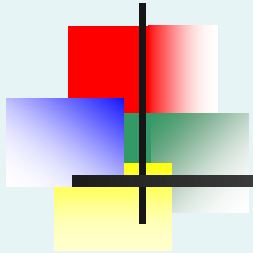


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Chapter 7

Sampling Distributions



Learning Objectives

In this chapter, you learn:

- The concept of the sampling distribution
- To compute probabilities related to the sample mean and the sample proportion
- The importance of the Central Limit Theorem



Sampling Distributions

- A sampling distribution is a distribution of all of the possible values of a sample statistic for a given size sample selected from a population.
- For example, suppose you sample 50 students from your college regarding their mean GPA. If you obtained many different samples of 50, you will compute a different mean for each sample. We are interested in the distribution of all potential mean GPAs we might calculate for any given sample of 50 students.

Developing a Sampling Distribution

- Assume there is a population ...
- Population size $N=4$
- Random variable, X , is **age** of individuals
- Values of X : 18, 20, 22, 24 (years)



Developing a Sampling Distribution

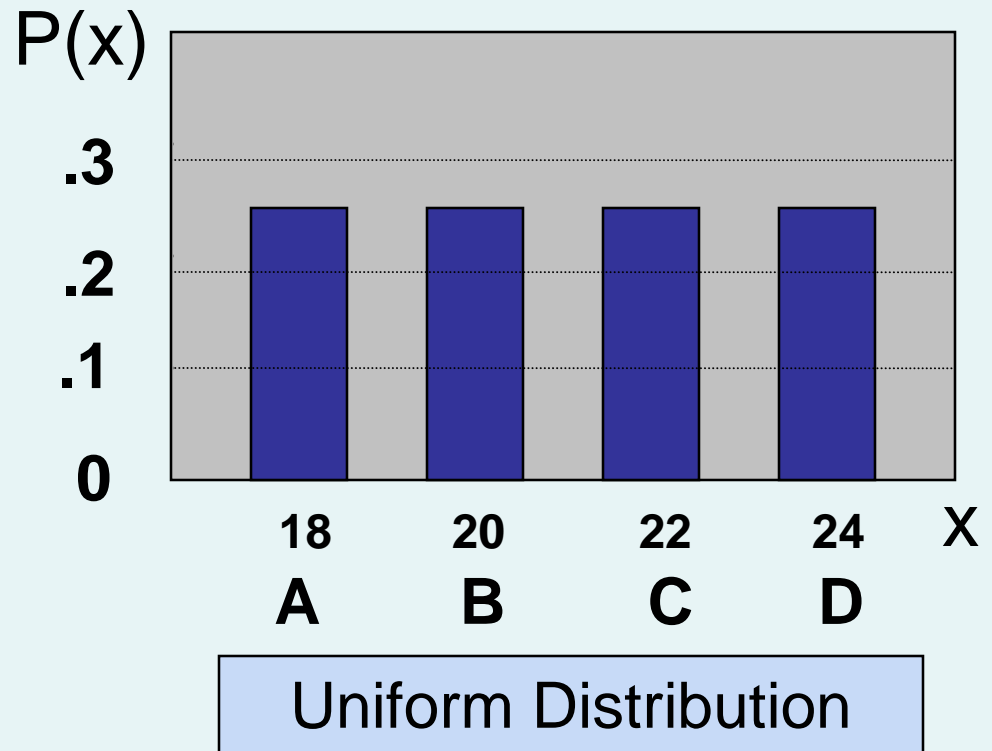
(continued)

DCOVA

Summary Measures for the Population Distribution:

$$\begin{aligned}\mu &= \frac{\sum X_i}{N} \\ &= \frac{18 + 20 + 22 + 24}{4} = 21\end{aligned}$$

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



Developing a Sampling Distribution

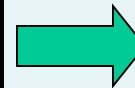
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Now consider all possible samples of size $n=2$

DCOVA

1 st Obs	2 nd Observation			
	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 Obs	2 nd 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

