## DISCRETE AND CONTINUOUS PROBABILITY DISTRIBUTIONS

## PROBABILITY DISTRIBUTION

- It describes how the outcomes of an experiment are expected to vary


## Probability Distributions



## DISCRETE PROBABILITY DISTRIBUTIONS

- A discrete random variable is a variable that can assume only a countable number of values

Many possible outcomes:

- number of complaints per day
- number of TV's in a household
- number of rings before the phone is answered

Only two possible outcomes:

- gender: male or female
- defective: yes or no
- spreads peanut butter first vs. spreads jelly first


## CONTINUOUS PROBABILITY DISTRIBUTIONS

- A continuous random variable is a variable that can assume any value on a continuum (can assume an uncountable number of values)
- thickness of an item
- time required to complete a task
- temperature of a solution
- height, in inches
- These can potentially take on any value, depending only on the ability to measure accurately.


## The Binomial Distribution



## THE BINOMIAL DISTRIBUTION

- Characteristics of the Binomial Distribution:
- A trial has only two possible outcomes - "success" or "failure"
- There is a fixed number, $n$, of identical trials
- The trials of the experiment are independent of each other
- The probability of a success, p , remains constant from trial to trial
- If $p$ represents the probability of a success, then $(1-p)=q$ is the probability of a failure


## BINOMIAL DISTRIBUTION SETTINGS

- A manufacturing plant labels items as either defective or acceptable
- A firm bidding for a contract will either get the 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


## COUNTING RULE FOR COMBINATIONS

$\circ$ A combination is an outcome of an experiment where x objects are selected from a group of n objects

$$
C_{x}^{n}=\frac{n!}{x!(n-x)!}
$$

where:

$$
\begin{aligned}
& n!=n(n-1)(n-2) \ldots(2)(1) \\
& x!=x(x-1)(x-2) \ldots(2)(1) \\
& 0!=1 \text { (by definition) }
\end{aligned}
$$

## BINOMIAL DISTRIBUTION FORMULA

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

$\mathrm{P}(\mathrm{x})=$ probability of x successes in n trials, with probability of success p on each trial
$x$ = number of 'successes' in sample, ( $x=0,1,2, \ldots, n$ )
$p$ = probability of "success" per trial
q = probability of "failure" $=(1-p)$
$\mathrm{n}=$ number of trials (sample size)

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

$$
\begin{gathered}
n=4 \\
p=0.5 \\
q=(1-.5)=.5 \\
x=0,1,2,3,4
\end{gathered}
$$

## BINOMIAL DISTRIBUTION

- The shape of the binomial distribution depends on the values of p and n
Here, $\mathrm{n}=5$ and $\mathrm{p}=.1$


Here, $\mathrm{n}=5$ and $\mathrm{p}=.5$


## BINOMIAL DISTRIBUTION CHARACTERISTICS

- Mean

$$
\mu=E(x)=n p
$$

## Variance and Standard Deviation

$$
\begin{aligned}
& \sigma^{2}=\mathrm{npq} \\
& \hline \sigma=\sqrt{\mathrm{npq}}
\end{aligned}
$$

Where $n=$ sample size
$\mathrm{p}=$ probability of success
$\mathrm{q}=(1-\mathrm{p})=$ probability of failure

## BINOMIAL CHARACTERISTICS

Examples

$$
\begin{aligned}
\mu=n p= & (5)(.1)=0.5 \\
\sigma=\sqrt{n p q} & =\sqrt{(5)(.1)(1-.1)} \\
& =0.6708
\end{aligned}
$$


