

GLOBAL
EDITION 



Business Statistics

A First Course

SEVENTH EDITION

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Chapter 5

Discrete Probability Distributions

Objectives

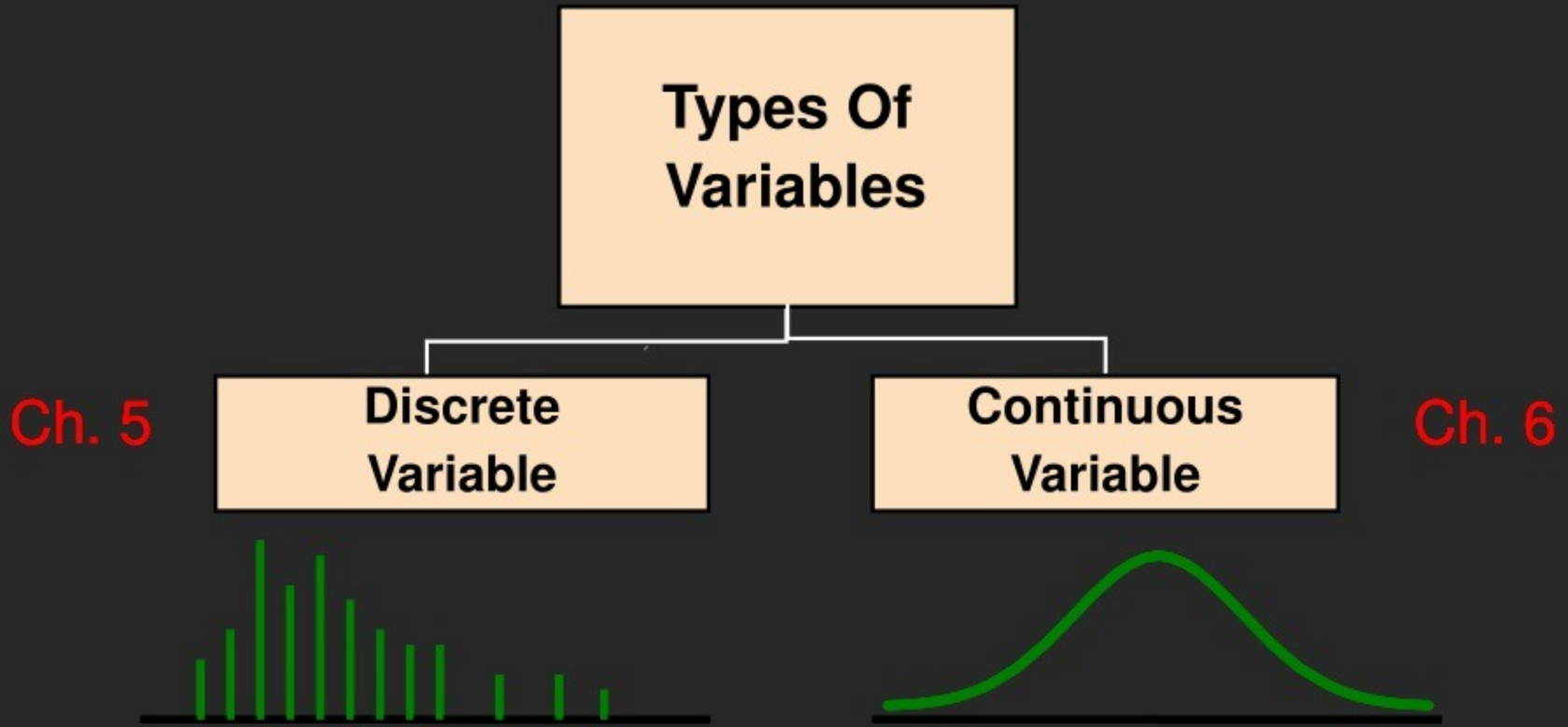
In this chapter, you learn:

- ↔ The properties of a probability distribution.
- ↔ To compute the expected value and variance of a probability distribution.
- ↔ To compute probabilities from binomial, and Poisson distributions.
- ↔ To use the binomial, and Poisson distributions to solve business problems

Definitions

- ↔ **Discrete** variables produce outcomes that come from a counting process (e.g. number of classes you are taking).
- ↔ **Continuous** variables produce outcomes that come from a measurement (e.g. your annual salary, or your weight).

