

## Two-Sample Tests

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# Learning Objectives

In this chapter, you learn how to use hypothesis testing for comparing the difference between:

- The means of two independent populations



# Two-Sample Tests Overview

## Two Sample Tests

Independent  
Population  
Means

Means,  
Related  
Populations

Independent  
Population  
Proportions

Independent  
Population  
Variances

Examples

Group 1 vs.  
Group 2

Same group  
before vs. after  
treatment

Proportion 1 vs.  
Proportion 2

Variance 1 vs.  
Variance 2



# Two-Sample Tests

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

Goal: Test hypothesis or form a confidence interval for the difference between two population means,  $\mu_1 - \mu_2$

The point estimate for the difference between sample means:

$$\bar{X}_1 - \bar{X}_2$$



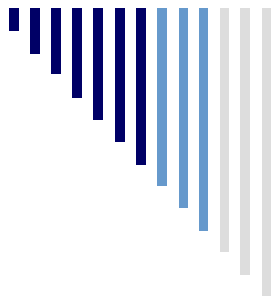
# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

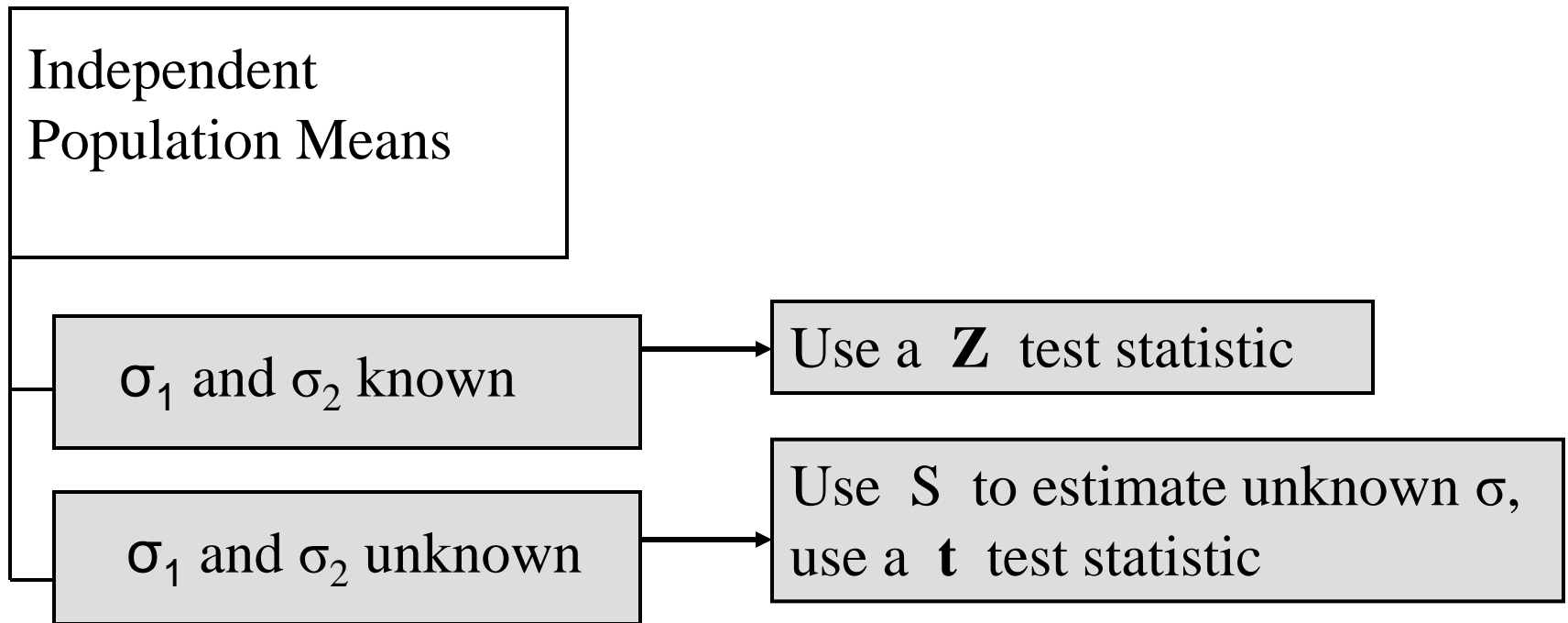
$\sigma_1$  and  $\sigma_2$  unknown

- Different data sources
  - Independent: Sample selected from one population has no effect on the sample selected from the other population
- Use the difference between 2 sample means
- Use Z test, pooled variance t test, or separate-variance t test



