

# Statistics for Business and Economics



## **Chapter 2**

### **Describing Data: Numerical**

# Measures of Central Tendency

## Overview

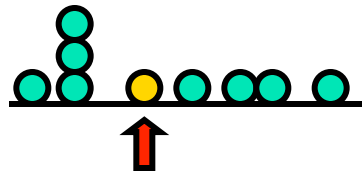
### Central Tendency

Mean

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

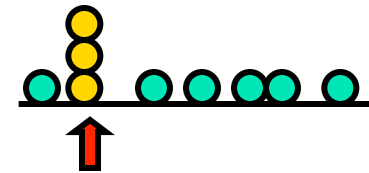
Arithmetic  
average

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:

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

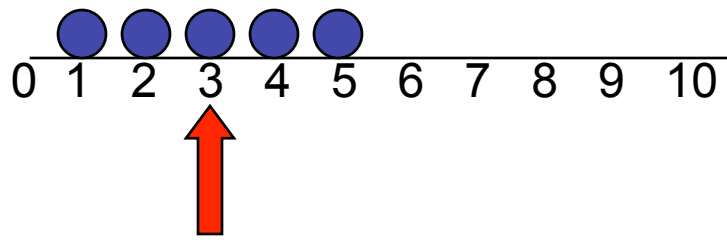
Observed values

Sample size

# Arithmetic Mean

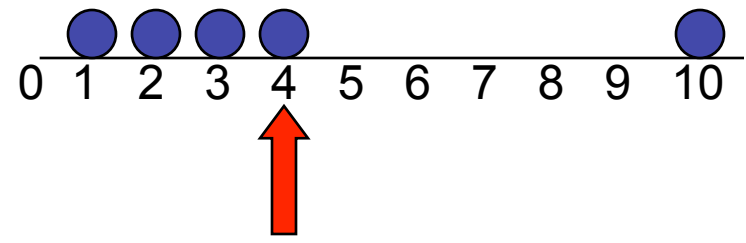
(continued)

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



Mean = 3

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

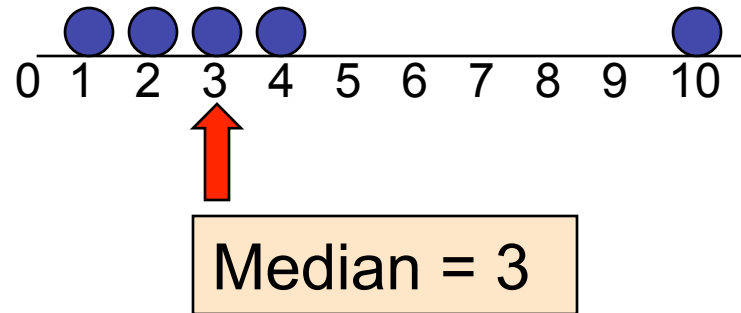
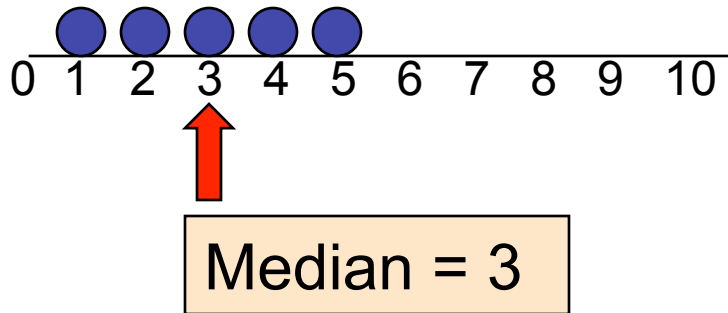


Mean = 4

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

# 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:

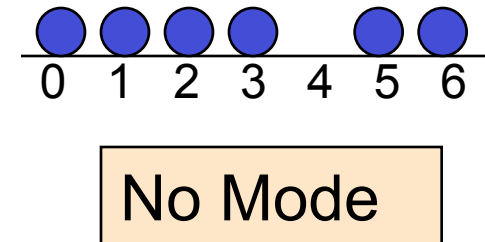
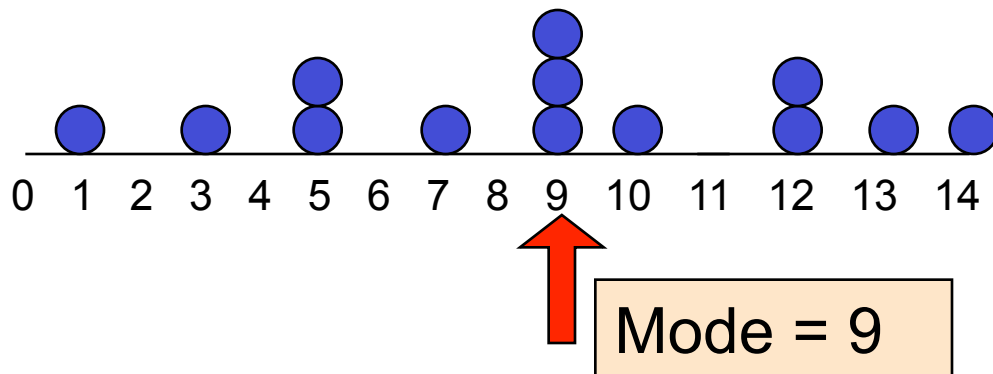
$$\text{Median position} = \frac{n+1}{2} \text{ position in the ordered data}$$

