



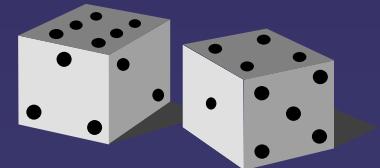
# Random Variables And Probability Distributions

DEPARTMENT OF APPLIED SCIENCES  
PARUL UNIVERSITY



# Topics

- Probability Distributions
- Expected Value and Variance of a RV



# Topics

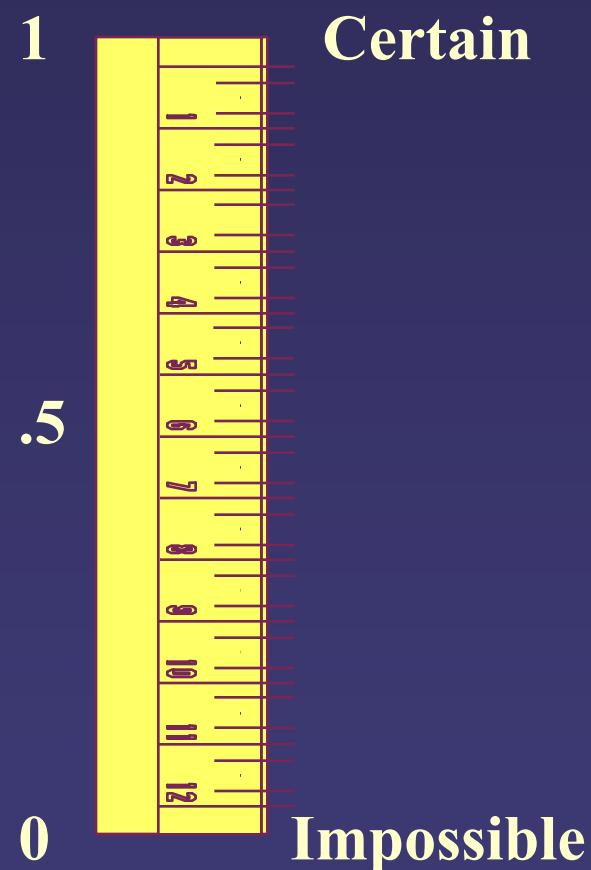
- Discrete Probability Distributions
  - Bernoulli and Binomial Distributions
  - Poisson Distributions
- Continuous Probability Distributions
  - Uniform
  - Normal

# Topics

- Random Sampling and Probability Distributions
  - Random Numbers
  - Sampling from Probability Distributions

# Probability

- Probability is the likelihood that the event will occur.
- Two Conditions:
  - Value is between 0 and 1.
  - Sum of the probabilities of all events must be 1.

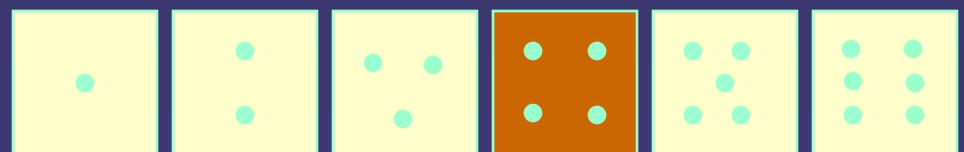


# Random Variable

- A numerical description of the outcome of an experiment

Example: Discrete RV: countable # of outcomes

Throw a die twice: Count the number of times 4 comes up (0, 1, or 2 times)



# Discrete Random Variable

- **Discrete Random Variable:**

- Obtained by Counting (0, 1, 2, 3, etc.)
- Usually finite by number of different values

e.g.

Toss a coin 5 times. Count the number of tails. (0, 1, 2, 3, 4, or 5 times)

# Random Variable

- A numerical description of the outcome of an experiment

*Example:* Continuous RV:

- The Value of the DJIA
- Time to repair a failed machine
- RandomVariable: Given by Capital Letters X & Y.
- Specific Values Given by lower case

# Probability Distribution

**Characterization of the possible values that a RV may assume along with the probability of assuming these values.**

# Discrete Probability Distribution

- List of all possible [  $x_i$ ,  $p(x_i)$  ] pairs
  - $X_i$  = value of random variable
  - $P(x_i)$  = probability associated with value
- Mutually exclusive (nothing in common)
- Collectively exhaustive (nothing left out)
  - $0 \leq p(x_i) \leq 1$
  - $\sum P(x_i) = 1$

# Weekly Demand of a Slow-Moving Product

## Probability Mass Function

Demand, $x$	Probability, $p(x)$
0	0.1
1	0.2
2	0.4
3	0.3
4 or more	0

# Weekly Demand of a Slow-Moving Product

A Cumulative Distribution Function:

Probability that RV assume a value  $\leq$  a given value,  $x$

Demand, $x$	Cumulative Probability, $P(x)$
0	0.1
1	0.3
2	0.7
3	1

# Discrete Probability Distribution Example

Event: Toss 2 Coins.

Count # Tails.



Probability distribution  
probability

Values

0

$1/4 = .25$

1

$2/4 = .50$

2

$1/4 = .25$

# Discrete Random Variable Summary Measures

## Expected value (The mean)

Weighted average of the probability distribution

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

In slow-moving product demand example,  
the expected value is :

$$E(X) = 0 \times 0.1 + 1 \times .2 + 2 \times .4 + 3 \times .3 = 1.9$$

The average demand on the long run is 1.9

# Discrete Random Variable Summary Measures

## Variance

Weighted average squared deviation about mean

$$\text{Var}[X] = \sigma^2 = E [ (x_i - E(X))^2 ] = \sum (x_i - E(X))^2 p(x_i)$$

For the Product demand example, the variance is:

$$\begin{aligned}\text{Var}[X] = \sigma^2 = & (0 - 1.9)^2(.1) + (1 - 1.9)^2(.2) + \\ & (2 - 1.9)^2(.4) + (3 - 1.9)^2(.3) = .89\end{aligned}$$

