

# Discrete Probability Distributions

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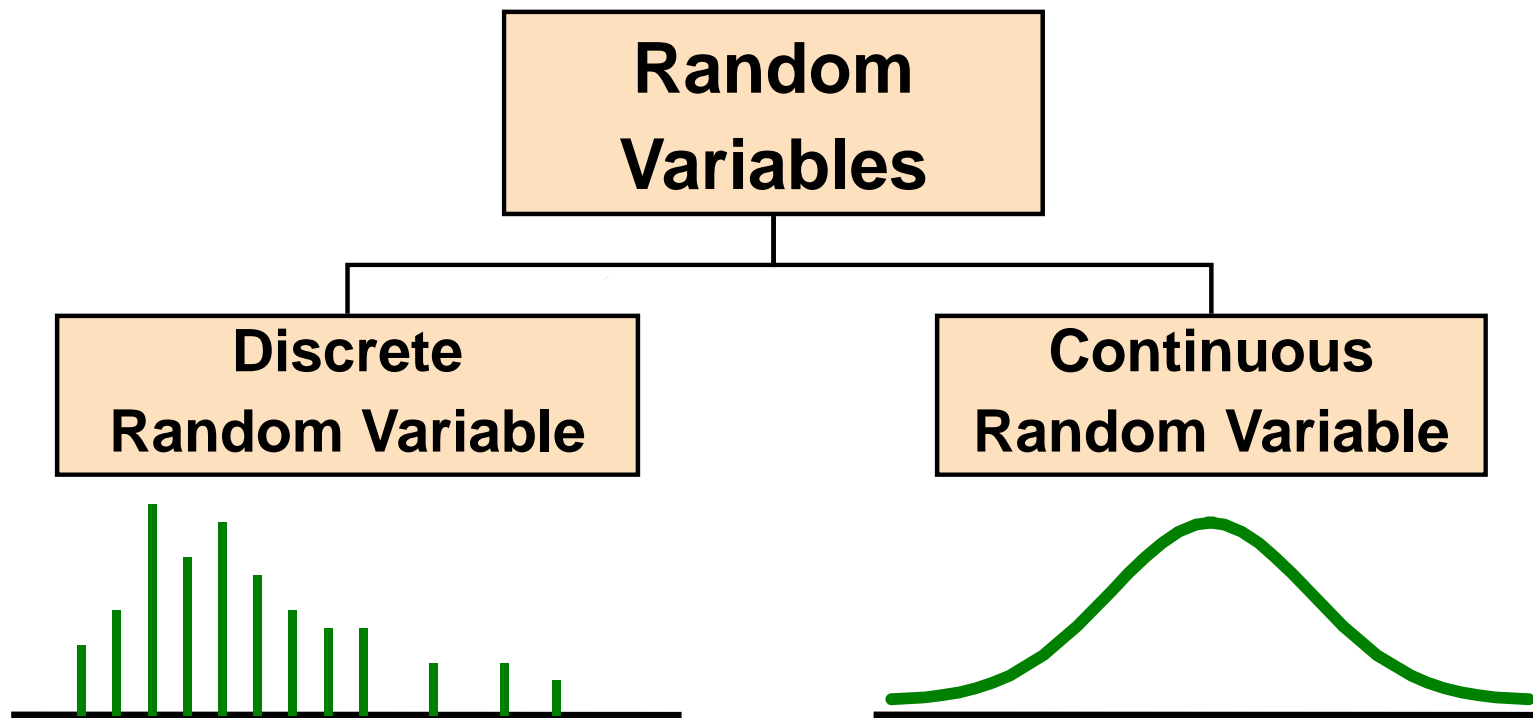
# Learning Objectives

In this lecture, you learn:

- The properties of a probability distribution
- To calculate the expected value, variance, and standard deviation of a probability distribution
- To calculate probabilities from binomial and Poisson distributions
- How to use the binomial and Poisson distributions to solve business problems

# Introduction to Probability Distributions

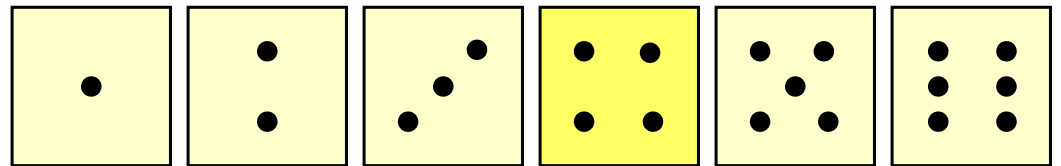
- **Random Variable**
  - **Represents a possible numerical value from an uncertain event**



# Discrete Random Variables

- Can only assume a countable number of values

Examples:



- Roll a die twice  
Let  $X$  be the number of times 4 comes up  
(then  $X$  could be 0, 1, or 2 times)