# Two-Sample Tests

## Independent Populations





# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

Assumptions:

- Samples are randomly and independently drawn
- population distributions are normal



# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

When  $\sigma_1$  and  $\sigma_2$  are known and both populations are normal, the test statistic is a Z-value and the standard error of  $\bar{X}_1 - \bar{X}_2$  is

$$\sigma_{\bar{X}_1 - \bar{X}_2} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$





# Two-Sample Tests Independent Populations

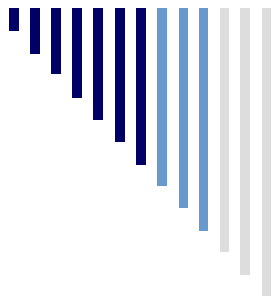
Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

The test statistic is:

$$Z = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}$$



# Two-Sample Tests

## Independent Populations

Two Independent Populations, Comparing Means

Lower-tail test:

$$H_0: \mu_1 \geq \mu_2$$

$$H_1: \mu_1 < \mu_2$$

i.e.,

$$H_0: \mu_1 - \mu_2 \geq 0$$

$$H_1: \mu_1 - \mu_2 < 0$$

Upper-tail test:

$$H_0: \mu_1 \leq \mu_2$$

$$H_1: \mu_1 > \mu_2$$

i.e.,

$$H_0: \mu_1 - \mu_2 \leq 0$$

$$H_1: \mu_1 - \mu_2 > 0$$

Two-tail test:

$$H_0: \mu_1 = \mu_2$$

$$H_1: \mu_1 \neq \mu_2$$

i.e.,

$$H_0: \mu_1 - \mu_2 = 0$$

$$H_1: \mu_1 - \mu_2 \neq 0$$

# Two-Sample Tests

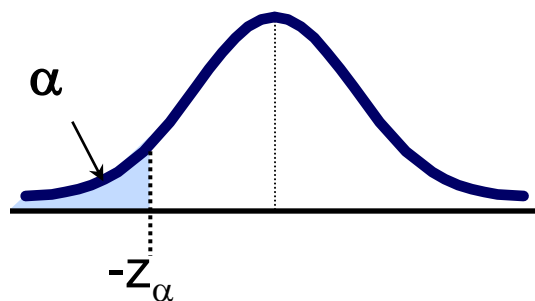
## Independent Populations

Two Independent Populations, Comparing Means

Lower-tail test:

$$H_0: \mu_1 - \mu_2 \geq 0$$

$$H_1: \mu_1 - \mu_2 < 0$$

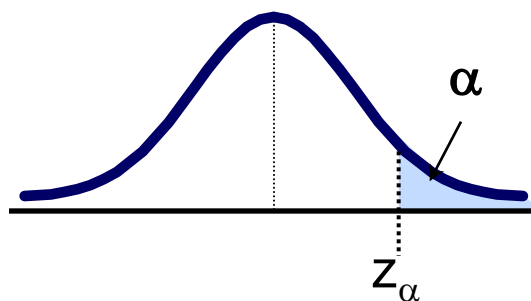


Reject  $H_0$  if  $Z < -Z_\alpha$

Upper-tail test:

$$H_0: \mu_1 - \mu_2 \leq 0$$

$$H_1: \mu_1 - \mu_2 > 0$$

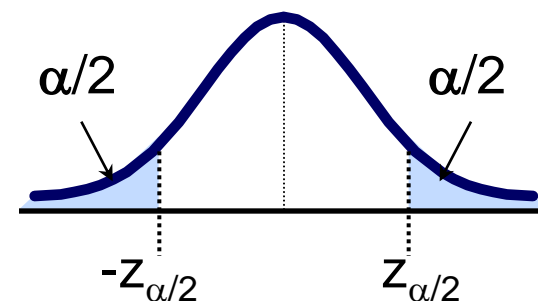


Reject  $H_0$  if  $Z > Z_\alpha$

Two-tail test:

$$H_0: \mu_1 - \mu_2 = 0$$

$$H_1: \mu_1 - \mu_2 \neq 0$$



Reject  $H_0$  if  $Z < -Z_{\alpha/2}$   
or  $Z > Z_{\alpha/2}$



# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

Assumptions:

- Samples are randomly and independently drawn
- Populations are normally distributed
- Population variances are unknown but assumed equal



# Two-Sample Tests

## Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

Forming interval estimates:

- The population variances are assumed equal, so use the two sample standard deviations and pool them to estimate  $\sigma$
- the test statistic is a t value with  $(n_1 + n_2 - 2)$  degrees of freedom



# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

The pooled standard  
deviation is:

$$S_p = \sqrt{\frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{(n_1 - 1) + (n_2 - 1)}}$$



# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

The test statistic is:

$$t = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}}$$

Where  $t$  has  $(n_1 + n_2 - 2)$  d.f., and

$$S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{(n_1 - 1) + (n_2 - 1)}$$



# Two-Sample Tests

## Independent Populations

- You are a financial analyst for a brokerage firm. Is there a difference in dividend yield between stocks listed on the NYSE & NASDAQ? You collect the following data:

	<u>NYSE</u>	<u>NASDAQ</u>
<b>Number</b>	<b>21</b>	<b>25</b>
<b>Sample mean</b>	<b>3.27</b>	<b>2.53</b>
<b>Sample std dev</b>	<b>1.30</b>	<b>1.16</b>

Assuming both populations are approximately normal with equal variances, is there a difference in average yield ( $\alpha = 0.05$ )?





# Two-Sample Tests Independent Populations

The test statistic is:

$$t = \frac{(\bar{X}_1 - \bar{X}_2) - (\mu_1 - \mu_2)}{\sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}} = \frac{(3.27 - 2.53) - 0}{\sqrt{1.5021 \left( \frac{1}{21} + \frac{1}{25} \right)}} = 2.040$$

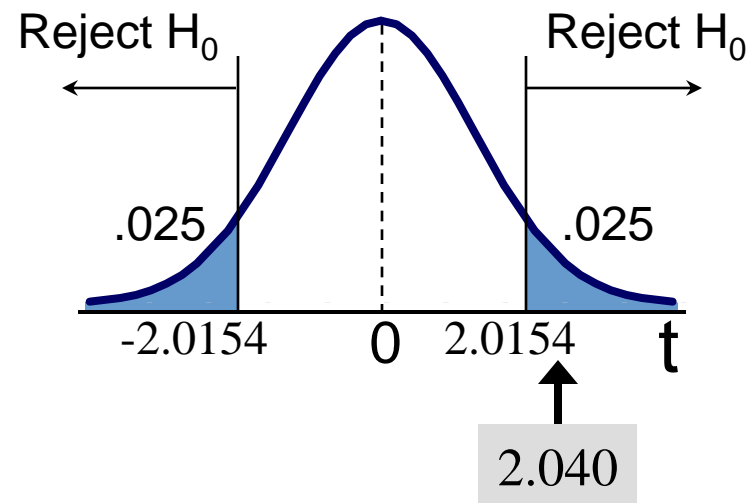
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$$S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{(n_1 - 1) + (n_2 - 1)} = \frac{(21 - 1)1.30^2 + (25 - 1)1.16^2}{(21 - 1) + (25 - 1)} = 1.5021$$

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# Two-Sample Tests Independent Populations

- $H_0: \mu_1 - \mu_2 = 0$  i.e.  $(\mu_1 = \mu_2)$
- $H_1: \mu_1 - \mu_2 \neq 0$  i.e.  $(\mu_1 \neq \mu_2)$
- $\alpha = 0.05$
- $df = 21 + 25 - 2 = 44$
- **Critical Values:  $t = \pm 2.0154$**
- **Test Statistic: 2.040**



Decision: Reject  $H_0$  at  $\alpha = 0.05$

Conclusion: There is evidence of a difference in the means.



# Independent Populations Unequal Variance

- If you cannot assume population variances are equal, the pooled-variance t test is inappropriate
- Instead, use a separate-variance t test, which includes the two separate sample variances in the computation of the test statistic
- The computations are complicated



# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

The confidence interval for  
 $\mu_1 - \mu_2$  is:

$$\left( \bar{X}_1 - \bar{X}_2 \right) \pm Z \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$



# Two-Sample Tests Independent Populations

Independent  
Population Means

$\sigma_1$  and  $\sigma_2$  known

$\sigma_1$  and  $\sigma_2$  unknown

The confidence interval for  
 $\mu_1 - \mu_2$  is:

$$\left( \bar{X}_1 - \bar{X}_2 \right) \pm t_{n_1+n_2-2} \sqrt{S_p^2 \left( \frac{1}{n_1} + \frac{1}{n_2} \right)}$$

Where

$$S_p^2 = \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{(n_1 - 1) + (n_2 - 1)}$$



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# Chapter Summary

In this chapter, we have

- Compared two independent samples
    - Performed Z test for the differences in two means
    - Performed pooled variance t test for the differences in two means
    - Formed confidence intervals for the differences between two means
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