MATH 4720 / MSCS 5720

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Chapter 9



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MULTIPLE COMPARISON (POST HOC ANOVA)



In the ANOVA we test

$$H_0$$
: $\mu_1 = \mu_2 = \cdots = \mu_t$
 H_a : $\mu_i \neq \mu_j$ for some pairs (i, j)

- If we reject H_0 in favor of H_a , then the question arises that for what pairs $\mu_i \neq \mu_j$.
- Now, we want to test

$$H_0^{ij}: \mu_i = \mu_j$$
$$H_a^{ij}: \mu_i \neq \mu_j$$

for all pairs using, say, two-sample t-statistics.

MULTIPLE COMPARISON CONT'D



$$t_{ij} = \frac{\bar{y}_i - \bar{y}_j}{s_p \sqrt{\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}} \text{ for all pairs } (i, j)$$

- We will be testing many hypotheses. For example, if there are 5 treatments, there will be 10 hypotheses.
- If each hypothesis is tested at $\alpha=0.05$, there is a good chance that one hypothesis will be falsely rejected.
- For illustration, suppose you are testing 100 hypothesis each at $\alpha=0.05$, and suppose null H_0 is true for all hypotheses, then 5 (i.e. 5%) hypothesis will be falsely rejected.
- This problem always arises due to multiplicity of hypotheses.

MULTIPLE COMPARISON CONT'D



- The problem of multiplicity is serious when you are testing many hypotheses, say thousands of hypotheses.
- Example: Gene Expression Analysis
- A data is collected on gene expressions on 10,000 genes. This kind of research is done if you are trying to find genes responsible for cancer. If those genes are correctly detected, you can study their structure and come up with a cure.

•
$$H_0^i: \mu_i = 0$$
 vs. $H_a^i: \mu_i \neq 0$, $i = 1, 2, ..., 10000$

• If we test each of the hypotheses at $\alpha=0.05$, then 5% of them (500 hypotheses) will be falsely discovered even if all nulls are true.

FAMILYWISE ERROR RATE (FEW)



 Instead of using Type-I error rate, we should use a different type of error rate

- α_F = P(Falsely reject at least one hypotheses)
- If there are m hypotheses each tested at α_I , then approximately

• $\alpha_F \leq 1 - (1 - \alpha_I)^m$

• Or we can say: $\alpha_F \leq m * \alpha_I$

- This implies that
- if $\alpha_I = \frac{0.05}{m}$, then $\alpha_F \leq 0.05$.

TABLE 9.5 A comparison of the familywise error rate α_F for m independent	m, Number of	α _I Probability of a Type I Error on an Individual Test		
multiple tests	multiple tests	.10	.05	.0
	1	.100	.050	.0
	2.	190	097	0

10

.271

.410

.651

.143

.185

.226

.401

.01

.010

.030

.039

.049

.096

BONFERRONI METHOD



- If there are m hypotheses, then we can test individual hypothesis at $\alpha=\frac{0.05}{m}$ using standard method such as t-test. This will guarantee that
- α_F = P(Falsely reject at least one null) ≤ 0.05 .
- Problem with this approach: If m is large, $\alpha=\frac{0.05}{m}$ will be very small, and the chance of rejecting H_0 will be small. In other word, the power of Bonferroni is very poor.

BEYOND BONFERRONI



- In addition to Bonferroni, there are many methods of Multiple Comparisons. Some of them are
 - 1. Fisher's Least Significant Difference (LSD)
 - 2. Tukey's Method
 - 3. Dunnette's Method
- Recall that the purpose is to test

$$\bullet \qquad H_0^{ij}: \mu_i = \mu_j$$

•
$$H_a^{ij}: \mu_i \neq \mu_j$$

- In all of the methods we reject H_0^{ij} if
- $|\bar{y}_i \bar{y}_j| > C$, a constant

BOOK EXAMPLE 9.3



• A study was done to test 5 different agents used to control weeds. Each of these agents were applied to sample of 6 one-acre plots. The hay was harvested and the total yield was recorded.