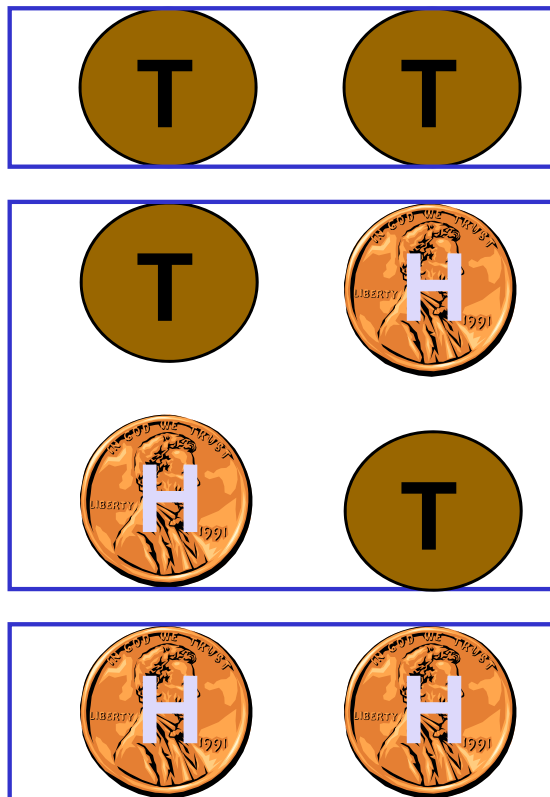
- Toss a coin 5 times.  
Let  $X$  be the number of heads  
(then  $X = 0, 1, 2, 3, 4, \text{ or } 5$ )



# Discrete Probability Distribution

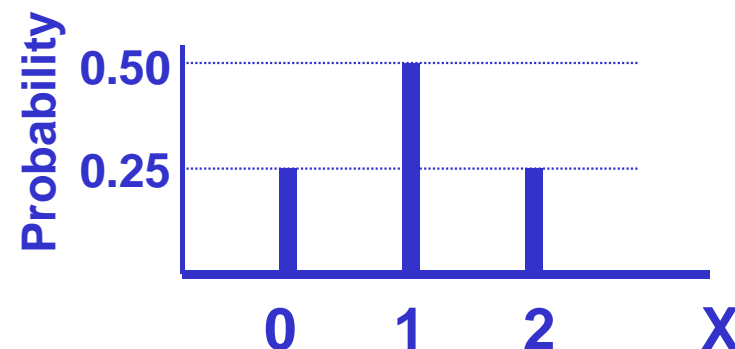
Experiment: Toss 2 Coins. Let  $X = \#$  heads.

4 possible outcomes



## Probability Distribution

<u>X Value</u>	<u>Probability</u>
0	$1/4 = 0.25$
1	$2/4 = 0.50$
2	$1/4 = 0.25$



# Discrete Random Variable Summary Measures

- Expected Value (or mean) of a discrete distribution (Weighted Average)

$$\mu = E(X) = \sum_{i=1}^N X_i P(X_i)$$

- Example: Toss 2 coins,**  
**X = # of heads,**  
**compute expected value of X:**

$$E(X) = (0 \times 0.25) + (1 \times 0.50) + (2 \times 0.25) \\ = 1.0$$

X	P(X)
0	0.25
1	0.50
2	0.25

# Discrete Random Variable Summary Measures

*(continued)*

- Variance of a discrete random variable

$$\sigma^2 = \sum_{i=1}^N [X_i - E(X)]^2 P(X_i)$$

- Standard Deviation of a discrete random variable

$$\sigma = \sqrt{\sigma^2} = \sqrt{\sum_{i=1}^N [X_i - E(X)]^2 P(X_i)}$$

where:

$E(X)$  = Expected value of the discrete random variable  $X$

$X_i$  = the  $i^{\text{th}}$  outcome of  $X$

$P(X_i)$  = Probability of the  $i^{\text{th}}$  occurrence of  $X$

# Discrete Random Variable Summary Measures

(continued)

- **Example:** Toss 2 coins,  $X = \#$  heads, compute standard deviation (recall  $E(X) = 1$ )