Developing a Sampling Distribution

DCOVA

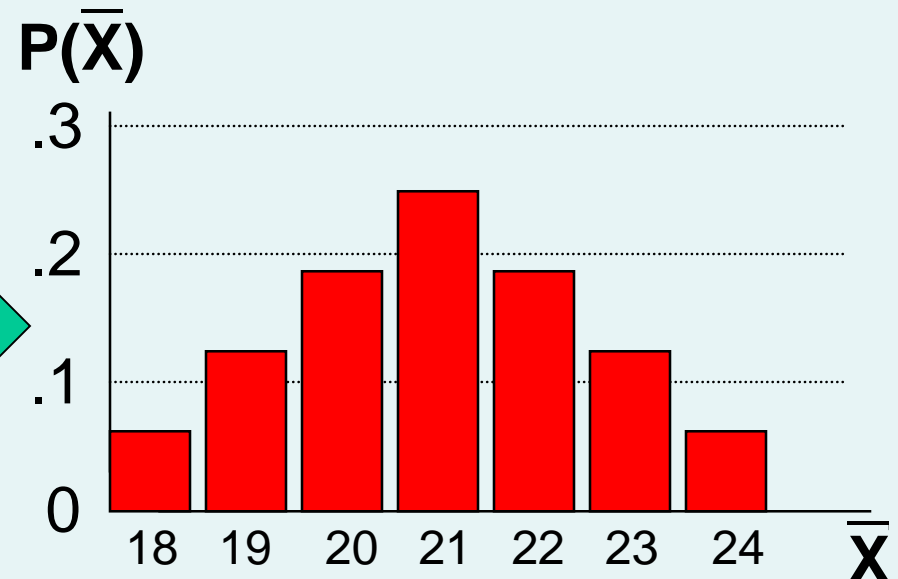
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Sampling Distribution of All Sample Means

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



(no longer uniform)

Developing a Sampling Distribution

DCOVA

(continued)

Summary Measures of this Sampling Distribution:

$$\mu_{\bar{X}} = \frac{18 + 19 + 19 + \dots + 24}{16} = 21$$

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

Note: Here we divide by 16 because there are 16 different samples of size 2.

Comparing the Population Distribution to the Sample Means Distribution

DCOVA

Population

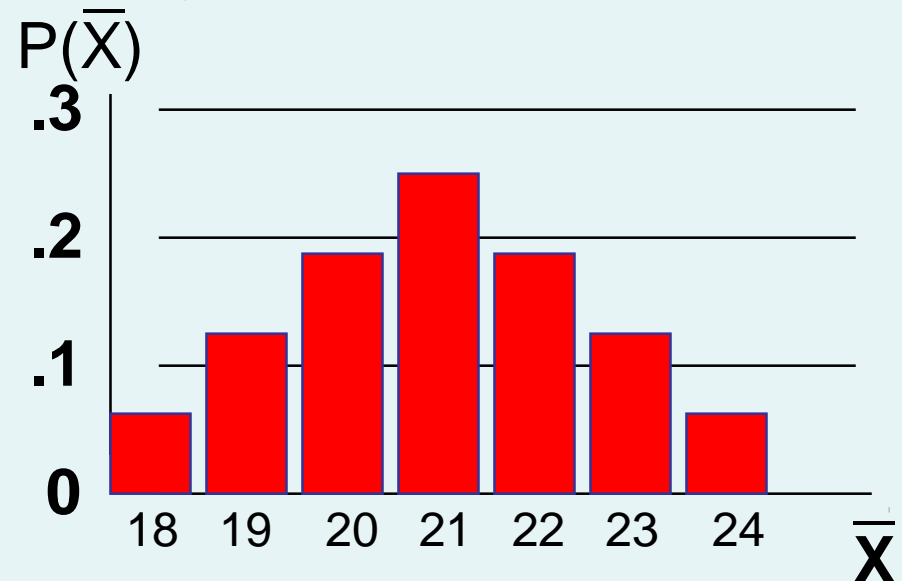
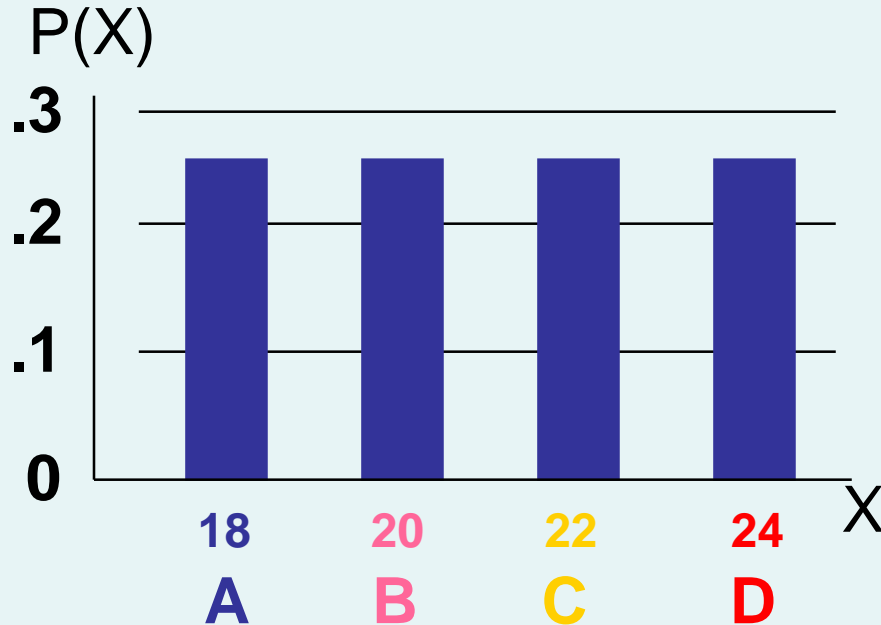
$$N = 4$$