# Important Discrete Probability Distribution Models

Discrete Probability Distributions

Binomial

Poisson

# Binomial Distribution

- ‘N’ identical trials
  - Example: 15 tosses of a coin, 10 light bulbs taken from a warehouse
- 2 mutually exclusive outcomes on each trial
  - Example: Heads or tails in each toss of a coin, defective or not defective light bulbs

# Binomial Distributions

- Constant Probability for each Trial
  - Example: Probability of getting a tail is the same each time we toss the coin and each light bulb has the same probability of being defective
- 2 Sampling Methods:
  - Infinite Population Without Replacement
  - Finite Population With Replacement
- Trials are Independent:
  - The Outcome of One Trial Does Not Affect the Outcome of Another

# Binomial Probability Distribution Function

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

**P(X)** = probability that X successes given a knowledge of n and p

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

**p** = probability of each ‘success’

**n** = sample size

*Tails in 2 Tosses of Coin*

$\underline{X}$	$\underline{P(X)}$
0	$1/4 = .25$

1	$2/4 = .50$
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2	$1/4 = .25$
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# Binomial Distribution Characteristics

## Mean

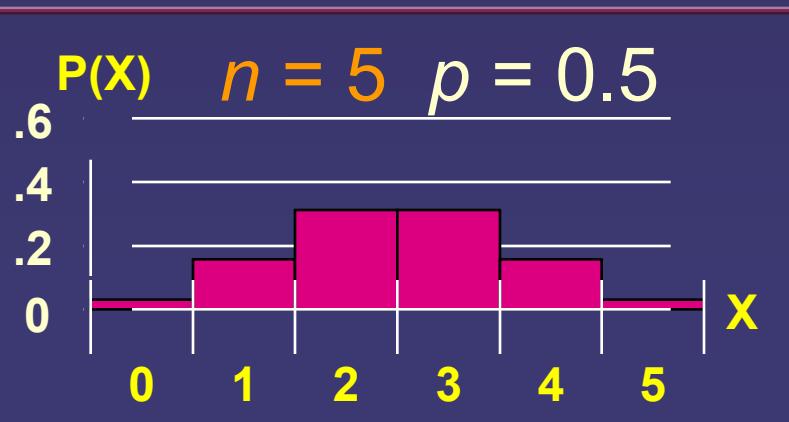
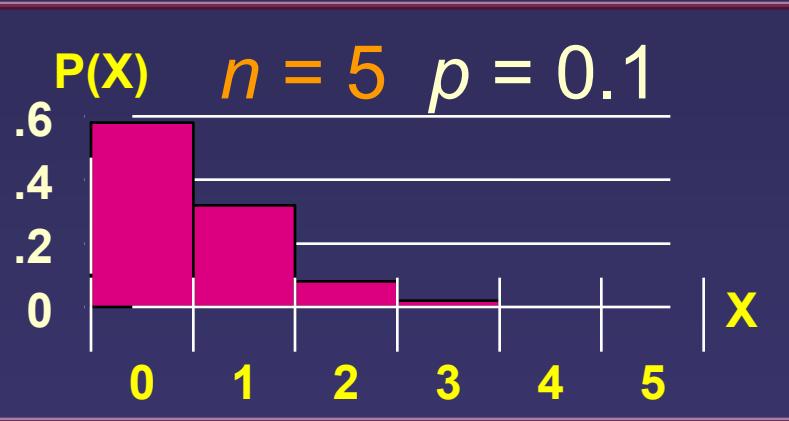
$$\mu = E(X) = np$$

e.g.  $\mu = 5 (.1) = .5$

## Standard Deviation

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

e.g.  $\sigma = \sqrt{5(.5)(1 - .5)} = 1.118$



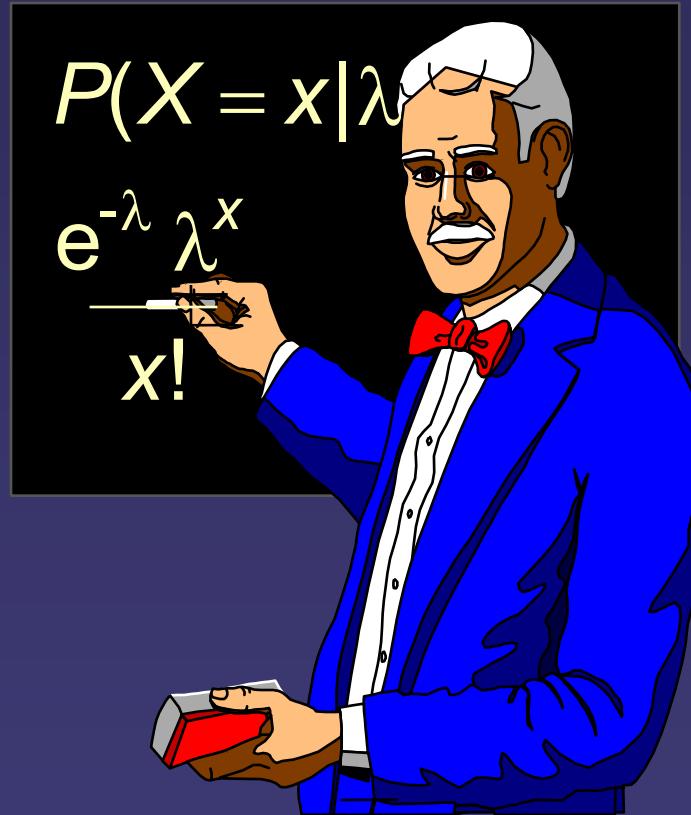
## Computing Binomial Probabilities using Excel Function BINOMDIST

Binomial Probabilities	
n	10
p	0.8
x	p(x)
0	0.000000
1	0.000004
2	0.000074
3	0.000786
4	0.005505
5	0.026424
6	0.088080
7	0.201327
8	0.301990
9	0.268435
10	0.107374

# Poisson Distribution

## Poisson process:

- Discrete events in an ‘interval’
  - The probability of one success in an interval is stable
  - The probability of more than one success in this interval is 0
- Probability of success is Independent from interval to Interval