Agent	1	2	3	4	5
Туре	None	Bio1	Bio2	Chm1	Chm2
$\overline{\mathcal{y}}$	1.175	1.293	1.328	1.415	1.500
S	.1204	.1269	.1196	.1249	.1265
n	6	6	6	6	6

 The question is that if there is a difference between the agents, which agent provides the best yield.

EXAMPLE 9.3: ASSUMPTIONS



First, we check the validity of assumptions among the agents.

• H_0 : $\sigma_1 = \sigma_2 = \sigma_3 = \sigma_4 = \sigma_5$

Test for Equal Variances: yield versus Agent

95% Bonferroni confidence intervals for standard deviations

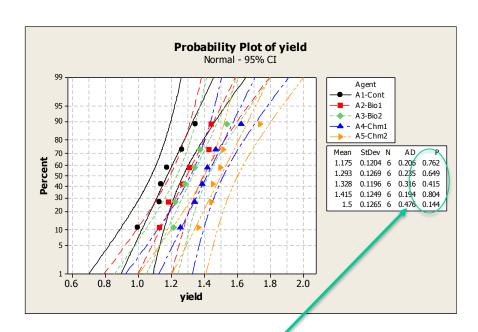
Agent	N	Lower	StDev	Upper
A1-Cont	6	0.0657789	0.120394	0.419543
A2-Bio1	6	0.0693292	0.126892	0.442187
A3-Bio2	6	0.0653430	0.119596	0.416762
A4-Chm1	6	0.0682488	0.124914	0.435296
A5-Chm2	6	0.0691068	0.126485	0.440768

Bartlett's Test (Normal Distribution)
Test statistic = 0.03, p-value = 1.000

Levene's Test (Any Continuous Distribution)
Test statistic = 0.04, p-value = 0.996

• Levene's test p-value is 0.996. Fail to reject equality of the variances

• H_0 : Data is generated from normal distribution for each type of food.



• Large p-values. Fail to reject Normality assumption.

EXAMPLE 9.3: ANOVA



 Next, we perform ANOVA to test if there is a difference among the agents.

•
$$H_0$$
: $\mu_1 = \mu_2 = \cdots = \mu_5$
 H_a : $\mu_i \neq \mu_j$ for some pairs (i, j)

One-way ANOVA: yield versus Agent

• **TS.**
$$F = \frac{SS_B/df_B}{SS_E/df_E} = 5.96$$

•
$$p - value = P(F > 5.96) = 0.0016$$

• Reject H_0 , and conclude that there is a difference among the agents.

POST HOC ANOVA



- In the ANOVA, we only conclude that yields are different under different agents.
- How do we say which agent is the best? Are chemical agents better than the biological agents?
- All of these questions can be answered by all pairwise comparisons.

$$H_0^{ij}: \mu_i = \mu_j$$

$$H_a^{ij}: \mu_i \neq \mu_j$$

There are 10 pairs.

WHAT TO DO?



(a) We need to control

$$\alpha_F = P(\text{false rejections of even one null}) = 0.05$$

- (b) To achieve (a), we might lose power of true discovery.
- We will discuss four methods
 - 1. Bonferroni
 - 2. Fisher's LSD (Least Significant Difference)
 - 3. Tukey's
 - 4. Dunnette's
- Later we will look into the pros and cons for these methods.
 Keep two things in mind.

BONFERRONI'S METHOD



• Use TS.
$$t_{ij}=\frac{\bar{y}_i-\bar{y}_j}{s_p\sqrt{\left(\frac{1}{n_i}+\frac{1}{n_j}\right)}}$$
 with Type-I error of $\frac{\alpha}{m}$.