- 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

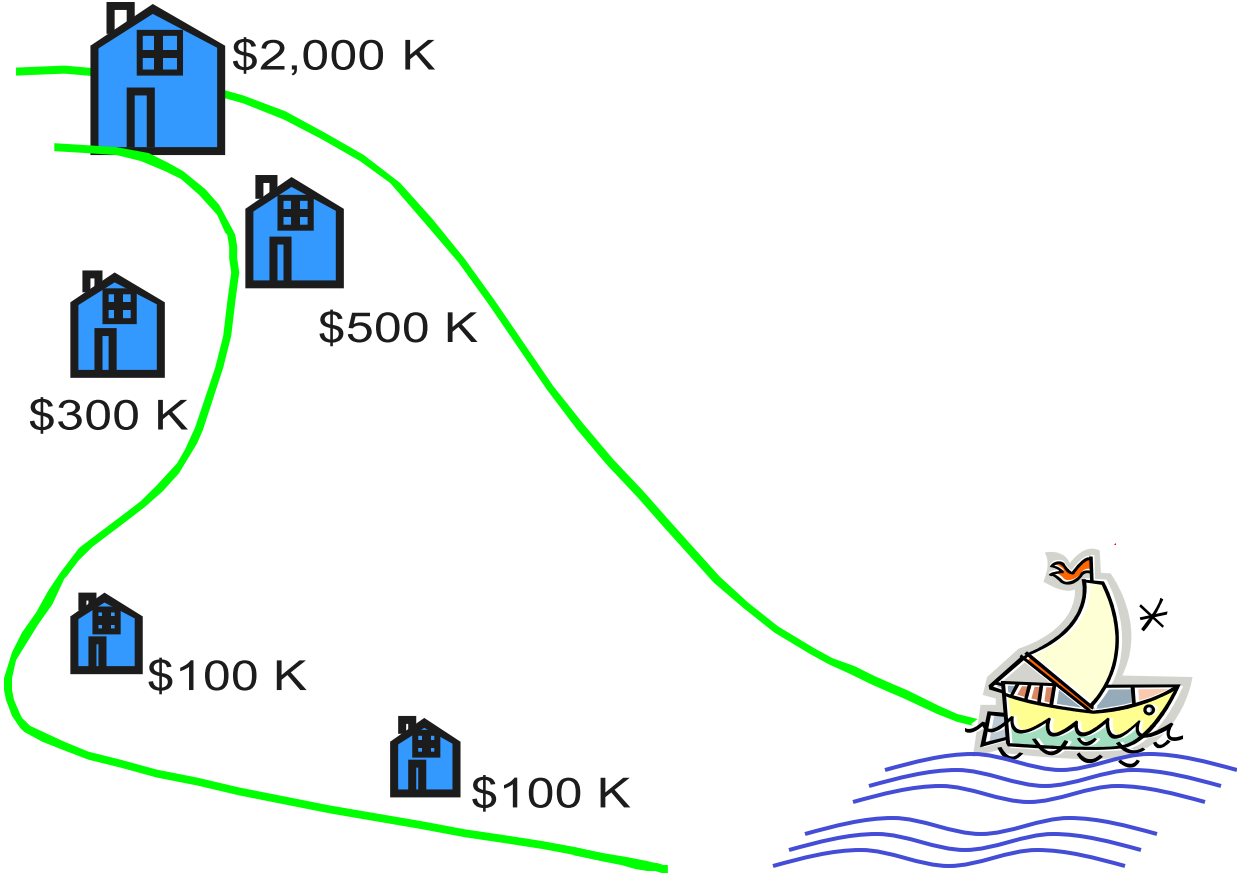


# Review Example

- 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”?

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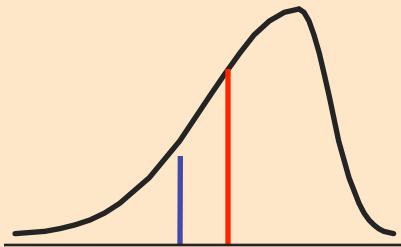
- **Mean** is generally used, unless extreme values (outliers) exist . . .
- Then **median** is often used, since the median is not sensitive to extreme values.
  - **Example:** Median home prices may be reported for a region – less sensitive to outliers

# Shape of a Distribution

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

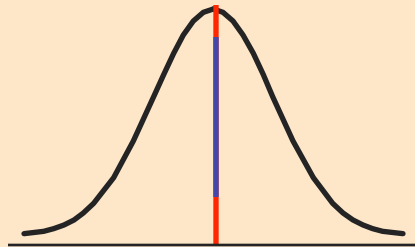
## Left-Skewed

Mean < Median



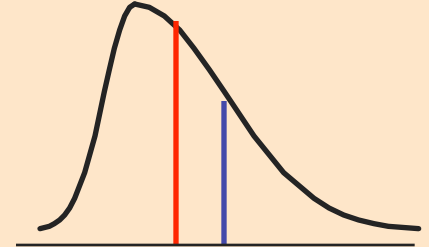
## Symmetric

Mean = Median



## Right-Skewed

Median < Mean





# Geometric Mean

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- Geometric mean

$$\bar{X}_g = \sqrt[n]{(X_1 \times X_2 \times \cdots \times X_n)} = (X_1 \times X_2 \times \cdots \times X_n)^{1/n}$$

