

ONE WAY ANALYSIS OF VARIANCE

Let us suppose that 'N' observations y_{ij} ; ($i=1, 2, \dots, k$, $j=1, 2, \dots, n_i$) of a random variable Y are grouped, on some basis, into k -classes of sizes n_1, n_2, \dots, n_k respectively,
$$N = \sum_{i=1}^k n_i.$$

The linear model for analysis of one way classified data is

$$y_{ij} = \mu + \alpha_i + e_{ij}; \quad \begin{matrix} i=1, 2, \dots, k \\ j=1, 2, \dots, n_i \end{matrix}$$

... (*)

where, y_{ij} = j^{th} observation in the i^{th} class

μ = general effect.

α_i = effect due to the i^{th} class.

e_{ij} = errors $\sim N(0, \sigma_e^2)$ independently.

Also
$$\sum_{i=1}^k n_i \alpha_i = 0$$

Estimation of the parameters:

(2)

μ and α_i are obtained by the method of least squares by minimizing $E = \sum_i \sum_j e_{ij}^2$. The ~~estimates~~

are two normal equations

$$\frac{\partial E}{\partial \mu} = 0 \quad \dots (i)$$

$$\text{and } \frac{\partial E}{\partial \alpha_i} = 0 \quad \dots (ii)$$

$$(i) \text{ gives } \hat{\mu} = \bar{y}_{..}$$

$$\text{and } (ii) \text{ gives } \hat{\alpha}_i = \bar{y}_{i.} - \bar{y}_{..}$$

$$\text{where, } \bar{y}_{..} = \frac{1}{N} \sum_i \sum_j y_{ij} = \text{Grand mean}$$

$$\text{and } \bar{y}_{i.} = \frac{1}{n_i} \sum_j y_{ij} = \text{mean of the } i^{\text{th}} \text{ class}$$

Partitioning Total Sum of Squares

Putting the value of $\hat{\mu}$, $\hat{\alpha}_i$ and \hat{e}_{ij} in the model (*)

We get. (3)

$$y_{ij} = \bar{y}_{..} + (\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})$$

Now,

$$(y_{ij} - \bar{y}_{..}) = (\bar{y}_{i.} - \bar{y}_{..}) + (y_{ij} - \bar{y}_{i.})$$

Squaring both sides and taking summation over 'i' and 'j' we get.

$$\sum_i \sum_j (y_{ij} - \bar{y}_{..})^2 = \sum_i n_i (\bar{y}_{i.} - \bar{y}_{..})^2 + \sum_i \sum_j (y_{ij} - \bar{y}_{i.})^2$$

[Product term vanishes due to sum of deviation from mean is zero]

∴ Total Sum of Squares (TSS) = Sum of Squares due to Class (SSC) + Sum of Squares due to Error (SSE)

Partitioning the degrees of freedom: (4)

The total d.f. is also partitioned as

$$\text{Total d.f.} = \text{Class d.f.} + \text{Error d.f.}$$

i.e. $(N-1) = (K-1) + (N-K)$

Null Hypothesis:

$H_0: \alpha_1 = \alpha_2 = \dots = \alpha_K = 0$ i.e. there is no significant difference between the class means.

Alternative Hypothesis:

At least two of the class means are different from each other.

Mean Squares:

$$\text{Mean Square due to class (MSC)} = \frac{SSC}{K-1}$$

$$\text{Mean Square due to error (MSE)} = \frac{SSE}{N-K}$$

Test statistic:

$$F = \frac{MSC}{MSE} \sim F_{K-1, N-K} \text{ will give}$$

us a test statistic for testing the null hypothesis. H_0 will be rejected at $\alpha\%$ level of significance if Calculated F value is greater than the tabulated value of F for $(K-1, N-K)$ d-f. at $\alpha\%$ level i.e. $F \geq F_{\alpha, K-1, N-K}$. otherwise H_0 may be accepted.

ANOVA TABLE

Source of Variation	d-f	SS	M-S	F
Between Classes	$K-1$	SS_B	MSC	$F = \frac{MSC}{MSE}$
Error	$N-K$	SSE	MSE	
Total	$N-1$	TSS	—	—