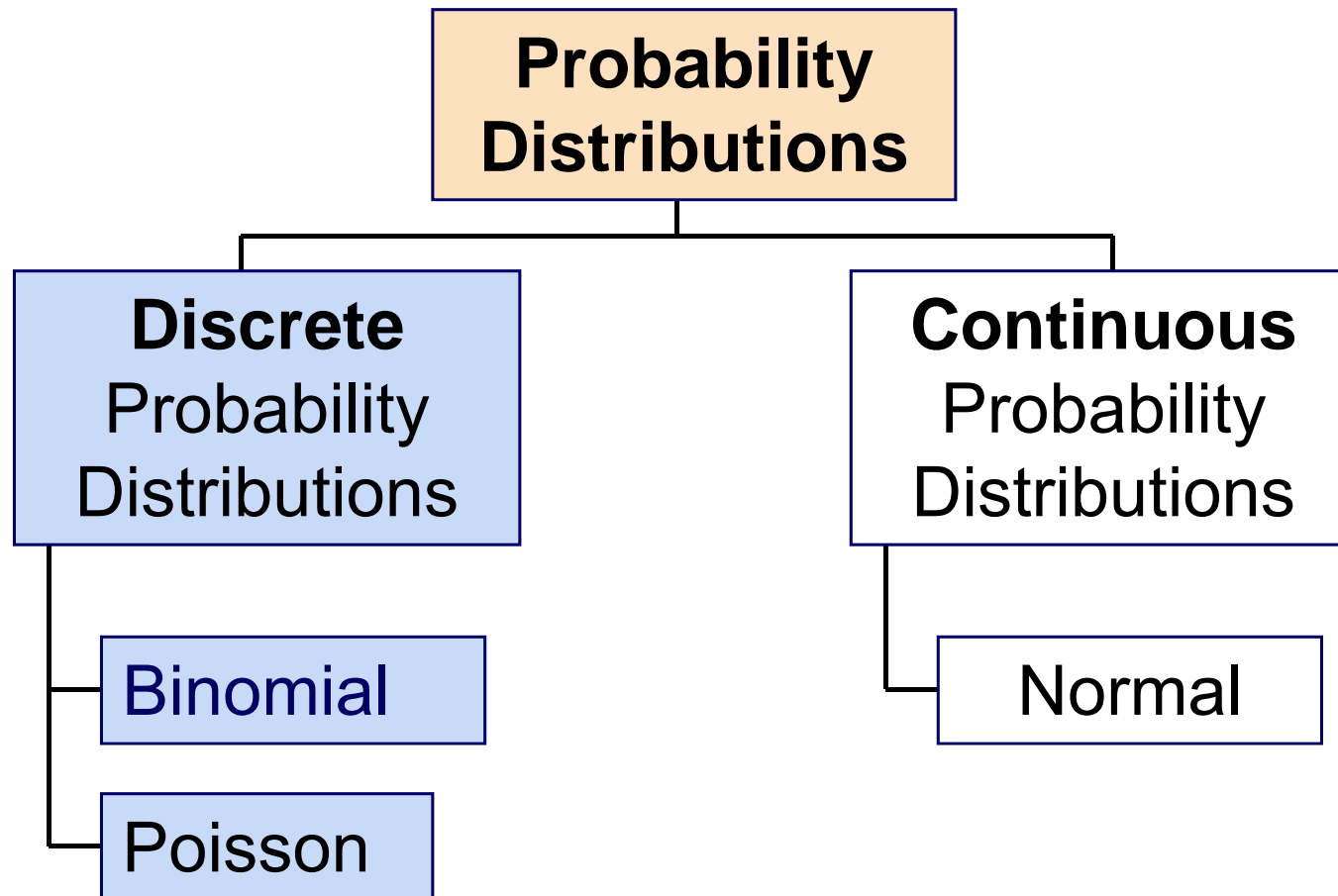
$$\sigma = \sqrt{\sum [X_i - E(X)]^2 P(X_i)}$$

$$\sigma = \sqrt{(0-1)^2(0.25) + (1-1)^2(0.50) + (2-1)^2(0.25)} = \sqrt{0.50} = 0.707$$