$\mu=n p=(5)(.5)=2.5$
$\sigma=\sqrt{\mathrm{npq}}=\sqrt{(5)(.5)(1-.5)}$ $=1.118$

## USING BINOMIAL TABLES

| $\mathbf{g l \|}$ |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\mathbf{x}$ | $\mathrm{p}=.15$ | $\mathrm{p}=.20$ | $\mathrm{p}=.25$ | $\mathrm{p}=.30$ | $\mathbf{p}=. \mathbf{3 5}$ | $\mathrm{p}=.40$ | $\mathrm{p}=.45$ | $\mathrm{p}=.50$ |  |
| 0 | 0.1969 | 0.1074 | 0.0563 | 0.0282 | 0.0135 | 0.0060 | 0.0025 | 0.0010 | 10 |
| 1 | 0.3474 | 0.2684 | 0.1877 | 0.1211 | 0.0725 | 0.0403 | 0.0207 | 0.0098 | 9 |
| 2 | 0.2759 | 0.3020 | 0.2816 | 0.2335 | 0.1757 | 0.1209 | 0.0763 | 0.0439 | 8 |
| $\mathbf{3}$ | 0.1298 | 0.2013 | 0.2503 | 0.2668 | $\mathbf{0 . 2 5 2 2}$ | 0.2150 | 0.1665 | 0.1172 | 7 |
| 4 | 0.0401 | 0.0881 | 0.1460 | 0.2001 | 0.2377 | 0.2508 | 0.2384 | 0.2051 | 6 |
| 5 | 0.0085 | 0.0264 | 0.0584 | 0.1029 | 0.1536 | 0.2007 | 0.2340 | 0.2461 | 5 |
| 6 | 0.0012 | 0.0055 | 0.0162 | 0.0368 | 0.0689 | 0.1115 | 0.1596 | 0.2051 | 4 |
| 7 | 0.0001 | 0.0008 | 0.0031 | 0.0090 | 0.0212 | 0.0425 | 0.0746 | 0.1172 | 3 |
| 8 | 0.0000 | 0.0001 | $\mathbf{0 . 0 0 0 4}$ | 0.0014 | 0.0043 | 0.0106 | 0.0229 | 0.0439 | $\mathbf{2}$ |
| 9 | 0.0000 | 0.0000 | 0.0000 | 0.0001 | 0.0005 | 0.0016 | 0.0042 | 0.0098 | 1 |
| 10 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0000 | 0.0001 | 0.0003 | 0.0010 | 0 |

## Examples:

$$
n=10, p=.35, x=3: \quad P(x=3 \mid n=10, p=.35)=.2522
$$

$$
n=10, p=.75, x=2: \quad P(x=2 \mid n=10, p=.75)=.0004
$$

## The Poisson Distribution

## Probability Distributions

## Discrete <br> Probability <br> Distributions

Binomial
Poisson
Hypergeometric

## THE POISSON DISTRIBUTION

- Characteristics of the Poisson Distribution:
- The outcomes of interest are rare relative to the possible outcomes
- The average number of outcomes of interest per time or space interval is ${ }^{66} \lambda$ "
- The number of outcomes of interest are random, and the occurrence of one outcome does not influence the chances of another outcome of interest
- The probability of that an outcome of interest occurs in a given segment is the same for all segments


## POISSON DISTRIBUTION FORMULA

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

where:
$t=$ size of the segment of interest
$x=$ number of successes in segment of interest
$\lambda=$ expected number of successes in a segment of unit size
$\mathrm{e}=$ base of the natural logarithm system (2.71828...)

## POISSON DISTRIBUTION CHARACTERISTICS

- Mean

$$
\mu=\lambda t
$$

## Variance and Standard Deviation

$$
\begin{aligned}
& \sigma^{2}=\lambda t \\
& \sigma=\sqrt{\lambda t}
\end{aligned}
$$

where
$\lambda=$ number of successes in a segment of unit size
$\mathrm{t}=$ the size of the segment of interest

## USING POISSON TABLES

| $\mathbf{X}$ | $\boldsymbol{\lambda}$ |  | 0.10 | 0.20 | 0.30 | 0.40 | $\mathbf{0 . 5 0}$ | 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 |  |
| $\mathbf{2}$ | 0.0045 | 0.0164 | 0.0333 | 0.0536 | $\mathbf{0 . 0 7 5 8}$ | 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 $\lambda=.05$ and $t=10$

$$
P(x=2)=\frac{(\lambda t)^{x} e^{-\lambda t}}{x!}=\frac{(0.50)^{2} e^{-0.50}}{2!}=.0758
$$