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

Sample Means Distribution

$$n = 2$$

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



Sample Mean Sampling Distribution: Standard Error of the Mean

DCOVA A

- 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:**

(This assumes that sampling is with replacement or sampling is without replacement from an infinite population)

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

- Note that the standard error of the mean decreases as the sample size increases

Sample Mean Sampling Distribution: If the Population is Normal

DCOVA

- 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

DCOVA

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

$$Z = \frac{(\bar{X} - \mu_{\bar{X}})}{\sigma_{\bar{X}}} = \frac{(\bar{X} - \mu)}{\frac{\sigma}{\sqrt{n}}}$$

where:

- \bar{X} = sample mean
- μ = population mean
- σ = population standard deviation
- n = sample size

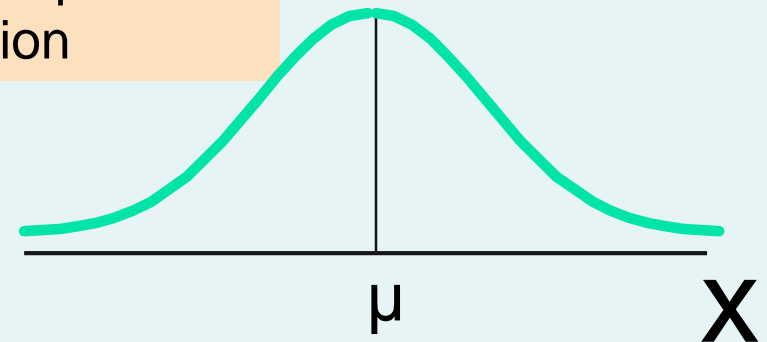
