

Difference of Two Population means



• $\overline{X}_1 - \overline{X}_2$ is Random Variable

- $\overline{X}_1 \overline{X}_2$ is a point estimator for $\mu 1 \mu 2$
- The standard deviation is given by the formula $\sqrt{\frac{{\sigma_1}^2}{n_1} + \frac{{\sigma_2}^2}{n_2}}$
- If n1 and n2 are sufficiently large, follows a normal distribution.

5



Difference between two means – known population variances

- If both σ_1 and σ_2 are known and the two populations are independently selected, this test can be run.
- Test Statistic:

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



Example 1

- Are larger houses more likely to have pools?
- The housing data square footage (size) was split into two groups by pool (Y/N).
- Test the hypothesis that the homes with pools have more square feet than the homes without pools. Let $\alpha=.01$

7



EXAMPLE 1 - Design

$$H_o: \mu_1 \le \mu_2 \quad H_a: \mu_1 > \mu_2$$

 $H_o: \mu_1 - \mu_2 \le 0 \quad H_a: \mu_1 - \mu_2 > 0$

 $\alpha = .01$

$$Z = (\overline{X}_1 - \overline{X}_2)/(\sqrt{\sigma_1/n_1 + \sigma_2/n_2})$$

 H_0 is rejected if Z > 2.326

8



EXAMPLE 1 Data

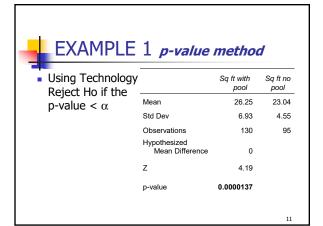
- Population 1 Size with pool
- Population 2 Size without pool
- Sample size = 130
- Sample size = 95
- Sample mean = 26.25
- Sample mean = 23.04
- Pop Std Dev = 6.93
- Pop Std Dev = 4.55

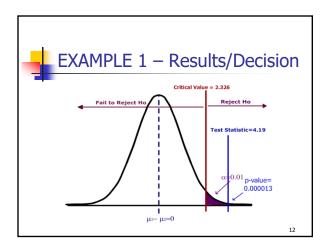


EXAMPLE 1 DATA

$$Z = \frac{(26.25 - 23.04) - 0}{\sqrt{\frac{6.93^2}{130} + \frac{4.55^2}{95}}} = 4.19$$

- Decision: Reject Ho
- Conclusion: Homes with pools have more mean square footage.







Pooled variance t-test

- To conduct this test, three assumptions are required:
 - The populations must be normally or approximately normally distributed (or central limit theorem must apply).
 - The sampling of populations must be independent.
 - The population variances must be equal.

13



Pooled Sample Variance and Test Statistic

- Pooled Sample Variance: $s^2 = \frac{(n_1 1)s_1^2 + (n_2)}{n_1^2 + n_2^2}$
- Test Statistic:

$$t = \frac{(\overline{X}_1 - \overline{X}_2) - (\mu_1 - \mu_2)}{s_p \sqrt{\frac{1}{n_1} + \frac{1}{n_2}}}$$

 $df = n_1 + n_2 - 2$

14



EXAMPLE 2

- A recent EPA study compared the highway fuel economy of domestic and imported passenger cars.
- A sample of 12 imported cars revealed a mean of 35.76 mpg with a standard deviation of 3.86.
- A sample of 15 domestic cars revealed a mean of 33.59 mpg with a standard deviation of 2.16 mpg
- At the .05 significance level can the EPA conclude that the mpg is higher on the imported cars? (Let subscript 2 be associated with domestic cars.)



EXAMPLE 2 - critical value method

- $H_o: H_o: \mu_1 \le \mu_2 \quad H_a: \mu_1 > \mu_2$
- $\alpha = .05$
- : $t = (\overline{X}_1 \overline{X}_2)/(s_p \sqrt{1/n_1 + 1/n_2})$: H_0 is rejected if t > 1.708, df = 25
- : $t=1.85 H_0$ is rejected. Imports have a higher mean mpg than domestic cars.



t-test when variances are not equal.

