

Chapter:13

Analysis of Variance (ANOVA)

INTRODUCTION

Analysis of variance is a powerful tool in the test of significance. In the test of significance ' t -test' is used to test the equality of two population means. Whereas ' F -test' is used to test the equality of several means. So F -test plays a vital role in the analysis of variance.

Definition

According to Prof. R.A. Fisher, "separation of variances ascribable to one group of causes from the variance ascribable to another group of causes" is known as analysis of variance.

The following are the important assumptions of analysis of variance.

1. All the observations are independent.
2. The parent population from which observations are drawn should be a normal population.
3. The various class effects are additive in nature.

Uses

1. Analysis of variance is used to test the equality of several means.
2. It is used to have a linearity regression lines.
3. It is used to test the significance of correlation ratio.

These are two types of analysis of classification.

1. One-way classification

If the data is divided into some classes according to one factor of causes, then the classification is called one way-classification.

2. Two-way classification

If the data is divided into some classes according to two factors of causes, then the classification is called two-way classification.

ANALYSIS OF VARIANCE—ONE-WAY CLASSIFICATION

n units of the variable X are divided into k classes A_1, A_2, \dots, A_k of sizes n_1, n_2, \dots, n_k . Such that $\sum_{i=1}^k n_i = n$.

Let Y_{ij} denote the value of j th unit in the i th class ($i = 1, 2, \dots, k; j = 1, 2, \dots, n_i$). These values can be put in the tabular form.

| A_1 | A_2 | | A_i | | A_k |
|-----------|-----------|--|-----------|--|-----------|
| x_{11} | x_{21} | | x_{i1} | | x_{k1} |
| x_{12} | x_{22} | | x_{i2} | | x_{k2} |
| | | | | | |
| x_{1j} | x_{2j} | | x_{ij} | | x_{kj} |
| | | | | | |
| x_{1n1} | x_{2n2} | | x_{ini} | | x_{knk} |
| $x_{1.}$ | $x_{2.}$ | | $x_{i.}$ | | $x_{k.}$ |

Mathematical Linear Model

The observations x_{ij} follow mathematical linear model as

$$X_{ij} = \mu + \alpha_i + \epsilon_{ij}$$

where $i = 1, 2, \dots, k; j = 1, 2, \dots, n_i$

μ = General mean effect

α_i = The mean effect of i th class

ϵ_{ij} = Error follows $N(0, \sigma^2)$

Assumptions

1. The sample observation x_{ij} should be independent.
2. Different class effects are additive in nature.
3. ϵ_{ij} follows $N(0, \sigma^2)$

Testing of Hypothesis

Now we set up null hypothesis as

$$H_0: \alpha_1 = \alpha_2 = \dots = \alpha_k$$

i.e.

H_0 : All the class effects are equal

Steps of Calculation

Calculate total G of all the observations is all samples

$$G = \Sigma x_1 + \Sigma x_2 + \Sigma x_3 + \dots + \Sigma x_k$$

Correction factor is (CF) = $\frac{(G)^2}{N}$

Calculate the total sum of squares (TSS) = $\sum \sum X_{ij}^2 - \frac{(G)^2}{N}$

Sum of squares of treatments

$$(SST) = \left[\frac{\sum X_1^2}{n_1} + \frac{\sum X_2^2}{n_2} + \frac{\sum X_3^2}{n_3} + \dots + \frac{\sum X_k^2}{n_k} \right] - \frac{(G)^2}{N}$$

Error sum of squares (SSE) = TSS – SST

Mean sum of squares of treatments (MST) = $\frac{SST}{k-1}$

Mean sum of squares of errors (MSE) = $\frac{SSE}{N-k}$

Calculated F value = $\frac{MST}{MSE}$

Table F value = $F_{(k-1), (N-k)}$

Analysis of Variance Table

| <i>Source</i> | <i>Degrees of freedom</i> | <i>Sum of squares</i> | <i>Mean sum of squares</i> | <i>F-calculated</i> | <i>F-table</i> |
|---------------|---------------------------|-----------------------|----------------------------|-----------------------|--------------------|
| Treatments | $k - 1$ | SST | $MST = \frac{SST}{k - 1}$ | $F = \frac{MST}{MSE}$ | $F_{(k-1), (N-K)}$ |
| Error | $N - k$ | SSE | $MSE = \frac{SSE}{N - k}$ | | |
| Total | $N - 1$ | TSS | | | |

At $\alpha\%$ level of significance if calculated value \leq table value then we accept H_0 .

