

Statistics for Business and Economics



Chapter 6

Sampling and Sampling Distributions

Tools of Business Statistics

- **Descriptive statistics**
 - Collecting, presenting, and describing data

- **Inferential statistics**
 - Drawing conclusions and/or making decisions concerning a population based only on sample data



Populations and Samples

- A **Population** is the set of all items or individuals of interest

- **Examples:**
 - All likely voters in the next election
 - All parts produced today
 - All sales receipts for November

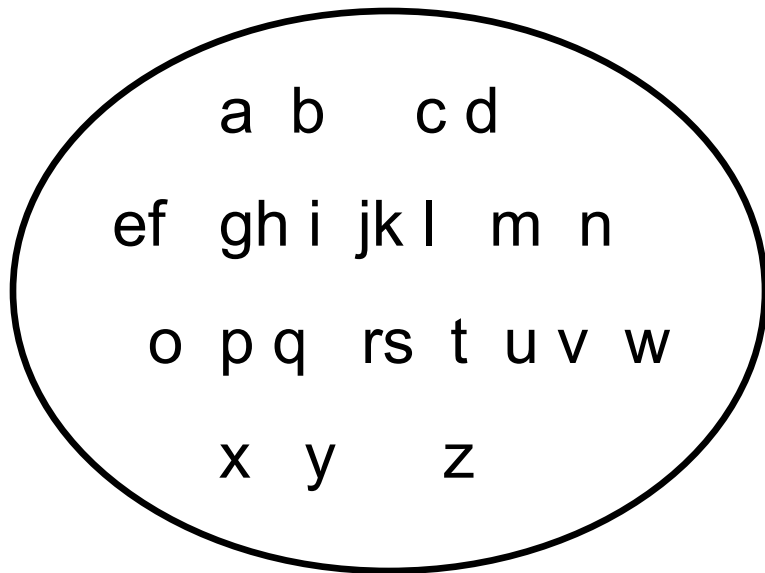
- A **Sample** is a subset of the population

- **Examples:**
 - 1000 voters selected at random for interview
 - A few parts selected for destructive testing
 - Random receipts selected for audit

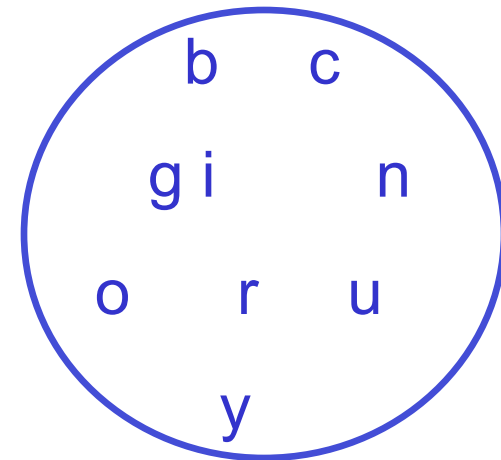


Population vs. Sample

Population



Sample



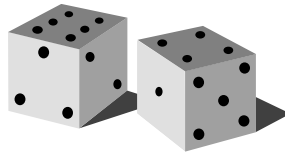


Why Sample?

- Less time consuming than a census
- Less costly to administer than a census
- It is possible to obtain statistical results of a sufficiently high precision based on samples.

Simple Random Samples

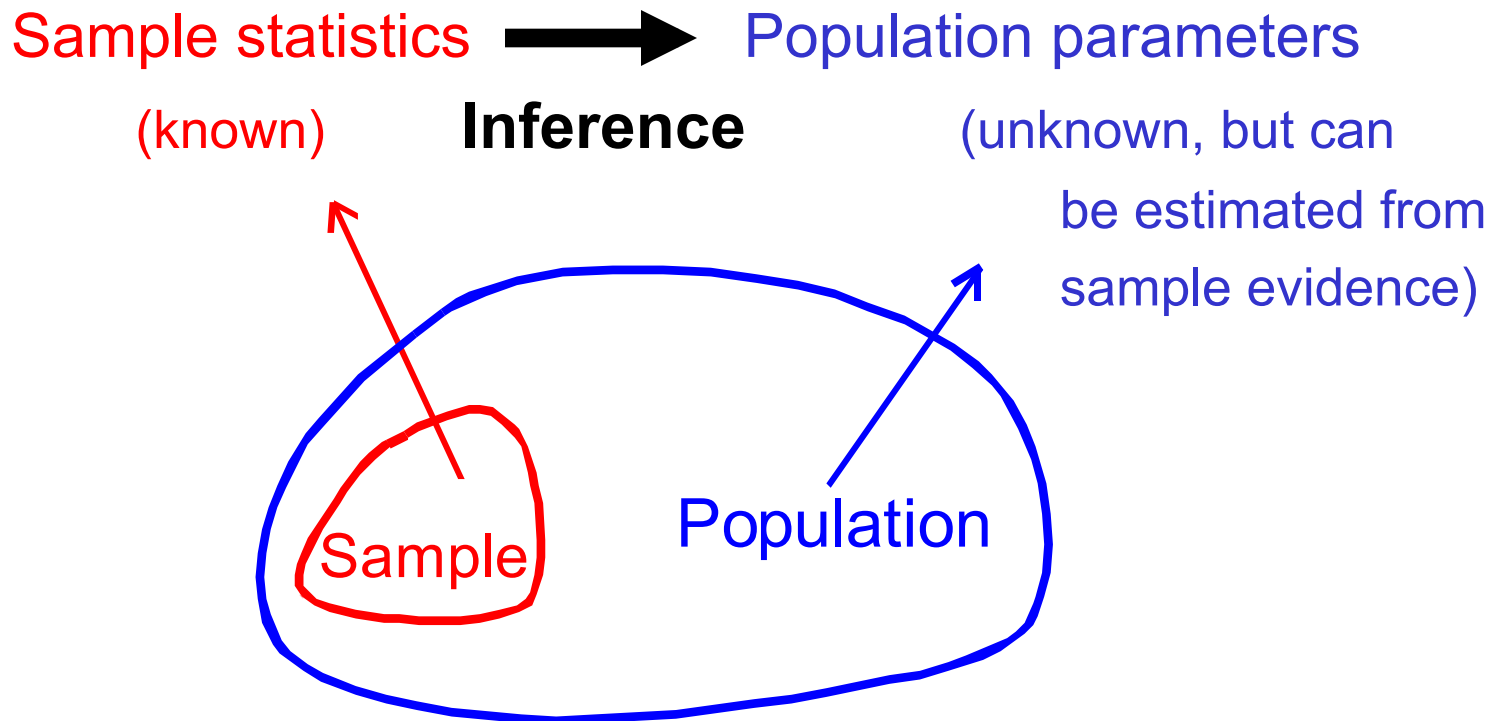
- Every object in the population has an **equal chance** of being selected
- Objects are selected independently
- Samples can be obtained from a table of random numbers or computer random number generators



- A simple random sample is the ideal against which other sample methods are compared

Inferential Statistics

- Making statements about a population by examining sample results



Inferential Statistics

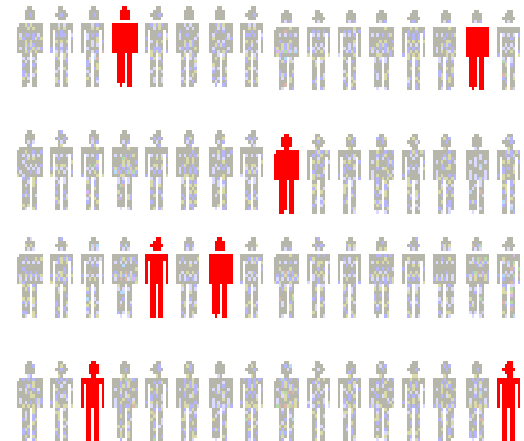
Drawing conclusions and/or making decisions concerning a **population** based on **sample** results.