- Test statistic:
- Degrees of freedom: df =
- This test (also known as the Welch-Aspin Test) has **less power** then the prior test and should only be used when it is clear the population variances are different.



EXAMPLE 2

- : $H_o: \mu_1 \le \mu_2$ $H_a: \mu_1 > \mu_2$
- : $\alpha = .05$
- : t' test
- : *H*₀ is rejected if t>1.746, *df*=16
- : $t=1.74 H_0$ is not rejected. There is insufficient sample evidence to claim a higher mpg on the imported cars.



Using Technology

- Decision Rule: Reject H_o if pvalue<α
- Megastat: Compare Two Independent Groups
- Use Equal Variance or Unequal Variance Test
- Use Original Data or Summarized Data

 $\textbf{domestic}\ 29.8\ 33.3\ 34.7\ 37.4\ 34.4\ 32.7\ 30.2\ 36.2\ 35.5\ 34.6\ 33.2\ 35.1\ 33.6\ 31.3\ 31.9$

import 39.0 35.1 39.1 32.2 35.6 35.5 40.8 34.7 33.2 29.4 42.3 32.2



Pooled Variance t-test

Minitab output

Two-sample T for MPG

• p-value = 0.038

TYPE N Mean StDev SE Mean import 12 35.76 3.86 1.1 US car 15 33.59 2.16 0.56

• p-value $< \alpha = .05$

Difference = μ (import) - μ (US car) Estimate for difference: 2.16 9% lower bound for difference: 0.16 T-Test of difference = 0 (vs >): T-Value = 1.85 P-Value = 0.038 DF = 25 Both use Pooled StDev = 3.0264

· Reject Ho



Unequal Variances t-test

Minitab output

Two-sample T for MPG

• p-value = 0.051

TYPE N Mean StDev SE Mean import 12 35.76 3.86 1.1 US car 15 33.59 2.16 0.56

• p-value $< \alpha = .05$

Difference = µ (import) - µ (US car)
Estimate for difference: 2.16
95% lower bound for difference: -0.01
T-Test of difference = 0 (vs >): T-Value = 1.74
P-Value = 0.051 DF = 16

· Fail to Reject Ho



Hypothesis Testing - Paired Observations

- Independent samples are samples that are not related in any way.
- Dependent samples are samples that are paired or related in some fashion.
 - For example, if you wished to buy a car you would look at the same car at two (or more) different dealerships and compare the prices.
- Use the following test when the samples are dependent:

22



Hypothesis Testing Involving Paired Observations

$$t = \frac{\overline{X}_d - \mu_d}{s_d / \sqrt{n}}$$

- where $\overline{X}_{\boldsymbol{d}}$ is the average of the differences
- S_d is the standard deviation of the differences
- n is the number of pairs (differences)

23



EXAMPLE 3

- An independent testing agency is comparing the daily rental cost for renting a compact car from Hertz and Avis.
- A random sample of 15 cities is obtained and the following rental information obtained.
- At the .05 significance level can the testing agency conclude that there is a difference in the rental charged?



Example 3 – continued

Data for Hertz

 $\overline{X}_1 = 46.67$ $s_1 = 5.23$

Data for Avis

 $\overline{X}_2 = 44.87$

 $s_2 = 5.62$

| City | Hertz | Avis |
|---------------|-------|------|
| Atlanta | 42 | 40 |
| Baltimore | 51 | 47 |
| Boston | 46 | 42 |
| Chicago | 56 | 52 |
| Cleveland | 45 | 43 |
| Denver | 48 | 48 |
| Dallas | 56 | 54 |
| Honolulu | 37 | 32 |
| Los Angeles | 51 | 48 |
| Kansas City | 45 | 48 |
| Miami | 41 | 39 |
| New York | 44 | 42 |
| San Francisco | 48 | 45 |
| Seattle | 46 | 50 |
| Washington DC | 44 | 43 |

25



Example 3 - continued

By taking the difference of each pair, variability (measured by standard deviation) is reduced.