If calculated value $>$ table value then we reject H_0 .

EXAMPLE 13.1: The following data relating to the prices of commodities in different months in five cities.

| <i>A</i> | <i>B</i> | <i>C</i> | <i>D</i> | <i>E</i> |
|----------|----------|----------|----------|----------|
| 29 | 31 | 26 | 19 | 28 |
| 24 | 35 | 28 | 21 | 28 |
| 27 | 28 | 24 | 23 | 25 |
| 26 | 28 | 25 | 29 | 26 |
| 21 | 22 | 20 | 25 | 20 |

Test whether the difference between mean prices of commodities in cities is significant or not.

Solution: H_0 : There is no significant difference between the mean prices of commodities in five cities.

| | | | | | | |
|-------------|----|----|----|----|----|-----|
| <i>A</i> | 29 | 24 | 27 | 26 | 21 | 127 |
| <i>B</i> | 31 | 35 | 28 | 28 | 22 | 144 |
| <i>C</i> | 26 | 28 | 24 | 25 | 20 | 123 |
| <i>D</i> | 19 | 21 | 23 | 29 | 25 | 117 |
| <i>E</i> | 28 | 28 | 25 | 26 | 20 | 127 |
| Grand Total | | | | | | 638 |

No. of cities (treatments) (k) = 5

No. of columns (n) = 5

Total values or observation in table (N) = 25

Grand total (G) = 638

$$\text{Correction factor (CF)} = \frac{G^2}{N} = \frac{(683)^2}{25} = 16281.76$$

$$\text{TSS} = \sum \sum X_{ij}^2 - \text{CF}$$

$$\begin{aligned} \text{TSS} &= [(29)^2 + (24)^2 + (27)^2 + \dots + (25)^2 + (26)^2 + (20)^2] - 16281.76 \\ &= 16628 - 16281.76 = 346.24 \end{aligned}$$

$$\text{SST} = \left[\frac{(127)^2}{5} + \frac{(144)^2}{5} + \frac{(123)^2}{5} + \frac{(117)^2}{5} + \frac{(127)^2}{5} \right] - 16281.76 = 80.64$$

$$\text{SSE} = 346.24 - 80.64 = 265.6$$

$$\text{MST} = \frac{80.64}{5-1} = 20.16$$

$$\text{MSE} = \frac{265.6}{25-5} = 13.28$$

$$F = \frac{20.16}{13.28} = 1.518072$$

F table value at 5% level of significance $F_{(5-1), (25-5)} = F_{(4, 20)} = 2.87$

Anova Table

| Source | Degrees of freedom | SS | MSS | F-calculated | F-table value |
|-----------|--------------------|--------|-------|--------------|---------------|
| Treatment | 4 | 80.64 | 20.16 | 1.518072 | 2.87 |
| Error | 20 | 256.6 | 13.28 | | |
| Total | 24 | 346.24 | | | |

Calculated value < table value then we accept H_0 .

We conclude that there is no significant difference between mean price of commodities in five cities.

$$\text{Correction factor (CF)} = \frac{(36.68)^2}{20} = 67.27112$$

$$\text{TSS} = [(1.60)^2 + (1.81)^2 + (1.63)^2 + \dots + (2.40)^2 + (2.50)^2] - 67.27112 = 1.64848$$

$$\text{SST} = \left[\frac{(11.92)^2}{7} + \frac{(10.57)^2}{6} + \frac{(4.09)^2}{3} + \frac{(9.30)^2}{4} \right] - 67.27112 = 1.239544$$

$$\text{SSE} = 1.64848 - 1.239544 = 0.408936$$

$$\text{MST} = \frac{1.239544}{4-1} = 0.413181$$

$$\text{MSE} = \frac{0.413181}{20-4} = 0.0255585$$

$$F = \frac{0.413181}{0.0255585} = 16.1660896$$

F table value at 5% level of significance $F_{(4-1), (20-4)} = F_{3, 16} = 3.24$

Anova Table

| <i>Source</i> | <i>Degrees of freedom</i> | <i>SS</i> | <i>MSS</i> | <i>F-calculated</i> | <i>F-table value</i> |
|---------------|---------------------------|-----------|------------|---------------------|----------------------|
| Treatment | 3 | 1.239544 | 0.413181 | 16.1660896 | 3.24 |
| Error | 16 | 0.408936 | 0.0255585 | | |
| Total | 19 | 1.64848 | | | |

Calculated value > table value then we reject H_0 .