• Here s_p is the pooled standard deviation that estimates the common standard deviation σ . Here, we can estimate σ by

•
$$\hat{\sigma} = \sqrt{MSE}$$
, where $MSE = \frac{SS_E}{df_E}$

- We say that pair (μ_i, μ_i) are significantly different if
- Formula

$$\left|\bar{y}_i - \bar{y}_j\right| > t_{\alpha/2m} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$$
 (*m* is # of hypotheses)

EXAMPLE 9.3: BONFERRONI'S METHOD



Be The Difference.

 Agent
 1
 2
 3
 4
 5

 Type
 None
 Bio1
 Bio2
 Chm1
 Chm2

 \bar{y} 1.175
 1.293
 1.328
 1.415
 1.500

 s .1204
 .1269
 .1196
 .1249
 .1265

6

6

ANOVA

One-way ANOVA: yield versus Agent						
Source	DF	SS	MS	F	P	
Agent	4	0.3647	0.0912	5.96	0.002	
Error	25	0.3825	0.0153			
Total	29	0.7472				
S = 0.1	237	R-Sa =	48.81%	R-Sa	(adi) =	40.62%

•
$$SS_E = \sum (n_i - 1)s_i^2 = 0.3825$$
, $df_E = 25$

6

•
$$MSE = \frac{SSE}{df_e} = \frac{0.3825}{25} = 0.0153$$

6

6

n

•
$$\frac{\alpha}{2m} = \frac{0.05}{2*10} = 0.0025$$
, $t_{\alpha/2m} = 3.361$

•
$$\mu_i \neq \mu_j$$
 if $|\bar{y}_i - \bar{y}_j| > 3.361 \sqrt{0.0153 \left(\frac{1}{6} + \frac{1}{6}\right)} = 0.24002$

$$\checkmark$$
 Agent 1 vs. 5: |1.500 - 1.175| = 0.325 > 0.24002

$$\checkmark$$
 Agent 1 vs. 4: $|1.415 - 1.175| = 0.240 < 0.24002$

$$\checkmark$$
 Agent 1 vs. 3: |1.328 - 1.175| = 0.153 < 0.24002

$$\checkmark$$
 Agent 1 vs. 2: |1.293 - 1.175| = 0.118 < 0.24002

• Thus, we only find that Chm2(Agent 5) is different from Control(Agent 1).

FISHER'S LSD



• Formula:
$$\left| \overline{y}_i - \overline{y}_j \right| > t_{\alpha/2} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$$

• **Example 9.3**

Agent 1 2 3 4 5
$$\bar{y}$$
 1.175 1.293 1.328 1.415 1.500

•
$$\frac{\alpha}{2} = 0.025$$
, $MSE = 0.0153$, $t_{\alpha/2} = 2.060$, $n_j = n_j = 6$

•
$$|\bar{y}_i - \bar{y}_j| > 2.06 * \sqrt{0.0153 \left(\frac{1}{6} + \frac{1}{6}\right)} = 0.14711$$

Minitab

- \checkmark 1 vs. 5: |1.500 1.175| = 0.325 > 0.14711
- \checkmark 1 vs. 4: |1.415 1.175| = 0.240 > 0.14711
- \checkmark 1 vs. 3: |1.328 1.175| = 0.153 > 0.14711
- \checkmark 1 vs. 2: |1.293 1.175| = 0.118 < 0.14711
- \checkmark 2 vs. 5: |1.500 1.293| = 0.207 > 0.14711
- \checkmark 2 vs. 4: |1.415 1.293| = 0.122 < 0.14711
- \checkmark 2 vs. 3: |1.328 1.293| = 0.035 < 0.14711

Agent 3,4 and 5 are different from control

Agent 2 and 5 are different

Grouping Information Using Fisher Method

Agent N Mean Grouping
A5-Chm2 6 1.5000 A
A4-Chm1 6 1.4150 A B
A3-Bio2 6 1.3280 B
A2-Bio1 6 1.2930 B C
A1-Cont 6 1.1750 C

Means that do not share a letter are significantly different.

