

# **Statistical Analysis in Fin Mkts**

MSF 502

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ILLINOIS INSTITUTE OF TECHNOLOGY

# 5

# Discrete Probability Distributions

C H A P T E R



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# Chapter 5 Learning Objectives (LOs)

- LO 5.1: Distinguish between discrete and continuous random variables.
- LO 5.2: Describe the probability distribution of a discrete random variable.
- LO 5.3: Calculate and interpret summary measures for a discrete random variable.
- LO 5.4: Differentiate among risk neutral, risk averse, and risk loving consumers.
- LO 5.5: Compute summary measures to evaluate portfolio returns.



# Chapter 5 Learning Objectives (LOs)

- LO 5.6: Describe the binomial distribution and compute relevant probabilities.
- LO 5.7: Describe the Poisson distribution and compute relevant probabilities.



# 5.1 Random Variables and Discrete Probability Distributions

**LO 5.1 Distinguish between discrete and continuous random variables.**

- **Random variable**
  - **A function that assigns numerical values to the outcomes of a random experiment.**
  - **Denoted by uppercase letters (e.g.,  $X$ ).**
- **Values of the random variable are denoted by corresponding lowercase letters.**
  - **Corresponding values of the random variable:**  
 $x_1, x_2, x_3, \dots$



# 5.1 Random Variables and Discrete Probability Distributions

- **Random variables may be classified as:**
  - **Discrete**
    - The random variable assumes a countable number of distinct values.
  - **Continuous**
    - The random variable is characterized by (infinitely) uncountable values within any interval.



**LO 5.1**

# 5.1 Random Variables and Discrete Probability Distributions

- Consider an experiment in which two shirts are selected from the production line and each can be defective (D) or non-defective (N).
  - Here is the sample space:
  - The random variable  $X$  is the number of defective shirts.
  - The possible number of defective shirts is the set  $\{0, 1, 2\}$ .
- Since these are the only possible outcomes, this is a *discrete* random variable.



# 5.1 Random Variables and Discrete Probability Distributions

**LO 5.2 Describe the probability distribution of a discrete random variable.**

- **Every random variable is associated with a probability distribution that describes the variable completely.**
  - ❑ **A probability mass function is used to describe discrete random variables.**
  - ❑ **A probability density function is used to describe continuous random variables.**
  - ❑ **A cumulative distribution function may be used to describe both discrete and continuous random variables.**





## 5.1 Random Variables and Discrete Probability Distributions

- The probability mass function of a discrete random variable  $X$  is a list of the values of  $X$  with the associated probabilities, that is, the list of all possible pairs

$$(x, P(X = x))$$

- The cumulative distribution function of  $X$  is defined as

$$P(X \leq x)$$



# 5.1 Random Variables and Discrete Probability Distributions

- **Two key properties of discrete probability distributions:**
  - The probability of each value  $x$  is a value between 0 and 1, or equivalently

$$0 \leq P(X = x) \leq 1$$

- The sum of the probabilities equals 1. In other words,

$$\sum P(X = x_i) = 1$$

where the sum extends over all values  $x$  of  $X$ .



**LO 5.2**

## 5.1 Random Variables and Discrete Probability Distributions

- A discrete probability distribution may be viewed as a table, algebraically, or graphically.
- For example, consider the experiment of rolling a six-sided die. A tabular presentation is:

$x$	1	2	3	4	5	6
$P(X = x)$	1/6	1/6	1/6	1/6	1/6	1/6

- Each outcome has an associated probability of 1/6. Thus, the pairs of values and their probabilities form the probability mass function for  $X$ .



## 5.1 Random Variables and Discrete Probability Distributions

- Another tabular view of a probability distribution is based on the cumulative probability distribution.
  - For example, consider the experiment of rolling a six-sided die. The cumulative probability distribution is

$X$	1	2	3	4	5	6
$P(X \leq x)$	1/6	2/6	3/6	4/6	5/6	6/6

- The cumulative probability distribution gives the probability of  $X$  being less than or equal to  $x$ . For example,  $P(X \leq 4) = 4/6$



**LO 5.2**

## 5.1 Random Variables and Discrete Probability Distributions

- A probability distribution may be expressed algebraically.
- For example, for the six-sided die experiment, the probability distribution of the random variable  $X$  is:

$$P(X = x) = \begin{cases} 1/6 & \text{if } x = 1, 2, 3, 4, 5, 6 \\ 0 & \text{otherwise} \end{cases}$$