We conclude that all the class effects are not equal.

ANALYSIS OF VARIANCE—TWO-WAY CLASSIFICATION

Let A and B denote two factors. Factor A is divided into k class like A_1, A_2, \dots, A_k and factor B is divided in to n classes B_1, B_2, \dots, B_n .

Let X_{ij} denote the value of j th unit of factor B in the i th class of factor A . These values are arranged in the two-way table.

| A/B | B_1 | B_2 | | B_j | | B_n | Total |
|-------|----------|----------|--|----------|--|----------|---------|
| A_1 | X_{11} | X_{21} | | X_{1j} | | X_{1n} | X_1 |
| A_2 | X_{21} | X_{22} | | X_{2j} | | X_{2n} | X_2 |
| | | | | | | | |
| A_i | X_{i1} | X_{i2} | | X_{ij} | | X_{in} | X_i |
| | | | | | | | |
| A_k | X_{k1} | X_{k2} | | X_{kj} | | X_{kn} | X_k |
| Total | X_1 | X_2 | | X_j | | X_n | $X = G$ |

Mathematical Linear Model

The observations x_{ij} follows mathematical linear model as

$$X_{ij} = \mu + \alpha_i + \beta_j + \epsilon_{ij} \quad \text{where } i = 1, 2, \dots, k; j = 1, 2, \dots, n$$

where μ = General mean effect

α_i = Mean effect of i th class

β_j = Mean effect of j th class

ϵ_{ij} = Error follows $N(0, \sigma^2)$.

Assumptions

1. The sample observation x_{ij} should be independent.
2. Different class effects are additive in nature.
3. ϵ_{ij} follows $N(0, \sigma^2)$
4. There is no interaction effect between two factors A and B .

Testing of Hypothesis

H_{01} : All the class effects of factor A are equal.

H_{02} : All the class effects of factor B are equal.

Steps of Calculation

Calculate total G of all the observations in all samples

$$G = \Sigma x_{11} + \Sigma x_{21} + \Sigma x_{31} + \dots + \Sigma x_{kn}$$

$$\text{Correction factor (CF)} = \frac{(G)^2}{N}$$

$$\text{Calculate total sum of squares (TSS)} = \Sigma \Sigma X_{ij}^2 - \frac{(G)^2}{N}$$

$$\text{Sum of squares of treatments (SST)} = \left[\frac{\Sigma X_{1.}^2}{n_1} + \frac{\Sigma X_{2.}^2}{n_2} + \dots + \frac{\Sigma X_{k.}^2}{n_k} \right] - \frac{(G)^2}{N}$$

$$\text{Sum of squares of blocks (SSB)} = \left[\frac{\Sigma X_{.1}^2}{k_1} + \frac{\Sigma X_{.2}^2}{k_2} + \dots + \frac{\Sigma X_{.n}^2}{k_n} \right] - \frac{(G)^2}{N}$$

$$\text{Error sum of squares (SSE)} = \text{TSS} - \text{SST} - \text{SSB}$$

$$\text{Mean sum of squares of treatments (MST)} = \frac{\text{SST}}{k - 1}$$

$$\text{Mean sum of squares of blocks (MSB)} = \frac{\text{SSB}}{n - 1}$$

$$\text{Mean sum of squares of errors (MSE)} = \frac{\text{SSE}}{(k - 1)(n - 1)}$$