 $\overline{X}_d = 1.80$ $s_d = 2.513$

n = 15

| City | Hertz | Avis | Difference |
|---------------|-------|------|------------|
| Atlanta | 42 | 40 | 2 |
| Baltimore | 51 | 47 | 4 |
| Boston | 46 | 42 | 4 |
| Chicago | 56 | 52 | 4 |
| Cleveland | 45 | 43 | 2 |
| Denver | 48 | 48 | 0 |
| Dallas | 56 | 54 | 2 |
| Honolulu | 37 | 32 | 5 |
| Los Angeles | 51 | 48 | 3 |
| Kansas City | 45 | 48 | -3 |
| Miami | 41 | 39 | 2 |
| New York | 44 | 42 | 2 |
| San Francisco | 48 | 45 | 3 |
| Seattle | 46 | 50 | -4 |
| Washington DC | 44 | 43 | 1 |

26



EXAMPLE 3 continued

- $H_0: \mu_d = 0$ $H_1: \mu_d \neq 0$
- α=.05
- Matched pairs t test, df=14
- H_0 is rejected if t<-2.145 or t>2.145
- $t = (1.80)/[2.513/\sqrt{15}] = 2.77$
- Reject H₀.
- There is a difference in mean price for compact cars between Hertz and Avis. Avis has lower mean prices.



Megastat Output – Example 3

Hypothesis Test: Paired Observations

0.000 hypothesized value 46.667 mean Hertz

44.867 mean Avis

1.800 mean difference (Hertz - Avis)

2.513 std. dev.

0.649 std. error

15 n

14 df

2.77 t

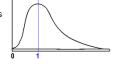
.0149 p-value (two-tailed)



Characteristics of F-Distribution

- There is a "family" of F Distributions.
 Each member of the family is determined by two parameters: the numerator degrees of freedom and the denominator degrees of freedom.
 F cannot be negative, and it is a continuous distribution.
 The F distribution is positively skewed.
 Its values range from 0 to ...

- Its values range from 0 $to \infty$. As $F \rightarrow \infty$ the curve approaches the X-axis.





Test for Equal Variances

• For the two tail test, the test statistic is given by:

- s_i^2 and s_j^2 are the sample variances for the two populations.
- There are 2 sets of degrees of freedom: n_i-1 for the numerator, n_i-1 for the denominator



EXAMPLE 4

- A stockbroker at brokerage firm, reported that the mean rate of return on a sample of 10 software stocks was 12.6 percent with a standard deviation of 4.9 percent.
- The mean rate of return on a sample of 8 utility stocks was 10.9 percent with a standard deviation of 3.5 percent.
- At the .05 significance level, can the broker conclude that there is more variation in the software stocks?



Test Statistic depends on Hypotheses

Hypotheses

Test Statistic

 $H_o: \sigma_1 \geq \sigma_2$

 $F = \frac{s_2^2}{s_1^2} \quad use \ \alpha \ table$ $H_a: \sigma_1 < \sigma_2$

 $H_o: \sigma_1 \leq \sigma_2$

 $F = \frac{s_1^2}{s_2^2} \quad use \ \alpha \ table$

 $H_a: \sigma_1 > \sigma_2$

 $H_o: \sigma_1 = \sigma_2$

 $H_a: \sigma_1 \neq \sigma_2$

 $F = \frac{\max(s_1^2, s_2^2)}{\min(s_1^2, s_2^2)}$ use $\alpha/2$ table



EXAMPLE 4 continued

- $: H_o: \sigma_1 \leq \sigma_2 \quad H_a: \sigma_1 > \sigma_2$
- : $\alpha = .05$
- F-test
- : H₀ is rejected if F>3.68, df=(9,7)
- : F=4.9 2 /3.5 2 =1.96 \rightarrow Fail to Reject H_0 .
- There is insufficient evidence to claim more variation in the software stock.

