

Chapter 13

• One-way ANOVA. Assumptions: k samples of sizes n_1, n_2, \dots, n_k (from k populations) are independent and normally distributed with means $\mu_1, \mu_2, \dots, \mu_k$ and common variance σ^2 .

Hypothesis

$$H_0 : \mu_1 = \mu_2 = \dots = \mu_k$$

$$H_1 : \mu_i \neq \mu_j \text{ for at least one pair } (i, j)$$

• Model. Data

$$Y_{ij} = \mu_i + \epsilon_{ij}, \quad i = 1, \dots, k, \quad j = 1, \dots, n_i, \quad \epsilon_{ij} \sim N(0, \sigma^2).$$

An alternative form

$$Y_{ij} = \mu + \alpha_i + \epsilon_{ij}, \quad \mu = \frac{1}{k} \sum_{i=1}^k \mu_i$$

(α_i is called the effect of the i -th treatment)

$$H_0 : \alpha_1 = \alpha_2 = \dots = \alpha_k = 0$$

$$H_1 : \alpha_i \neq 0 \text{ for at least one } i$$

• Means and sums of squares:

$$\bar{Y}_{i.} = \frac{1}{n_i} \sum_{j=1}^{n_i} Y_{ij}, \quad \bar{Y}_{..} = \frac{1}{N} \sum_{i=1}^k n_i \bar{Y}_{i.} \quad \left(N = \sum_{i=1}^k n_i \right),$$

$$SST = \sum_{i=1}^k \sum_{j=1}^{n_i} (Y_{ij} - \bar{Y}_{..})^2$$

(total sum of squares),

$$SSA = \sum_{i=1}^k \sum_{j=1}^{n_i} (\bar{Y}_{i.} - \bar{Y}_{..})^2$$

(treatment sum of squares),

$$SSE = \sum_{i=1}^k \sum_{j=1}^{n_i} (Y_{ij} - \bar{Y}_{i.})^2$$

(error sum of squares).

• Properties:

- 1) $SST = SSA + SSE$,
- 2) SSA and SSE are independent,
- 3) $SSE/\sigma^2 \sim \chi_{N-k}^2$,
- 4) $SSA/\sigma^2 \sim \chi_{k-1}^2$ under H_0 .

• Test statistic

$$F = \frac{SSA/(k-1)}{SSE/(N-k)}.$$

Under H_0 F has the Fisher distribution with $k - 1$ and $N - k$ degrees of freedom.
 Test: if $F \geq f_{\alpha, k-1, N-k}$, then H_0 is rejected.

- Bartlett's test (testing equality of the variances). Sample variances:

$$S_i^2 = \frac{1}{n_i - 1} \sum_{j=1}^{n_i} (Y_{ij} - \bar{Y}_i)^2$$

Pooled variance

$$S^2 = \frac{1}{N - k} \sum_{i=1}^k (n_i - 1) S_i^2.$$

Test statistic

$$B = \frac{[(S_1^2)^{n_1-1} (S_2^2)^{n_2-1} \dots (S_k^2)^{n_k-1}]^{1/(N-k)}}{S^2}.$$

If $n_1 = n_2 = \dots = n_k = n$, then H_0 is rejected if

$$B \leq b_k(\alpha; n)$$

where $b_k(\alpha; n)$ is taken from the table of the Bartlett distribution. If n_i are different, then the critical value is approximative (see the book).

- A contrast

$$w = \sum_{i=1}^k c_i \mu_i$$

where $\sum_{i=1}^k c_i = 0$. Many hypotheses about μ -s can be formulated as

$$H_0 : w = 0 \quad H_1 : w \neq 0.$$

Test: if

$$\frac{(\sum_{i=1}^k c_i \bar{Y}_i)^2}{S_p^2 \sum_{i=1}^k (c_i^2/n_i)} \geq f_{\alpha, 1, N-k},$$

then H_0 is rejected.