- Using this formula we can find

$$P(X = 5) = 1/6$$

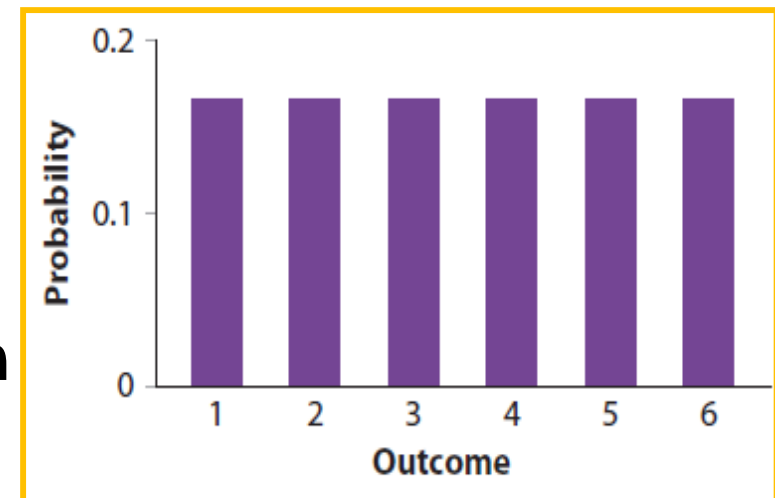
$$P(X = 7) = 0$$



**LO 5.2**

# 5.1 Random Variables and Discrete Probability Distributions

- A probability distribution may be expressed graphically.
  - The values  $x$  of  $X$  are placed on the horizontal axis and the associated probabilities on the vertical axis.
  - A line is drawn such that its height is associated with the probability of  $x$ .
  - For example, here is the graph representing the six-sided die experiment:
  - This is a uniform distribution since the bar heights are all the same.



## LO 5.2

# 5.1 Random Variables and Discrete Probability Distributions

- **Example:** Consider the probability distribution which reflects the number of credit cards that [Bankrate.com's](http://Bankrate.com) readers carry:

- ❑ Is this a valid probability distribution?
- ❑ What is the probability that a reader carries no credit cards?
- ❑ What is the probability that a reader carries less than two?
- ❑ What is the probability that a reader carries at least two credit cards?

Number of Credit Cards	Percentage
0	2.5%
1	9.8
2	16.6
3	16.5
4*	54.6

\*denotes 4 or more credit cards.

SOURCE: [www.bankrate.com](http://www.bankrate.com), Financial Literacy Series, 2007.



# 5.1 Random Variables and Discrete Probability Distributions

- Consider the probability distribution which reflects the number of credit cards that Bankrate.com's readers carry:

- Yes, because  $0 \leq P(X = x) \leq 1$  and  $\sum P(X = x) = 1$ .

- $P(X = 0) = 0.025$

- $P(X < 2) = P(X = 0) + P(X = 1) = 0.025 + 0.098 = 0.123$ .

- $P(X \geq 2) = P(X = 2) + P(X = 3) + P(X = 4^*) = 0.166 + 0.165 + 0.546 = 0.877$ .

Alternatively,  $P(X \geq 2) = 1 - P(X < 2) = 1 - 0.123 = 0.877$ .

Number of Credit Cards	Percentage
0	2.5%
1	9.8
2	16.6
3	16.5
4*	54.6

\*denotes 4 or more credit cards.

SOURCE: www.bankrate.com, Financial Literacy Series, 2007.





## 5.2 Expected Value, Variance, and Standard Deviation

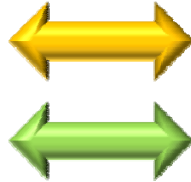
**LO 5.3 Calculate and interpret summary measures for a discrete random variable.**

- **Summary measures for a random variable include the**
  - **Mean (Expected Value)**
  - **Variance**
  - **Standard Deviation**



## 5.2 Expected Value, Variance, and Standard Deviation

### LO 5.3

- **Expected Value**  $E(X)$   **Population Mean**  $\mu$ 
  - $E(X)$  is the long-run average value of the random variable over infinitely many independent repetitions of an experiment.
  - For a discrete random variable  $X$  with values  $x_1, x_2, x_3, \dots$  that occur with probabilities  $P(X = x_i)$ , the expected value of  $X$  is

$$E(X) = \mu = \sum x_i P(X = x_i)$$



**LO 5.3**

## 5.2 Expected Value, Variance, and Standard Deviation

### ■ Variance and Standard Deviation

- For a discrete random variable  $X$  with values  $x_1, x_2, x_3, \dots$  that occur with probabilities  $P(X = x)$ ,

$$\begin{aligned} \text{Var}(X) = \sigma^2 &= \sum (x_i - \mu)^2 P(X = x_i) \\ &= \sum x_i^2 P(X = x_i) - \mu^2 \end{aligned}$$