# Examples of Poisson Distribution



# Poisson Distribution Function

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

$P(X)$  = probability of  $X$  successes given  $\lambda$

$\lambda$  = expected (mean) number of ‘successes’

$e$  = 2.71828 (base of natural logs)

$X$  = number of ‘successes’ per unit

e.g. Find the probability of 4 customers arriving in 3 minutes when the mean is 3.6

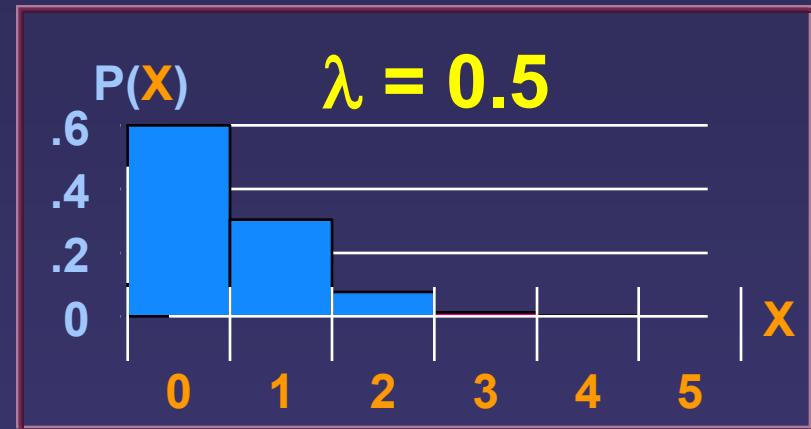
$$P(X) = \frac{e^{-3.6} 3.6^4}{4!} = .1912$$

# Poisson Distribution Characteristics

## Mean

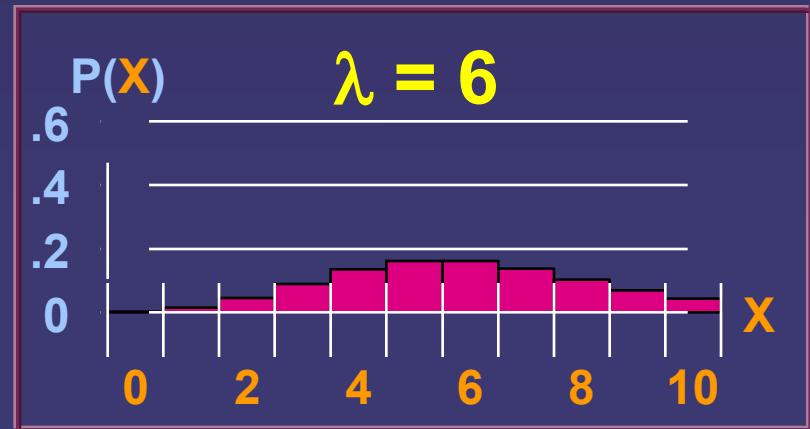
$$\mu = E(X) = \lambda$$

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



## Standard Deviation

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



# Computing Poisson Probabilities using Excel Function POISSON

Poisson Distribution

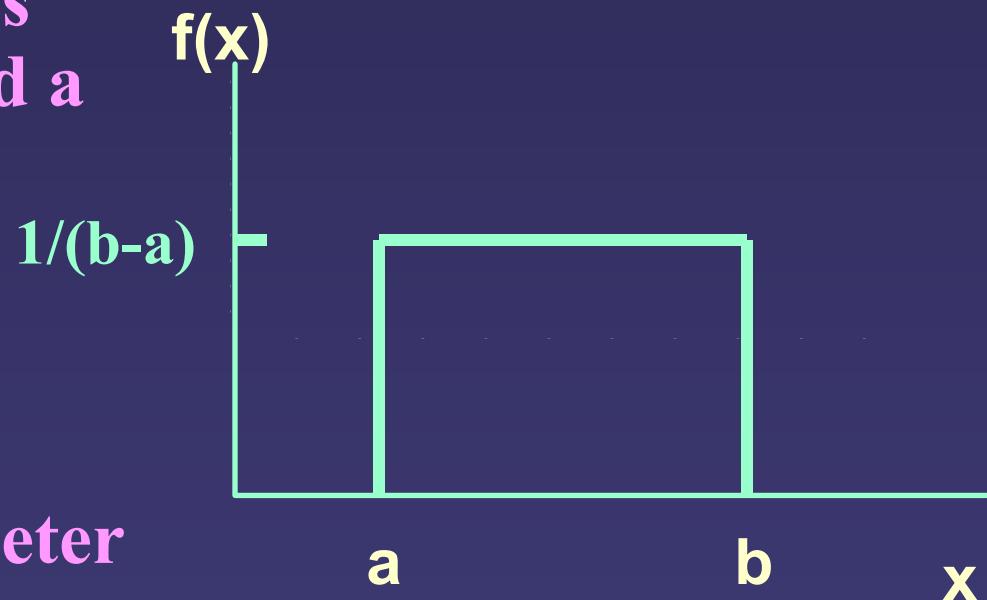
Mean	12
x	p(x)
1	0.00007
2	0.00044
3	0.00177
4	0.00531
5	0.01274
6	0.02548
7	0.04368
8	0.06552
9	0.08736
10	0.10484
11	0.11437
12	0.11437

# Continuous Probability Distributions

- Uniform
- Normal

# The Uniform Distribution

- Equally Likely chances of occurrences of RV values between a maximum and a minimum
- Mean =  $(b+a)/2$
- Variance =  $(b-a)^2/12$
- ‘a’ is a location parameter
- ‘ $b-a$ ’ is a scale parameter
- no shape parameter