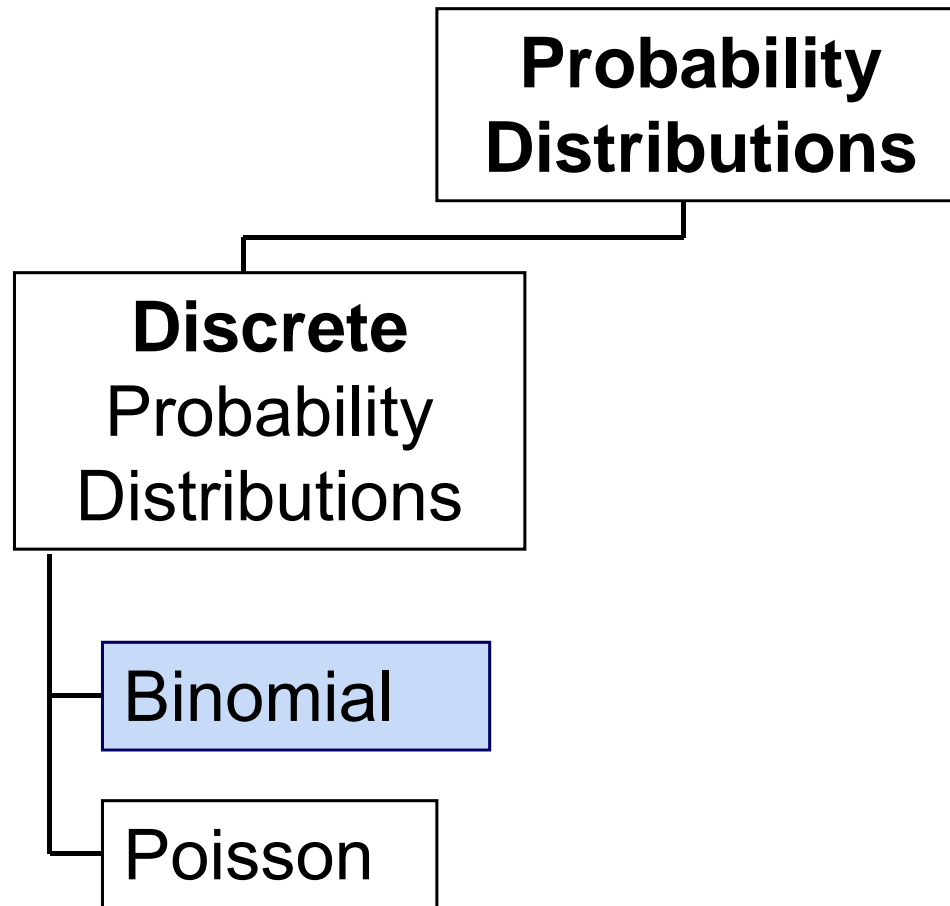
Possible number of heads  
= 0, 1, or 2



# Probability Distributions



# The Binomial Distribution



# Binomial Probability Distribution

- **A fixed number of observations,  $n$** 
  - e.g., 15 tosses of a coin; ten light bulbs taken from a warehouse
- **Two mutually exclusive and collectively exhaustive categories**
  - e.g., head or tail in each toss of a coin; defective or not defective light bulb
  - Generally called “success” and “failure”
  - Probability of success is  $p$ , probability of failure is  $1 - p$
- **Constant probability for each observation**
  - e.g., Probability of getting a tail is the same each time we toss the coin

# Binomial Probability Distribution

*(continued)*

- **Observations are independent**
  - The outcome of one observation does not affect the outcome of the other
- **Two sampling methods**
  - Infinite population without replacement
  - Finite population with replacement

# Possible Binomial Distribution Settings

- A patient either cured or died after treatment
- A manufacturing plant labels items as either defective or acceptable
- A firm bidding for contracts will either get a contract or not
- A marketing research firm receives survey responses of “yes I will buy” or “no I will not”
- New job applicants either accept the offer or reject it

# Rule of Combinations

- The number of combinations of selecting  $X$  objects out of  $n$  objects is

$${}_n C_x = \frac{n!}{X!(n-X)!}$$

where:

$$n! = (n)(n-1)(n-2) \dots (2)(1)$$

$$X! = (X)(X-1)(X-2) \dots (2)(1)$$

$$0! = 1 \quad (\text{by definition})$$

# Binomial Distribution Formula

$$P(X) = \frac{n!}{X!(n-X)!} p^X (1-p)^{n-X}$$

$P(X)$  = probability of  $X$  successes in  $n$  trials,  
with probability of success  $p$  on each trial

$X$  = number of 'successes' in sample,  
( $X = 0, 1, 2, \dots, n$ )

$n$  = sample size (number of trials  
or observations)

$p$  = probability of "success"

**Example:** Flip a coin four  
times, let  $x = \#$  heads:

$$n = 4$$

$$p = 0.5$$

$$1 - p = (1 - 0.5) = 0.5$$

$$X = 0, 1, 2, 3, 4$$

# Example: Calculating a Binomial Probability

What is the probability of one success in five observations if the probability of success is .1?

$$X = 1, n = 5, \text{ and } p = 0.1$$

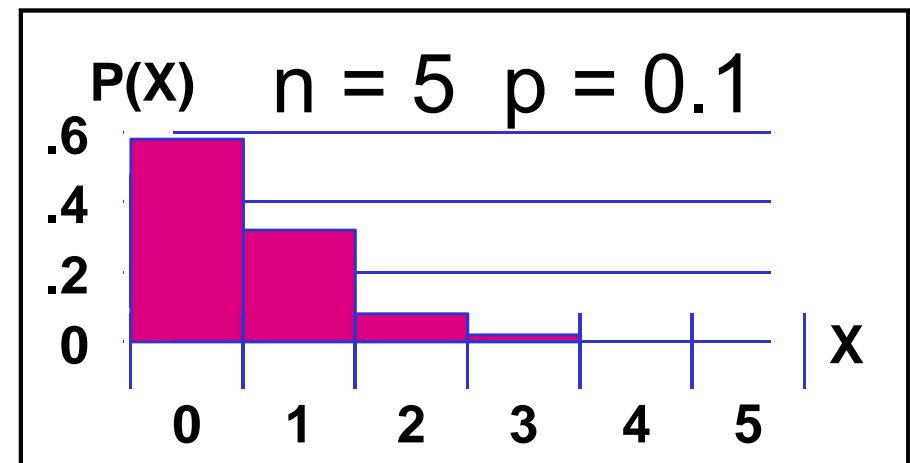
$$\begin{aligned} P(X = 1) &= \frac{n!}{X!(n-X)!} p^X (1-p)^{n-X} \\ &= \frac{5!}{1!(5-1)!} (0.1)^1 (1-0.1)^{5-1} \\ &= (5)(0.1)(0.9)^4 \\ &= 0.32805 \end{aligned}$$