■ Estimation

- e.g., Estimate the population mean weight using the sample mean weight

■ Hypothesis Testing

- e.g., Use sample evidence to test the claim that the population mean weight is 120 pounds

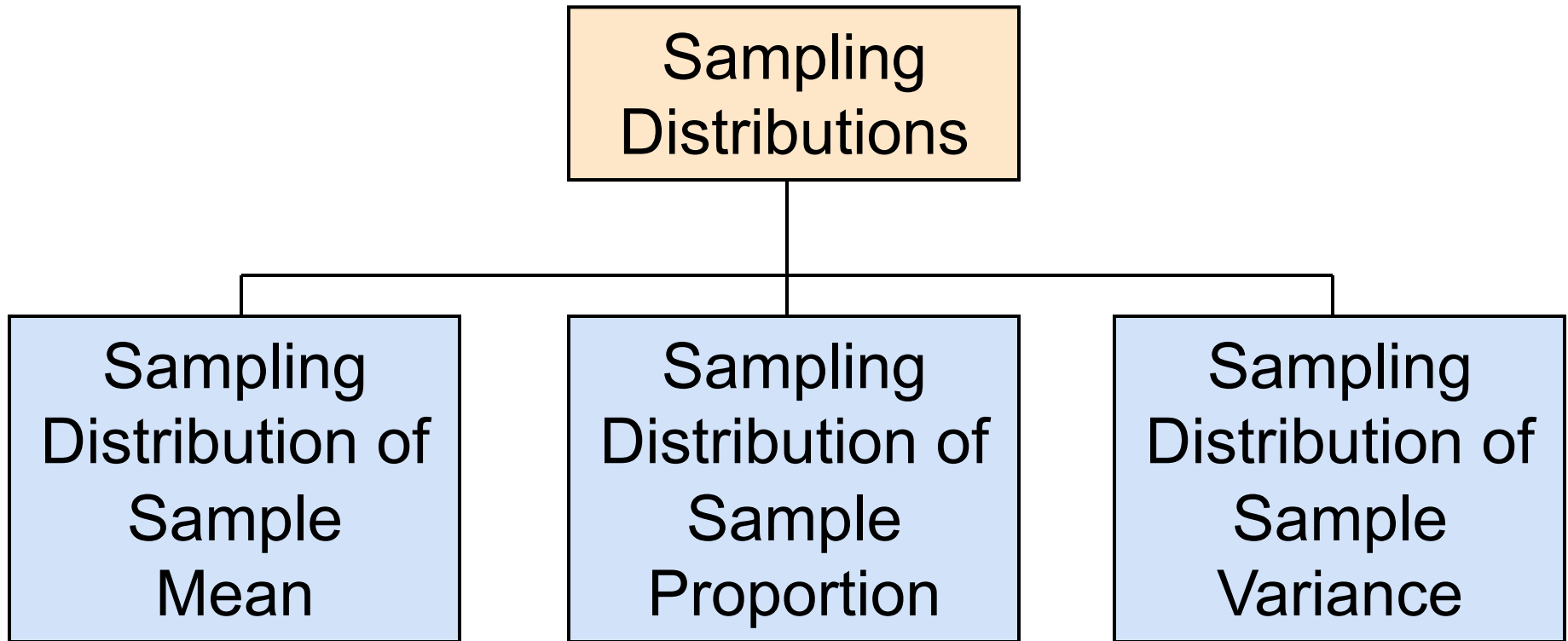


Sampling Distributions

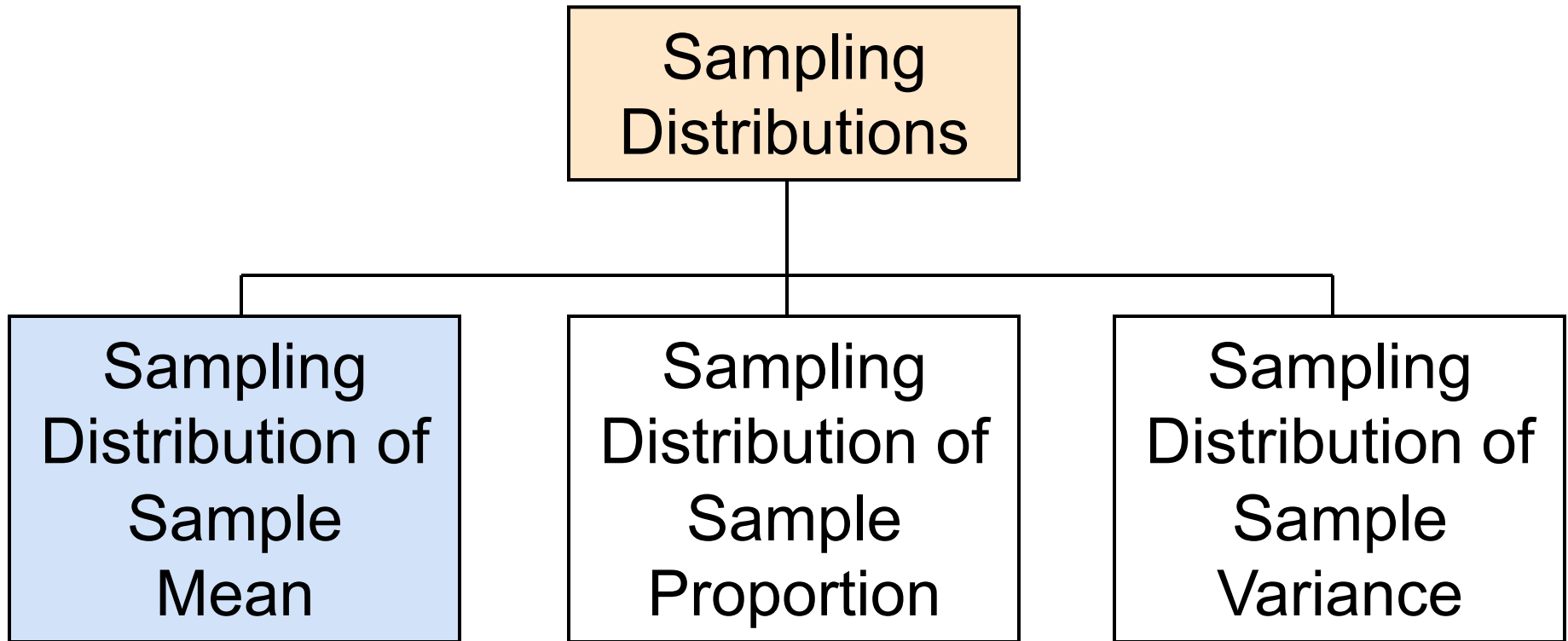
- A **sampling distribution** is a distribution of all of the possible values of a statistic for a given size sample selected from a population



Chapter Outline

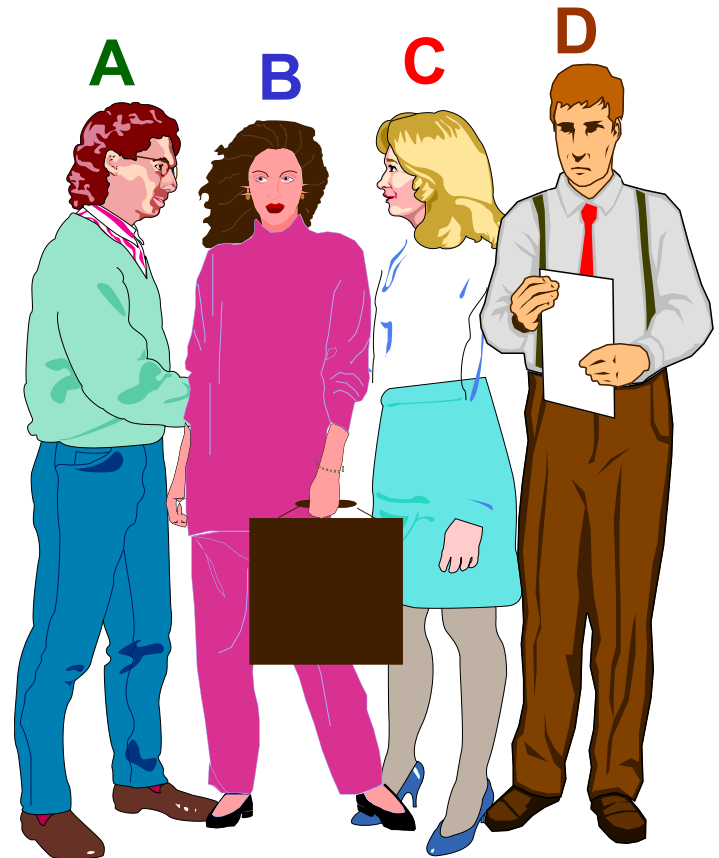


Sampling Distributions of Sample Means



Developing a Sampling Distribution

- Assume there is a population ...
- Four types of people
- Random variable, X , is age of individuals
- Possible Values of X :
18, 20, 22, 24 (years)