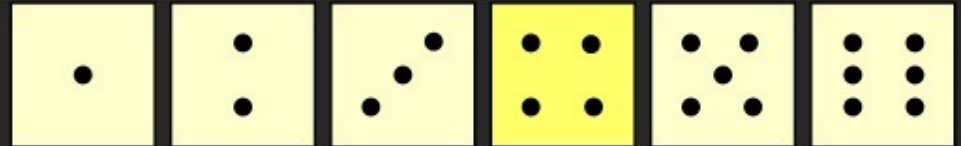
Types Of Variables



Discrete Variables

↔ Can only assume a countable number of values

Examples:



↔ Roll a die twice

Let X be the number of times 4 occurs
(then X could be 0, 1, or 2 times)

↔ Toss a coin 5 times.

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

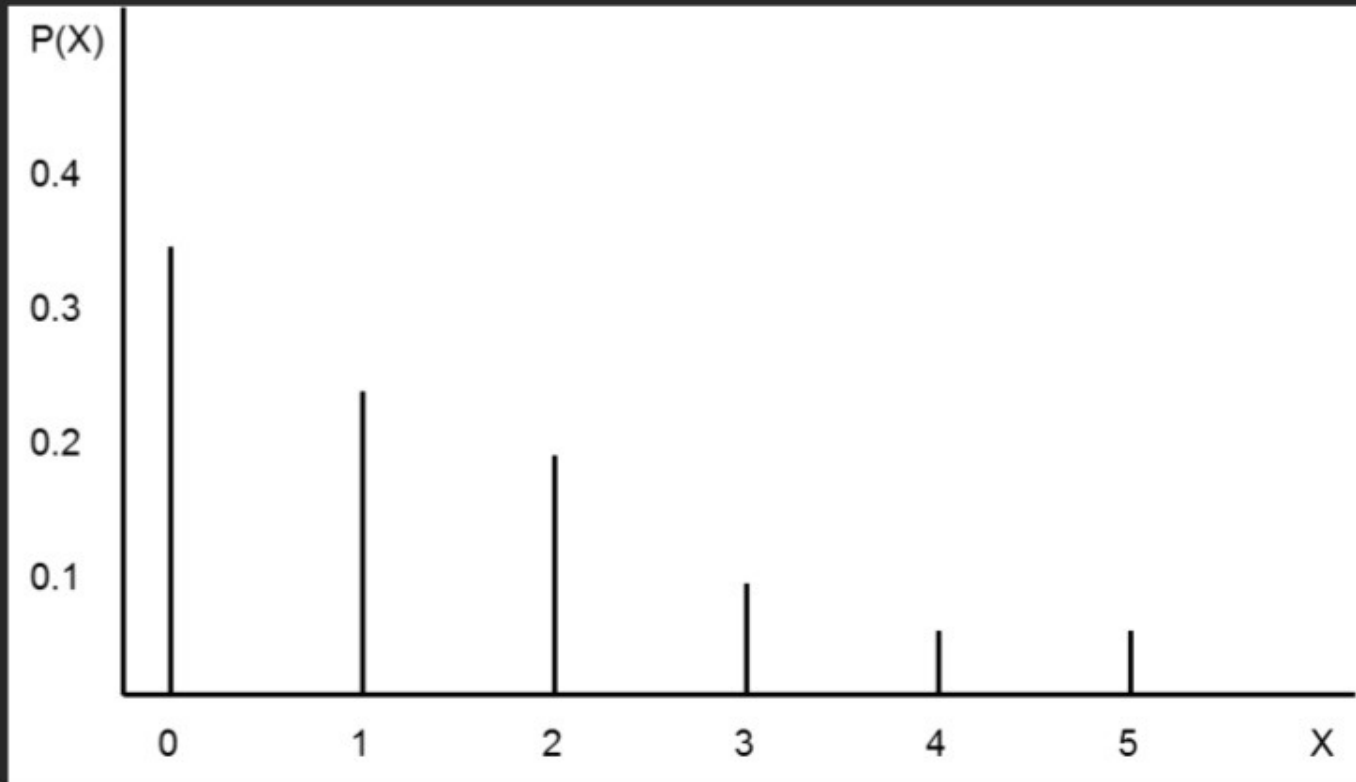


Probability Distribution For A Discrete Variable

- ↔ A **probability distribution for a discrete variable** is a mutually exclusive listing of all possible numerical outcomes for that variable and a probability of occurrence associated with each outcome.

Interruptions Per Day In Computer Network	Probability
0	0.35
1	0.25
2	0.20
3	0.10
4	0.05
5	0.05

Probability Distributions Are Often Represented Graphically



Discrete Variables Expected Value (Measuring Center)

↔ Expected Value (or mean) of a discrete variable (Weighted Average)

