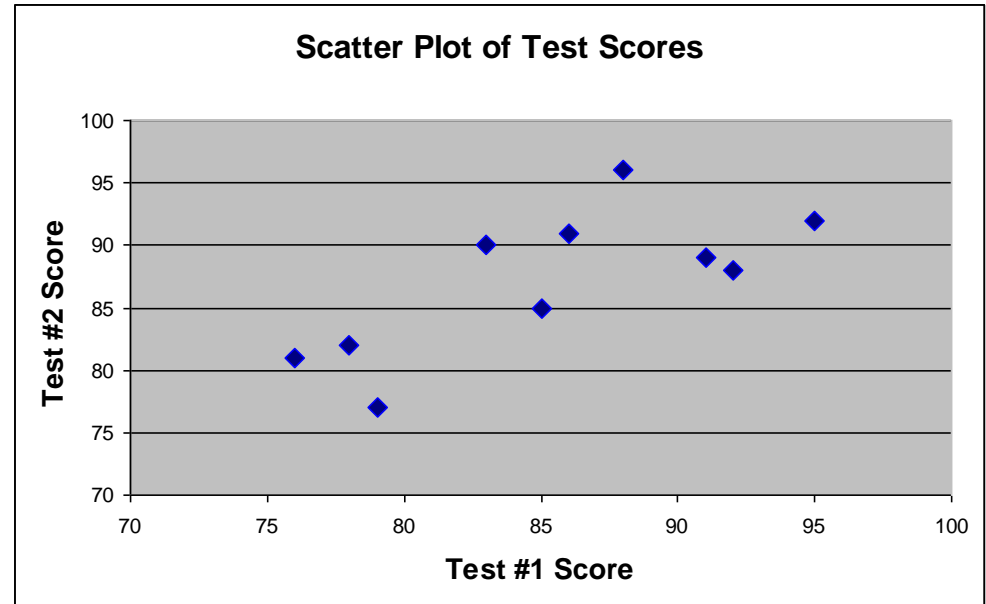
- Input data range and select appropriate options
- Click OK to get output

	A	B	C
1		Test #1 Score	Test #2 Score
2	Test #1 Score	1	
3	Test #2 Score	0.733243705	1
4			



Interpreting the Result

- $r = .733$
- There is a **relatively strong positive linear relationship** between test score #1 and test score #2
- Students who scored high on the first test tended to score high on second test





Obtaining Linear Relationships

- An equation can be fit to show the best linear relationship between two variables:

$$Y = \beta_0 + \beta_1 X$$

Where Y is the **dependent variable** and X is the **independent variable**



Least Squares Regression

- Estimates for coefficients β_0 and β_1 are found to minimize the sum of the squared residuals
- The least-squares regression line, based on sample data, is

$$\hat{y} = b_0 + b_1x$$

- Where b_1 is the slope of the line and b_0 is the y-intercept:

$$b_1 = \frac{\text{Cov}(x, y)}{s_x^2} = r \frac{s_y}{s_x}$$

$$b_0 = \bar{y} - b_1\bar{x}$$



Chapter Summary

- Described measures of central tendency
 - Mean, median, mode
- Illustrated the shape of the distribution
 - Symmetric, skewed
- Described measures of variation
 - Range, interquartile range, variance and standard deviation, coefficient of variation
- Discussed measures of grouped data
- Calculated measures of relationships between variables
 - covariance and correlation coefficient