## GRAPH OF POISSON PROBABILITIES

Graphically:
$\lambda=.05$ and $t=100$

| $\mathbf{X}$ | $\mathbf{\lambda t}=$ |
| :---: | :---: |
| $\mathbf{0 . 5 0}$ |  |
| 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 parameters $\lambda$ and t :

$$
\lambda t=0.50
$$



$$
\lambda t=3.0
$$



## The Hypergeometric Distribution

## Probability Distributions

## Discrete <br> Probability <br> Distributions

Binomial
Poisson
Hypergeometric

## THE HYPERGEOMETRIC DISTRIBUTION

- "n" trials in a sample taken from a finite population of size N
- Sample taken without replacement
- Trials are dependent
- Concerned with finding the probability of " x " successes in the sample where there are "X" successes in the population


## HYPERGEOMETRIC DISTRIBUTION FORMULA

(Two possible outcomes per trial)

$$
P(x)=\frac{C_{n-x}^{N-X} \cdot C_{x}^{x}}{C_{n}^{N}}
$$

Where
$\mathrm{N}=$ Population size
$X=$ number of successes in the population
$\mathrm{n}=$ sample size
$x=$ number of successes in the sample
$n-x=$ number of failures in the sample

## HYPERGEOMETRIC DISTRIBUTION FORMULA

- Example: 3 Light bulbs were selected from 10. Of the 10 there were 4 defective. What is the probability that 2 of the 3 selected are defective?

$$
\begin{array}{ll}
\mathrm{N}=10 & \mathrm{n}=3 \\
\mathrm{X}=4 & \mathrm{x}=2
\end{array}
$$

$$
P(x=2)=\frac{C_{n-x}^{N-x} C_{x}^{x}}{C_{n}^{N}}=\frac{C_{1}^{6} C_{2}^{4}}{C_{3}^{10}}=\frac{(6)(6)}{120}=0.3
$$

## The Normal Distribution

## Probability Distributions

Continuous
Probability
Distributions

Normal
Uniform
Exponential

# THE NORMAL DISTRIBUTION 

- 'Bell Shaped'
- Symmetrical
- Mean, Median and Mode are Equal
Location is determined by the mean, $\boldsymbol{\mu}$
Spread is determined by the standard deviation, $\sigma$


Mean
= Median
= Mode

The random variable has an infinite theoretical range:

The general formula for the probability density function of the normal distribution is

$$
f(x)=\frac{e^{-(x-\mu)^{2} /\left(2 \sigma^{2}\right)}}{\sigma \sqrt{2 \pi}}
$$

Where $\mu$ is mean and $\sigma$ is standard deviation

## Many Normal Distributions



By varying the parameters $\mu$ and $\sigma$, we obtain different normal distributions

## THE NORMAL DISTRIBUTION SHAPE



## FINDING NORMAL PROBABILITIES

Probability is measured by the area under the curve


## PROBABILITY AS AREA UNDER THE CURVE

The total area under the curve is 1.0 , and the curve is symmetric, so half is above the mean, half is below


## EMPIRICAL RULES

What can we say about the distribution of values around the mean? There are some general rules:


## THE EMPIRICAL RULE

$\mu \pm 2 \sigma$ covers about $95 \%$ of $\times$ 's
$\mu \pm 3 \sigma$ covers about $99.7 \%$ of $\mathbf{x}$ 's



## THE STANDARD NORMAL DISTRIBUTION

- Also known as the "z" distribution - Mean is defined to be 0
- Standard Deviation is 1


Values above the mean have positive $z$-values, values below the mean have negative $z$-values

## THE STANDARD NORMAL

- Any normal distribution (with any mean and standard deviation combination) can be transformed into the standard normal distribution (z)
- Need to transform x units into z units


## TRANSLATION TO THE STANDARD NORMAL DISTRIBUTION

- Translate from x to the standard normal (the " z " distribution) by subtracting the mean of x and dividing by its standard deviation:

$$
z=\frac{x-\mu}{\sigma}
$$

## EXAMPLE

- If x is distributed normally with mean of 100 and standard deviation of 50 , the $z$ value for $x=250$ is

$$
z=\frac{x-\mu}{\sigma}=\frac{250-100}{50}=3.0
$$

- This says that $\mathrm{x}=250$ is three standard deviations (3 increments of 50 units) above the mean of 100 .


