AID-521 Mathematics for Data Science

Module: Statistics | Lecture: 6

HYP. TESTING FOR MULTIPLE SAMPLES

testing two samples, analysis of variance (ANOVA)

Hypothesis Test for Two Samples

- → Comparing the means and variances of two populations
- \rightarrow Let $X_{11},...,X_{1n_1}$ be a random sample from population 1 with mean μ_1 and variance σ_1^2 , and $X_{21},...,X_{2n_2}$ be a random sample from population 2 with mean μ_2 and variance σ_2^2 .
- \rightarrow Here, we study **for the case when** samples are independent and $n_1, n_2 \ge 30$.



Hypothesis Test for Two Samples

SUMMARY OF HYPOTHESIS TEST FOR $\mu_1 - \mu_2$ FOR LARGE SAMPLES ($n_1 \& n_2 \ge 30$)

To test

$$H_0: \mu_1 - \mu_2 = D_0$$

versus

$$H_a: \begin{cases} \mu_1 - \mu_2 > D_0, & \text{upper tailed test} \\ \mu_1 - \mu_2 < D_0, & \text{lower tailed test} \\ \mu_1 - \mu_2 \neq D_0, & \text{two-tailed test}. \end{cases}$$

The test statistic is

$$Z = \frac{\overline{X}_1 - \overline{X}_2 - D_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}.$$

Replace σ_i by S_i , if σ_i , i = 1,2 are not known. Rejection region is

$$RR: \begin{cases} z > z_{lpha}, & ext{upper tail RR} \\ z < -z_{lpha}, & ext{lower tail RR} \\ |z| > z_{lpha/2}, & ext{two tail RR}, \end{cases}$$

where z is the observed test statistic given by

$$z = \frac{\overline{x}_1 - \overline{x}_2 - D_0}{\sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}}.$$



Introduction to ANOVA

- → Tests to analyze data from more than two populations.
- → The hypothesis that the population means are equal is considered equivalent to the hypothesis that there is no difference in treatment effects (as in experiments).
- → Can be considered as an extension of the test of hypothesis for the equality of two means.

Introduction to ANOVA

- → Assume 4 populations. Why do we need a new method to test for differences among these 4 population means?
- → Can't we use z- or t-tests for all possible pairs and test for differences in each pair?
 - → If any one of these tests leads to the rejection of the hypothesis of equal means, then we might conclude that at least two of the four population means differ.
- → Actual Type I error becomes amplified than what we might think!
 - \rightarrow For $\binom{4}{2} = 6$ tests, let $\alpha = 0.10$ be the significance level.
 - \rightarrow Probability that at least one of the six tests leads to the conclusion that there is a difference leads to an error 1– $(0.9)^6=0.46856$.
 - $\,\rightarrow\,\,$ Hence, one is likely to declare significance when there is none.



Introduction to ANOVA -- Common Terms Used

- \rightarrow *Total SS* = total sums of squares of values
- \rightarrow SST = sum of squares for treatment
- \rightarrow SSE = Total SS SST = sum of squares of errors
- $\rightarrow MSE = \frac{SSE}{N-k} = \text{mean square error}$
- $\rightarrow MST = \frac{SSE}{N-k} = \text{mean square treatment}$



Int. to ANOVA -- Important Result in Use

If χ_1^2 and χ_2^2 are independent, and have ν_1 and ν_2 d.o.f.s respectively, then

$$F = \frac{\chi_1^2/\nu_1}{\chi_2^2/\nu_2}$$

has a F-distribution with ν_1 numerator d.o.f. and ν_2 denominator d.o.f.



ANOVA for Two Treatments

- → The simplest form of the analysis of variance procedure, the case of studying the means of two populations I and II.
- → For comparing only two means, the ANOVA will result in the same conclusions as the t-test for independent random samples.
- → This section will help to introduce the concept of ANOVA in simpler terms.