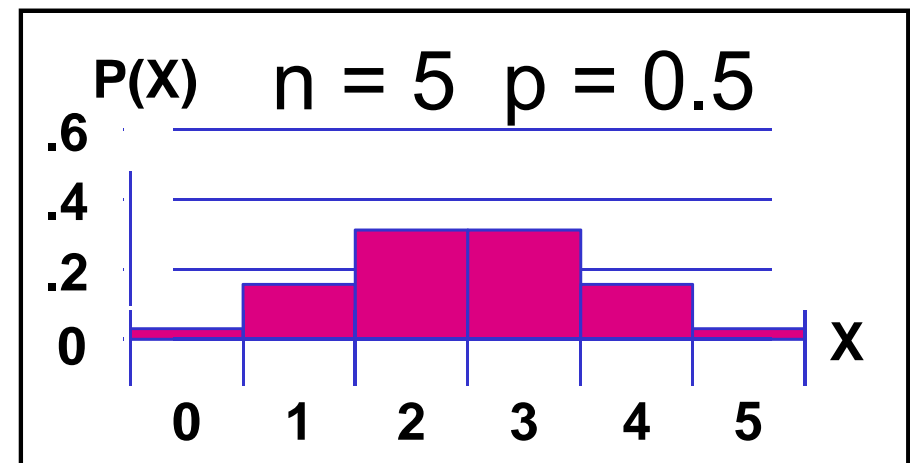
# Binomial Distribution

- The shape of the binomial distribution depends on the values of  $p$  and  $n$

- Here,  $n = 5$  and  $p = 0.1$



- Here,  $n = 5$  and  $p = 0.5$



# Binomial Distribution Characteristics

- Mean

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

- Variance and Standard Deviation

$$\sigma^2 = np(1-p)$$

$$\sigma = \sqrt{np(1-p)}$$

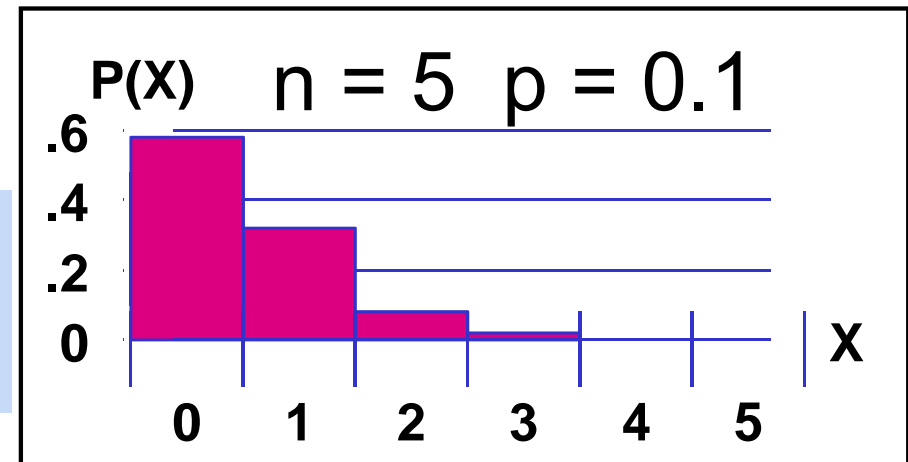
Where  $n$  = sample size  
 $p$  = probability of success  
 $(1 - p)$  = probability of failure