Sampling Distribution Properties

DCOVA

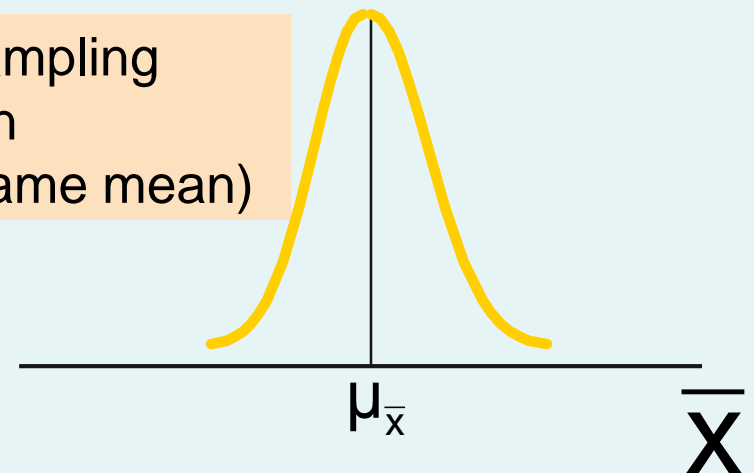
$$\mu_{\bar{X}} = \mu$$

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

Normal Population
Distribution



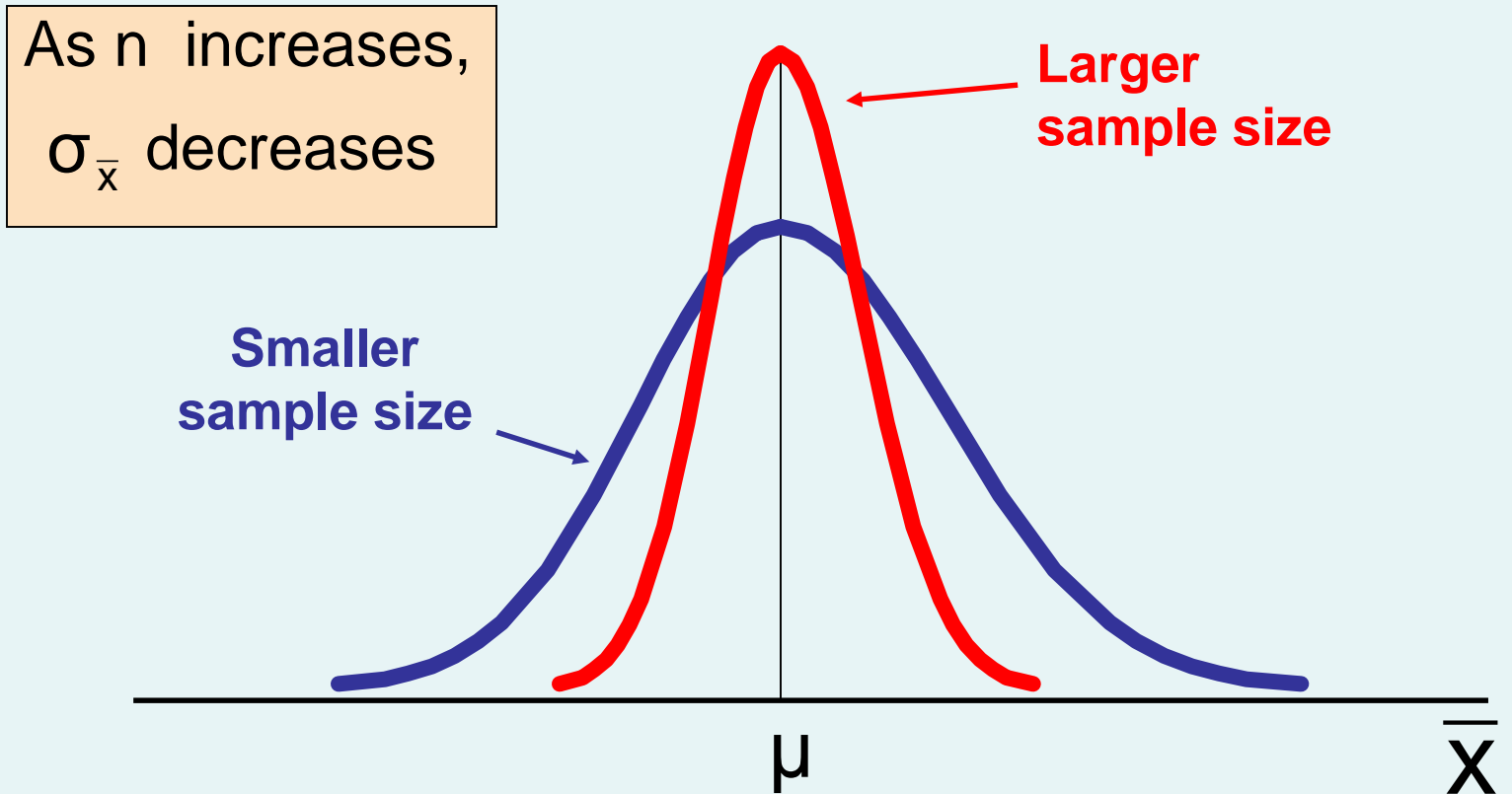
Normal Sampling
Distribution
(has the same mean)

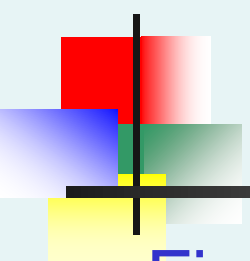


Sampling Distribution Properties

(continued)

DCOVA





Determining An Interval Including A Fixed Proportion of the Sample Means

DCOVA A

Find a symmetrically distributed interval around μ that will include 95% of the sample means when $\mu = 368$, $\sigma = 15$, and $n = 25$.

- Since the interval contains 95% of the sample means 5% of the sample means will be outside the interval
- Since the interval is symmetric 2.5% will be above the upper limit and 2.5% will be below the lower limit.
- From the standardized normal table, the Z score with 2.5% (0.0250) below it is -1.96 and the Z score with 2.5% (0.0250) above it is 1.96.