TUKEY'S METHOD



• For the common sample sizes, $n_i = n$

Formula:
$$\left|\bar{y}_i - \bar{y}_j\right| > q_{\alpha}(t, df_E) \sqrt{\frac{MSE}{n}}$$

- where $q_{\alpha}(t, df_E)$ is the upper-tail value of the studentized range statistics (Given in Table 10)
- **Example 9.3**,
 - $t = 5, df_E = 25, n = 6,$
 - From Table 10, $q_{0.05}(5, 25) = 4.158$

•
$$q_{\alpha}(t, df_E) \sqrt{\frac{MSE}{n}} = 4.158 * \sqrt{\frac{0.0153}{6}} = 0.20997$$

- \checkmark 1 vs. 5: |1.500 1.175| = 0.325 > 0.20997
- \checkmark 1 vs. 4: |1.415 1.175| = 0.240 > 0.20997
- \checkmark 1 vs. 3: |1.328 1.175| = 0.153 < 0.20997
- \checkmark 1 vs. 2: |1.293 1.175| = 0.118 < 0.20997
- \checkmark 2 vs. 5: |1.500 1.293| = 0.207 < 0.20997
- \checkmark 2 vs. 4: |1.415 1.293| = 0.122 < 0.20997
- \checkmark 2 vs. 3: |1.328 1.293| = 0.035 < 0.20997

Agents 4 and 5 are different from control

No Significant difference between agents 2,3,4 and 5

16

Means that do not share a letter are significantly different.



TUKEY'S METHOD CONT'D

Remark for unequal sample sizes:

• If sample sizes $n_i s$ are not same, then a modified Tukey's test is

$$\left|\bar{y}_i - \bar{y}_j\right| > \frac{q_{\alpha}(t, df_E)}{\sqrt{2}} \sqrt{MSE\left(\frac{1}{n_i} + \frac{1}{n_j}\right)}$$

• This is also called **Tukey-Kramer's Method**.

DUNNETT'S METHOD



- Comparing with a control
- Assume that the sample sizes are same, i.e., $n_i = n$

$$|\bar{y}_i - \bar{y}_c| > d_\alpha (t - 1, df_E) \sqrt{\frac{2MSE}{n}}$$

- Note that this method is only used for comparing with a control.
- d_{α} $(t-1, df_E)$ can be obtained from Table 11.
- For the book example 9.3, Agent 1 is the control.

$$-t=5, df_E=25, n=6,$$

- From Table 11, $d_{0.05}(5-1,24) = 2.28$

•
$$d_{\alpha} (t-1, df_e) \sqrt{\frac{2MSE}{n}} = 2.28 * \sqrt{2 * \frac{0.0153}{6}} = 0.16282$$

DUNNETT'S METHOD: EXAMPLE 9.3



•
$$H_0^i: \mu_i = \mu_c$$

•
$$H_a^i: \mu_i \neq \mu_c$$

• Reject H_0^i if $|\bar{y}_i - \bar{y}_c| > 0.16282$

Agent 1 2 3 4 5
$$\bar{y}$$
 1.175 1.293 1.328 1.415 1.500

```
\checkmark 5 vs. 1: |1.500 - 1.175|=0.325 > 0.16282
```

$$\checkmark$$
 4 vs. 1: $|1.415 - 1.175| = 0.240 > 0.16282$

$$\checkmark$$
 3 vs. 1: $|1.328 - 1.175| = 0.153 < 0.16282$

Only the chemical agents 4 and 5 are different from control. No biological agents are different from control.

Minitab -

Means not labeled with letter A are significantly different from control level mean.

COMPARISON OF BONFERRONI, LSD, TUKEY'S AND DUNNETT'S METHODS.