- The standard deviation is the square root of the variance.

$$SD(X) = \sigma = \sqrt{\sigma^2}$$



**LO 5.3**

## 5.2 Expected Value, Variance, and Standard Deviation

- **Example: Brad Williams, owner of a car dealership in Chicago, decides to construct an incentive compensation program based on performance.**

Bonus (in \$1,000s)	Performance Type	Probability
\$10	Superior	0.15
\$6	Good	0.25
\$3	Fair	0.40
\$0	Poor	0.20

- ❑ **Calculate the expected value of the annual bonus amount.**
- ❑ **Calculate the variance and standard deviation of the annual bonus amount.**



**LO 5.3**

## 5.2 Expected Value, Variance, and Standard Deviation

- **Solution:** Let the random variable  $X$  denote the bonus amount (in \$1,000s) for an employee.

Value, $x_i$	Probability, $P(X = x_i)$	Weighted Value, $x_i P(X = x_i)$	Weighted Squared Deviation, $(x_i - \mu)^2 P(X = x_i)$
10	0.15	$10 \times 0.15 = 1.5$	$(10 - 4.2)^2 \times 0.15 = 5.05$
6	0.25	$6 \times 0.25 = 1.5$	$(6 - 4.2)^2 \times 0.25 = 0.81$
3	0.40	$3 \times 0.40 = 1.2$	$(3 - 4.2)^2 \times 0.40 = 0.58$
0	0.20	$0 \times 0.20 = 0$	$(0 - 4.2)^2 \times 0.20 = 3.53$
		Total = 4.2	Total = 9.97

- $E(X) = \mu = \sum x_i P(X = x_i) = 4.2$  or \$4,200
- $Var(X) = \sigma^2 = \sum (x_i - \mu)^2 P(X = x_i) = 9.97$  (in \$1,000s)<sup>2</sup>.
- $SD(X) = \sqrt{\sigma^2} = \sqrt{9.97} = 3.158$  or \$3,158.



## 5.2 Expected Value, Variance, and Standard Deviation

**LO 5.4 Differentiate among risk neutral, risk averse, and risk loving consumers.**

- **Risk Neutrality and Risk Aversion**
  - **Risk averse consumers:**
    - Expect a reward for taking a risk.
    - May decline a risky prospect even if it offers a positive expected gain.
  - **Risk neutral consumers:**
    - Completely ignore risk.
    - Always accept a prospect that offers a positive expected gain.



**LO 5.4**

## 5.2 Expected Value, Variance, and Standard Deviation

### ■ Risk Neutrality and Risk Aversion

#### □ Risk loving consumers:

- May accept a risky prospect even if the expected gain is negative.

### ■ Application of Expected Value to Risk

#### □ Suppose you have a choice of receiving \$1,000 in cash

or receiving a beautiful painting from your grandmother.

- #### □ The actual value of the painting is uncertain. Here is a probability distribution of the possible worth of the painting. What should you do?

$x$	$P(X = x)$
\$2,000	0.20
\$1,000	0.50
\$500	0.30

**LO 5.4**

## 5.2 Expected Value, Variance, and Standard Deviation

### ■ Application of Expected Value to Risk

- First calculate the expected value:

$x$	$P(X = x)$
\$2,000	0.20
\$1,000	0.50
\$500	0.30

$$\begin{aligned} E(X) &= \sum x_i P(X = x_i) \\ &= \$2,000 \times 0.20 + \$1,000 \times 0.50 + \$500 \times 0.30 \\ &= \$1,050 \end{aligned}$$

- Since the expected value is more than \$1,000 it may seem logical to choose the painting over cash.
- However, we have not taken into account risk.





# 5.3 Portfolio Returns

**LO 5.5 Compute summary measures to evaluate a portfolio returns.**

- **Investment opportunities often use both:**
  - Expected return as a measure of reward.
  - Variance or standard deviation of return as a measure of risk.
- **Portfolio is defined as a collection of assets such as stocks and bonds.**
  - Let  $X$  and  $Y$  represent two random variables of interest, denoting, say, the returns of two assets.
  - Since an investor may have invested in both assets, we would like to evaluate the portfolio return formed by a linear combination of  $X$  and  $Y$ .