Determining An Interval Including A Fixed Proportion of the Sample Means

(continued)

DCOVA

- Calculating the lower limit of the interval

$$\bar{X}_L = \mu + Z \frac{\sigma}{\sqrt{n}} = 368 + (-1.96) \frac{15}{\sqrt{25}} = 362.12$$

- Calculating the upper limit of the interval

$$\bar{X}_U = \mu + Z \frac{\sigma}{\sqrt{n}} = 368 + (1.96) \frac{15}{\sqrt{25}} = 373.88$$

- 95% of all sample means of sample size 25 are between 362.12 and 373.88

Sample Mean Sampling Distribution: If the Population is **not** Normal

DCOVA

- 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:

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

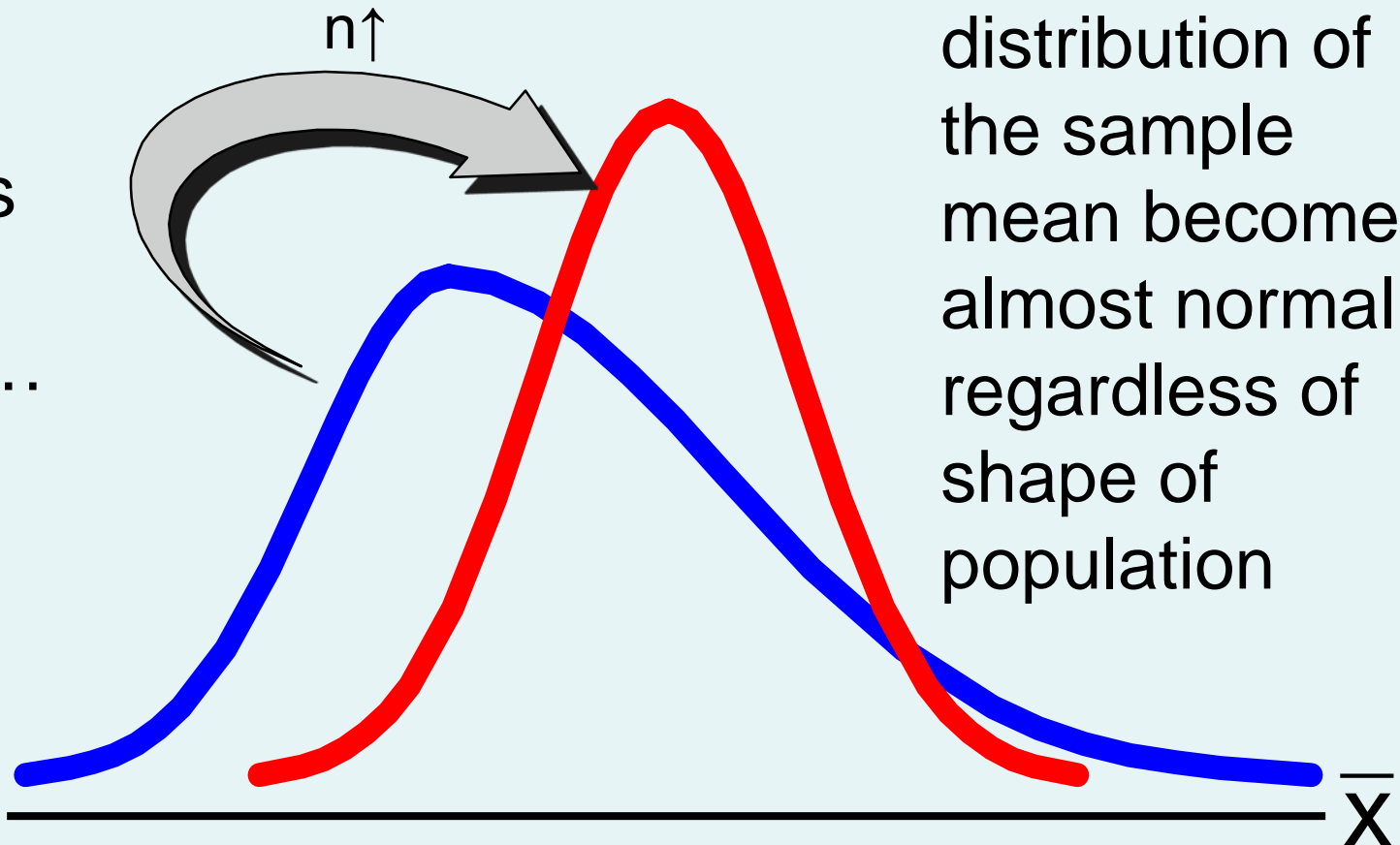
and

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

Central Limit Theorem

DCOVA A

As the sample size gets large enough...