## COMPARING X AND Z UNITS



Note that the distribution is the same, only the scale has changed. We can express the problem in original units ( $\mathbf{x}$ ) or in standardized units ( $\mathbf{z}$ )

## THE STANDARD NORMAL TABLE

The column gives the value of $z$ to the second decimal point

The row shows the value of $z$ to the first decimal point

The value within the table gives the probability from $z=0$ up to the desired z value

## GENERAL PROCEDURE FOR FINDING PROBABILITIES

To find $\mathrm{P}(\mathrm{a}<\mathrm{x}<\mathrm{b})$ when x is distributed normally:

- Draw the normal curve for the problem in terms of $x$
- Translate x -values to z -values
- Use the Standard Normal Table


## TABLE EXAMPLE

Suppose x is normal with mean 8.0 and standard deviation 5.0. Find $\mathrm{P}(8<\mathrm{x}<8.6)$

Calculate z-values:

$$
z=\frac{x-\mu}{\sigma}=\frac{8-8}{5}=0
$$

$$
z=\frac{x-\mu}{\sigma}=\frac{8.6-8}{5}=0.12
$$



## TABLE EXAMPLE

Suppose x is normal with mean 8.0 and standard deviation 5.0. Find $\mathrm{P}(8<\mathrm{x}<8.6)$


## SOLUTION: FINDING $\mathrm{P}(0<\mathrm{Z}<$ 0.12)

|  |  |  |  |
| :---: | :---: | :---: | :---: |
| z | 00 | 01 |  |
| 0.0 | . 0000 | . 0040 | . 0080 |
|  | . 0398 | . 0438 | . 04 |
| 0.2 | . 0793 | . 0832 | . 08 |
| 0.3 | . 1179 | . 1217 | . 12 |



## FINDING NORMAL PROBABILITIES

-Suppose x is normal with mean 8.0 and standard deviation 5.0.
o Now Find $\mathrm{P}(\mathrm{x}<8.6)$


## FINDING NORMAL PROBABILITIES

(continued)
-Suppose x is normal with mean 8.0 and standard deviation 5.0.
o Now Find $\mathrm{P}(\mathrm{x}<8.6)$

$$
\begin{aligned}
& \mathrm{P}(\mathrm{x}<8.6) \\
= & \mathrm{P}(\mathrm{z}<0.12) \\
= & \mathrm{P}(\mathrm{z}<0)+\mathrm{P}(0<\mathrm{z}<0.12) \\
= & .5+.0478=.5478
\end{aligned}
$$



## UPPER TAIL PROBABILITIES

-Suppose x is normal with mean 8.0 and standard deviation 5.0.
o Now Find $\mathrm{P}(\mathrm{x}>8.6)$


## UPPER TAIL PROBABILITIES

(continued)
o Now Find $\mathrm{P}(\mathrm{x}>8.6)$...

$$
\begin{aligned}
P(x>8.6)=P(z>0.12) & =P(z>0)-P(0<z<0.12) \\
& =.5-.0478=.4522
\end{aligned}
$$




## LOWER TAIL PROBABILITIES

-Suppose x is normal with mean 8.0 and standard deviation 5.0.

- Now Find $\mathrm{P}(7.4<\mathrm{x}<8)$



## LOWER TAIL PROBABILITIES

## Now Find $\mathrm{P}(7.4<\mathrm{x}<8) . .$.

The Normal distribution is symmetric, so we use the same table even if $z$-values are negative:

$$
\begin{aligned}
& \mathrm{P}(7.4<\mathrm{x}<8) \\
& =\mathrm{P}(-0.12<\mathrm{z}<0) \\
& =.0478
\end{aligned}
$$