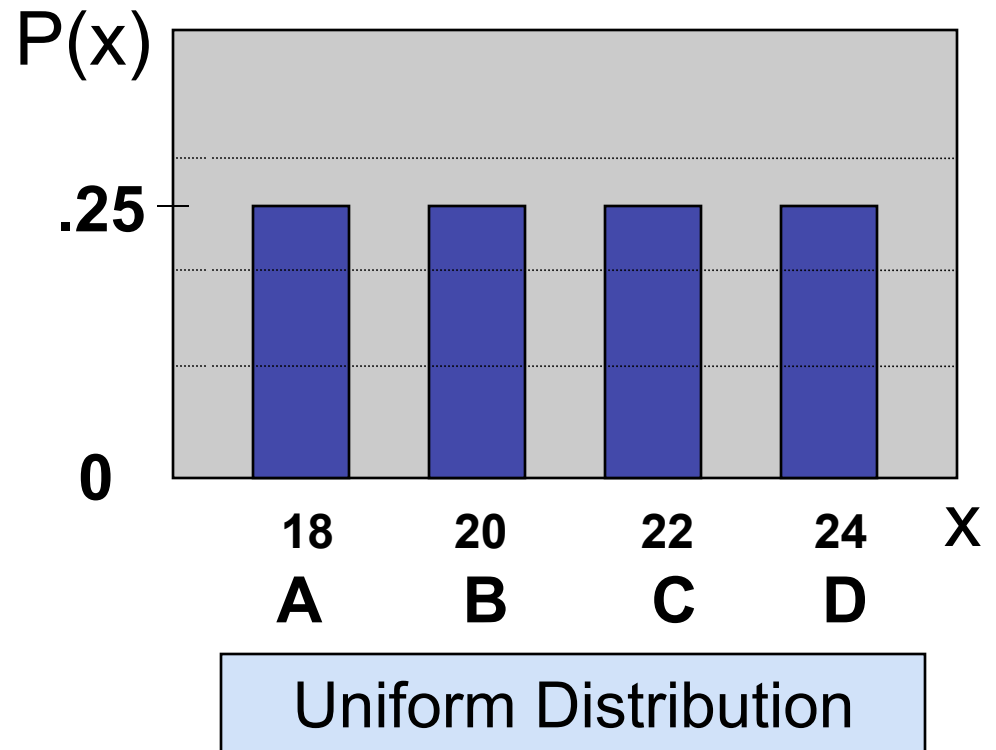
Developing a Sampling Distribution

(continued)

Summary Measures for the **Population** Distribution:

$$\mu = \frac{18 + 20 + 22 + 24}{4} = 21$$

$$\sigma^2 = \sqrt{\frac{\sum (X_i - \mu)^2}{4}} = 2.236$$



Developing a Sampling Distribution

(continued)

Now consider all possible samples of size $n = 2$

1 st	2 nd Observation			
Obs	18	20	22	24
18	18,18	18,20	18,22	18,24
20	20,18	20,20	20,22	20,24
22	22,18	22,20	22,22	22,24
24	24,18	24,20	24,22	24,24

16 possible samples
(sampling with
replacement)

16 Sample
Means

1 st	2 nd Observation			
Obs	18	20	22	24
18	18	19	20	21
20	19	20	21	22
22	20	21	22	23
24	21	22	23	24

Developing a Sampling Distribution

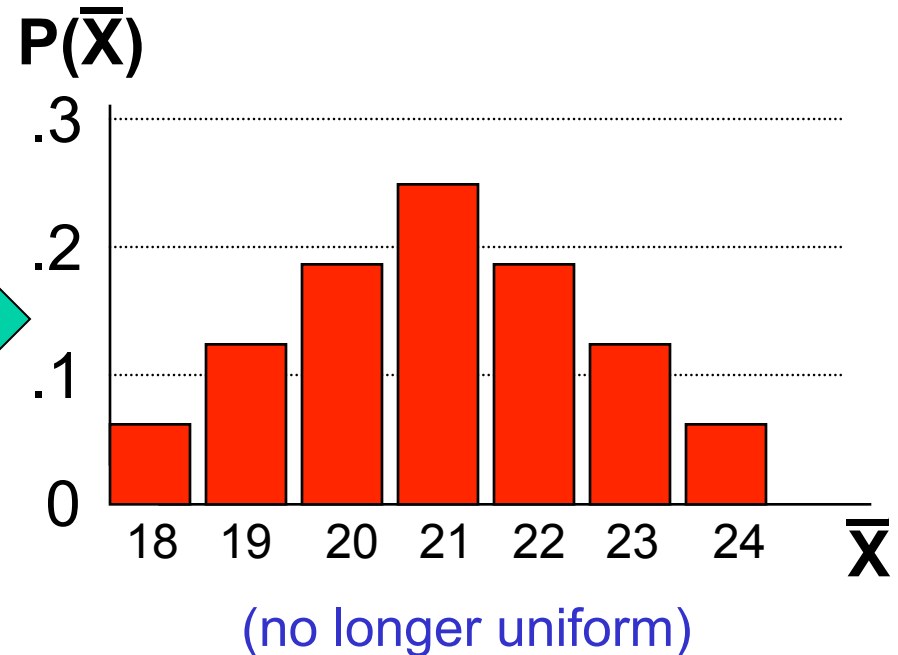
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Sampling Distribution of All Sample Means

16 Sample Means

1st Obs	2nd Observation			
	18	20	22	24
18	18	19	20	21
20	19	20	21	22
22	20	21	22	23
24	21	22	23	24

Sample Means
Distribution





Developing a Sampling Distribution

(continued)

Summary Measures of this Sampling Distribution:

$$\mu_{\bar{X}} = \frac{18 + 2 \times 19 + 3 \times 20 + \cdots + 2 \times 23 + 24}{16} = 21$$

$$\sigma_{\bar{X}} = \sqrt{\frac{(18-21)^2 + 2 \times (19-21)^2 + \cdots + (24-21)^2}{16}} = 1.58$$