**LO 5.5**

## 5.3 Portfolio Returns

- **Properties of random variables useful in evaluating portfolio returns.**

- **Given two random variables  $X$  and  $Y$ ,**
  - **The expected value of  $X$  and  $Y$  is**

$$E(X + Y) = E(X) + E(Y)$$

- **The variance of  $X$  and  $Y$  is**

$$Var(X + Y) = Var(X) + Var(Y) + 2Cov(X, Y)$$

where  $Cov(X, Y)$  is the covariance between  $X$  and  $Y$ .

- **For constants  $a$ ,  $b$ , the formulas extend to**

$$E(aX + bY) = aE(X) + bE(Y)$$

$$Var(aX + bY) = a^2Var(X) + b^2Var(Y) + 2abCov(X, Y)$$



**LO 5.5**

## 5.3 Portfolio Returns

- Expected return, variance, and standard deviation of portfolio returns.
- Given a portfolio with two assets, Asset *A* and Asset *B*, the expected return of the portfolio  $E(R_p)$  is computed as:

$$E(R_p) = w_A E(R_A) + w_B E(R_B)$$

- where
  - $w_A$  and  $w_B$  are the portfolio weights
  - $w_A + w_B = 1$
  - $E(R_A)$  and  $E(R_B)$  are the expected returns on assets *A* and *B*, respectively.



**LO 5.5**

## 5.3 Portfolio Returns

- **Expected return, variance, and standard deviation of portfolio returns.**
  - **Using the covariance or the correlation coefficient of the two returns, the portfolio variance of return is:**

$$\text{Var}(R_p) = W_A^2 \sigma_A^2 + W_B^2 \sigma_B^2 + 2W_A W_B \rho_{AB} \sigma_A \sigma_B$$

where  $\sigma_A^2$  and  $\sigma_B^2$  are the variances of the returns for Asset A and Asset B, respectively,

$\sigma_{AB}$  is the covariance between the returns for Assets A and B

$\rho_{AB}$  is the correlation coefficient between the returns for Asset A and Asset B.



**LO 5.5**

## 5.3 Portfolio Returns

- **Example:** Consider an investment portfolio of \$40,000 in Stock A and \$60,000 in Stock B.
  - Given the following information, calculate the expected return of this portfolio.

Stock A	Stock B
$E(R_A) = \mu_A = 9.5\%$	$E(R_B) = \mu_B = 7.6\%$
$SD(R_A) = \sigma_A = 12.93\%$	$SD(R_B) = \sigma_B = 8.20\%$
$Cov(R_A, R_B) = \sigma_{AB} = 18.60\%$	

- **SOLUTION:** First we compute the portfolio weights. Since \$40,000 is invested in Stock A and \$60,000 in Stock B, we compute

$$w_A = \frac{40,000}{100,000} = 0.40 \quad \text{and} \quad w_B = \frac{60,000}{100,000} = 0.60.$$

Thus, using the formula for portfolio expected return, we solve:

$$E(R_p) = (0.40 \times 9.5\%) + (0.60 \times 7.6\%) = 3.80\% + 4.56\% = 8.36\%.$$



**LO 5.5**

## 5.3 Portfolio Returns

- **Example:** Consider an investment portfolio of \$40,000 in Stock A and \$60,000 in Stock B.

Stock A	Stock B
$E(R_A) = \mu_A = 9.5\%$	$E(R_B) = \mu_B = 7.6\%$
$SD(R_A) = \sigma_A = 12.93\%$	$SD(R_B) = \sigma_B = 8.20\%$
$Cov(R_A, R_B) = \sigma_{AB} = 18.60\%$	

- **Calculate the correlation coefficient between the returns on Stocks A and B.**

- **Solution:**  $\rho_{AB} = \frac{\sigma_{AB}}{\sigma_A \sigma_B} = \frac{18.60}{12.93 \times 8.20} = 0.1754$



**LO 5.5**

## 5.3 Portfolio Returns

- **Example:** Consider an investment portfolio of \$40,000 in Stock A and \$60,000 in Stock B.