the sampling distribution of the sample mean becomes almost normal regardless of shape of population

Sample Mean Sampling Distribution: If the Population is **not** Normal

(continued)

Sampling distribution
properties:

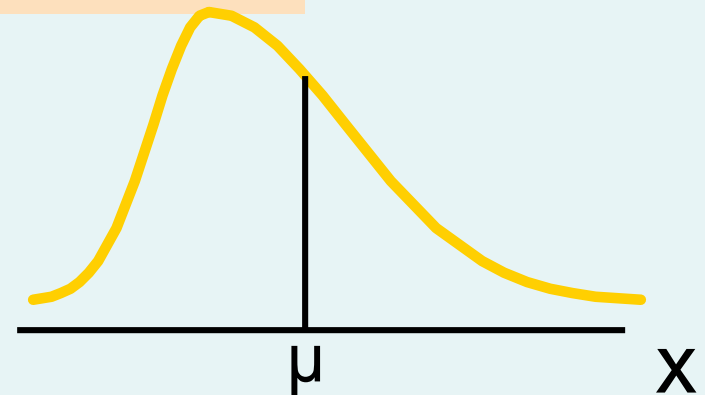
Central Tendency

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

Variation

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

Population Distribution

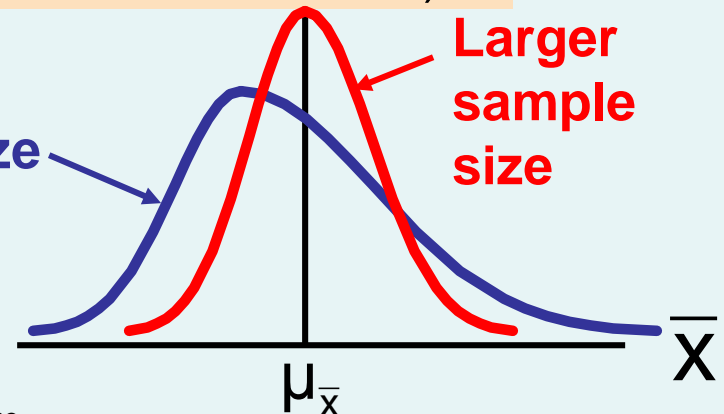


DCOVA A

Sampling Distribution
(becomes normal as n increases)

Smaller
sample size

Larger
sample size





How Large is Large Enough?

DCOVA A

- For most distributions, $n > 30$ will give a sampling distribution that is nearly normal
- For fairly symmetric distributions, $n > 15$
- For normal population distributions, the sampling distribution of the mean is always normally distributed



Example

DCOVA

- Suppose a 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)

DCOVA A

Solution:

- Even if the population is not normally distributed, the central limit theorem can be used ($n > 30$)
- ... so 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

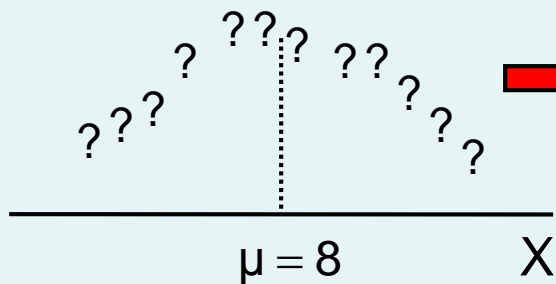
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Solution (continued):

DCOVA

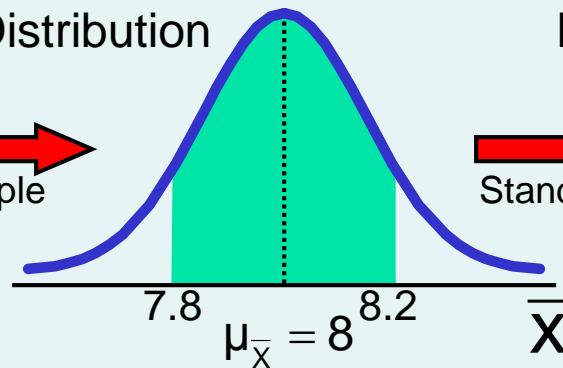
$$\begin{aligned} P(7.8 < \bar{X} < 8.2) &= P\left(\frac{7.8 - 8}{3/\sqrt{36}} < \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} < \frac{8.2 - 8}{3/\sqrt{36}}\right) \\ &= P(-0.4 < Z < 0.4) = 0.6554 - 0.3446 = \boxed{0.3108} \end{aligned}$$