# The Uniform Distribution

Probability Density Function

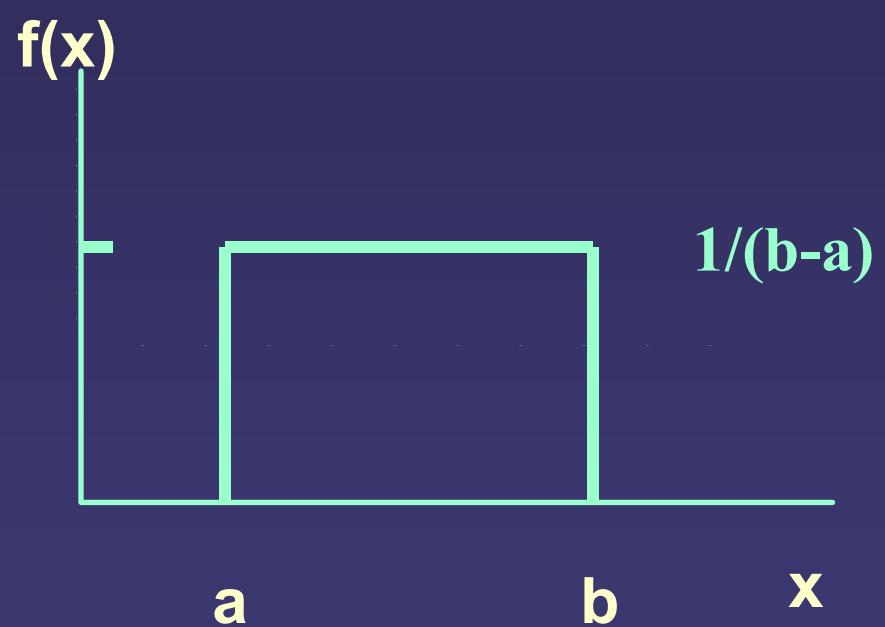
$$f(x) = \frac{1}{b-a} \quad \text{if } a \leq x \leq b$$

Distribution Function

$$F(x) = 0 \quad \text{if } x < a$$

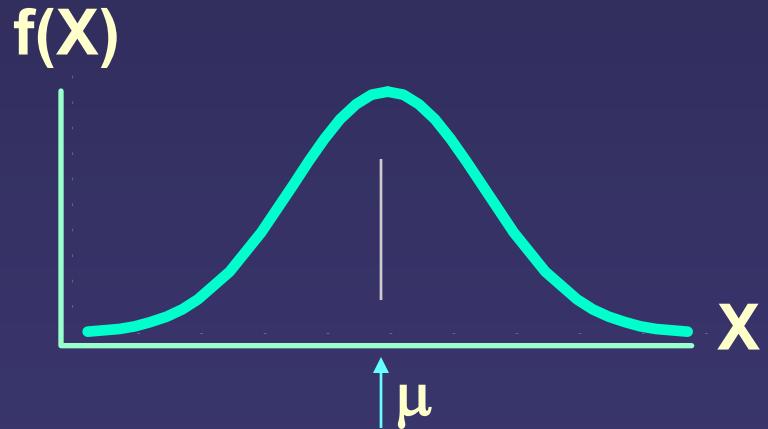
$$F(x) = \frac{x-a}{b-a} \quad \text{if } a \leq x \leq b$$

$$F(x) = 1 \quad \text{if } b < x$$



# The Normal Distribution

- ‘Bell Shaped’
- Symmetrical
- Mean, Median and Mode are Equal
- ‘Middle Spread’  
Equals  $1.33 \sigma$
- Random Variable has Infinite Range



Mean  
Median  
Mode

# The Mathematical Model

$$f(X) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(X-\mu)^2}$$

$f(X)$  = frequency of random variable  $X$

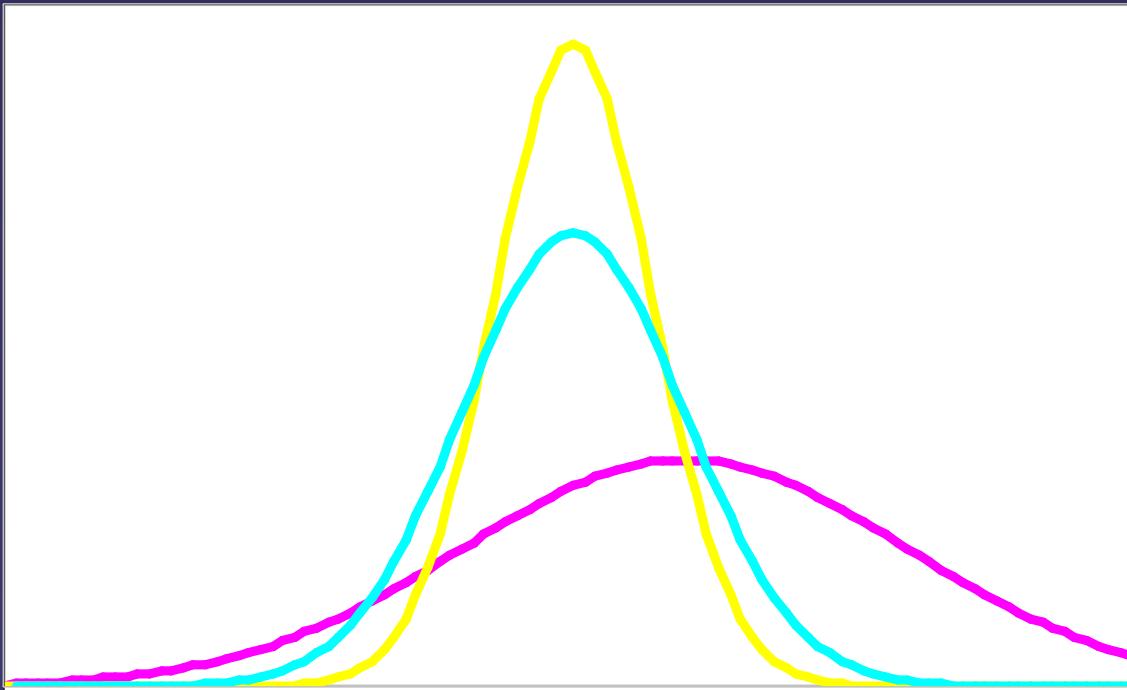
$\pi$  = 3.14159;  $e$  = 2.71828

$\sigma$  = population standard deviation

$X$  = value of random variable ( $-\infty < X < \infty$ )