- Geometric mean rate of return

$$\bar{r}_g = (X_1 \times X_2 \times \cdots \times X_n)^{1/n} - 1$$



# Example

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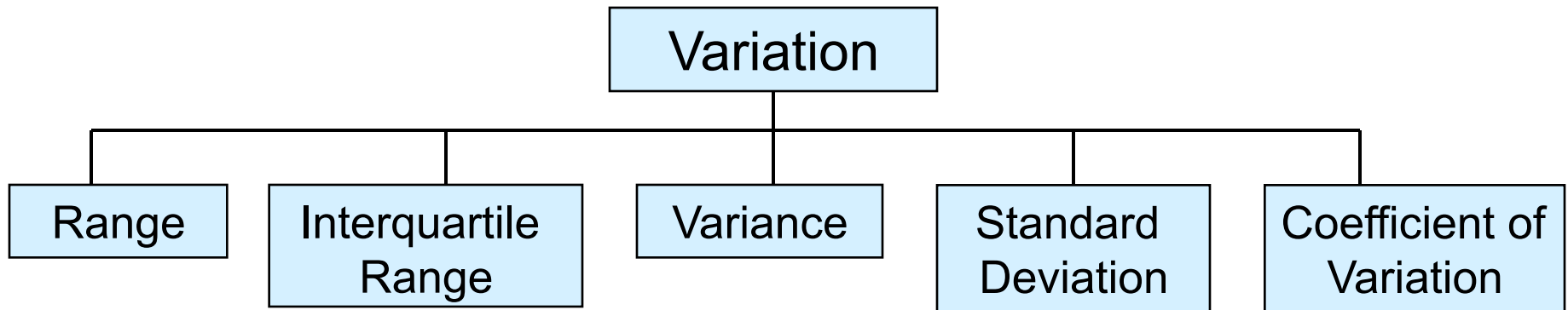
Initial Investment: \$100

After 5 years: \$125

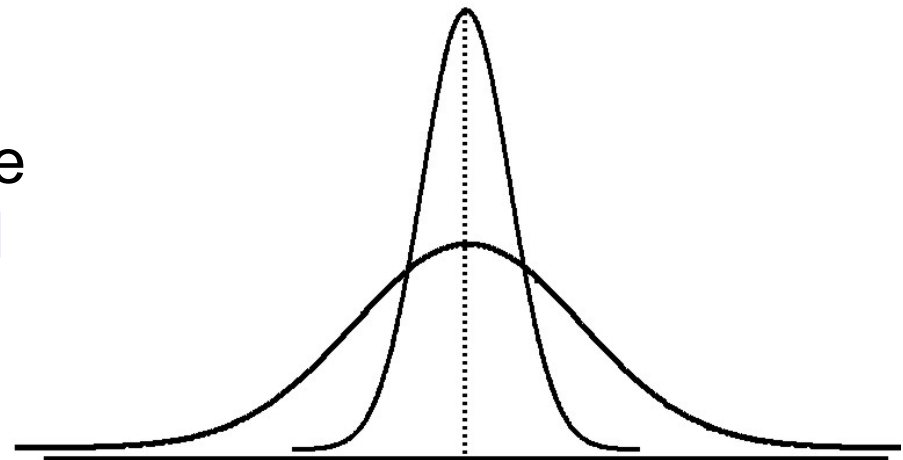
Question: the **average annual** rate of returns?

- 25 % divided by 5 years = 5%? This is Wrong!
- ``Compound Interest'' over 5 years  
 $\$100 \times (1+0.05)^5 = \$127.63 > \$125$  after 5 years
- Answer:  
 $\$100 \times (1+r)^5 = \$125 \rightarrow r = 4.6\%$

# Measures of Variability



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



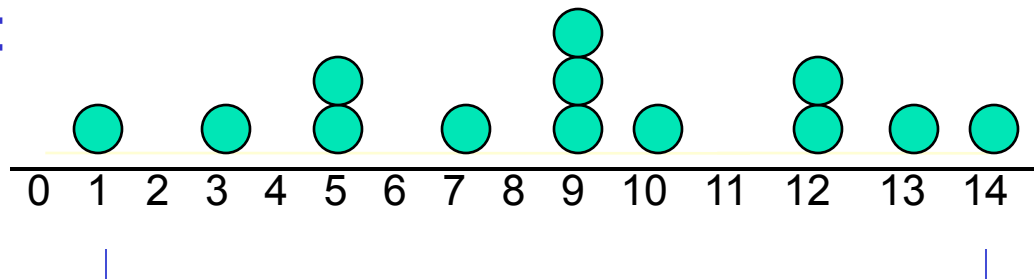
Same center,  
different variation

# Range

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

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

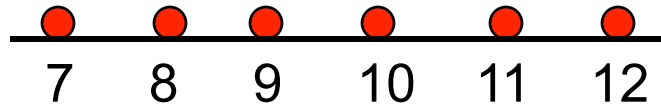
Example:



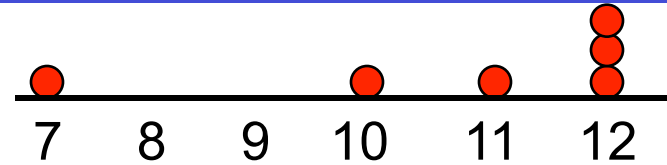
$$\text{Range} = 14 - 1 = 13$$

# 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, 2, 2, 3, 3, 3, 3, 3, 4, 5

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

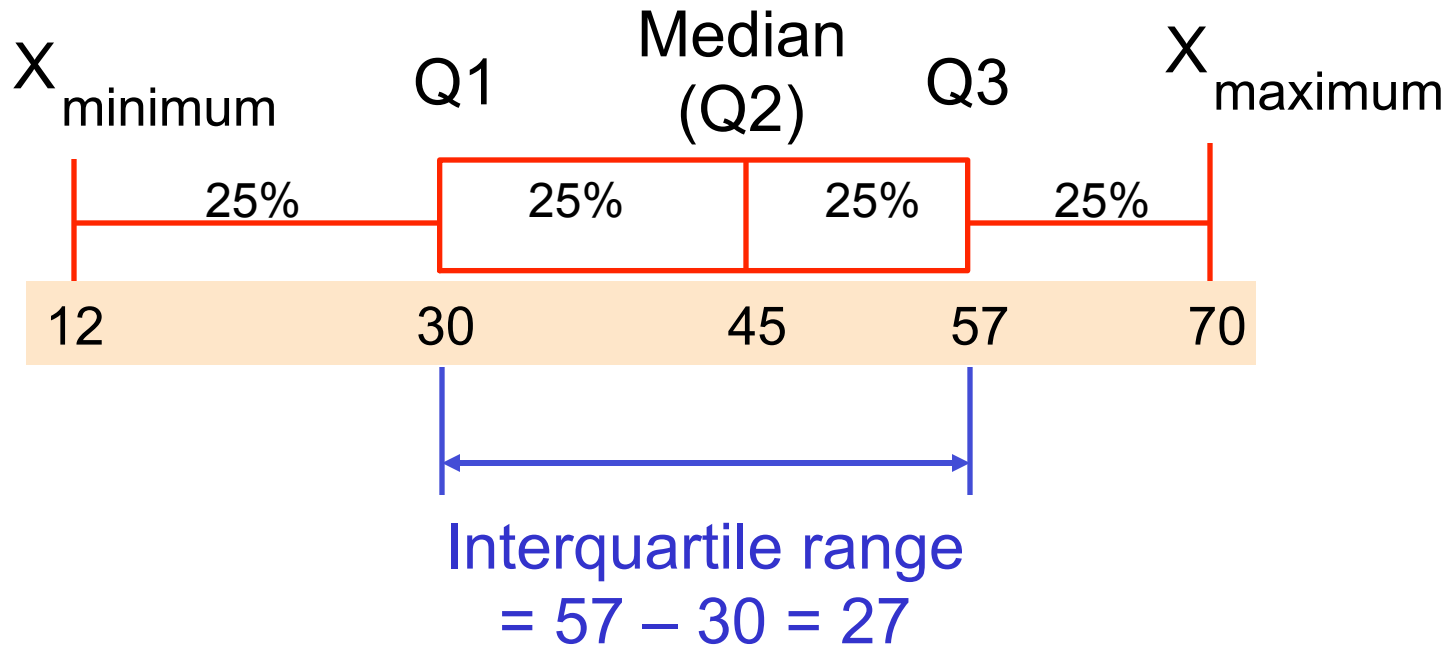
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 3, 3, 3, 3, 4, 120