ANOVA for Two Treatments

ANALYSIS OF VARIANCE PROCEDURE FOR TWO TREATMENTS

For equal sample sizes $n=n_1=n_2$, assume $\sigma_1^2=\sigma_2^2$. We test

$$H_0: \mu_1 = \mu_2 \text{ vs. } H_a: \mu_1 \neq \mu_2.$$

1. Calculate: $\overline{y_1}$, $\overline{y_2}$, $\sum_{ij} y_{ij}^2$, $\sum_{ij} y_{ij}$, and find

$$SST = \sum_{i=1}^{2} n_i \left(\overline{y}_i - \overline{y} \right)^2.$$

Also calculate

Total SS =
$$\sum_{i} \sum_{j} y_{ij}^{2} - \frac{\left(\sum_{i} \sum_{j} y_{ij}\right)^{2}}{n_{1} + n_{2}}$$
.

Then

$$SSE = Total SS - SST.$$



ANOVA for Two Treatments

2. Compute

$$MST = \frac{SST}{1}$$

$$MSE = \frac{SSE}{n_1 + n_2 - 2}.$$

3. Compute the test statistic,

$$F = \frac{MST}{MSE}.$$

4. For a given α , find the rejection region as

$$RR: F > F_{\alpha}$$

based on 1 numerator and $(n_1 + n_2 - 2)$ denominator degrees of freedom.

5. Conclusion: If the test statistic F falls in the rejection region, conclude that the sample evidence supports the alternative hypothesis that the means are indeed different for the two treatments.
Assumptions: Populations are normal with equal but unknown variances.

- Hypothesis testing problem of comparing population means of more than two independent populations
- → Data are about several independent groups
- \rightarrow Let $\mu_1, ..., \mu_k$ be the means of k <u>normal</u> populations with unknown but equal variance σ^2 .
- → Are the means of these groups are different, or are all equal?
- ightarrow Overall variability: (1) between-groups, (2) within-groups
- If between groups is much larger than that within groups, this will indicate that differences between the groups are real, not merely due to the random nature of sampling.

- \rightarrow Let independent samples be drawn of sizes n_i , i = 1, 2, ..., k, and let $N = n_1 + \cdots + n_k$.
- → Let y_{ij} be the measured response on the jth experimental unit in the ith sample. That is, Y_{ij} is the jth observation from population i, i = 1, 2, ..., k, and $j = 1, 2, ..., n_i$.
- \rightarrow Let \bar{y} be the overall mean of all observations.
- → The problem can be formulated as a hypothesis testing problem, where we need to test

$$H_0: \mu_1 = \mu_2 = \cdots = \mu_k$$
 vs. $H_1:$ Not all μ_i s are equal.

1. Compute

$$\begin{split} T_{I} &= \sum_{j=1}^{n_{j}} y_{ij}, T = \sum_{i=1}^{k} \sum_{j=1}^{n_{j}} y_{ij}, \text{ and } \sum_{i=1}^{k} \sum_{j=1}^{n_{j}} y_{ij}^{2}, \\ CM &= \frac{\left(\sum_{i=1}^{k} \sum_{j=1}^{n_{j}} y_{ij}\right)^{2}}{N} = \frac{T^{2}}{N}, \text{ where } N = \sum_{i=1}^{k} n_{i}, \\ \overline{T_{I}} &= \overline{T_{I_{I}}}, \end{split}$$

and

Total SS =
$$\sum_{i=1}^{k} \sum_{j=1}^{n_i} y_{ij}^2 - CM$$
.

2. Compute the sum of squares between samples (treatments),

$$SST = \sum_{i=1}^{k} \frac{T_i^2}{n_i} - CM$$
$$= \sum_{i=1}^{k} \overline{T_i} - CM.$$

and the sum of squares within samples,

$$SSE = Total SS - SST$$

Let

$$MST = \frac{SST}{k-1},$$

and

$$MSE = \frac{SSE}{n-k}.$$

3. Compute the test statistic:

$$F = \frac{MST}{MSE}.$$

4. For a given α , find the rejection region as

$$RR: F > F_{\alpha}$$

with $v_1 = (k-1)$ numerator degrees of freedom and $v_2 = \left(\sum_{i=1}^k n_i\right) - k = N-k$ denominator degrees of freedom, where $N = \sum_{i=1}^k n_i$.

5. Conclusion: If the test statistic F falls in the rejection region, conclude that the sample evidence supports the alternative hypothesis that the means are indeed different for the k treatments and are not all equal.

Assumptions: The samples are randomly selected from the k populations in an independent manner. The populations are assumed to be normally distributed with equal variances σ^2 and means μ_1,\ldots,μ_k .