$\mu$  = population mean

# Many Normal Distributions

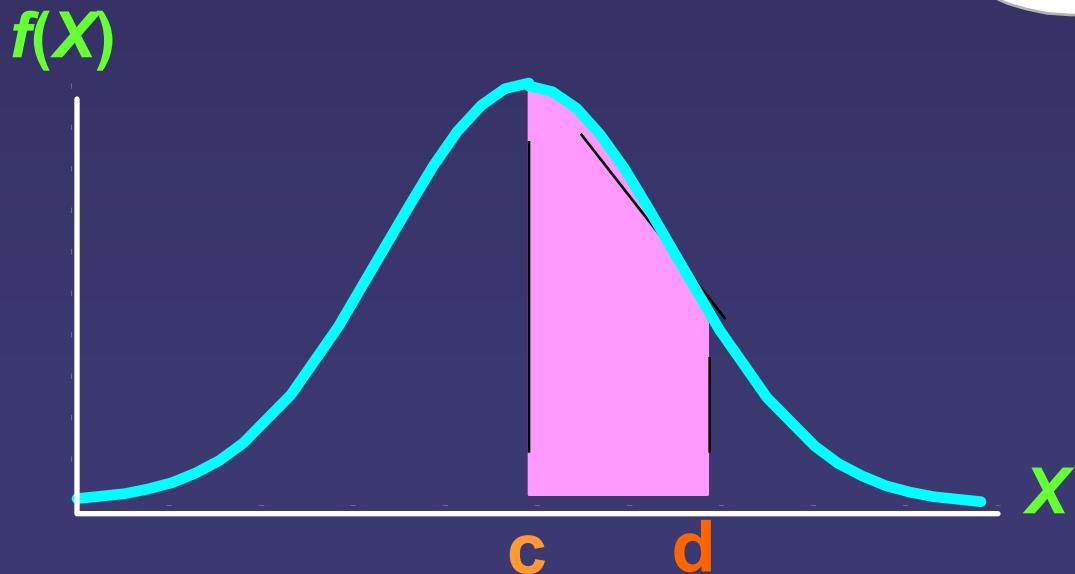


There are  
an Infinite  
Number

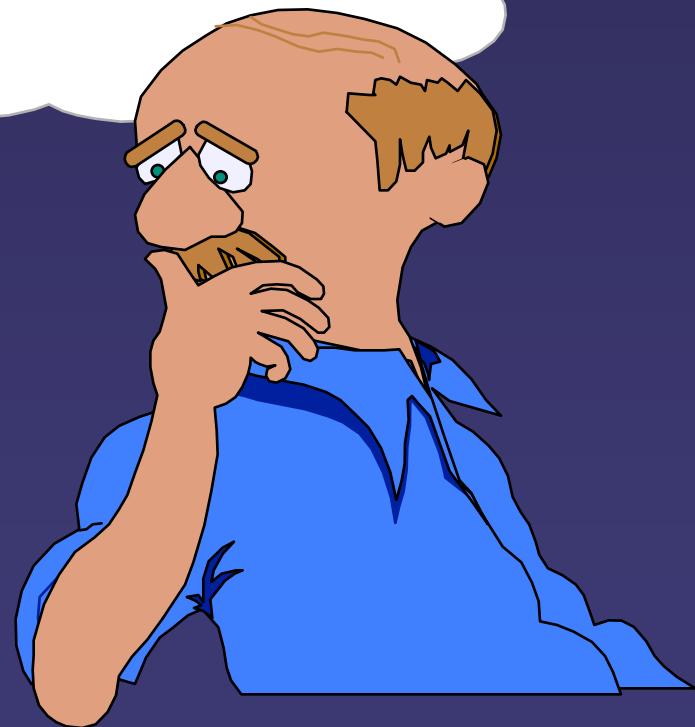
Varying the Parameters  $\sigma$  and  $\mu$ , we obtain  
Different Normal Distributions.

# Normal Distribution: Finding Probabilities

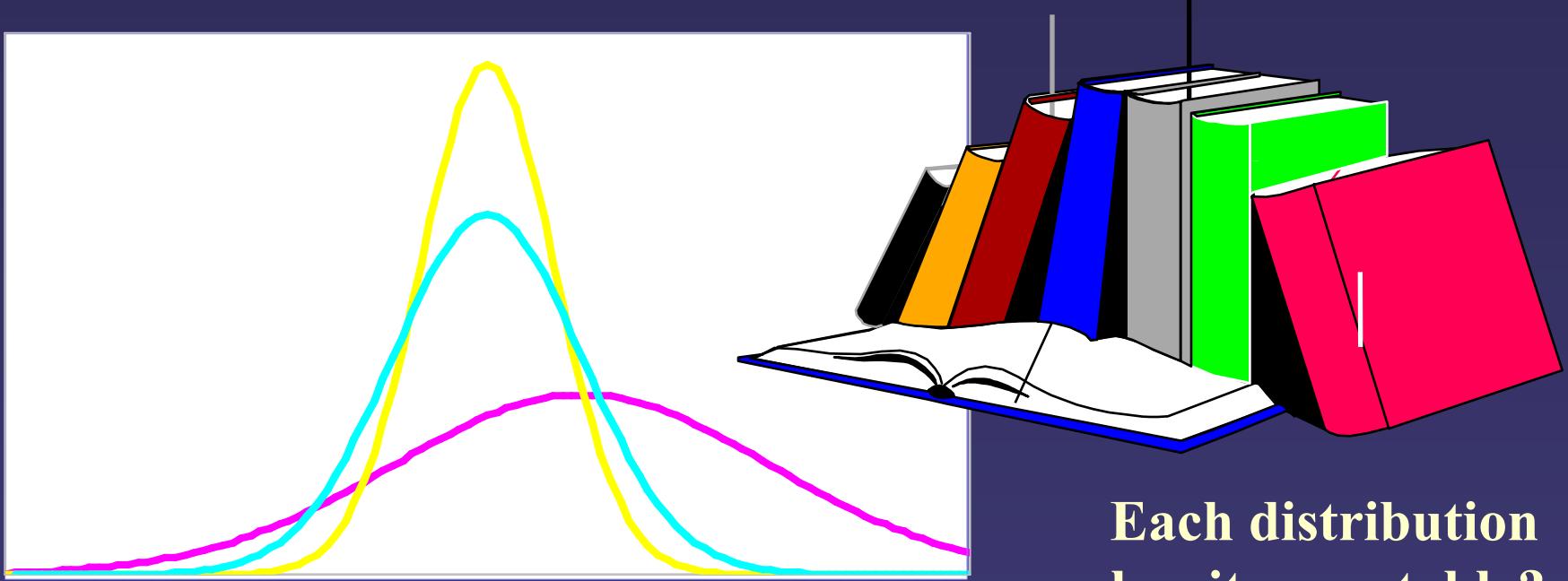
Probability is the  
area under the  
curve!



