

## Ch. 28 - Comparing Several Means

Use  $t$ -tools? NO!

→ Reason? *Compound uncertainty*

- In any test, there is uncertainty such that we reject  $H_0$  when it's true, or Type I error. By comparing multiple means and using ONE  $t$ -test for each pair, the "overall" Type I error will compound.
- For example, consider 3 means that are equal and each  $t$ -test uses  $\alpha = 0.05$ . Thus, there's a 5% chance to show a difference when there isn't (recall  $H_0$  assumes no diff.). The chance of detecting *at least* one difference among the three means is roughly  $1 - 0.95^3 = 0.143$  or 14.3% when the means are EQUAL! (Note: 14.3% is the "overall" Type I error.)
- For 5 means, the "overall" Type I error becomes approximately 40%.

Def'n: ANalysis Of VAriance (ANOVA) is a procedure to test the equality of three or more population means. NOTE: the name of the test refers to comparing different sources of variability; it WILL test differences among means.

Test requires the following assumptions:

1. The populations are all normally distributed.
2. The populations all have the same standard deviation.
3. The samples from different populations are random and independent.

*Checking Assumptions:*

- Assumption #1 is checked with histograms/boxplots for each group.
- Assumption #2 is more critical but harder to assess. Still, we can use side-by-side boxplots (or the informal rule from Ch. 24).
- Assumption #3 by analyzing the experiment design.

*Notation:*

- $y_{ij}$  = observation for  $i^{\text{th}}$  subject in  $j^{\text{th}}$  group
- $j = 1, \dots, k$  indexes groups
- $i = 1, \dots, n_j$  indexes subjects within groups
- $n_j$  = # of observations in  $j^{\text{th}}$  group;  $N = \sum_j n_j$  = total # of observations
- $\bar{y}_j$  and  $s_j^2$  are sample mean and variance for the  $j^{\text{th}}$  group
- $\bar{\bar{y}}$  = grand mean = mean for combined sample:

$$\bar{\bar{y}} = \frac{1}{N} \sum_j \sum_i y_{ij} = \frac{1}{N} \sum_j n_j \bar{y}_j$$

*Statistical model, parameters, hypotheses:*

Each observation can be represented by

$$Y_{ij} = \mu + \tau_j + \varepsilon_{ij}$$

where  $Y_{ij}$  are independent random observations,  $\mu$  is the *overall mean*,  $\tau_j$  is a parameter associated with the  $j^{\text{th}}$  group called the  *$j^{\text{th}}$  treatment effect*, and  $\varepsilon_{ij}$  is a random error.

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

$H_A$ : the  $\mu_j$  are not all equal

(OR, at least 2  $\mu_j$  are different from each other; OR at least 1  $\mu_j$  is different)

*ANOVA F test statistic:*

For sources of variability in the model above, the *ANOVA identity* is

$$\sum_{j=1}^k \sum_{i=1}^{n_j} (y_{ij} - \bar{y})^2 = \sum_{j=1}^k \sum_{i=1}^{n_j} (\bar{y}_j - \bar{y})^2 + \sum_{j=1}^k \sum_{i=1}^{n_j} (y_{ij} - \bar{y}_j)^2$$

$$SS_Y = SS_T + SS_E$$

where  $SS_Y$  is the total sum of squares,  $SS_T$  is the treatment sum of squares and  $SS_E$  is the error sum of squares. Also, the presence of “squares” suggest a ratio test. Thus, for the above  $H_0$ , we have

$$SS_T = \sum_j n_j (\bar{y}_j - \bar{y})^2 \quad (\text{variability between samples})$$

$$SS_E = \sum_j (n_j - 1) s_j^2 \quad (\text{variability within samples})$$

$$F_0 = \frac{MS_T}{MS_E} = \frac{SS_T / (k - 1)}{SS_E / (N - k)}$$

Under  $H_0$ , the  $F_0$  test statistic has an  $F$ -distribution with  $df_1 = k - 1$  and  $df_2 = N - k$ .

Def'n: The  $F$  distribution has the following properties:

1. It is continuous and skewed to the right.
2. It has two parameters:  $df$  for the numerator and  $df$  for the denominator.
3. The units of an  $F$  distribution are nonnegative.

Reject  $H_0$  when  $F_0$  is large (greater than 10 works for most values of  $\alpha$ )

- Rationale:

- If  $H_0$  is true, then both types of variability are identical.  $\rightarrow F_0 \approx 1$
- If  $H_A$  is true, then  $MS_T$  should be larger.  $\rightarrow F_0 > 1$

*ANOVA Table:*

Source	df	SS	MS	F	P-value
Between	$k - 1$	$SS_T$	$MS_T$	$MS_T / MS_E$	?
Within	$N - k$	$SS_E$	$MS_E$		
<b>Total</b>	<b><math>N - 1</math></b>	<b><math>SS_Y</math></b>			

Note that  $SS_Y = SS_T + SS_E$ ;  $df(\text{total}) = df(\text{Between}) + df(\text{Within})$ ;  $MS = SS/df$

Calculating SS is tedious! More important to understand values and how they relate to other values in the ANOVA table. If you received an incomplete table, you should be able to fill it in.

Ex28.1) Consider a  $k$ -mean problem. Five observations are gathered for each group. Use the given information in the table below to answer the questions. Assume all populations are normal with some common variance.

(a) What is  $k$ ?

(b) What is the test statistic for the test to determine if any of the  $k$  groups are different?

Source	df	SS	MS	$F$	$P$ -value
Between					
Within			6		
<b>Total</b>	<b>54</b>	<b>360</b>			

(filled out in class)

(Additional example shown in class with full hypothesis test.)

#### Summary

- check assumptions.
- rejecting  $H_0$  does NOT mean all means are different, AT LEAST ONE is.
- not rejecting  $H_0$  finishes the analysis, rejecting  $H_0$  requires subsequent determination of which means are significantly different from the rest.