Stock A	Stock B
$E(R_A) = \mu_A = 9.5\%$	$E(R_B) = \mu_B = 7.6\%$
$SD(R_A) = \sigma_A = 12.93\%$	$SD(R_B) = \sigma_B = 8.20\%$
$Cov(R_A, R_B) = \sigma_{AB} = 18.60\%$	

- **Calculate the portfolio variance.**

- **Solution:**

$$\begin{aligned} Var(R_p) &= w_A^2 \sigma_A^2 + w_B^2 \sigma_B^2 + 2w_A w_B \sigma_{AB} \\ &= (0.40)^2 (12.93)^2 + (0.60)^2 (8.20)^2 + 2(0.40)(0.60)(18.60) \\ &= 26.75 + 24.21 + 8.93 \\ &= 59.89 \end{aligned}$$



**LO 5.5**

## 5.3 Portfolio Returns

- **Example:** Consider an investment portfolio of \$40,000 in Stock A and \$60,000 in Stock B.

Stock A	Stock B
$E(R_A) = \mu_A = 9.5\%$	$E(R_B) = \mu_B = 7.6\%$
$SD(R_A) = \sigma_A = 12.93\%$	$SD(R_B) = \sigma_B = 8.20\%$
$Cov(R_A, R_B) = \sigma_{AB} = 18.60\%$	

- **Calculate the portfolio standard deviation.**
- **Solution:**

$$SD(R_p) = \sqrt{59.89} = 7.74, \text{ or } 7.74\%$$





## 5.4 The Binomial Probability Distribution

**LO 5.6 Describe the binomial distribution and compute relevant probabilities.**

- A binomial random variable is defined as the number of successes achieved in the  $n$  trials of a Bernoulli process.
- A Bernoulli process consists of a series of  $n$  independent and identical trials of an experiment such that on each trial:
  - There are only two possible outcomes:  
 $p$  = probability of a success  
 $1-p = q$  = probability of a failure
  - Each time the trial is repeated, the probabilities of success and failure remain the same.



**LO 5.6**

## 5.4 The Binomial Probability Distribution

- A binomial random variable  $X$  is defined as the number of successes achieved in the  $n$  trials of a Bernoulli process.
- A binomial probability distribution shows the probabilities associated with the possible values of the binomial random variable (that is,  $0, 1, \dots, n$ ).
  - For a binomial random variable  $X$ , the probability of  $x$  successes in  $n$  Bernoulli trials is

$$P(X = x) = \binom{n}{x} p^x q^{n-x} = \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

for  $x = 0, 1, 2, \dots, n$ . By definition,  $0! = 1$ .



**LO 5.6**

## 5.4 The Binomial Probability Distribution

- **For a binomial distribution:**

- **The expected value ( $E(X)$ ) is:**

$$E(X) = \mu = np$$

- **The variance ( $Var(X)$ ) is:**

$$Var(X) = \sigma^2 = npq$$

- **The standard deviation ( $SD(X)$ ) is:**

$$SD(X) = \sigma = \sqrt{npq}$$



**LO 5.6**

## 5.4 The Binomial Probability Distribution

- **Example: Approximately 20% of U.S. workers are afraid that they will never be able to retire. Suppose 10 workers are randomly selected.**
  - **What is the probability that none of the workers is afraid that they will never be able to retire?**
    - **Solution: Let  $X = 10$ , then**

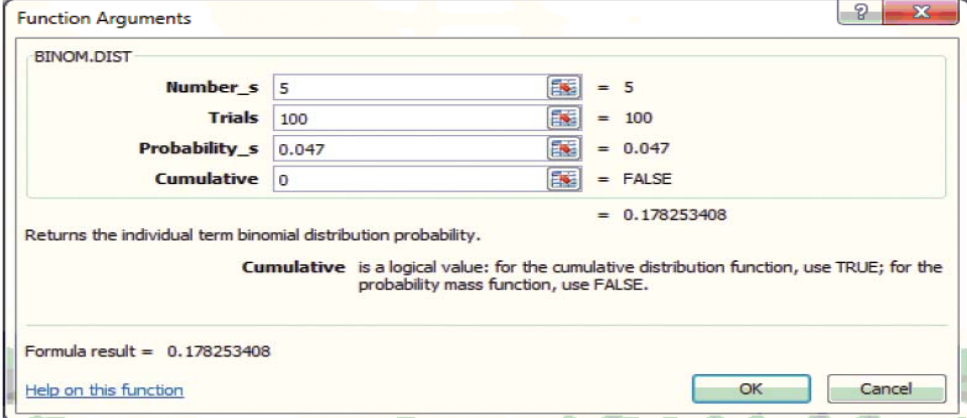
$$\begin{aligned}P(X = 0) &= \frac{10!}{0!(10 - 0)!} \times (0.20)^0 \times (0.80)^{10} \\&= \frac{10 \times 9 \times \cdots \times 1}{(1) \times (10 \times 9 \times \cdots \times 1)} \times 1 \times (0.80)^{10} = 1 \times 1 \times 0.1074 \\&= 0.1074.\end{aligned}$$