$$\text{Calculated } F \text{ treatment value} = \frac{\text{MST}}{\text{MSE}}$$

$$\text{Table } F \text{ (treatment) value} = F_{(k-1), (k-1)(n-1)}$$

$$\text{Calculated } F \text{ blocks value} = \frac{\text{MSB}}{\text{MSE}}$$

$$\text{Table } F \text{ (blocks) value} = F_{(n-1), (k-1)(n-1)}$$

Analysis of Variance Table

| <i>Source</i> | <i>Degrees of freedom</i> | <i>Sum of squares</i> | <i>Mean sum of squares</i> | <i>F-calculated</i> | <i>F-table value</i> |
|------------------------------|---------------------------|-----------------------|----------------------------|-----------------------|-------------------------|
| Factor <i>A</i> Treatment | $k - 1$ | SST | MST | $F = \frac{MST}{MSE}$ | $F_{(k-1), (k-1)(n-1)}$ |

| | | | | | |
|---------------------------|------------------|-----|-----|-----------------------|-------------------------|
| Factor <i>B</i> Blocks | $n - 1$ | SSB | MSB | $F = \frac{MSB}{MSE}$ | $F_{(n-1), (k-1)(n-1)}$ |
| Error | $(n - 1)(k - 1)$ | SSE | MSE | | |
| Total | $nk - 1$ | TSS | | | |

At $\alpha\%$ level of significance if calculated value \leq table value then we accept H_{01} .

If calculated value $>$ table value then we reject H_{01} .

At $\alpha\%$ level of significance if calculated value \leq table value then we accept H_{02} .

If calculated value $>$ table value then we reject H_{02} .

EXAMPLE 13.3: Two drugs were simultaneously administered to patients to control body temperature. The number of days required to come to normal temperature were recorded and presented. Test the significant difference between doses of two drugs A and B using two-way analysis of variance technique at 5 per cent level of significance.

| <i>Drug A</i> | <i>Drug B</i> | | |
|---------------|---------------|---------------|---------------|
| | <i>50 mg</i> | <i>100 mg</i> | <i>150 mg</i> |
| 50 mg | 10 | 8 | 7 |
| 100 mg | 9 | 6 | 5 |
| 150 mg | 8 | 4 | 3 |
| 200 mg | 7 | 3 | 2 |

Solution: H_{01} : In drug ' A ' all doses are equal in their effectiveness.

H_{02} : In drug ' B ' all doses are equal in their effectiveness.

Here number of drug $A(k) = 4$

Number of drug $B(n) = 3$

| <i>Drug A</i> | <i>Drug B</i> | | | |
|---------------|---------------|---------------|---------------|--------------|
| | <i>50 mg</i> | <i>100 mg</i> | <i>150 mg</i> | <i>Total</i> |
| 50 mg | 10 | 8 | 7 | 25 |
| 100 mg | 9 | 6 | 5 | 20 |
| 150 mg | 8 | 4 | 3 | 15 |
| 200 mg | 7 | 3 | 2 | 12 |
| Total | 34 | 21 | 17 | 72 |

$$\text{Correction factor (CF)} = \frac{(72)^2}{12} = 432$$

$$\text{TSS} = [(10)^2 + (8)^2 + (7)^2 + \dots + (7)^2 + (3)^2 + (2)^2] - 432 = 506 - 432 = 74$$

$$\text{SS Drug A} = \left[\frac{(25)^2}{3} + \frac{(20)^2}{3} + \frac{(15)^2}{3} + \frac{(12)^2}{3} \right] - 432 = 32.6667$$

$$\text{SS drug B} = \left[\frac{(34)^2}{4} + \frac{(21)^2}{4} + \frac{(17)^2}{4} \right] - 432 = 39.5$$

$$\text{ESS} = 74 - 32.6667 - 39.5 = 1.8333$$

$$\text{MSS drug A} = \frac{32.6667}{4-1} = 10.8889$$

$$\text{MSS drug B} = \frac{39.5}{3-1} = 19.75$$

$$\text{MSE} = \frac{1.8333}{(3-1)(4-1)} = 0.30555$$

$$F^A = \frac{10.8889}{0.30555} = 35.63705$$

$$F^B = \frac{19.75}{0.30555} = 64.63754$$

At 5% level of significance the table value $F_{(4-1), (3-1)(4-1)} = F_{3, 6} = 4.76$

$$F_{(3-1), (3-1)(4-1)} = F_{2, 6} = 5.14$$

Analysis of Variance Table

| <i>Source</i> | <i>Degrees of freedom</i> | <i>SS</i> | <i>MSS</i> | <i>F-calculated value</i> | <i>F-table value</i> |
|---------------|---------------------------|-----------|------------|---------------------------|----------------------|
| Drug <i>A</i> | $4 - 1 = 3$ | 32.6667 | 10.8889 | $F^A = 35.63705$ | $F^A = 4.76$ |
| Drug <i>B</i> | $3 - 1 = 2$ | 39.5 | 19.75 | | |
| Error | $(4 - 1)(3 - 1) = 6$ | 1.8333 | | $F^B = 64.63754$ | $F^B = 5.14$ |
| Total | $12 - 1 = 11$ | 74 | | | |

At 5% level of significance if calculated value $>$ table value then we reject H_{01} .