Comparing the Population with its Sampling Distribution

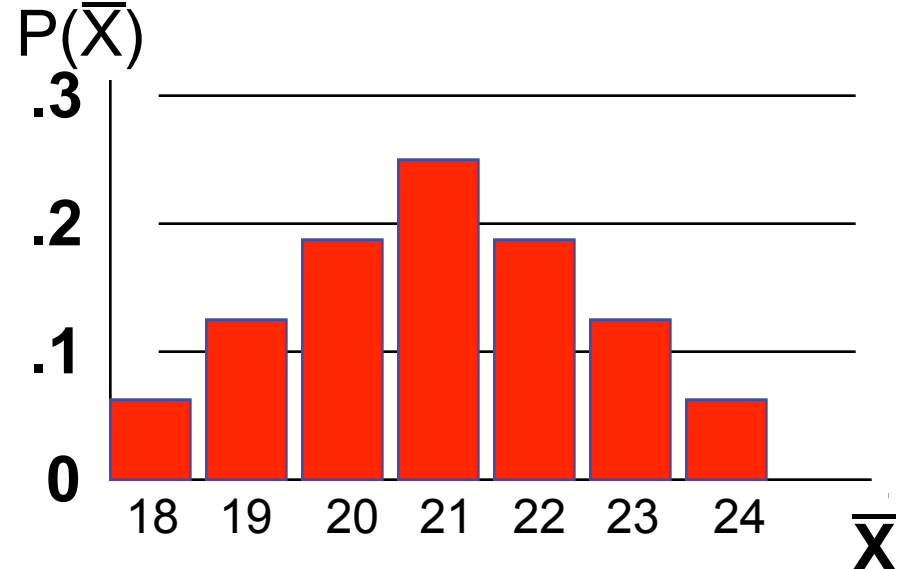
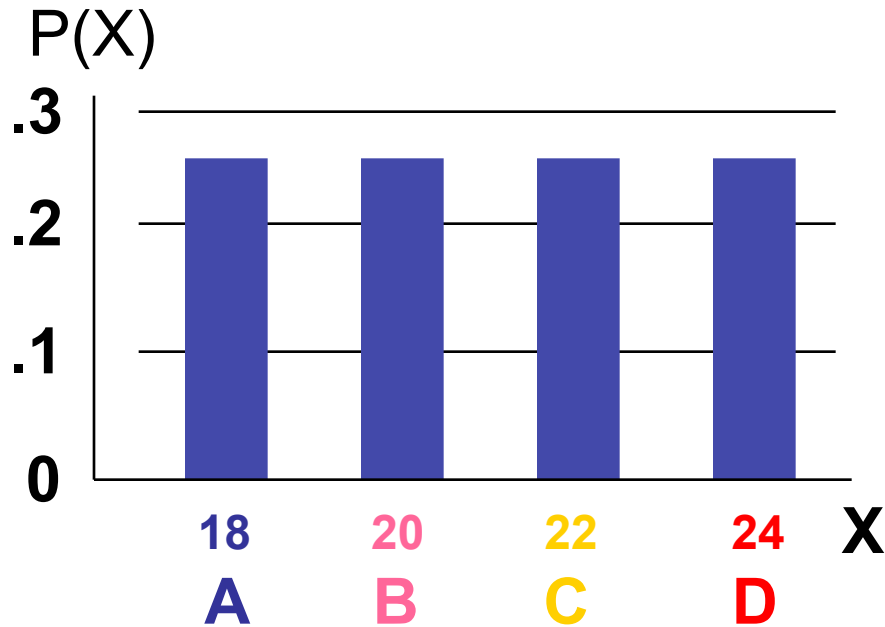
Population

Sample Means Distribution

$n = 2$

$$\mu = 21 \quad \sigma = 2.236$$

$$\mu_{\bar{X}} = 21 \quad \sigma_{\bar{X}} = 1.58$$





Expected Value of Sample Mean

- Let X_1, X_2, \dots, X_n represent a random sample from a population
- The **sample mean** value of these observations is defined as

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i$$



Standard Error of the Mean

- Different samples of the same size from the same population will yield different sample means
- A measure of the variability in the mean from sample to sample is given by the **Standard Error of the Mean:**

$$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$

- Note that the standard error of the mean decreases as the sample size increases
- If $n=2$, then $\sigma / \sigma_{\bar{X}} = \sqrt{2} = 1.4142$



If the Population is Normal

- If a population is **normal** with mean μ and standard deviation σ , the sampling distribution of \bar{X} is **also normally distributed** with

$$\mu_{\bar{X}} = \mu$$

and

$$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}}$$



Z-value for Sampling Distribution of the Mean

- Z-value for the sampling distribution of \bar{X} :

$$Z = \frac{(\bar{X} - \mu)}{\sigma_{\bar{X}}}$$

where:

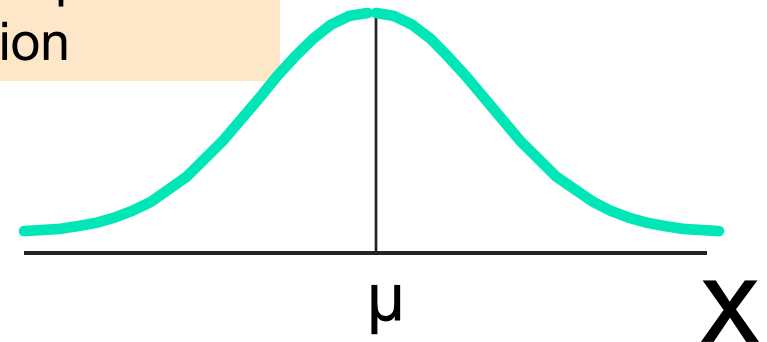
- \bar{X} = sample mean
- μ = population mean
- $\sigma_{\bar{X}}$ = standard error of the mean

Sampling Distribution Properties

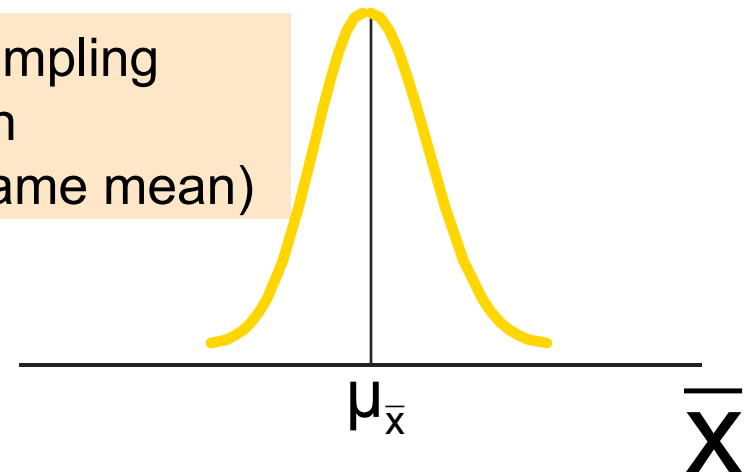
$$\mu_{\bar{X}} = \mu$$

(i.e. \bar{X} is unbiased)

Normal Population Distribution



Normal Sampling Distribution
(has the same mean)