**LO 5.6**

## 5.4 The Binomial Probability Distribution

- **Computing binomial probabilities with Excel:**
  - In 2007 approximately 4.7% of the households in the Detroit metropolitan area were in some stage of foreclosure. What is the probability that exactly 5 of these 100 mortgage-holding households in Detroit are in some stage of foreclosure?
  - **Solution:** Using the binomial function on Excel, enter the four arguments shown
  - Excel returns the formula result as 0.1783; thus,  
 $P(X = 5) = 0.1783$ .



The image shows the 'Function Arguments' dialog box for the BINOM.DIST function in Excel. The dialog box has a title bar with a question mark and a close button. Inside, the function name 'BINOM.DIST' is displayed. There are four input fields with their corresponding values and formulas:

Argument	Value	Formula
Number_s	5	= 5
Trials	100	= 100
Probability_s	0.047	= 0.047
Cumulative	0	= FALSE

Below the input fields, the text 'Returns the individual term binomial distribution probability.' is displayed. To the right of this text, the formula result is shown: '= 0.178253408'. Below this, a note states: 'Cumulative is a logical value: for the cumulative distribution function, use TRUE; for the probability mass function, use FALSE.' At the bottom of the dialog box, the 'Formula result' is displayed as '0.178253408'. There are two buttons at the bottom right: 'OK' and 'Cancel'. A link 'Help on this function' is located at the bottom left.



# 5.5 The Poisson Probability Distribution

**LO 5.7 Describe the Poisson distribution and compute relevant probabilities.**

- A binomial random variable counts the number of successes in a fixed number of Bernoulli trials.
- In contrast, a Poisson random variable counts the number of successes over a given interval of time or space.
- Examples of a Poisson random variable include
  - *With respect to time*—the number of cars that cross the Brooklyn Bridge between 9:00 am and 10:00 am on a Monday morning.
  - *With respect to space*—the number of defects in a 50-yard roll of fabric.



**LO 5.7**

## 5.5 The Poisson Probability Distribution

- **A random experiment satisfies a Poisson process if:**
  - **The number of successes within a specified time or space interval equals any integer between zero and infinity.**
  - **The numbers of successes counted in nonoverlapping intervals are independent.**
  - **The probability that success occurs in any interval is the same for all intervals of equal size and is proportional to the size of the interval.**



**LO 5.7**

## 5.5 The Poisson Probability Distribution

- **For a Poisson random variable  $X$ , the probability of  $x$  successes over a given interval of time or space is**

$$P(X = x) = \frac{e^{-\mu} \mu^x}{x!} \quad \text{for } x = 0, 1, 2, \dots,$$

**where  $\mu$  is the mean number of successes and  $e \approx 2.718$  is the base of the natural logarithm.**





**LO 5.7**

## 5.5 The Poisson Probability Distribution

- **For a Poisson distribution:**

- The expected value ( $E(X)$ ) is:

$$E(X) = \mu$$

- The variance ( $Var(X)$ ) is:

$$Var(X) = \sigma^2 = \mu$$

- The standard deviation ( $SD(X)$ ) is:

$$SD(X) = \sigma = \sqrt{\mu}$$



**LO 5.7**

## 5.5 The Poisson Probability Distribution

- **Example: Returning to the Starbucks example, Ann believes that the typical Starbucks customer averages 18 visits over a 30-day month.**

- **How many visits should Anne expect in a 5-day period from a typical Starbucks customer?**

Given the rate of 18 visits over a 30-day month, we can write the mean for the 30-day period as  $\mu_{30} = 18$ . For this problem, we compute the proportional mean for a 5-day period as  $\mu_5 = 3$  because  $\frac{18 \text{ visits}}{30 \text{ days}} = \frac{3 \text{ visits}}{5 \text{ days}}$ .

- **What is the probability that a customer visits the chain five times in a 5-day period?**

- $$P(X = 5) = \frac{e^{-3}3^5}{5!} = \frac{(0.0498)(243)}{120} = 0.1008$$



# End of Chapter



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