

13.2 : Analysis of variance : testing for the equality of K population means.

Analysis of variance can be used to test for the equality of K population means. The general form of the hypotheses tested is

$$H_0 : \mu_1 = \mu_2 = \mu_3 = \dots = \mu_K$$

$$H_1 : \text{Not all population means are equal.}$$

where

$$\mu_j = \text{mean of the } j\text{th population.}$$

* We assume that a simple random sample of size n_j has been selected from each of the K populations or treatments. For the resulting sample data, let

$$x_{ij} = \text{value of observation } i \text{ for treatment } j.$$

$$n_j = \text{number of observations for treatment } j.$$

$$\bar{x}_j = \text{sample mean for treatment } j.$$

$$s_j^2 = \text{sample variance for treatment } j.$$

$$s_j = \text{sample standard deviation for treatment } j.$$

→ Testing for the equality of K population means sample mean for Treatment j :

$$\bar{x}_j = \frac{\sum_{i=1}^{n_j} x_{ij}}{n_j}$$

→ sample variance for Treatment j :

$$s_j^2 = \frac{\sum_{i=1}^{n_j} (x_{ij} - \bar{x}_j)^2}{n_j - 1}$$

• The overall sample mean denoted \bar{x} is the sum of all the observations divided by the total number of observations, that is →

→ Over sample mean :

$$\bar{X} = \frac{\sum_{j=1}^k \sum_{i=1}^{n_j} x_{ij}}{n_T}$$

Where $n_T = n_1 + n_2 + \dots + n_K$

→ If the size of each sample is n , $n_T = kn$, in this case equation reduces to ?

$$\bar{X} = \frac{\sum_{j=1}^k \bar{x}_j}{k}$$

تجربة
table 13.1

In the words, whenever the sample sizes are equal, the overall sample mean is just the average of the k sample means.

→ Between-treatments estimate of population variance

We introduced the concept of a between-treatments estimate of σ^2 and showed how to compute it when the sample sizes were equal. This estimate of σ^2 is called the mean square due to treatments and is denoted **MSTR**. The general formula for computing MSTR is

$$MSTR = \frac{\sum_{j=1}^k n_j (\bar{x}_j - \bar{x})^2}{k-1}$$

$\sum_{j=1}^k n_j (\bar{x}_j - \bar{x})^2$: sum of squares due to treatments (denoted SSTR)

$k-1$: degrees of freedom associated with SSTR.

→ Mean square due to treatments

$$MSTR = \frac{SSTR}{k-1}$$

$$\text{where } SSTR = \sum_{j=1}^k n_j (\bar{x}_j - \bar{x})^2$$

Note:

If H_0 is true, MSTR provides an unbiased estimate of σ^2 .

If the means of k populations are not equal, MSTR is not an unbiased estimate of σ^2 .

→ Within-treatments estimate of population variance.

We introduced the concept of a within-treatments estimate of σ^2 and showed how to compute it when the sample sizes were equal. This estimate of σ^2 is called the mean square due to error and is denoted MSE. The general formula for computing MSE is

$$MSE = \frac{\sum_{j=1}^k (n_j - 1) S_j^2}{n_T - k}$$

$\sum_{j=1}^k (n_j - 1) S_j^2$: sum of squares due to error and is denoted SSE.

$n_T - k$: degrees of freedom associated with SSE.

→ Mean square due to error:

$$MSE = \frac{SSE}{n_T - k} \quad \text{where } SSE = \sum_{j=1}^k (n_j - 1) S_j^2$$

Note: MSE is based on the variation within each of the treatments; it is not influenced by whether the null hypothesis is true. Thus, MSE always provides an unbiased estimate of σ^2 .

→ Comparing the variance estimates : The F test.

- Test statistic for the equality of K population means

$$F = \frac{MSTR}{MSE}$$

The test statistic follows an F distribution with $K-1$ degrees of freedom in the numerator and $n_T - K$ degrees of freedom in the denominator.

$$df \text{ of } MSTR = K-1$$

$$df \text{ of } MSE = n_T - K$$

2.15 90% upper tail test

→ Rejection Rule :

- Reject H_0 if $p\text{-value} \leq \alpha$ (p-value approach).
- Reject H_0 if $F \geq F_\alpha$ (critical value approach).

By F-table

→ ANOVA table :

source of variation	df	sum of squares	mean square	F
Treatment	$K-1$	SSTR	$MSTR = \frac{SSTR}{df}$	F statistic = $\frac{MSTR}{MSE}$
Error	$n_T - K$	SSE	$MSE = \frac{SSE}{df}$	
Total	df associative with SST	SST		

★ ANOVA table :

→ Total sums of squares

$$SST = \sum_{j=1}^k \sum_{i=1}^n (x_{ij} - \bar{x})^2$$

→ partitioning of sum of squares

$$SST = SSTR + SSE$$

→ ANOVA table of exp :

Source of var.	df	SS	ME	F
Treatments	2	516	258	9
Error	15	430	28.67	
Total	17	946		

Reject if p-value < α \Rightarrow upper test table by ANOVA
and reject if $F > F_{\alpha}$.

→ p-value < 0.01 so reject H_0 ($\alpha = 0.05$)
Not all μ_j are equal ($\alpha = 0.05$).

By F-table

→ $F_{\alpha} = F_{0.05} = 3.68$
 $F > F_{\alpha}$ so reject H_0 ($\alpha = 0.05$)
Not all μ_j are equal ($\alpha = 0.05$).