$P(c \leq X \leq d) = ?$



# Which Table?



Infinitely Many Normal Distributions Means  
Infinitely Many Tables to Look Up!

# Solution (I): The Standardized Normal Distribution

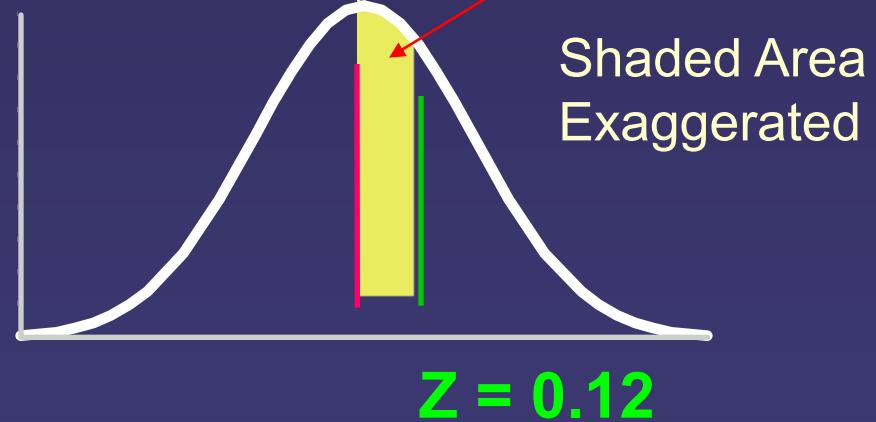
## Standardized Normal Distribution

### Table (Portion)

Z	.00	.01	.02
0.0	.0000	.0040	.0080
0.1	.0398	.0438	.0478
0.2	.0793	.0832	.0871
0.3	.0179	.0217	.0255

$$\mu_z = 0 \text{ and } \sigma_z = 1$$

.0478



Probabilities

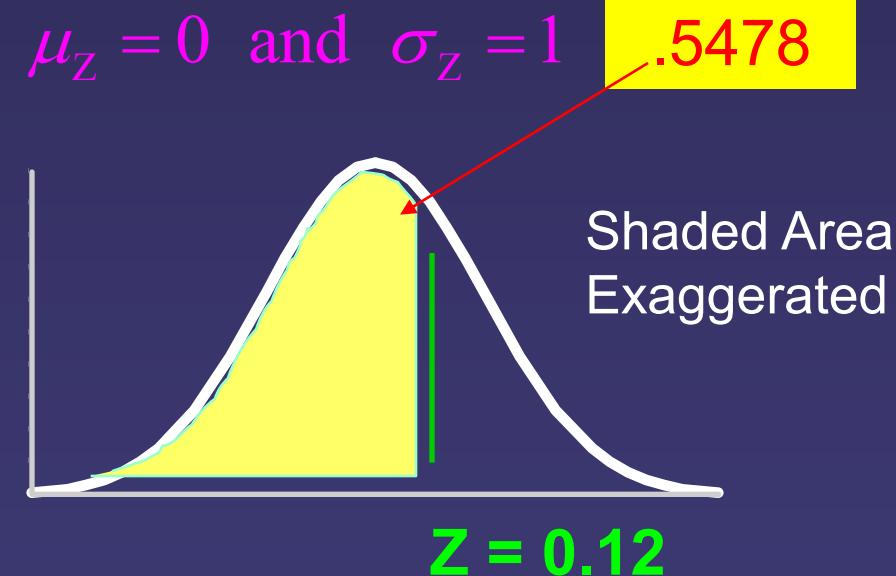
Only One Table is Needed

# Solution (II): The Cumulative Standardized Normal Distribution

## Cumulative Standardized Normal Distribution Table (Portion)

Z	.00	.01	.02