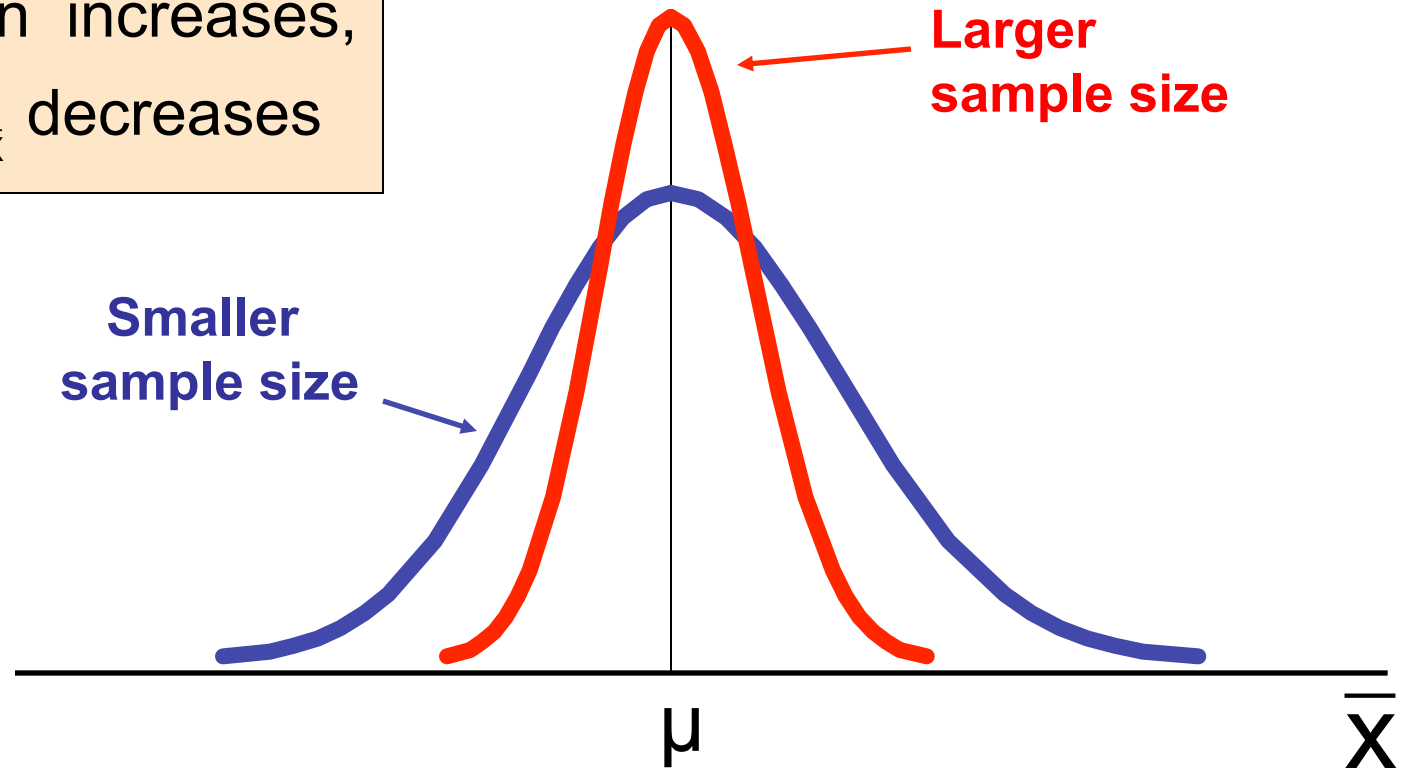


Sampling Distribution Properties

(continued)

- For sampling **with replacement**:

As n increases,
 $\sigma_{\bar{x}}$ decreases





If the Population is **not** Normal

- We can apply the **Central Limit Theorem**:
 - Even if the population is **not normal**,
 - ...sample means from the population **will be approximately normal** as long as the sample size is large enough.

Properties of the sampling distribution:

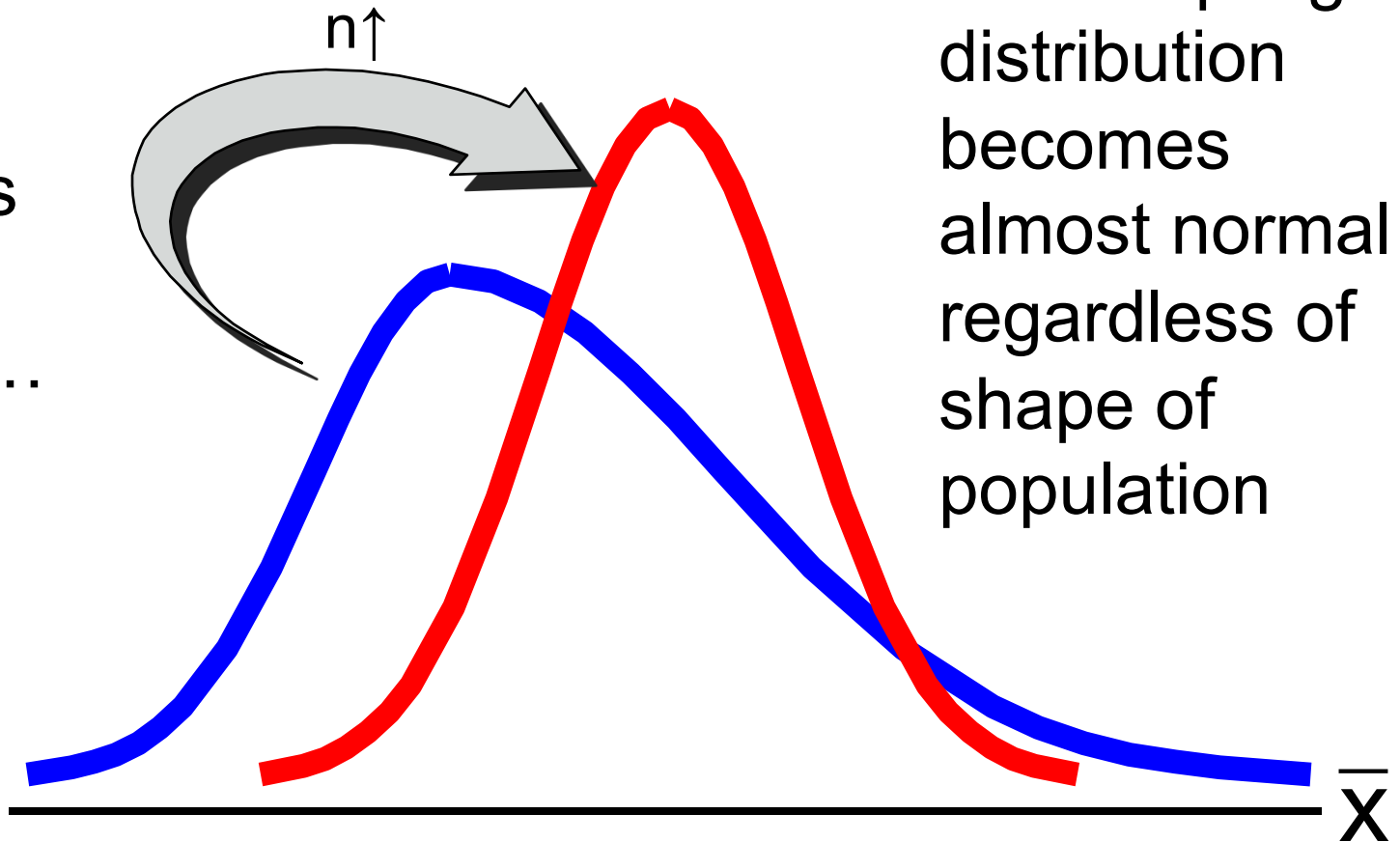
$$\mu_{\bar{x}} = \mu$$

and

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

Central Limit Theorem

As the sample size gets large enough...



the sampling distribution becomes almost normal regardless of shape of population

If the Population is not Normal

(continued)

Sampling distribution properties:

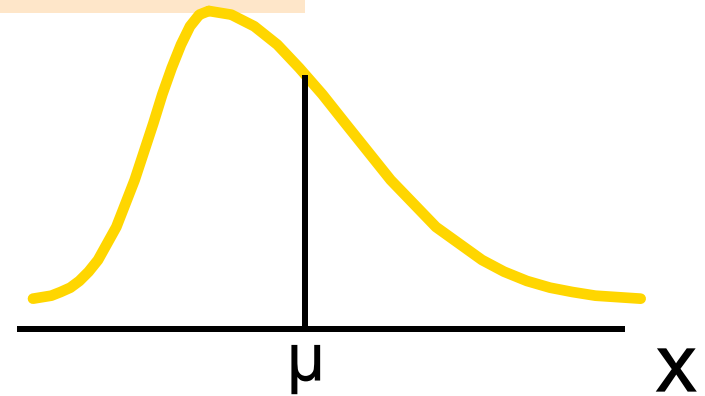
Central Tendency

$$\mu_{\bar{x}} = \mu$$

Variation

$$\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}}$$

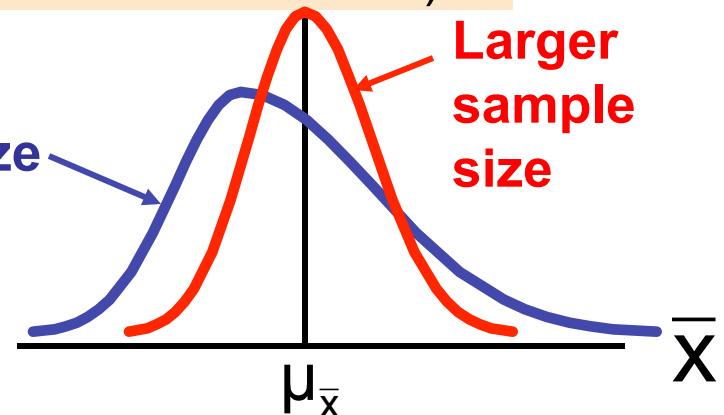
Population Distribution



Sampling Distribution
(becomes normal as n increases)

Smaller sample size

Larger sample size





Example

- Suppose a large population has mean $\mu = 8$ and standard deviation $\sigma = 3$. Suppose a random sample of size $n = 36$ is selected.
- What is the probability that the **sample mean** is between 7.8 and 8.2?