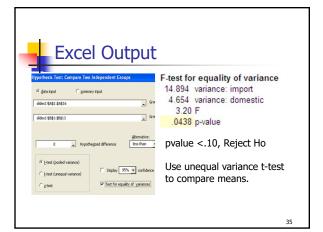
Chapter 10 Slides



Excel Example

- Using Megastat Test for equal variances under two population independent samples test and click the box to test for equality of variances
- The default p-value is a two-tailed test, so take onehalf reported p-value for one-tailed tests
- Example Domestic vs Import Data
- $H_o: \sigma_1 = \sigma_2$ $H_a: \sigma_1 \neq \sigma_2$
- α =.10
- Reject Ho means use unequal variance t-test
- FTR Ho means use pooled variance t-test

24





Comparing two proportions

- Suppose we take a sample of n₁ from population 1 and n₂ from population 2.
- Let X₁ be the number of success in sample 1 and X₂ be the number of success in sample 2.
- The sample proportions are then calculated for each group.

$$\hat{p}_1 = \frac{X_1}{n_1} \qquad \hat{p}_2 = \frac{X_2}{n_2}$$



Hypothesis testing for 2 Proportions

 In conducting a Hypothesis test where the Null hypothesis assumes equal proportions, it is best practice to pool or combine the sample proportions into a single estimated proportion, and use an estimated standard error.

$$\overline{p} = \frac{X_1 + X_2}{n_1 + n_2}$$
 $s_{\hat{p}_1 - \hat{p}_2} = \sqrt{\frac{\overline{p}(1 - \overline{p})}{n_1} + \frac{\overline{p}(1 - \overline{p})}{n_2}}$

37



Hypothesis testing for 2 Proportions

 The test statistic will have a Normal Distribution as long as there are at least 10 successes and 10 failures in both samples.

$$Z = \frac{\left(\hat{p}_1 - \hat{p}_2\right) - \left(p_1 - p_2\right)}{\sqrt{\frac{\overline{p}\left(1 - \overline{p}\right)}{n_1} + \frac{\overline{p}\left(1 - \overline{p}\right)}{n_2}}}$$

38



Example

- In an August 2016 Study, Pew Research asked the sampled Americans if background checks required at gun stores should be made universal extended to all sales of guns between private owners or at gun shows
- 772 out 990 men said yes, while 857 out of 1020 women said yes.
- Is there a difference in the proportion of men and women who support universal background checks for purchasing guns? Design and conduct the test with a significance level of 1%.



Example (Design)

Ho: p_m=p_w (There is no difference in the proportion of support for background checks by gender)
 Ha: p_m≠p_w (There is a difference in the proportion of support for background checks by gender)

Model:

Two proportion Z test. This is a two-tailed test with a = 0.01.

- Model Assumptions: for men there are 772 yes and 218 no.
 For women there are 857 yes and 16 no. Since all these numbers exceed 10, the model is appropriate.
- Decision Rules:
 Original Mark

Critical Value Method - Reject Ho if Z > 2.58 or Z < -2.58. P-value method - Reject Ho if p-value < 0.01

40



Example (Results)

$$\hat{p}_m = \frac{772}{990} = 0.780 \qquad \hat{p}_w = \frac{857}{1020} = 0.840$$

$$\overline{p} = \frac{772 + 857}{990 + 1020} = 0.810 \qquad Z = \frac{(0.780 - 0.840) - 0}{\sqrt{\frac{0.810(1 - 0.810)}{990} + \frac{0.810(1 - 0.810)}{1020}}} = -3.45$$

p-value = 0.0005 < α Reject Ho Under both methods.

Conclusion: There is a difference in the proportion of support for background checks by gender. Women are more likely to support background checks.