0.0	.5000	.5040	.5080
0.1	.5398	.5438	.5478
0.2	.5793	.5832	.5871
0.3	.5179	.5217	.5255

Probabilities

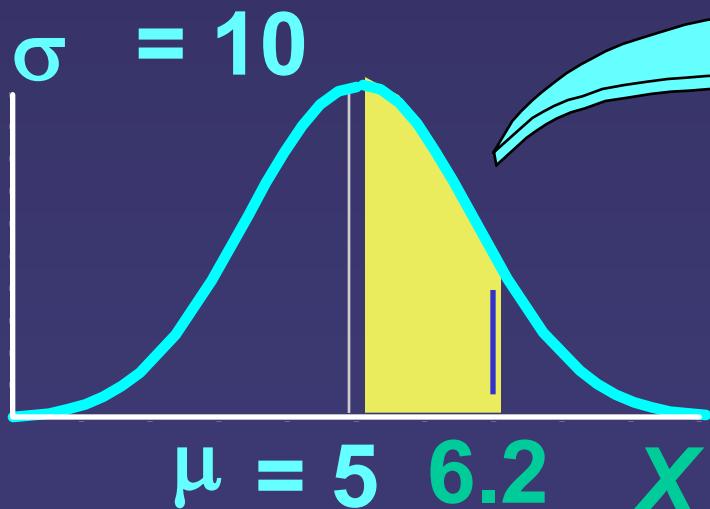


Only One Table is Needed

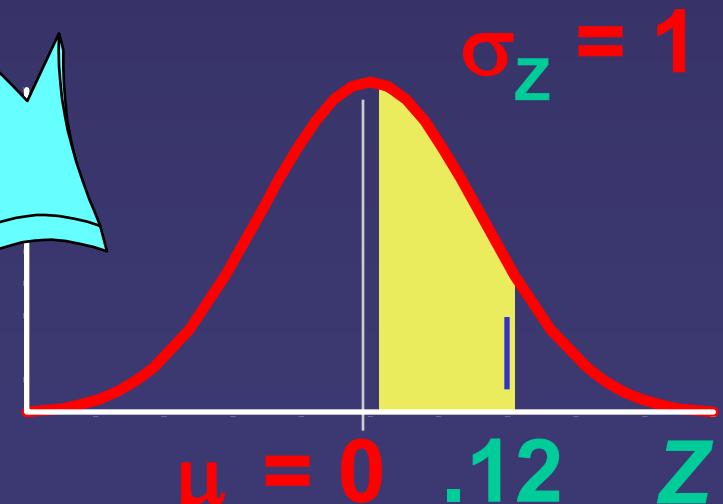
# Standardizing Example

$$Z = \frac{X - \mu}{\sigma} = \frac{6.2 - 5}{10} = 0.12$$

Normal  
Distribution



Standardized  
Normal Distribution



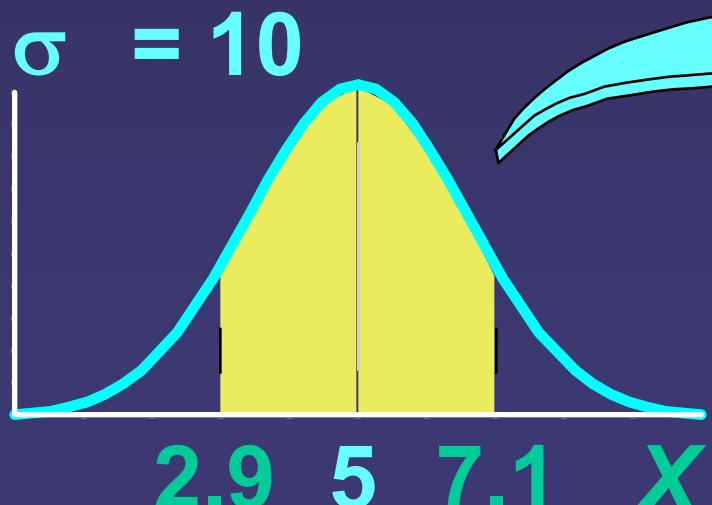
Shaded Area Exaggerated

# Example:

$$P(2.9 < X < 7.1) = .1664$$

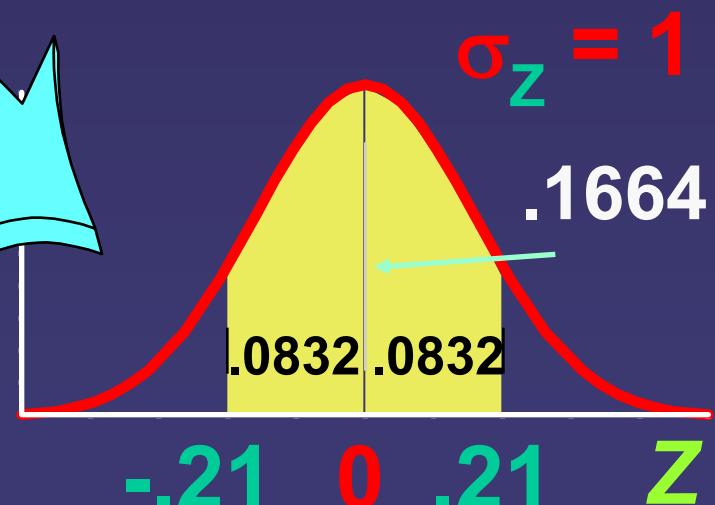
$$z = \frac{x - \mu}{\sigma} = \frac{2.9 - 5}{10} = -.21$$

