

# Practical Applications of Statistical Analysis

M. Mokhtari, Ph.D.

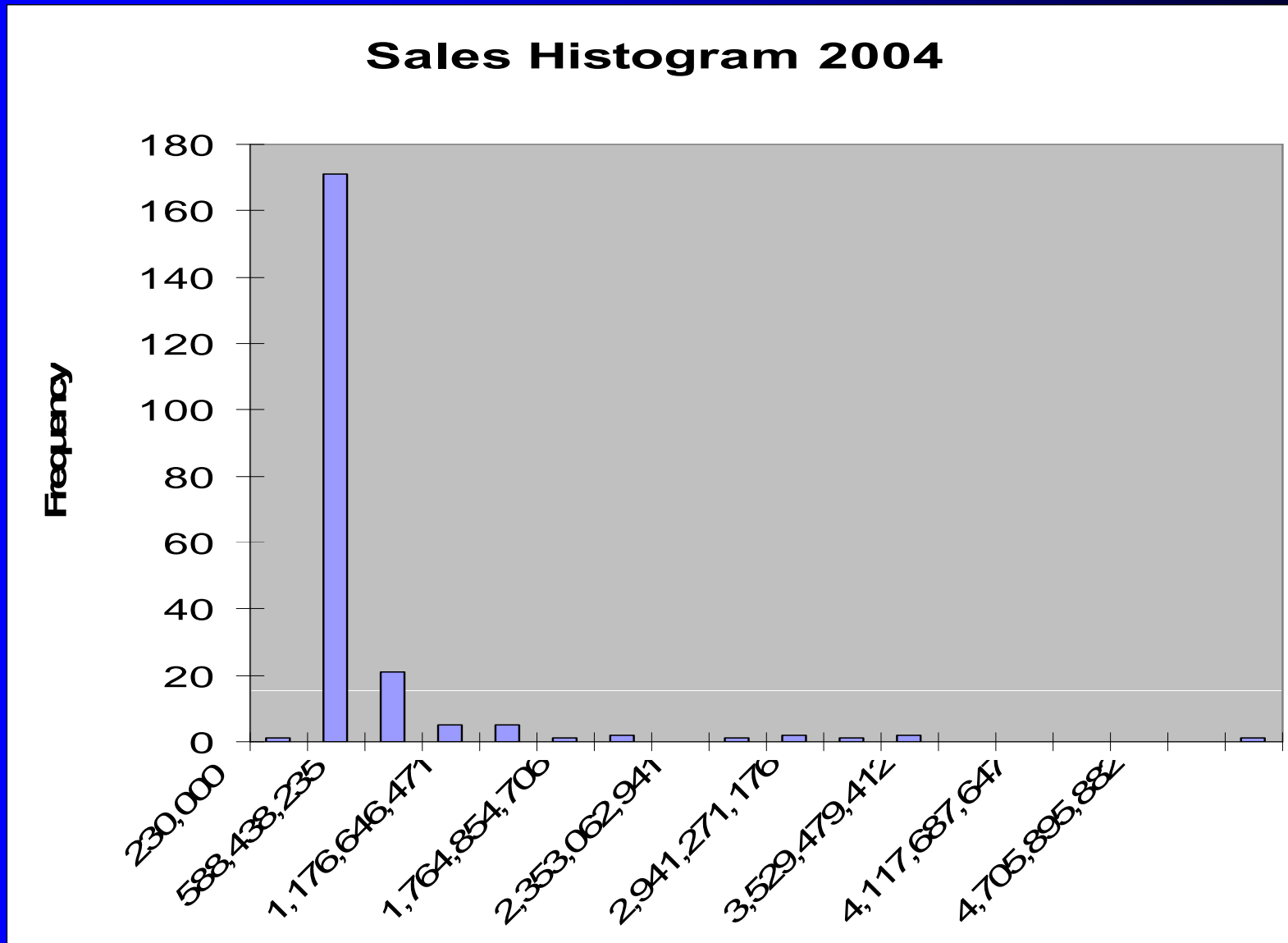
# Statistics and Types Of Statistics

- Statistics is the science of collecting, organizing, presenting, analyzing, and interpreting numerical data for the purpose of better problem solving
- **Descriptive statistics** is the procedures used to organize and summarize data; for example, ARCS
- **Inferential statistics** is the method used to find out something about the **population** (*all enterprises in Armenia*) using a **sample** (*300 enterprises in ARCS*).