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

Interruptions Per Day In Computer Network (x_i)	Probability $P(X = x_i)$	$x_i P(X = x_i)$
0	0.35	$(0)(0.35) = 0.00$
1	0.25	$(1)(0.25) = 0.25$
2	0.20	$(2)(0.20) = 0.40$
3	0.10	$(3)(0.10) = 0.30$
4	0.05	$(4)(0.05) = 0.20$
5	0.05	$(5)(0.05) = 0.25$
	1.00	$\mu = E(X) = 1.40$

Discrete Variables: Measuring Dispersion

(continued)

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

Interruptions Per Day In Computer Network (x_i)	Probability $P(X = x_i)$	$[x_i - E(X)]^2$	$[x_i - E(X)]^2 P(X = x_i)$
0	0.35	$(0 - 1.4)^2 = 1.96$	$(1.96)(0.35) = 0.686$
1	0.25	$(1 - 1.4)^2 = 0.16$	$(0.16)(0.25) = 0.040$
2	0.20	$(2 - 1.4)^2 = 0.36$	$(0.36)(0.20) = 0.072$
3	0.10	$(3 - 1.4)^2 = 2.56$	$(2.56)(0.10) = 0.256$
4	0.05	$(4 - 1.4)^2 = 6.76$	$(6.76)(0.05) = 0.338$
5	0.05	$(5 - 1.4)^2 = 12.96$	$(12.96)(0.05) = 0.648$
			$\sigma^2 = 2.04, \sigma = 1.4283$