Example

(continued)

Solution:

- Even though the population is not normally distributed, we use the central limit theorem to get an approximated solution
- ... the sampling distribution of \bar{X} is approximately normal
- ... with mean $\mu_{\bar{x}} = 8$
- ...and standard deviation $\sigma_{\bar{x}} = \frac{\sigma}{\sqrt{n}} = \frac{3}{\sqrt{36}} = 0.5$

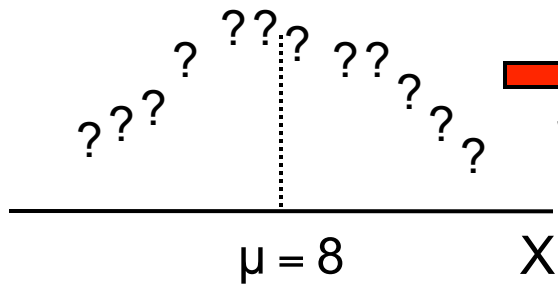
Example

(continued)

Solution (continued):

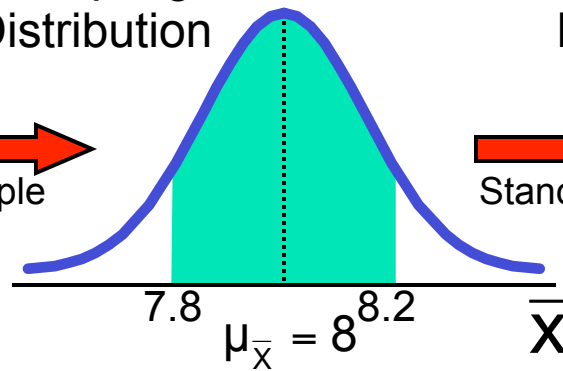
$$\begin{aligned} P(7.8 < \bar{X} < 8.2) &= P\left(\frac{7.8 - 8}{\frac{3}{\sqrt{36}}} < \frac{\bar{X} - \mu}{\frac{\sigma}{\sqrt{n}}} < \frac{8.2 - 8}{\frac{3}{\sqrt{36}}}\right) \\ &= P(-0.4 < Z < 0.4) = \boxed{0.3108} \end{aligned}$$

Population
Distribution



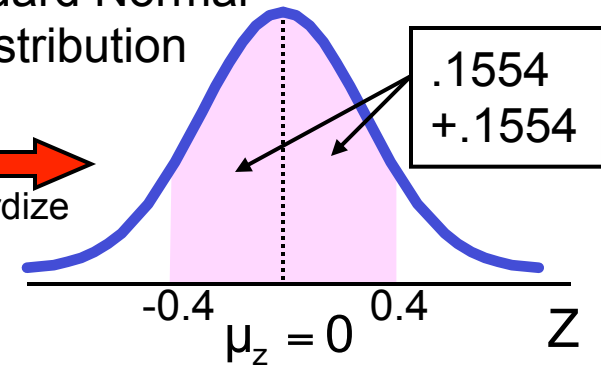
Sampling
Distribution

Sample



Standard Normal
Distribution

Standardize





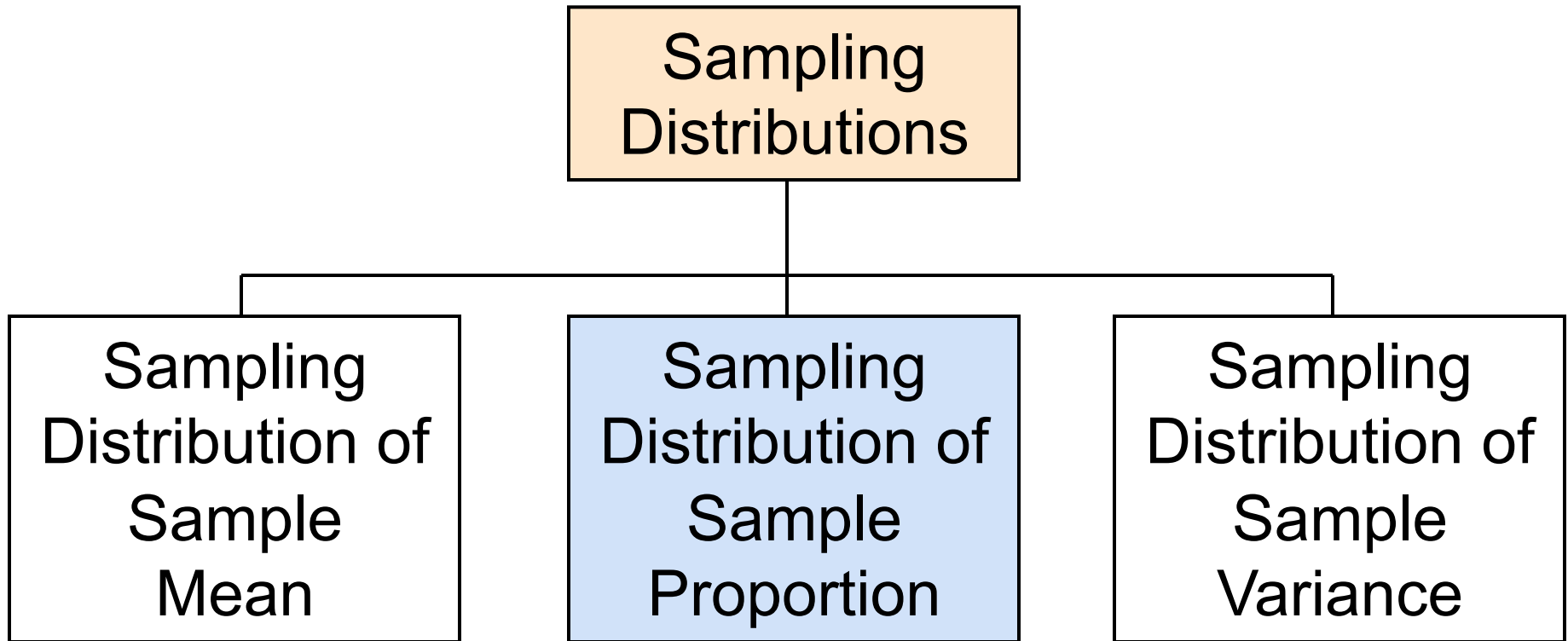
Acceptance Intervals

- Goal: determine a range within which sample means are likely to occur, given a population mean and variance
 - By the Central Limit Theorem, we know that the distribution of \bar{X} is approximately normal if n is large enough, with mean μ and standard deviation $\sigma_{\bar{X}}$
 - Let $z_{\alpha/2}$ be the z-value that leaves area $\alpha/2$ in the upper tail of the normal distribution (i.e., the interval $-z_{\alpha/2}$ to $z_{\alpha/2}$ encloses probability $1 - \alpha$)
 - Then

