

Univariate Statistical Analysis

Lecture 11

Analysis of Variance Chapter 15

Single-Factor ANOVA and the F Test

A **single-factor analysis of variance (ANOVA)** problem involves a comparison of k population or treatment means $\mu_1, \mu_2, \dots, \mu_k$. The objective is to test

$H_0: \mu_1 = \mu_2 = \mu_3$
against

$H_a: \text{At least two of the } \mu\text{'s are different from each other, } \mu_1 \neq \mu_2 \neq \mu_3$

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Single-Factor ANOVA and the F Test

Watch video:

<https://www.youtube.com/watch?v=WcmzS3nEUqo>

Understanding Analysis of Variance (ANOVA) including Excel - Statistics Help

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Refer to Text Example 15.2 (p.764) Heart Attack
(Data source p.762 Table 15.1).

$$\bar{x}_1 = 10.89, \bar{x}_2 = 11.25, \bar{x}_3 = 11.37, \bar{x}_4 = 11.75$$

$\bar{\bar{x}} = 11.315$ is the average of all samples.

$$s_1 = 0.69, s_2 = 0.74, s_3 = 0.91, s_4 = 1.07$$

$$n_1 = n_2 = n_3 = n_4 = 35$$

$k = 3$ is the number of groups

$$N = n_1 + n_2 + n_3 + n_4 = 35 + 35 + 35 + 35 = 140$$

is the total samples

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A measure of **differences among the sample means** is the **treatment sum of squares**, denoted by **SSTr** and given by

$$\begin{aligned} \text{SSTr} &= n_1(\bar{x}_1 - \bar{\bar{x}})^2 + n_2(\bar{x}_2 - \bar{\bar{x}})^2 + \dots + n_k(\bar{x}_k - \bar{\bar{x}})^2 \\ &= 35(10.89 - 11.315)^2 + 35(11.25 - 11.315)^2 + 35(11.37 - 11.315)^2 + 35(11.75 - 11.315)^2 \\ &= 6.322 + 0.148 + 0.106 + 6.623 \\ &= 13.199 \end{aligned}$$

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A measure of **variability within the k samples**, called **error sum of squares** and denoted by **SSE**, is

$$\begin{aligned} \text{SSE} &= (n_1 - 1)s_1^2 + (n_2 - 1)s_2^2 + \dots + (n_k - 1)s_k^2 \\ &= (35 - 1)(0.69)^2 + (35 - 1)(0.74)^2 + (35 - 1)(0.91)^2 + 35 - 1)(1.07)^2 \\ &= 101.888 \end{aligned}$$

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A **single-factor analysis of variance (ANOVA)** problem involves a comparison of k population or treatment means $\mu_1, \mu_2, \dots, \mu_k$. The objective is to test

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against

$$H_a: \mu_1 \neq \mu_2 \neq \mu_3$$

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Each sum of squares has an associated df:

$$\text{treatment } df_1 = k - 1 = 4 - 1 = 3$$

$$\text{error } df_2 = N - k = 140 - 4 = 136$$

A **mean square** is a sum of squares divided by its df. In particular,

$$\text{mean square for treatments} = \text{MStr} = \frac{\text{SSTr}}{k - 1} = \frac{13.199}{4 - 1} = 4.4$$

$$\text{mean square for error} = \text{MSe} = \frac{\text{SSE}}{N - k} = \frac{101.888}{140 - 4} = 0.749$$

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All F tests in this text are upper-tailed, so P -values are areas under the F curve to the right of the calculated values of F .

$$\text{Test value: } F = \frac{\text{MStr}}{\text{MSE}} = \frac{4.4}{0.749} = 5.87$$

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Critical value, given

$$df_1 = 3$$

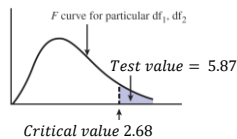
$$df_2 = 136 \rightarrow 120 \text{ (closest value in the table)}$$

$$\text{Level of significant} = 0.05$$

Refer to Appendix TABLE 6 (p.795 – 798)

The critical value is 2.68

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Conclusion:

Reject H_0 . There is **enough evidence** to conclude that at least one population mean is different that the other means.

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