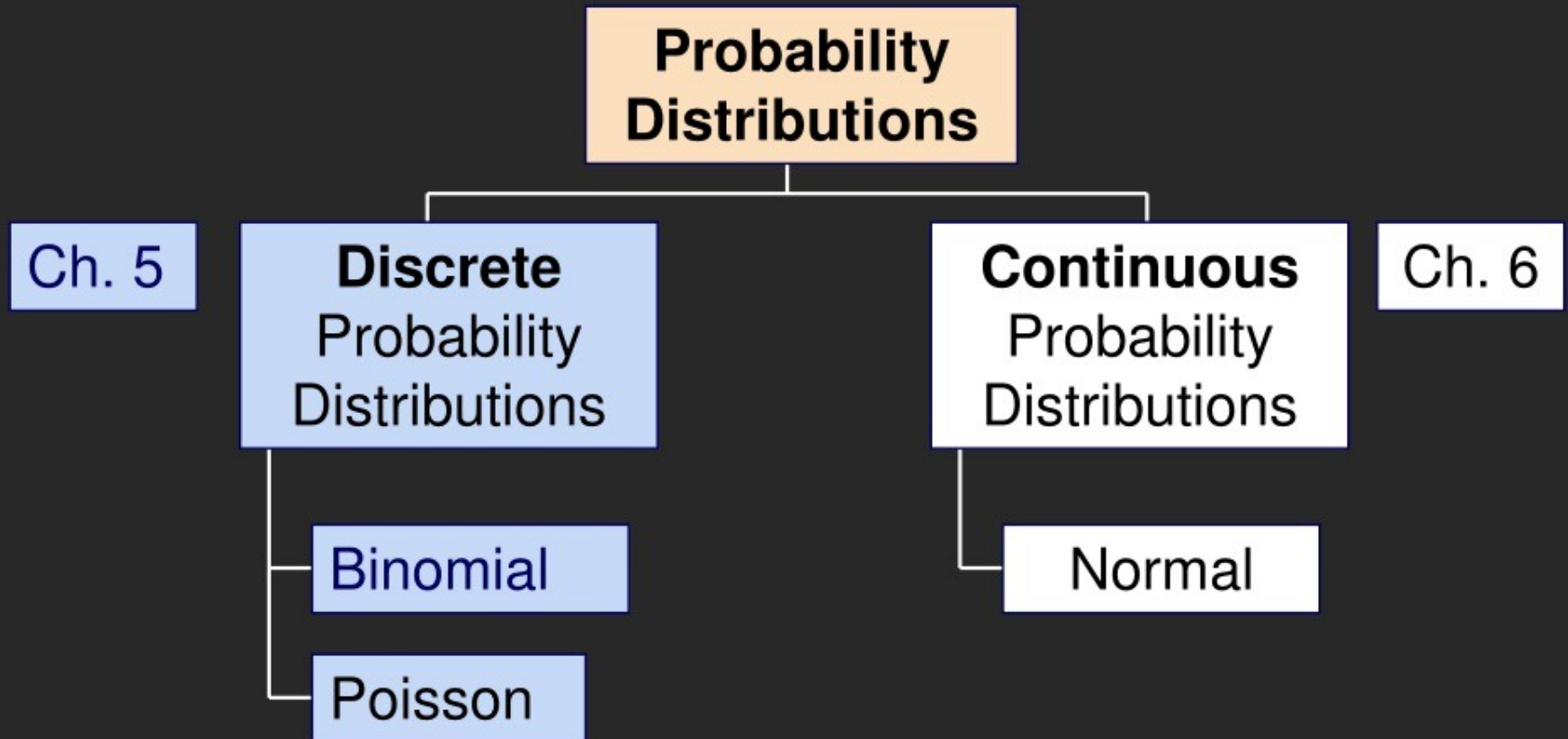
Discrete Variables: Measuring Dispersion

(continued)

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

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Probability Distributions



Binomial Probability Distribution

(continued)

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

Binomial Probability Distribution

(continued)

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Possible Applications for the Binomial Distribution

- ↔ 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

The Binomial Distribution

Counting Techniques

- ↔ Suppose the event of interest is obtaining heads on the toss of a fair coin. You are to toss the coin three times. In how many ways can you get two heads?
- ↔ Possible ways: HHT, HTH, THH, so there are three ways you can get two heads.
- ↔ This situation is fairly simple. We need to be able to count the number of ways for more complicated situations.

Counting Techniques

Rule of Combinations

- ↔ How many possible 3 scoop combinations could you create at an ice cream parlor if you have 31 flavors to select from and no flavor can be used more than once in the 3 scoops?
- ↔ The total choices is $n = 31$, and we select $X = 3$.

$${}_{31}C_3 = \frac{31!}{3!(31-3)!} = \frac{31!}{3!28!} = \frac{31 \cdot 30 \cdot 29 \cdot 28!}{3 \cdot 2 \cdot 1 \cdot 28!} = 31 \cdot 5 \cdot 29 = 4,495$$

Counting Techniques

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Binomial Distribution Formula

$$P(X=x | n, \pi) = \frac{n!}{x! (n-x)!} \pi^x (1-\pi)^{n-x}$$

$P(X=x|n,\pi)$ = probability of x events of interest in n trials, with the probability of an “event of interest” being π for each trial

x = number of “events of interest” in sample, ($x = 0, 1, 2, \dots, n$)

n = sample size (number of trials or observations)