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

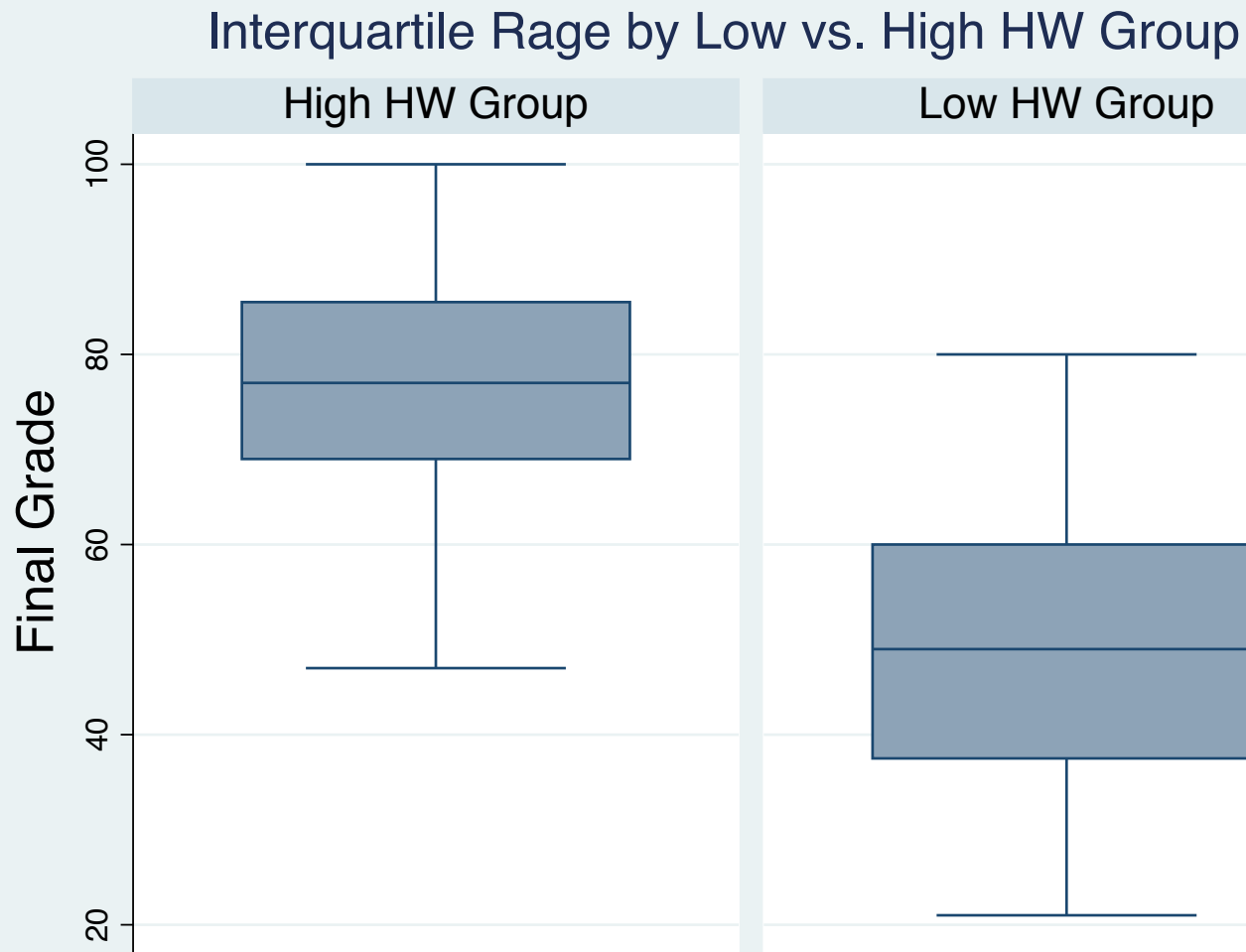


# Interquartile Range

Example:



# STATA Example





# Population Variance

- Average of squared deviations of values from the mean

- Population variance:

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

Where

$\mu$  = population mean

$N$  = population size

$x_i$  =  $i^{\text{th}}$  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

$X_i$  =  $i^{\text{th}}$  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:

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



# 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 =  $\bar{x} = 16$

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

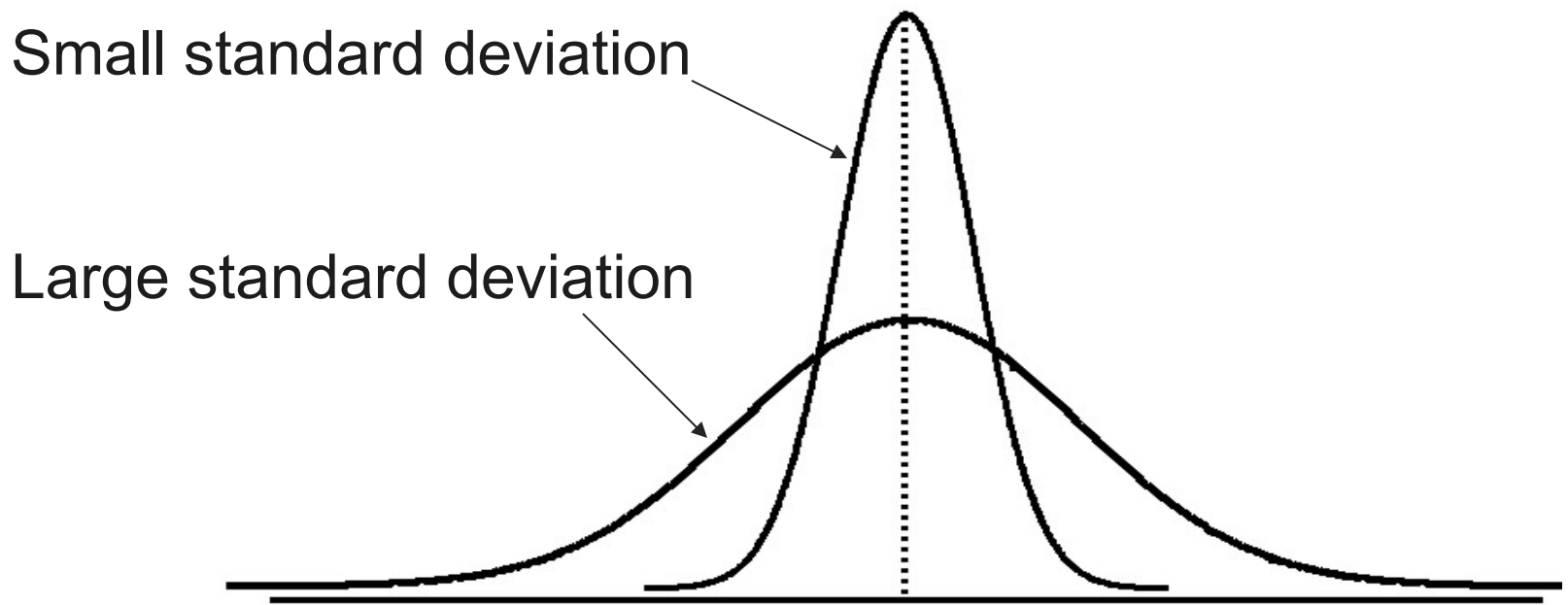
$$= \sqrt{\frac{(10 - 16)^2 + (12 - 16)^2 + (14 - 16)^2 + \dots + (24 - 16)^2}{8 - 1}}$$