Population
Distribution



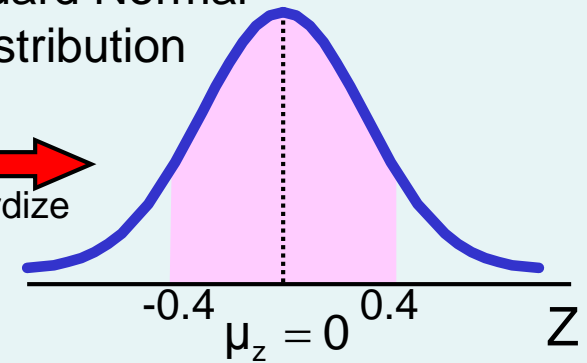
Sampling
Distribution

Sample



Standard Normal
Distribution

Standardize





Population Proportions

π = the proportion of the population having some characteristic

- **Sample proportion (p)** provides an estimate of π :

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

- $0 \leq p \leq 1$
- p is approximately distributed as a normal distribution when n is large

(assuming sampling with replacement from a finite population or without replacement from an infinite population)

Sampling Distribution of p

- Approximated by a normal distribution if:

- $n\pi \geq 5$
and
 $n(1 - \pi) \geq 5$

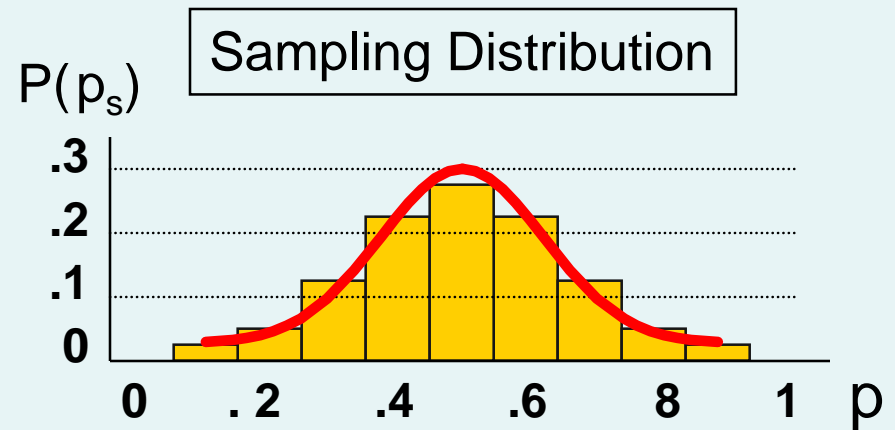
where

$$\mu_p = \pi$$

and

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

(where π = population proportion)





Z-Value for Proportions

DCOVA

Standardize p to a Z value with the formula:

$$Z = \frac{p - \pi}{\sigma_p} = \frac{p - \pi}{\sqrt{\frac{\pi(1 - \pi)}{n}}}$$



Example

- If the true proportion of voters who support Proposition A is $\pi = 0.4$, what is the probability that a sample of size 200 yields a sample proportion between 0.40 and 0.45?
- i.e.: **if $\pi = 0.4$ and $n = 200$, what is $P(0.40 \leq p \leq 0.45)$?**



Example

(continued)

DCOVA

- if $\pi = 0.4$ and $n = 200$, what is $P(0.40 \leq p \leq 0.45)$?

Find σ_p :

$$\sigma_p = \sqrt{\frac{\pi(1-\pi)}{n}} = \sqrt{\frac{0.4(1-0.4)}{200}} = 0.03464$$

Convert to
standardized
normal:

$$\begin{aligned} P(0.40 \leq p \leq 0.45) &= P\left(\frac{0.40 - 0.40}{0.03464} \leq Z \leq \frac{0.45 - 0.40}{0.03464}\right) \\ &= P(0 \leq Z \leq 1.44) \end{aligned}$$