π = probability of “event of interest”

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

$$n = 4$$

$$\pi = 0.5$$

$$1 - \pi = (1 - 0.5) = 0.5$$

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

Example: Calculating a Binomial Probability

What is the probability of one success in five observations if the probability of an event of interest is 0.1?

$$x = 1, n = 5, \text{ and } \pi = 0.1$$

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

The Binomial Distribution Example

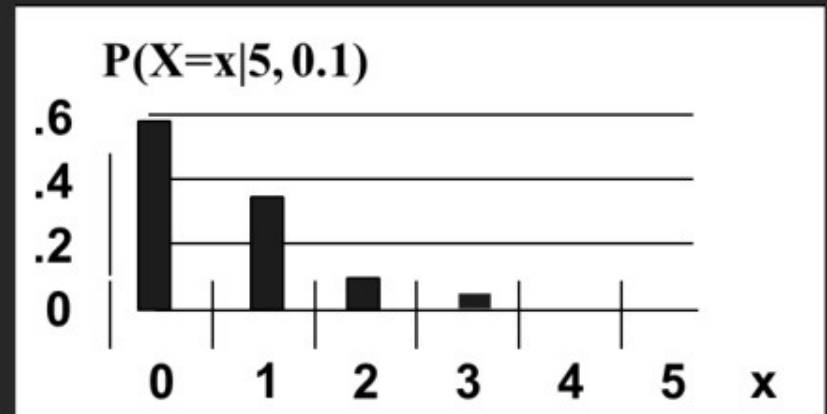
Suppose the probability of purchasing a defective computer is 0.02. What is the probability of purchasing 2 defective computers in a group of 10?

$$x = 2, n = 10, \text{ and } \pi = 0.02$$

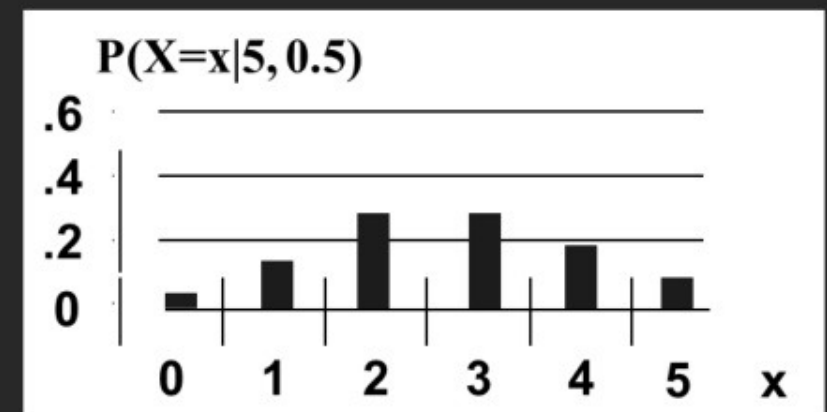
$$\begin{aligned} P(X = 2 | 10, 0.02) &= \frac{n!}{x!(n-x)!} \pi^x (1-\pi)^{n-x} \\ &= \frac{10!}{2!(10-2)!} (.02)^2 (1-.02)^{10-2} \\ &= (45)(.0004)(.8508) \\ &= .01531 \end{aligned}$$

The Binomial Distribution Shape

- ↔ The shape of the binomial distribution depends on the values of π and n
- ↔ Here, $n = 5$ and $\pi = .1$



- ↔ Here, $n = 5$ and $\pi = .5$



The Binomial Distribution Using Binomial Tables (Available On Line)

n = 10									
x	...	$\pi=.20$	$\pi=.25$	$\pi=.30$	$\pi=.35$	$\pi=.40$	$\pi=.45$	$\pi=.50$	
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	0.2522	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	0.0004	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
	...	$\pi=.80$	$\pi=.75$	$\pi=.70$	$\pi=.65$	$\pi=.60$	$\pi=.55$	$\pi=.50$	x