$$= \sqrt{\frac{126}{7}}$$

$$= 4.2426$$

→ A measure of the “average”  
scatter around the mean

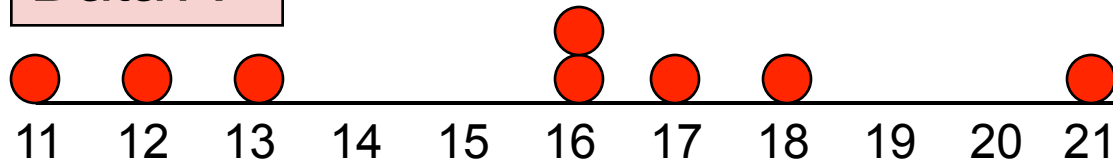
# Measuring variation





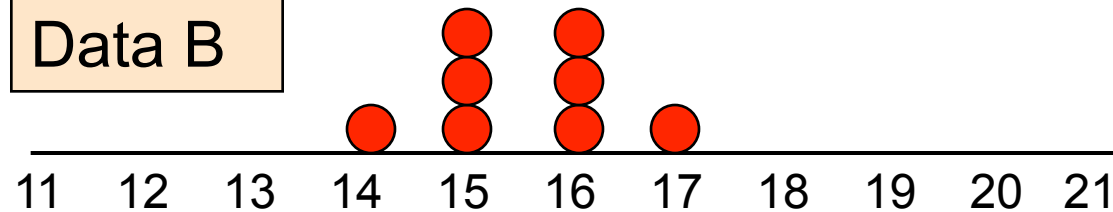
# Comparing Standard Deviations

Data A



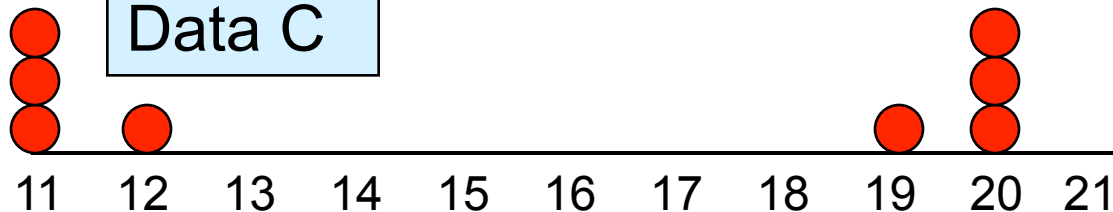
Mean = 15.5  
 $s = 3.338$

Data B



Mean = 15.5  
 $s = 0.926$

Data C



Mean = 15.5  
 $s = 4.570$



# Advantages of Variance and Standard Deviation

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- 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)



# Coefficient of Variation

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- 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:

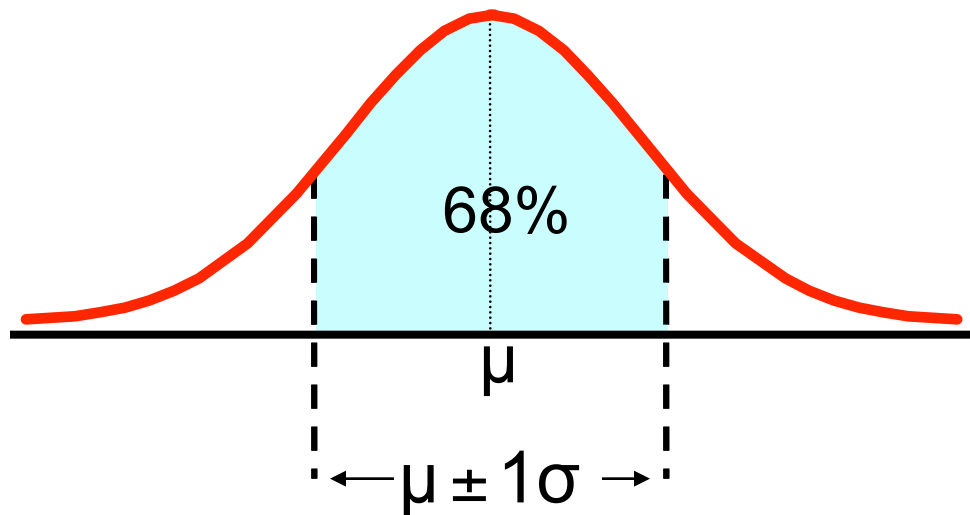
- Average price last year = \$100
- Standard deviation = \$5

