

Summary

If the assumptions

- ▶ independence (holds by design in a CRD)
- ▶ Normal populations
- ▶ equal variances

hold, use

ANOVA F-test

If the assumptions **do not hold**

- ▶ use **Randomization/Permutation test**
- ▶ use **Non-parametric test** (see Section 3)

1.3 Multiple Comparison of Means

If the ANOVA F-test null hypothesis

$$H_0 : \mu_1 = \cdots = \mu_k$$

is **rejected**, then it is of interest to discover which of the means are different. For k groups, there are $c = k(k - 1)/2$ pairs of group means that can be compared.

Consider a “family” of hypothesis tests - a collection of tests of different hypotheses carried out independently on different data sets. For each test in the family, we consider testing the hypothesis at significance level α .

Notation

Label the tests $i = 1, \dots, c$, and for each i , label

- ▶ the null hypotheses H_{0i}
- ▶ the test statistics T_i
- ▶ the rejection regions \mathcal{R}_i

that are potentially different for each i .

We specify for each i ,

$$\alpha = P[T_i \in \mathcal{R}_i | H_{0i} \text{ is } \mathbf{TRUE}]$$

which implicitly defines \mathcal{R}_i . Note that α is the

“Test Type-I Error Rate” or “Comparisonwise Error Rate”

Now consider the results of all tests in the family; what is the “Familywise” Type-I error rate ?

Using the laws of probability

$$P[T_i \in \mathcal{R}_i | H_{0i} \text{ is } \mathbf{TRUE}] = \alpha$$

means that

$$P[T_i \notin \mathcal{R}_i | H_{0i} \text{ is } \mathbf{TRUE}] = 1 - \alpha$$

giving the probability that the test **does not reject** H_{0i} , if H_{0i} is in fact true, is $1 - \alpha$.

Now we consider all tests together;

$$P[\text{Each } T_i \notin \mathcal{R}_i | \text{Each } H_{0i} \text{ is } \mathbf{TRUE}] = (1 - \alpha)^c$$

This is the probability that each test results in the null hypothesis **not** being rejected, that is, the probability that we **never** commit a Type-I error.

Therefore the probability of **at least one** Type-I error is

$$\alpha_F = 1 - (1 - \alpha)^c$$

α_F is the **Familywise Error Rate**.

	$\alpha = 0.05$	$\alpha = 0.01$
c	α_F	α_F
5	0.226	0.049
10	0.401	0.096
50	0.923	0.395
100	0.994	0.634

Therefore, whenever we carry out a “family” of tests, we should not use the traditional choices of $\alpha = 0.05$ or 0.01 on each test.

To fix $\alpha_F = 0.05$, say, we need to use α on each test where

$$\alpha_F = 1 - (1 - \alpha)^c \iff \alpha = 1 - (1 - \alpha_F)^{1/c}$$

For example, if $\alpha_F = 0.05$ and $c = 10$, use

$$\alpha = 1 - (1 - 0.05)^{1/10} = 0.0051$$

Bonferroni Method

It can be shown that

$$1 - (1 - \alpha)^c \approx c\alpha$$

Therefore, if α_F is the familywise error rate, we can set

$$\alpha = \frac{\alpha_F}{c}$$

to get the comparisonwise error rate.

α_F/c is known as the Bonferroni Correction.