Examples:

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

$$n = 10, \pi = 0.75, x = 8: \quad P(X = 8|10, 0.75) = 0.2816$$

Binomial Distribution Characteristics

↔ Mean

$$\mu = E(X) = n\pi$$

↔ Variance and Standard Deviation

$$\sigma^2 = n\pi(1-\pi)$$

$$\sigma = \sqrt{n\pi(1-\pi)}$$

Where n = sample size

π = probability of the event of interest for any trial

$(1 - \pi)$ = probability of no event of interest for any trial

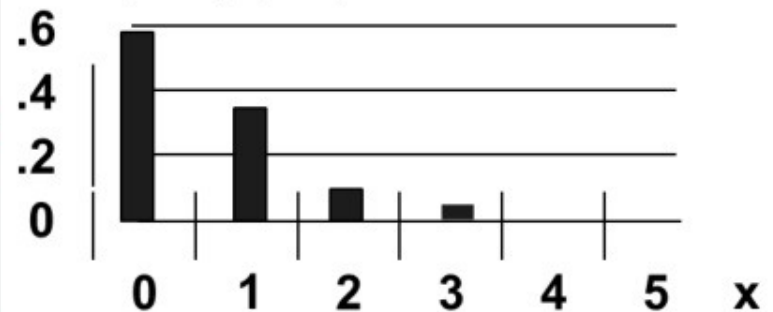
The Binomial Distribution Characteristics

Examples

$$\mu \bullet n\pi \bullet (5)(.1) \bullet 0.5$$

$$\sigma \bullet \sqrt{n\pi(1-\pi)} \bullet \sqrt{(5)(.1)(1-.1)} \bullet 0.6708$$

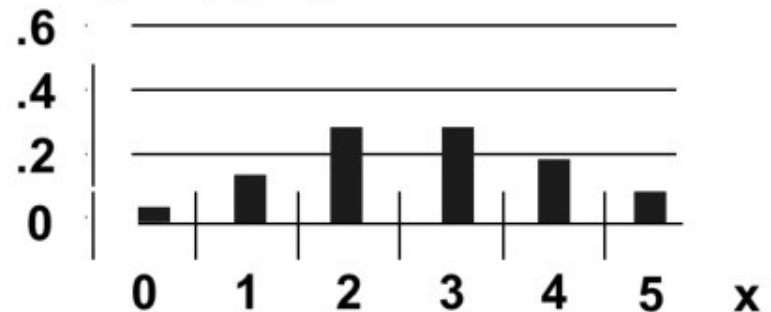
$P(X=x|5, 0.1)$



$$\mu \bullet n\pi \bullet (5)(.5) \bullet 2.5$$

$$\sigma \bullet \sqrt{n\pi(1-\pi)} \bullet \sqrt{(5)(.5)(1-.5)} \bullet 1.118$$

$P(X=x|5, 0.5)$



The Poisson Distribution

Definitions

- ↔ You use the **Poisson distribution** when you are interested in the number of times an event occurs in a given **area of opportunity**.
- ↔ An **area of opportunity** is a continuous unit or interval of time, volume, or such area in which more than one occurrence of an event can occur.
 - ↔ The number of scratches in a car's paint
 - ↔ The number of mosquito bites on a person
 - ↔ The number of computer crashes in a day

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 λ (lambda)

Poisson Distribution Formula

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

where:

x = number of events in an area of opportunity

λ = expected number of events

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

Poisson Distribution Characteristics

↔ Mean

$$\mu \bullet \lambda$$

↔ Variance and Standard Deviation

$$\sigma^2 \bullet \lambda$$

$$\sigma \bullet \sqrt{\lambda}$$

where λ = expected number of events

Using Poisson Tables (Available On Line)

X	λ								
	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	0.3098	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 $P(X = 2 \mid \lambda = 0.50)$