$$\mu \pm z_{\alpha/2} \sigma_{\bar{X}}$$

is the interval that includes \bar{X} with probability $1 - \alpha$

Sampling Distributions of Sample Proportions





Sampling Distributions of Sample Proportions

p = the proportion of the population having some characteristic

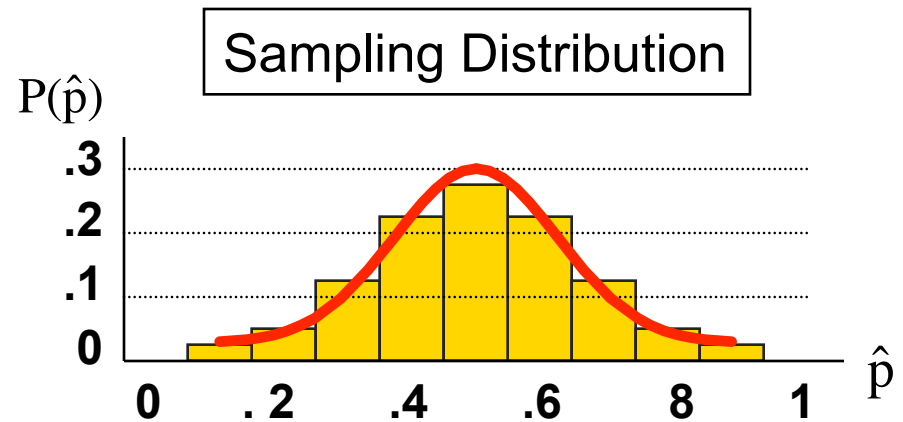
- **Sample proportion** (\hat{p}) provides an estimate of p :

$$\hat{p} = \frac{\text{number of items in the sample having the characteristic of interest}}{\text{sample size}}$$

- $0 \leq \hat{p} \leq 1$
- \hat{p} has a binomial distribution, but can be approximated by a normal distribution when n is large

Sampling Distribution of \hat{p}

- Normal approximation:



Properties:

$$E(\hat{p}) = p$$

and

$$\sigma_{\hat{p}}^2 = \text{Var}(\hat{p}) = \frac{p(1-p)}{n}$$

(where p = population proportion)



Z-Value for Proportions

Standardize \hat{p} to a Z value with the formula:

$$Z = \frac{\hat{p} - p}{\sigma_{\hat{p}}} = \frac{\hat{p} - p}{\sqrt{\frac{p(1-p)}{n}}}$$



Example

- If the true proportion of voters who support Proposition A is $p = .4$, what is the probability that a sample of size 200 yields a sample proportion between .40 and .45?

- i.e.: **if $p = .4$ and $n = 200$, what is $P(.40 \leq \hat{p} \leq .45)$?**



Example

(continued)

- if $p = .4$ and $n = 200$, what is $P(.40 \leq \hat{p} \leq .45)$?

Find $\sigma_{\hat{p}}$:

$$\sigma_{\hat{p}} = \sqrt{\frac{p(1-p)}{n}} = \sqrt{\frac{.4(1-.4)}{200}} = .03464$$

Convert to
standard
normal:

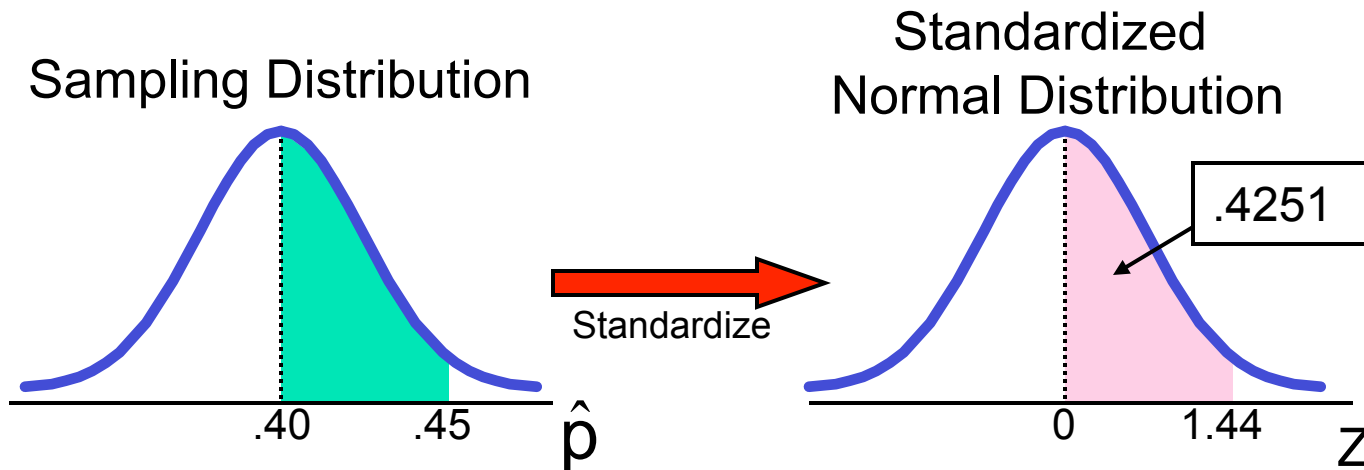
$$\begin{aligned} P(.40 \leq \hat{p} \leq .45) &= P\left(\frac{.40 - .40}{.03464} \leq Z \leq \frac{.45 - .40}{.03464}\right) \\ &= P(0 \leq Z \leq 1.44) \end{aligned}$$

Example

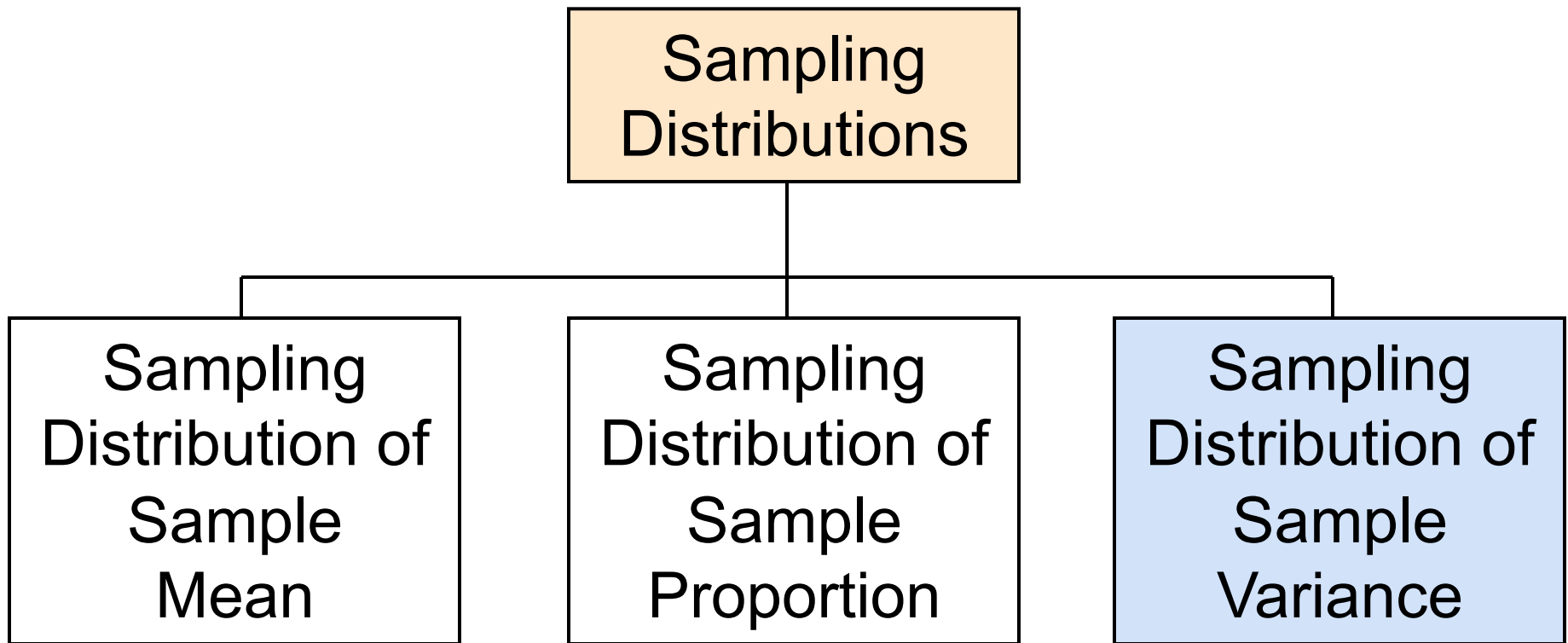
(continued)

- if $P = .4$ and $n = 200$, what is $P(.40 \leq \hat{p} \leq .45)$?