• For the Book example 9.3, we had $\mu_i \neq \mu_j$ if

• **Bonferroni**:
$$|\bar{y}_i - \bar{y}_i| > 0.24002$$

• Fisher's LSD:
$$|\bar{y}_i - \bar{y}_i| > 0.14711$$

• Tukey's:
$$|\bar{y}_i - \bar{y}_j| > 0.20997$$

• **Dunnett's:**
$$|\bar{y}_i - \bar{y}_i| > 0.16282$$

 Not all method control familywise error rate. Although the power of discovery for Fisher's LSD is better than other methods, you cannot say that

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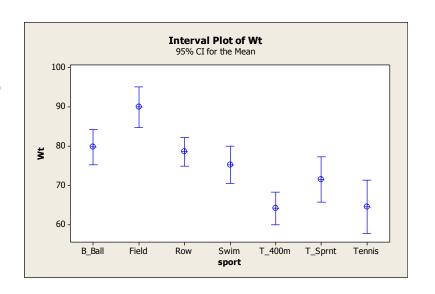
Confidence Interval of Wt for different Sports

Individual Confidence Interval

Does NOT consider the familywise error

$$- \overline{y}_i \pm t_{\alpha/2} * s_p \sqrt{\frac{1}{n_i}}$$

- where $s_p = \frac{\sum_{i=1}^{t} (ni-1)s_i^2}{(\sum_{i=1}^{t} n_i) - t}$



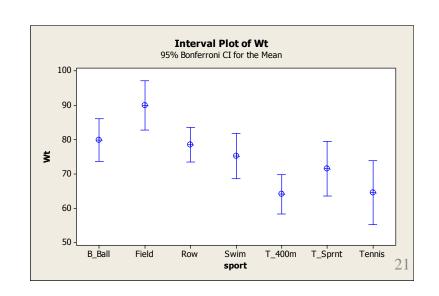
Bonferroni's Method

Conservative

$$- \quad \overline{y}_i \pm t_{\alpha/2m} * s_p \sqrt{\frac{1}{n_i}}$$

- where m=7

Minitab



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Pairwise comparison of Wt for different Sports

Fisher's Method

Does NOT consider the familywise error

Grouping Information Using Fisher Method

sport	N	Mean	Grouping
Field	19	89.97	A
B_Ball	25	79.78	В
Row	37	78.54	В
Swim	22	75.15	ВС
T_Sprnt	15	71.51	C D
Tennis	11	64.47	D E
T_400m	29	64.05	E

Means that do not share a letter are significantly different.

Tukey's Method

Consider the familywise error

Grouping Information Using Tukey Method

sport	N	Mean	Grouping
Field	19	89.97	A
B_Ball	25	79.78	A B
Row	37	78.54	В
Swim	22	75.15	ВС
T_Sprnt	15	71.51	BCD
Tennis	11	64.47	C D
T_400m			

Means that do not share a letter are significantly different.

Bonferroni's Method

Conservative

Grouping Information Using Bonferroni Method

sport	N	Mean	Grouping
Field	19	89.97	A
B_Ball	25	79.78	A B
Row	37	78.54	В
Swim	22	75.15	ВС
T_Sprnt	15	71.51	BCD
Tennis	11	64.47	C D
T 400m	29	64.05	D

Means that do not share a letter are significantly different.

Dunnett's Method

Comparing with a Control group

Grouping Information Using Dunnett Method

Level	N	Mean	Grouping
B_Ball (control)	25	79.78	A
Field	19	89.97	
Row	37	78.54	A
Swim	22	75.15	A
T_Sprnt	15	71.51	A
Tennis	11	64.47	
T_400m	29	64.05	

Means not labeled with letter A are significantly different from control level mean.

CONCLUSION



- Excluding Fisher LSD, the other methods control familywise error rate. However, Bonferroni has the poorest discovery rate. Discovery rate for the Tukey's is better than Bonferroni, but not as good as Dunnett's. However, Dunnett's used only to compare with the control.
- In other words, Bonferroni, Tukey and Dunnett's all have familywise error rate of 0.05. If the objective is to compare only with a control, then Dunnett's is more powerful among three. Otherwise, Tukey's is more powerful than Bonferroni.
- Although Bonferroni is not very powerful, it does have advantage that it can be used in any situation (whether it is one factor or multifactor analyses) whenever there are multiple hypotheses.