Example

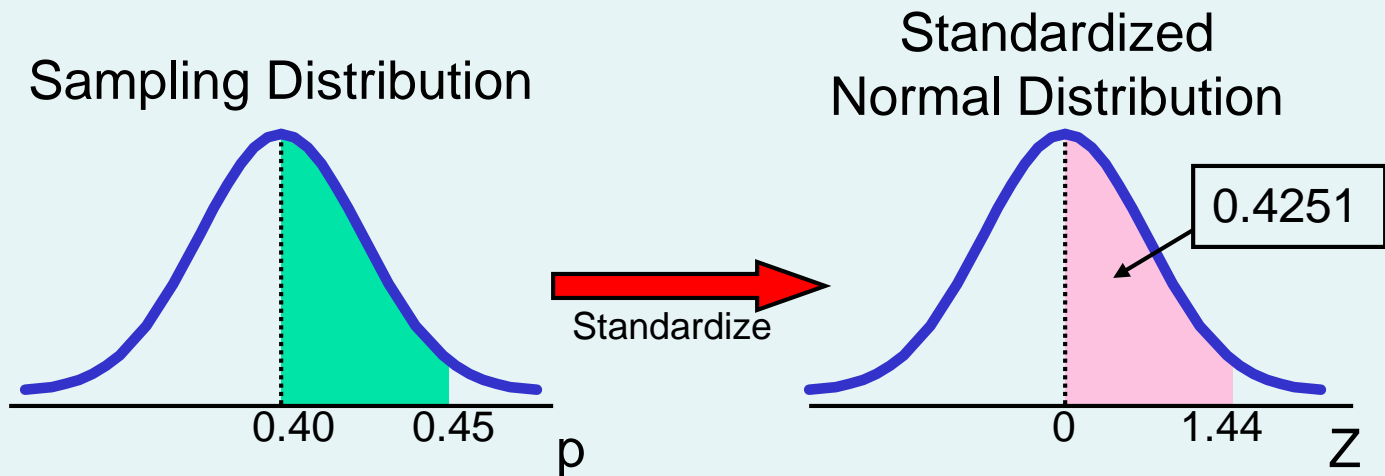
(continued)

DCOVA

- if $\pi = 0.4$ and $n = 200$, what is $P(0.40 \leq p \leq 0.45)$?

Utilize the cumulative normal table:

$$P(0 \leq Z \leq 1.44) = 0.9251 - 0.5000 = 0.4251$$



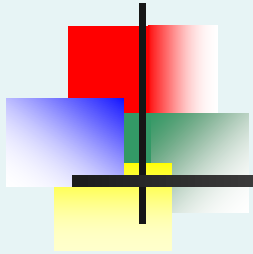


Chapter Summary

In this chapter we discussed

- Sampling distributions
- The sampling distribution of the mean
 - For normal populations
 - Using the Central Limit Theorem
- The sampling distribution of a proportion
- Calculating probabilities using sampling distributions

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Online Topic

Sampling From Finite Populations



Learning Objectives

In this topic, you learn:

- To know when finite population corrections are needed
- To know how to utilize finite population correction factors in calculating standard errors

Finite Population Correction Factors

DCOVA A

- Used to calculate the standard error of both the sample mean and the sample proportion
- Needed when the sample size, n , is more than 5% of the population size N (i.e. $n / N > 0.05$)
- The Finite Population Correction Factor Is:

$$f_{pc} = \sqrt{\frac{N - n}{N - 1}}$$

Using The fpc In Calculating Standard Errors

DCOVA

Standard Error of the Mean for Finite Populations

$$\sigma_{\bar{X}} = \frac{\sigma}{\sqrt{n}} \sqrt{\frac{N-n}{N-1}}$$

Standard Error of the Proportion for Finite Populations

$$\sigma_p = \sqrt{\frac{\pi(1-\pi)}{n}} \sqrt{\frac{N-n}{N-1}}$$

Using The fpc Reduces The Standard Error

DCOVA

- The fpc is always less than 1
- So when it is used it reduces the standard error
- Resulting in more precise estimates of population parameters

Using fpc With The Mean - Example

Suppose a random sample of size 100 is drawn from a population of size 1,000 with a standard deviation of 40.

Here $n=100$, $N=1,000$ and $100/1,000 = 0.10 > 0.05$.

So using the fpc for the standard error of the mean we get:

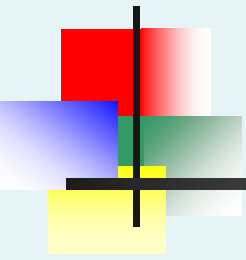
$$\sigma_{\bar{X}} = \frac{40}{\sqrt{100}} \sqrt{\frac{1000-100}{1000-1}} = 3.8$$



Topic Summary

In this topic we discussed

- When a finite population correction should be used.
- How to utilize a finite population correction factor in calculating the standard error of both a sample mean and a sample proportion



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