Use standard normal table: $P(0 \leq Z \leq 1.44) = .4251$



Sampling Distributions of Sample Variance





Sample Variance

- Let x_1, x_2, \dots, x_n be a random sample from a population. The **sample variance** is

$$s^2 = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})^2$$

- the square root of the sample variance is called the **sample standard deviation**
- the sample variance is different for different random samples from the same population



Sampling Distribution of Sample Variances

- The sampling distribution of s^2 has mean σ^2

$$E(s^2) = \sigma^2$$

- If the population distribution is normal, then

$$\text{Var}(s^2) = \frac{2\sigma^4}{n-1}$$

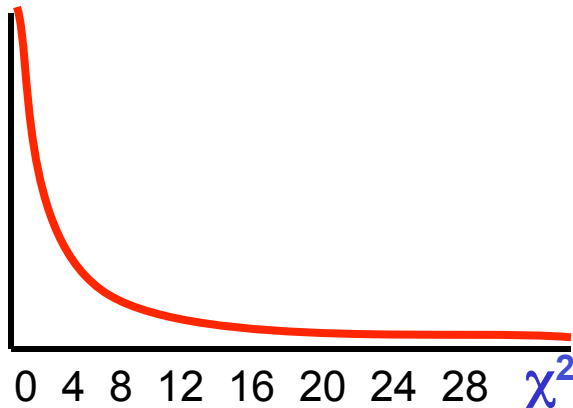
- If the population distribution is normal then

$$\frac{(n-1)s^2}{\sigma^2}$$

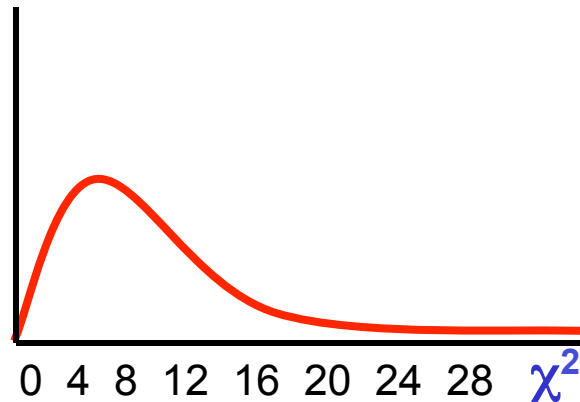
has a χ^2 distribution with $n - 1$ degrees of freedom

The Chi-square Distribution

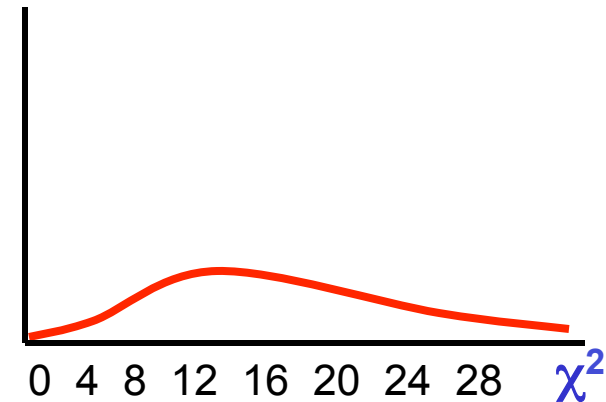
- The **chi-square distribution** is a family of distributions, depending on degrees of freedom:
- $d.f. = n - 1$



d.f. = 1



d.f. = 5



d.f. = 15

- Text **Table 7** contains chi-square probabilities



Degrees of Freedom (df)

Idea: Number of observations that are free to vary after sample mean has been calculated

Example: Suppose the mean of 3 numbers is 8.0

Let $X_1 = 7$
Let $X_2 = 8$
What is X_3 ?



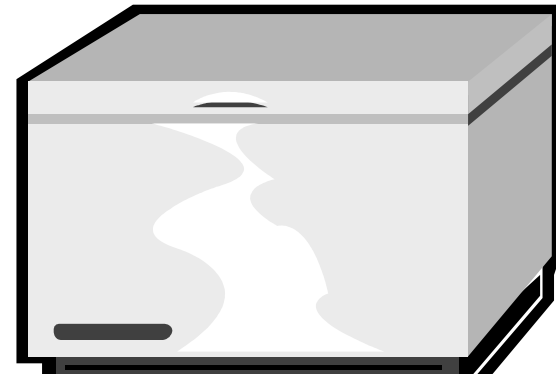
If the mean of these three values is 8.0,
then X_3 **must be 9**
(i.e., X_3 is not free to vary)

Here, $n = 3$, so degrees of freedom = $n - 1 = 3 - 1 = 2$

(2 values can be any numbers, but the third is not free to vary for a given mean)

Chi-square Example

- A commercial freezer must hold a selected temperature with little variation. Specifications call for a standard deviation of no more than 4 degrees (a variance of 16 degrees²).
- A sample of 14 freezers is to be tested
- What is the upper limit (K) for the sample variance such that the probability of exceeding this limit, given that the population standard deviation is 4, is less than 0.05?



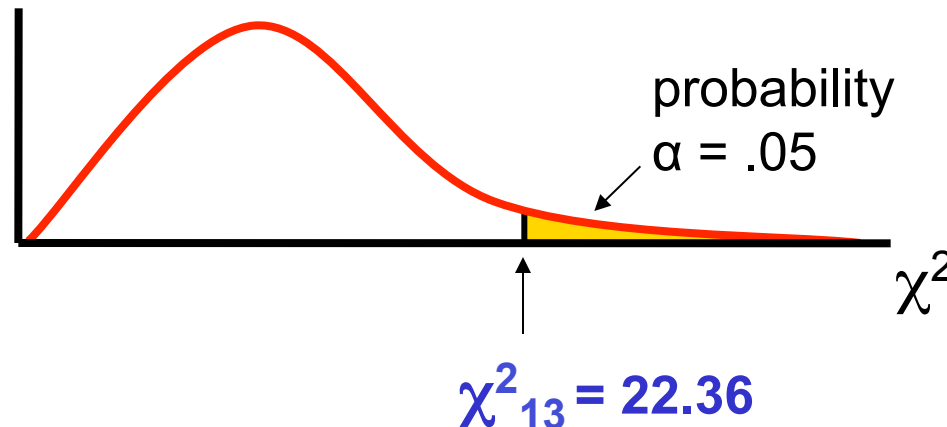
Finding the Chi-square Value

$$\chi^2 = \frac{(n-1)s^2}{\sigma^2}$$

Is chi-square distributed with $(n - 1) = 13$ degrees of freedom

- Use the the chi-square distribution with area 0.05 in the upper tail:

$$\chi^2_{13} = 22.36 \quad (\alpha = .05 \text{ and } 14 - 1 = 13 \text{ d.f.})$$



Chi-square Example

(continued)

$$\chi^2_{13} = 22.36 \quad (\alpha = .05 \text{ and } 14 - 1 = 13 \text{ d.f.})$$

So:

$$P(s^2 > K) = P\left(\frac{(n-1)s^2}{16} > \chi^2_{13}\right) = 0.05$$

$$\text{or} \quad \frac{(n-1)K}{16} = 22.36$$

(where $n = 14$)

$$\text{so} \quad K = \frac{(22.36)(16)}{(14-1)} = 27.52$$

If s^2 from the sample of size $n = 14$ is greater than 27.52, there is strong evidence to suggest the population variance exceeds 16.