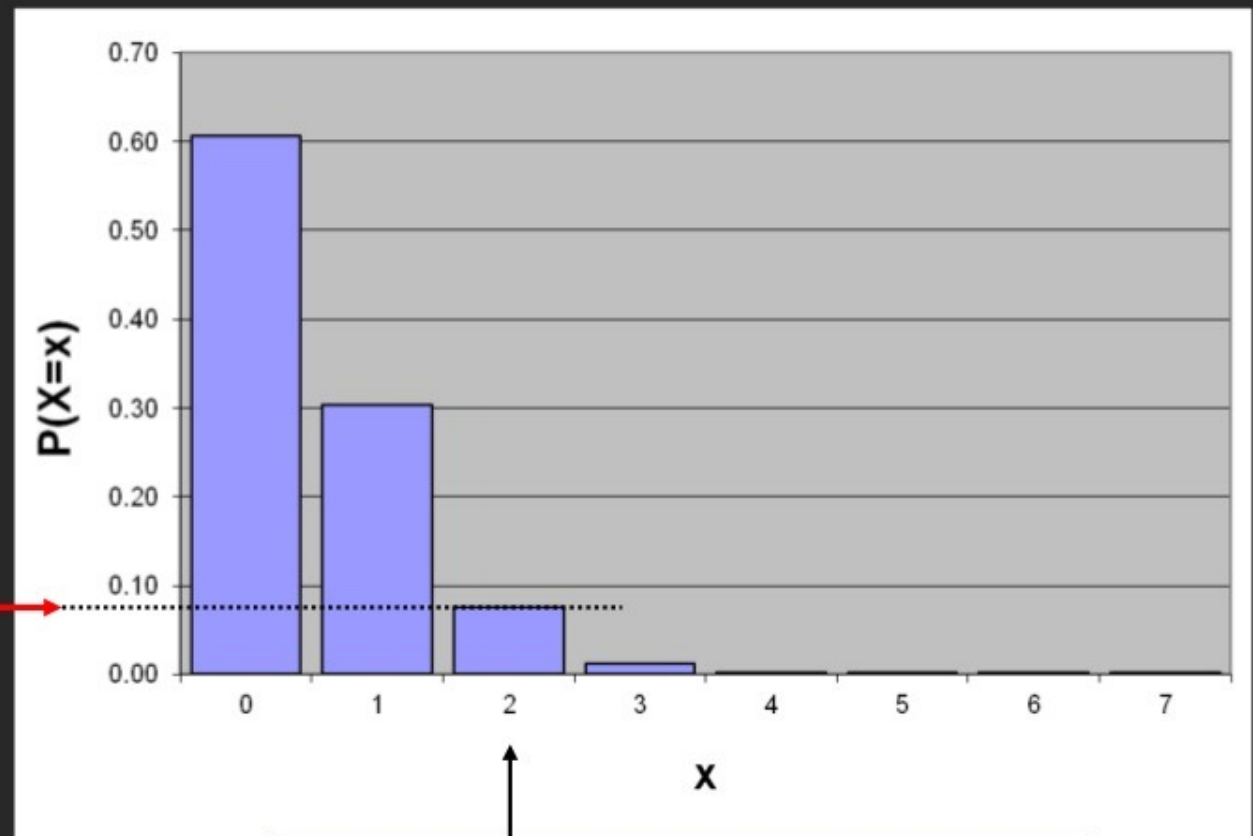
$$P(X = 2 \mid 0.50) = \frac{e^{-\lambda} \lambda^x}{x!} = \frac{e^{-0.50} (0.50)^2}{2!} = 0.0758$$

Graph of Poisson Probabilities

Graphically:

$$\lambda = 0.50$$

X	$\lambda = 0.50$
0	0.6065
1	0.3033
2	0.0758
3	0.0126
4	0.0016
5	0.0002
6	0.0000
7	0.0000