Normal  
Distribution



$$z = \frac{x - \mu}{\sigma} = \frac{7.1 - 5}{10} = .21$$

Standardized  
Normal Distribution

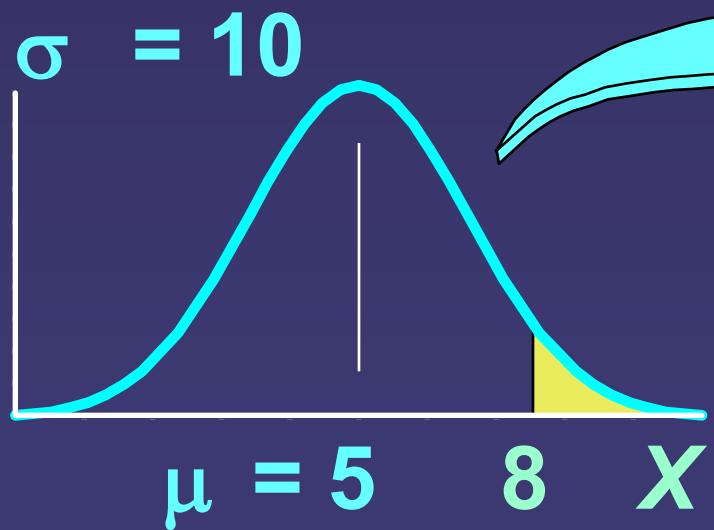


Shaded Area Exaggerated

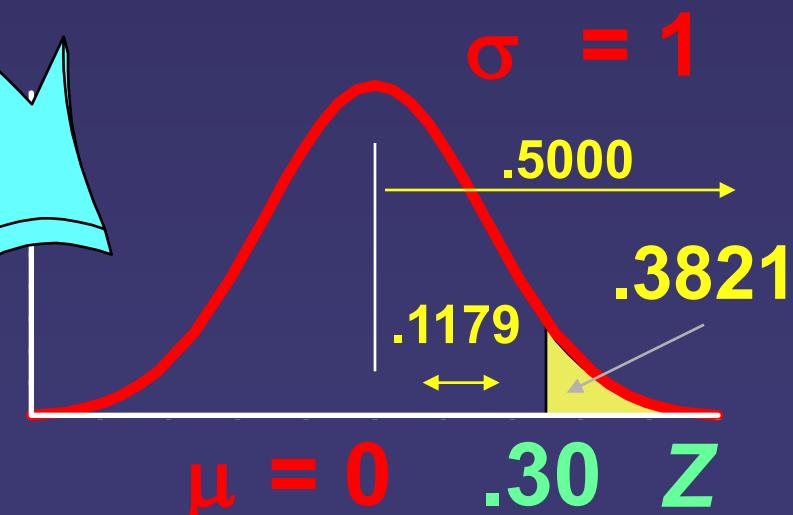
Example:  $P(X \geq 8) = .3821$

$$z = \frac{x - \mu}{\sigma} = \frac{8 - 5}{10} = .30$$

Normal  
Distribution



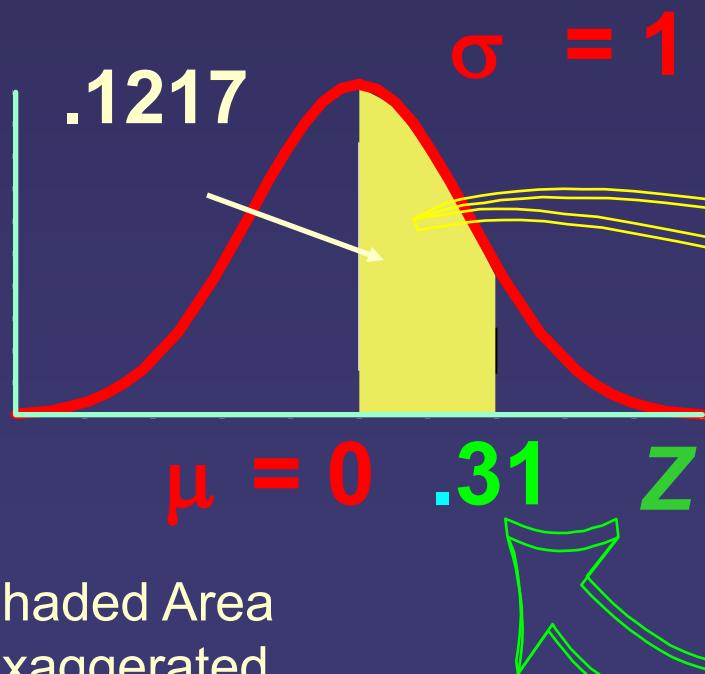
Standardized  
Normal Distribution



Shaded Area Exaggerated

# Finding Z Values for Known Probabilities

What Is Z Given  
Probability = 0.1217?

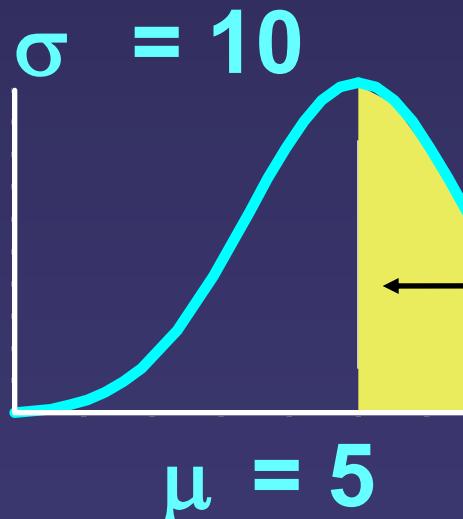


Standardized Normal  
Probability Table (Portion)

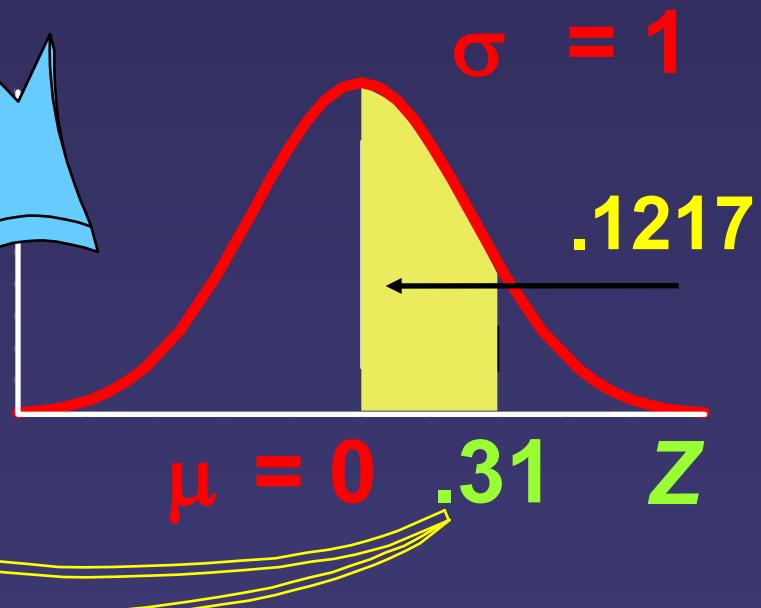
Z	.00	.01	0.2
0.0	.0000	.0040	.0080
0.1	.0398	.0438	.0478
0.2	.0793	.0832	.0871
0.3	.1179	<b>.1217</b>	.1255

# Recovering $X$ Values for Known Probabilities

Normal Distribution



Standardized Normal Distribution



$$X = \mu + Z\sigma = 5 + (0.31)(10) = 8.1$$

Shaded Area Exaggerated



# THANK YOU