# Binomial Characteristics

## Examples

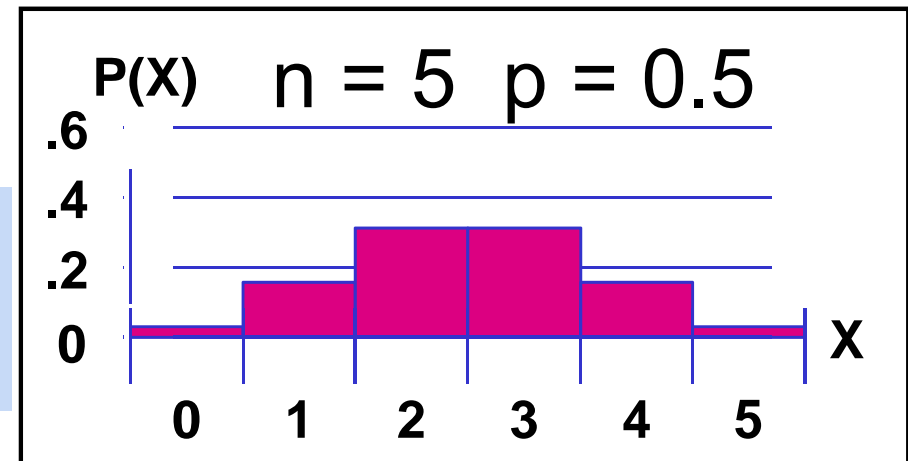
$$\mu = np = (5)(0.1) = 0.5$$

$$\sigma = \sqrt{np(1-p)} = \sqrt{(5)(0.1)(1-0.1)} = 0.6708$$



$$\mu = np = (5)(0.5) = 2.5$$

$$\sigma = \sqrt{np(1-p)} = \sqrt{(5)(0.5)(1-0.5)} = 1.118$$



# Using Binomial Tables

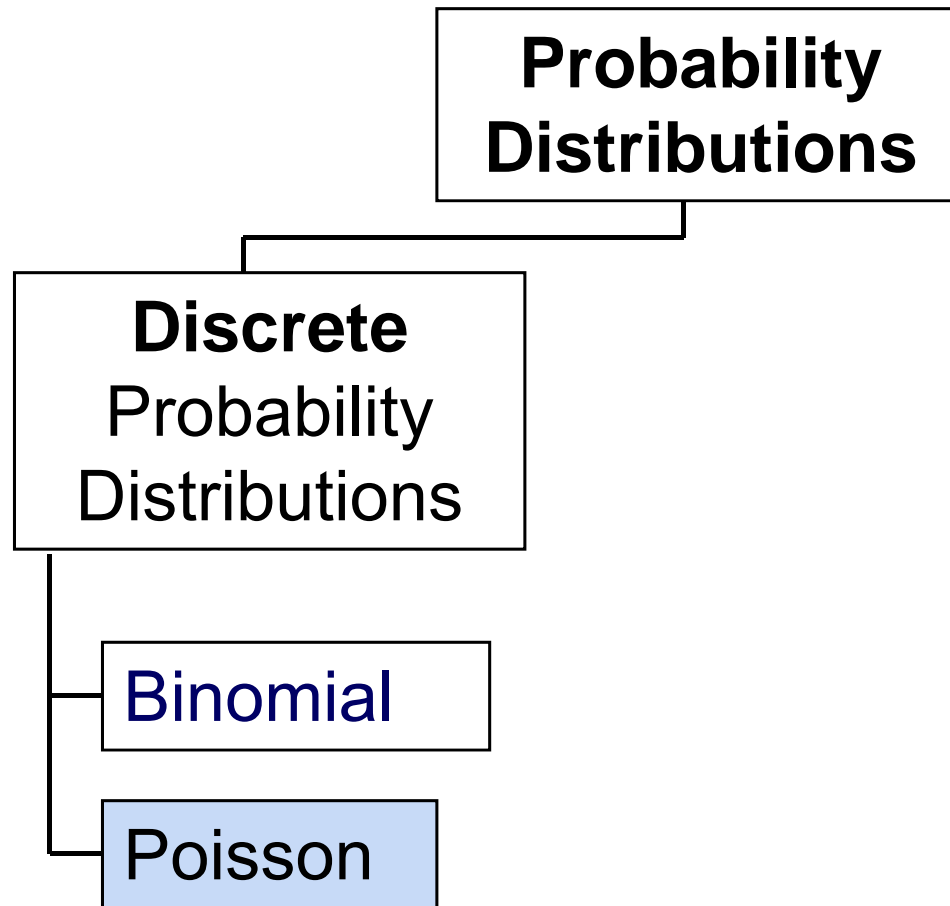
n = 10									
x	...	p=.20	p=.25	p=.30	p=.35	p=.40	p=.45	p=.50	
0	...	0.1074	0.0563	0.0282	0.0135	0.0060	0.0025	0.0010	10
1	...	0.2684	0.1877	0.1211	0.0725	0.0403	0.0207	0.0098	9
2	...	0.3020	0.2816	0.2335	0.1757	0.1209	0.0763	0.0439	8
3	...	0.2013	0.2503	0.2668	<b>0.2522</b>	0.2150	0.1665	0.1172	7
4	...	0.0881	0.1460	0.2001	0.2377	0.2508	0.2384	0.2051	6
5	...	0.0264	0.0584	0.1029	0.1536	0.2007	0.2340	0.2461	5
6	...	0.0055	0.0162	0.0368	0.0689	0.1115	0.1596	0.2051	4
7	...	0.0008	0.0031	0.0090	0.0212	0.0425	0.0746	0.1172	3
8	...	0.0001	<b>0.0004</b>	0.0014	0.0043	0.0106	0.0229	0.0439	2
9	...	0.0000	0.0000	0.0001	0.0005	0.0016	0.0042	0.0098	1
10	...	0.0000	0.0000	0.0000	0.0000	0.0001	0.0003	0.0010	0
	...	p=.80	p=.75	p=.70	p=.65	p=.60	p=.55	p=.50	x

## Examples:

$$n = 10, p = 0.35, x = 3: \quad P(x = 3|n = 10, p = 0.35) = 0.2522$$

$$n = 10, p = 0.75, x = 2: \quad P(x = 2|n = 10, p = 0.75) = 0.0004$$

# The Poisson Distribution



# The Poisson Distribution

- **Apply the Poisson Distribution when:**
  - You wish to count the number of times an event occurs in a given **area of opportunity**
  - The probability that an event occurs in one area of opportunity is the same for all areas of opportunity
  - The number of events that occur in one area of opportunity is independent of the number of events that occur in the other areas of opportunity
  - The probability that two or more events occur in an area of opportunity approaches zero as the area of opportunity becomes smaller
  - The **average number of events per unit is  $\lambda$  (lambda)**

# Poisson Distribution Formula

$$P(X) = \frac{e^{-\lambda} \lambda^x}{X!}$$

where:

$X$  = number of events in an area of opportunity

$\lambda$  = expected number of events

$e$  = base of the natural logarithm system (2.71828...)