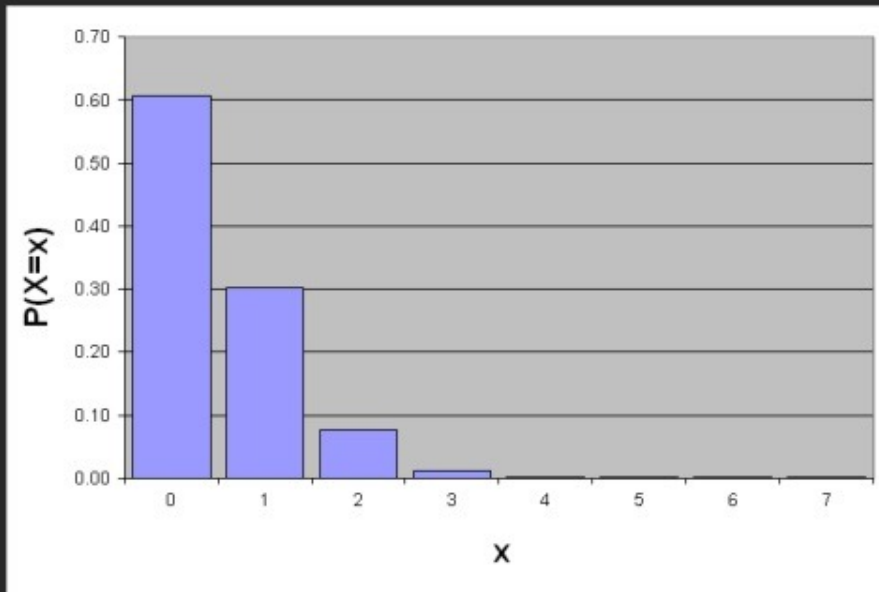


$$P(X = 2 \mid \lambda = 0.50) = 0.0758$$

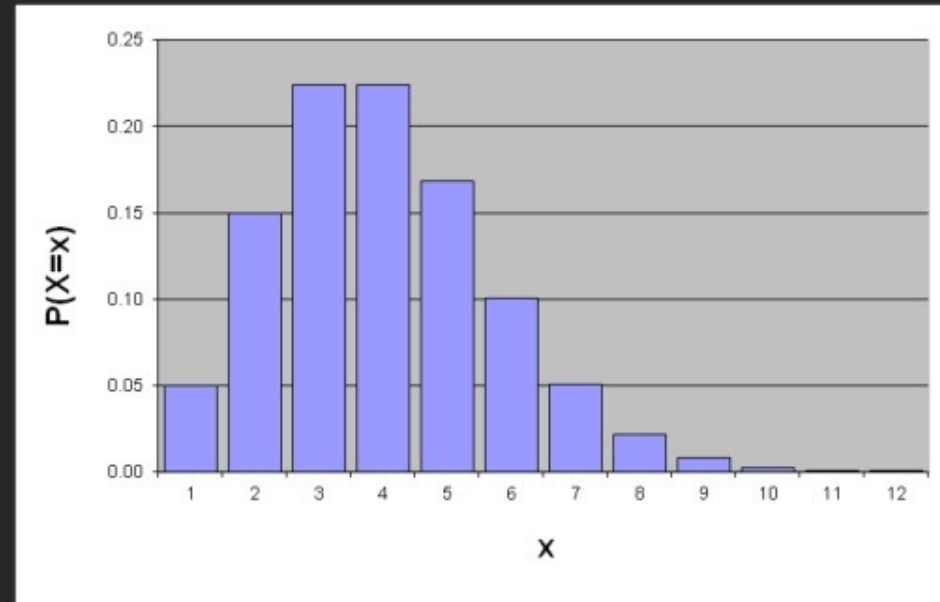
Poisson Distribution Shape

↔ The shape of the Poisson Distribution depends on the parameter λ :

$$\lambda = 0.50$$



$$\lambda = 3.00$$



Chapter Summary

In this chapter we covered:

- ↔ The properties of a probability distribution.
- ↔ To compute the expected value and variance of a probability distribution.
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- ↔ To use the binomial, and Poisson distributions to solve business problems