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

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

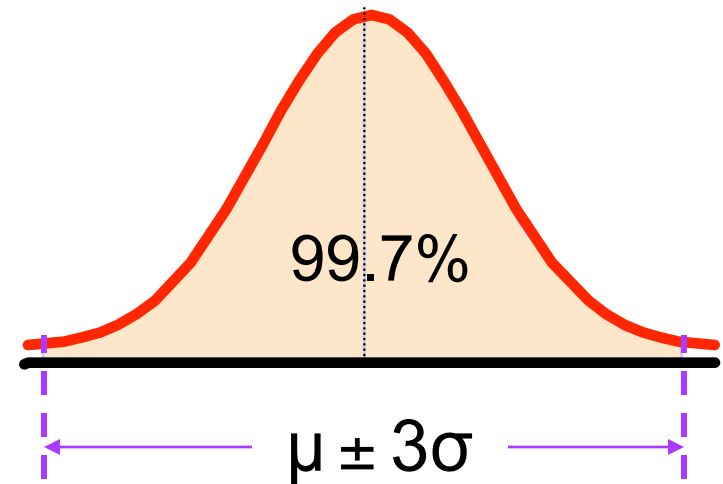
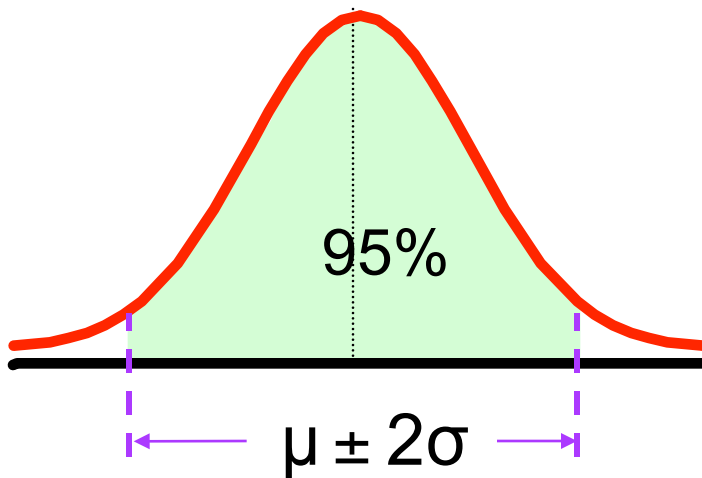
# The Empirical Rule

- If the data distribution is approximated by normal distribution, 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 **almost all** (about **99.7%**) of the values in the population or the sample



# Weighted Mean

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

$$\bar{X} = \sum_{i=1}^n w_i X_i = w_1 X_1 + w_2 X_2 + \cdots + w_n X_n$$

- Where  $w_i$  is the weight of the  $i^{\text{th}}$  observation  
and  $\sum w_i = 1$
- Use when data is already grouped into  $n$  classes, with  $w_i$  values in the  $i^{\text{th}}$  class

# Example

- Consider a student with the scores of assignment ( $x_1$ ), midterm ( $x_2$ ), and final exam ( $x_3$ ) given by

$$x_1 = 90, x_2 = 90, \text{ and } x_3 = 46.$$

- The weights:

- $w_1 = 0.1, w_2 = 0.3, \text{ and } w_3 = 0.6.$

- The final grade for this student is

$$\sum_{i=1}^3 w_i x_i = 0.1 \times 90 + 0.3 \times 90 + 0.6 \times 46 = 63.6$$



# 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
- Depends on the unit of measurement



# Interpreting Covariance

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## ■ **Covariance** between two variables:

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

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

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



# Coefficient of Correlation

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- 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:

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

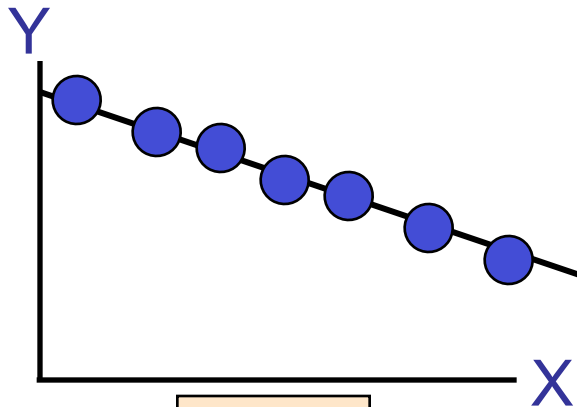


# Features of Correlation Coefficient, $r$

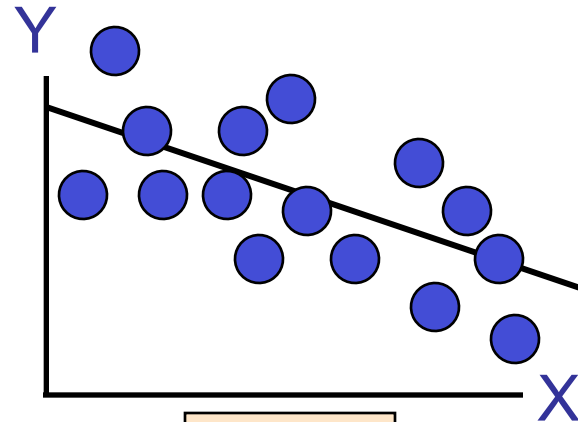
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- Unit free
- Ranges between  $-1$  and  $1$
- The closer to  $-1$ , the stronger the negative linear relationship
- The closer to  $1$ , the stronger the positive linear relationship
- The closer to  $0$ , the weaker any positive linear relationship

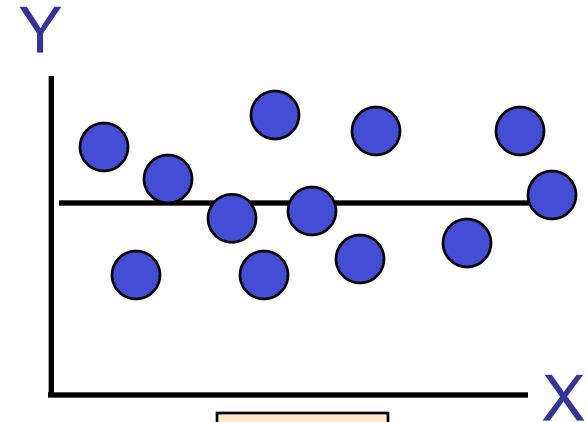
# Scatter Plots of Data with Various Correlation Coefficients