# Poisson Distribution Characteristics

- Mean

$$\mu = \lambda$$

- Variance and Standard Deviation

$$\sigma^2 = \lambda$$

$$\sigma = \sqrt{\lambda}$$

where  $\lambda$  = expected number of events



# Using Poisson Tables

x	$\lambda$								
	0.10	0.20	0.30	0.40	0.50	0.60	0.70	0.80	0.90
0	0.9048	0.8187	0.7408	0.6703	0.6065	0.5488	0.4966	0.4493	0.4066
1	0.0905	0.1637	0.2222	0.2681	0.3033	0.3293	0.3476	0.3595	0.3659
2	0.0045	0.0164	0.0333	0.0536	0.0758	0.0988	0.1217	0.1438	0.1647
3	0.0002	0.0011	0.0033	0.0072	0.0126	0.0198	0.0284	0.0383	0.0494
4	0.0000	0.0001	0.0003	0.0007	0.0016	0.0030	0.0050	0.0077	0.0111
5	0.0000	0.0000	0.0000	0.0001	0.0002	0.0004	0.0007	0.0012	0.0020
6	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0001	0.0002	0.0003
7	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000

Example: Find  $P(X = 2)$  if  $\lambda = 0.50$

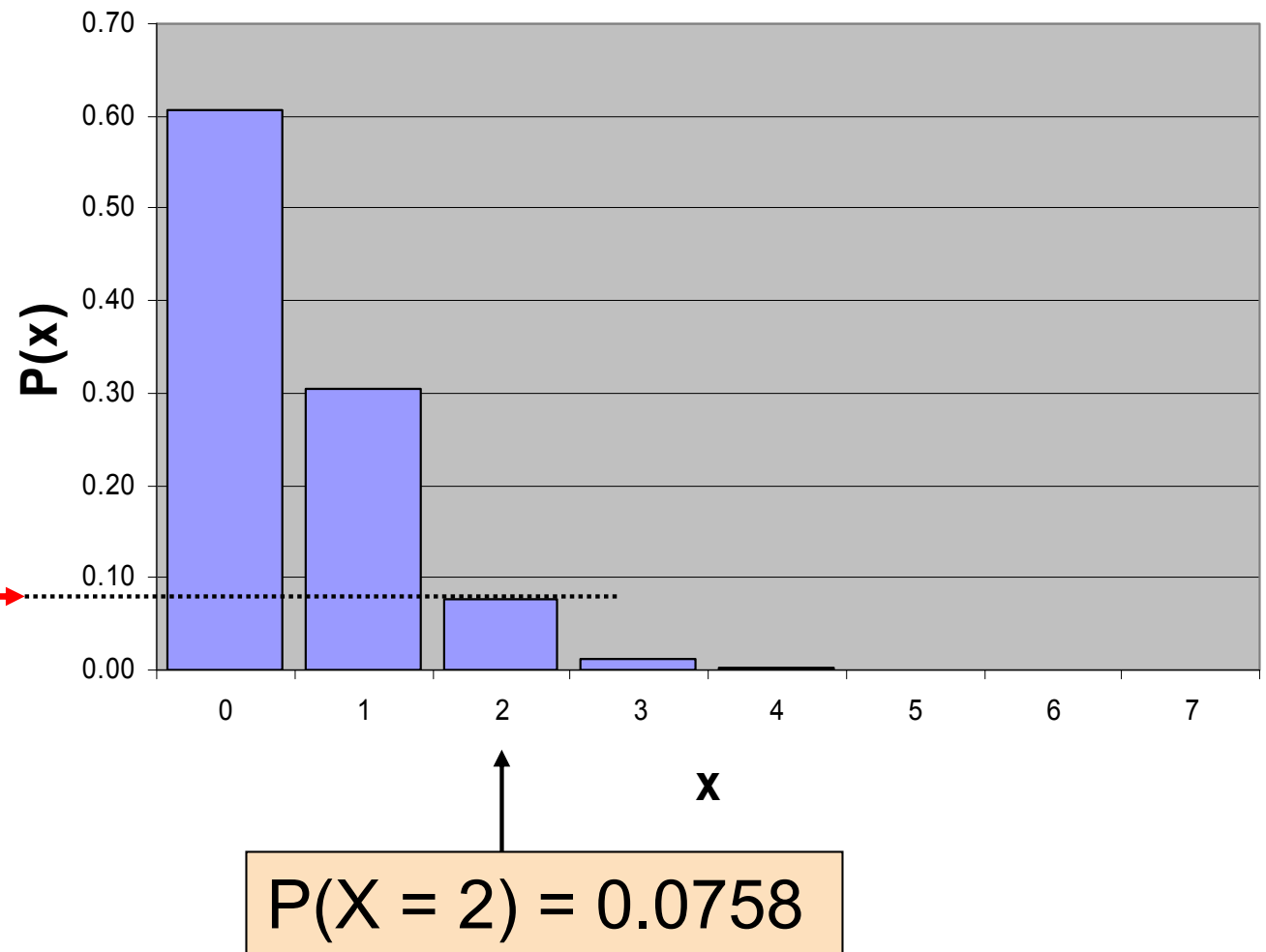
$$P(X = 2) = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-0.50} (0.50)^2}{2!} = 0.0758$$