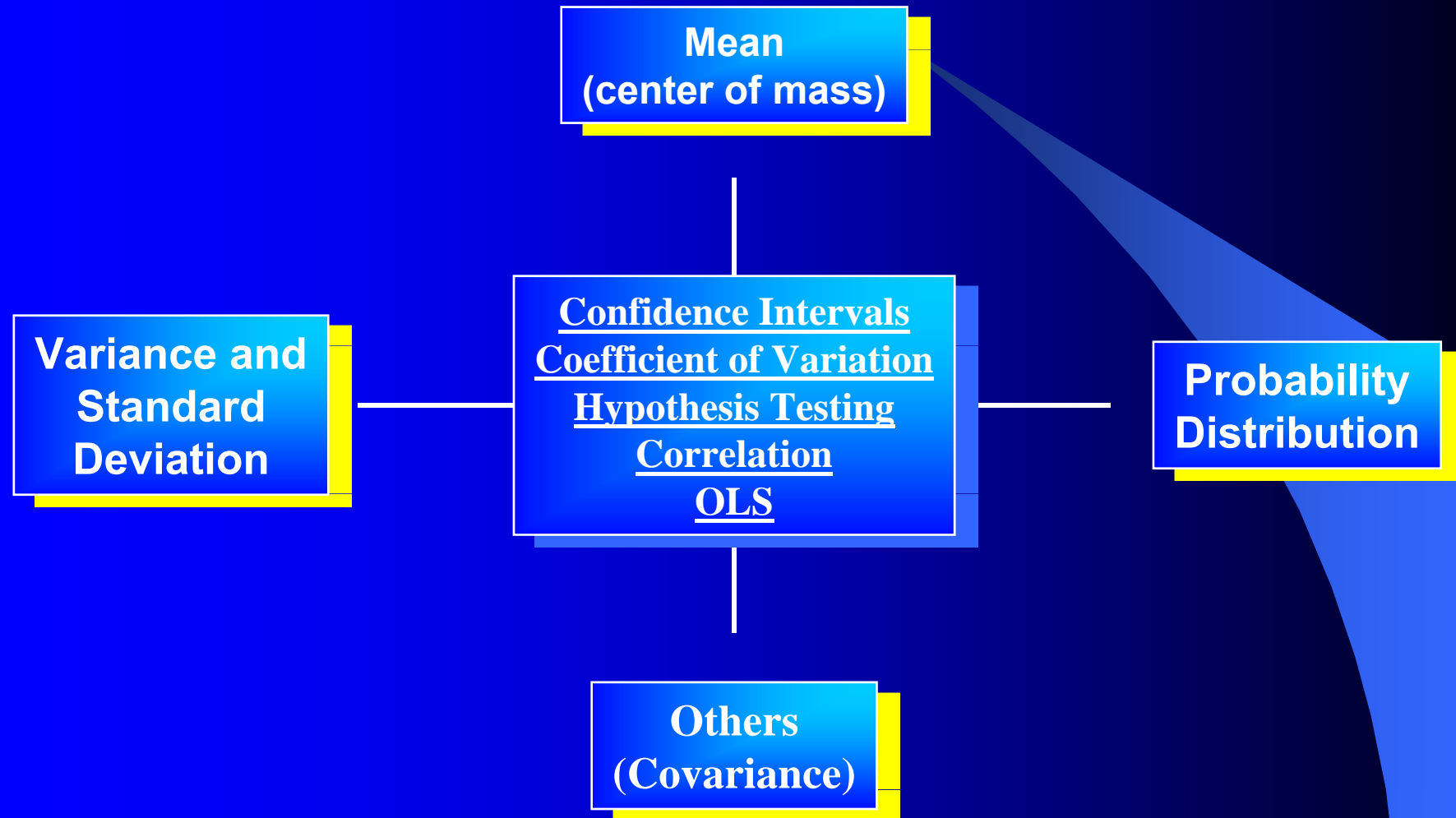
# How To Organize Data & Present Data (Descriptive Statistics)

- Arranging the data in ascending or descending order.
- Frequency Distribution
- Cumulative Frequency Distribution
  
- Bar Charts
- Pie Charts

# Histogram of Sales (ARCS-2004)



# What do we know and how we can use this knowledge:



# Quick Review: Mean Value

- **Mean** of the data is what you **expect** the data to be. *Mean could be a simple average value for the data: Life expectancy at birth = 72.7*

*GDP per capita (PPP) = \$5,900*

- **Mean** usually estimate by adding up the individual elements of a variable and dividing the sum by the total number of elements, i.e. average of 2 and 6 is then  $(2+6)/2$
- Sometimes we hear a phrase “**weighted average**”. It is a **mean** but instead of weighing each individual element equally we weigh them by specified weights (for example: food cannot be consumed in equal amount by a child and an adult, but we have to estimate an average consumption per family.)

# Mean

- Average = (Sum of all values) / (Number of values)

For example: 3, 4, 7, 3, 8

Average =  $(3 + 4 + 7 + 3 + 8) / 5 = 5$

In this example, there were 5 observations, that is,  $i=1, 2, \dots, 5$ .  
Hence, the total number of observations are 5 ( $n=5$ )

$$\text{Mean} = \frac{1}{n} \sum_{i=1}^n X_i$$

# Variance

- **Variance** is a measure of dispersion of the data around the mean
- Variance also serves as a measurement of risk, i.e. greater the variability around the mean greater the risk
- Since in the real world we rarely deal with entire population, we get to work with the **sample variance formula**:

$$S^2 = \frac{\sum_i^n (Y_i - \bar{Y})^2}{n - 1}$$

- **Standard Deviation** (S) is a standardized measure of dispersion around the mean. The value of standard deviation is just a square root of variance