If calculated value $>$ table value then we reject H_{02} .

We conclude that there is significant difference between doses of drug *A* and there is significant difference between doses of drug *B*.

EXAMPLE 13.4: A farmer applied three types of fertilizers on 4 separate plots. The figure of yield per acre are tabulated below.

| <i>Fertilizer</i> | <i>Yield</i> | | | |
|-------------------|--------------|----------|----------|----------|
| | <i>A</i> | <i>B</i> | <i>C</i> | <i>D</i> |
| Nitrogen | 6 | 4 | 8 | 6 |
| Potash | 7 | 6 | 6 | 9 |
| Phosphates | 8 | 5 | 10 | 9 |

Find out if the plots are materially different in fertility, as also, if the three fertilizers make any significant difference in the yield.

Solution: H_{01} : There is no significant difference between the fertilizers plots.

H_{02} : There is no significant difference between the yields.

Here $k = 4$ and $n = 3$

$$nk = 4 \times 3 = 12$$

| <i>Fertilizers plots</i> | <i>Yield</i> | | | | |
|--------------------------|--------------|----------|----------|----------|--------------|
| | <i>A</i> | <i>B</i> | <i>C</i> | <i>D</i> | <i>Total</i> |
| Nitrogen | 6 | 4 | 8 | 6 | 24 |
| Potash | 7 | 6 | 6 | 9 | 28 |
| Phosphates | 8 | 5 | 10 | 9 | 32 |
| Total | 21 | 15 | 24 | 24 | 84 |

$$\text{Correction factor (CF)} = \frac{(84)^2}{12} = 588$$

$$\text{TSS} = [(6)^2 + (4)^2 + (8)^2 + \dots + (10)^2 + (9)^2] - 588 = 624 - 588 = 36$$

$$\text{SSF} = \left[\frac{(21)^2}{3} + \frac{(15)^2}{3} + \frac{(24)^2}{3} + \frac{(24)^2}{3} \right] - 588 = 606 - 588 = 18$$

$$\text{SSY} = \left[\frac{(24)^2}{4} + \frac{(28)^2}{4} + \frac{(32)^2}{4} \right] - 588 = 596 - 588 = 8$$

$$\text{ESS} = 36 - 18 - 8 = 10$$

$$\text{MSF} = \frac{18}{4-1} = 6$$

$$\text{MSY} = \frac{8}{3-1} = 4$$

$$\text{MSE} = \frac{10}{(4-1)(3-1)} = 1.6667$$

$$F\text{-calculated value (Fert)} = \frac{6}{1.6667} = 3.6667$$

$$F\text{-table value} = F_{(4-1), (4-1)(3-1)} = F_{3, 6} = 4.76$$

$$F\text{-calculated value (Yield)} = \frac{4}{1.6667} = 2.4$$

$$F\text{-table value} = F_{(3-1), (4-1)(3-1)} = F_{2, 6} = 5.14$$

Analysis of Variance Table

| Source | Degrees of freedom | SS | MSS | F-calculated | F-table value |
|-----------|----------------------|----|--------|--------------|---------------|
| Treatment | $4 - 1 = 3$ | 18 | 6 | 3.6667 | 4.76 |
| Blocks | $3 - 1 = 2$ | 8 | 4 | | |
| Error | $(4 - 1)(3 - 1) = 6$ | 10 | 1.6667 | 2.4 | 5.14 |
| Total | $12 - 1 = 11$ | 36 | | | |

At 5% level of significance if calculated value < table value then we accept H_{01} .

if calculated value < table value then we accept H_{02} .

We conclude that the plots are equally effective and the fertilizers have the same effect.

END