# Graph of Poisson Probabilities

Graphically:

$\lambda = 0.50$

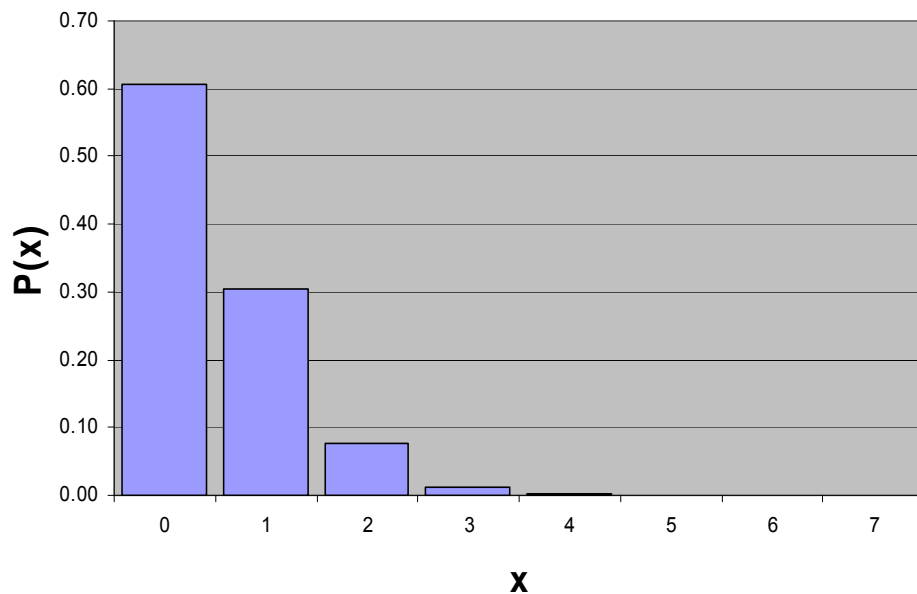
<b>X</b>	<b><math>\lambda = 0.50</math></b>
0	0.6065
1	0.3033
2	0.0758
3	0.0126
4	0.0016
5	0.0002
6	0.0000
7	0.0000



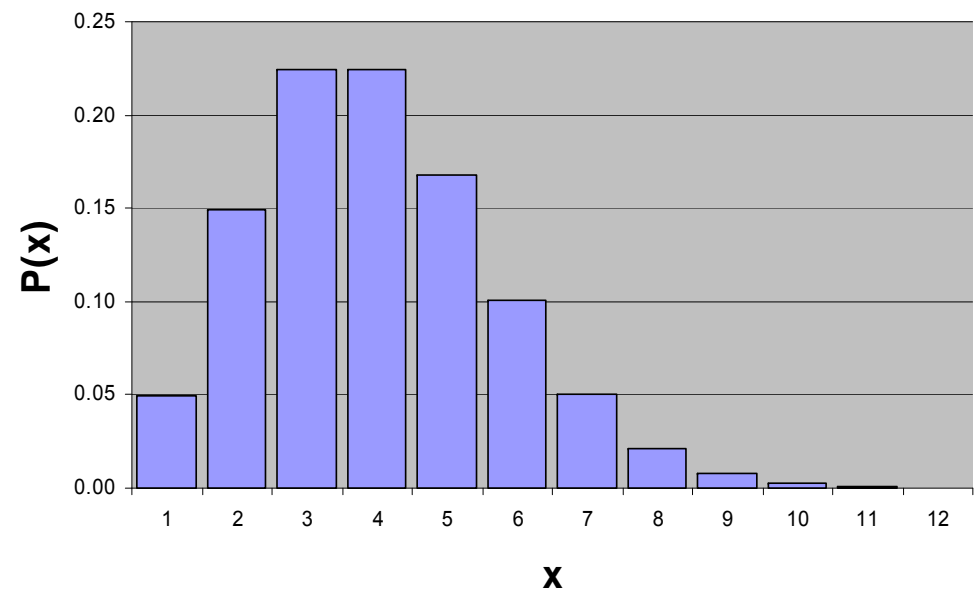
# Poisson Distribution Shape

- The shape of the Poisson Distribution depends on the parameter  $\lambda$  :

$$\lambda = 0.50$$



$$\lambda = 3.00$$



# Chapter Summary

- Addressed the probability of a discrete random variable
- Discussed the Binomial distribution
- Discussed the Poisson distribution