## An Example: Computation Of Variance (n=4)

$$S^2 = \frac{\sum (X_i - \bar{X})^2}{n-1}$$

$$S^2 = \frac{26}{3} = 8.67 \rightarrow S = 2.94$$

$X_i$	$\bar{X}$	$X_i - \bar{X}$	$(X_i - \bar{X})^2$
2	5	-3	9
8	5	3	9
3	5	-2	4
7	5	2	4

$\sum X_i$	$\bar{X}$	$\sum (X_i - \bar{X})$	$\sum (X_i - \bar{X})^2$
20	5	0	26

# Covariance

- A measure of co-movement
- It is just the average of the cross products of two random variables

$$\text{Covariance} = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{N - 2}$$

X-Average	Y-Average	(X-Average)(Y-Average)
1	-7.75	-7.25
11	8.75	96.25
-4	-2.25	9.00
-8	0.75	-6.00

Sum / (4-2) = 92/2 = 46 = Covariance

$$Co\ variance = \frac{\sum (X_i - \bar{X})(Y_i - \bar{Y})}{N - 2}$$

# Application of Variance and Covariance: Regression Model

Modeling Y: Assume that theory tells us that Y is influenced by X:

$$Y_t = f(X_t)$$

Also, assume that , the relationship between Y and X is linear; and that, “e” is a portion of Y that is not explained by X. Then, we have the following model:

$$Y_t = \beta_0 + \beta_1 X_t + e_t$$

# Application of Variance and Covariance :

*Modeling Y: Assume Y is influenced by X; also, assume that “e” is a portion of Y that is not explained by X)*

- Example:

$$Y_t = \beta_0 + \beta_1 X_t + e_t$$

Note :  $Y_t = \beta_0$  if  $X_t = 0$  or  $\beta_1 = 0$

Note:  $\frac{\Delta Y}{\Delta X} = \beta_1$  = Change in Y, if X changes by a unit

Note: *Elasticity* =  $E = \% \Delta Y / \% \Delta X = E = \beta_1 \frac{\bar{X}}{\bar{Y}}$



## Application - continued:

- Slope or parameter  $\beta_1$  is estimated using what we have already learned:

$$\hat{\beta}_1 = \frac{\text{Covariance}(Y, X)}{\text{Variance}(X)}$$

- This slope tells us the following: For one unit increase in explanatory variable  $X$ , how many units will the predicted variable  $Y$ .
- And  $\beta_0$  is estimated by:

$$\beta_0 = \bar{Y} - \beta_1 \bar{X}$$

## Application - continued:

- Slope or parameter  $\beta_1$  is estimated by:

$$\hat{\beta}_1 = \frac{\text{Covariance}(Y, X)}{\text{Variance}(X)}$$

$$\hat{\beta}_1 = \frac{90}{512.7} = 0.176$$

- This slope tells us the following: For one unit increase in  $X$ , we find 0.176 unit increase in  $Y$ . Or for 100 unit increase in  $X$ , then  $Y$  will go up by 17.6 unit.

- $\beta_0$  is estimated by:

$$\beta_0 = \bar{Y} - \beta_1 \bar{X}$$

$$\beta_0 = 6 - (0.176)(37.75) = -0.64$$

This shows that if  $X = 0$ , then  $Y = -0.64$



# Application of Variance and Covariance - continue

$$Y_t = \beta_0 + \beta_1 X_t + e_t$$

- *Elasticity* or  $E = \% \Delta Y / \% \Delta X$  is :

$$E = \beta_1 \frac{\bar{X}}{\bar{Y}}$$

$$E = (0.176) \frac{37.75}{6} = 1.107$$

*Therefore, “E= 1.107” shows that for every one percent rise in X, we will see Y to rise by 1.107 percent.*

# Univariate Descriptive Statistics - ARCS 2009

	<i>VI=Employment</i>	<i>V39=Exports</i>	<i>V69=Sales</i>
Mean	17	127,235,294	98,237,481
Standard Error	3	60,639,880	22,467,923
Median	7	30,000,000	27,000,000
Mode	3	30,000,000	7,000,000
Standard Deviation	49	250,024,631	187,980,132
Sample Variance	2,447	62,512,316,176,470,600	35,336,530,138,429,900
Kurtosis	212	5	9
Skewness	14	2	3
Range	799	800,000,000	879,500,000
Minimum	1	0	500,000
Maximum	800	800,000,000	880,000,000
Sum	5,038	2,163,000,000	6,876,623,700
Count	300	17	70

# Covariance - ARCS 2009

	<b>VI</b>	<b>V39</b>	<b>V69</b>
<b>VI=Employment</b>	2,439		
<b>V39=Export</b>	30,762,508,651	58,835,121,107,266,400	
<b>V69=Sales</b>	1,319,889,425	5,641,795,918,367,350	34,831,722,565,023,800

Correlation Coefficients =  $\text{Cov}(Y, X) / (S_y S_x)$   
- ARCS 2009

	<b>VI</b>	<b>V39</b>	<b>V69</b>
<b>VI=Employment</b>	1		
<b>V39=Export</b>	0.69	1.00	
<b>V69=Sales</b>	0.42	0.21	1