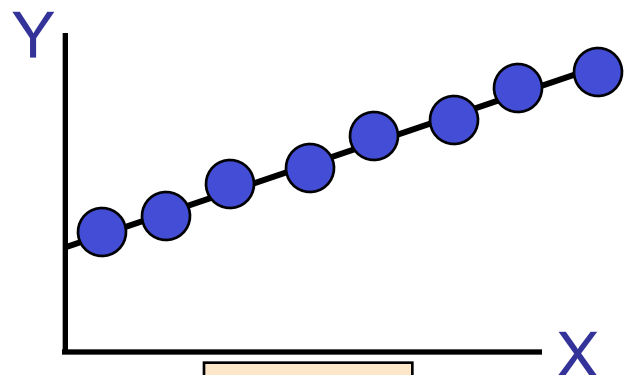
$r = -1$



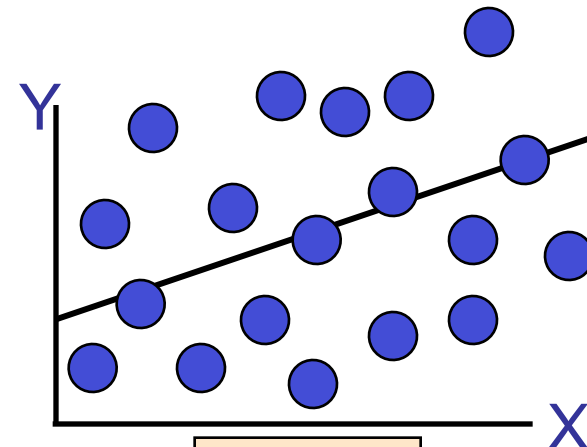
$r = -.6$



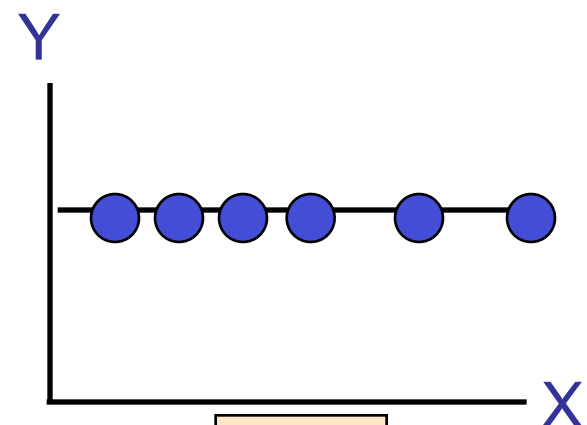
$r = 0$



$r = +1$



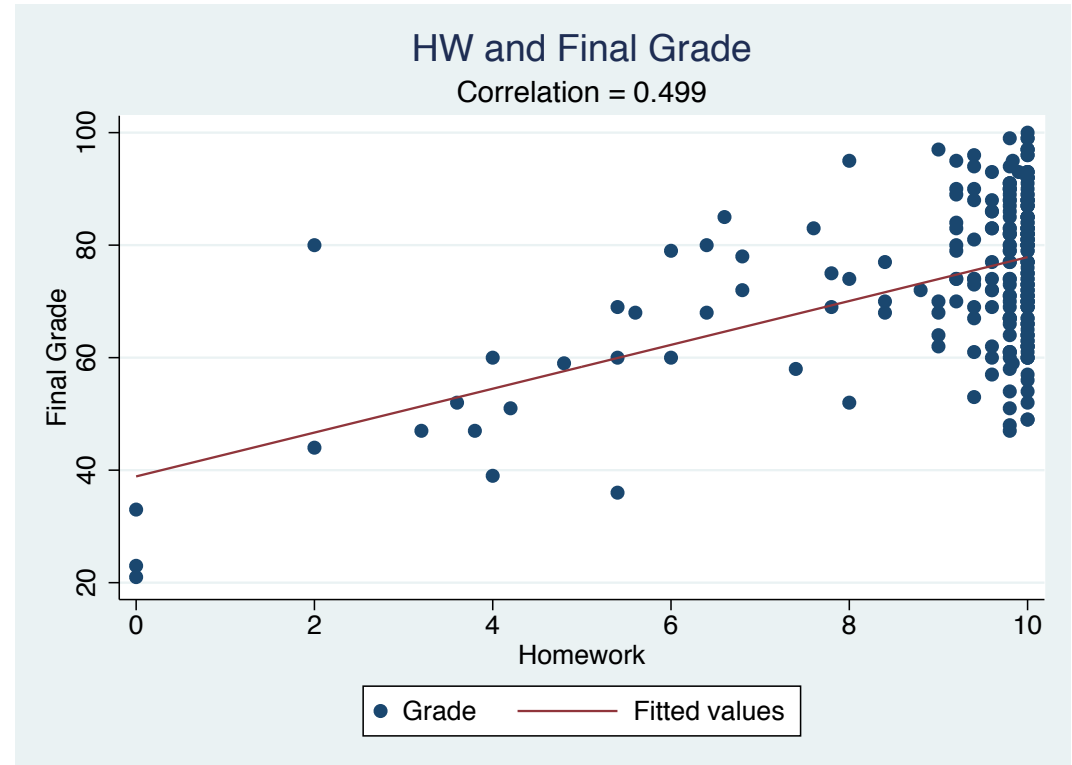
$r = +.3$



$r = 0$

# Example: HW and Final Grade

- $r = .0.499$
- There is a **relatively strong positive linear relationship** between HW scores and Final Grades



- Students who scored high on HW assignment tended to have high final grades