

*“We have to go to the deductions and the inferences,”  
said Lestrade, winking at me.*

*“I find it hard enough to tackle facts, Holmes, without  
flying away after theories and fancies.”*

**Inspector Lestrade to Sherlock Holmes**

*The Boscombe Valley Mystery*

# COHEN CHAP 13 MULTIPLE COMPARISON PROCEDURES

For EDUC/PSY 6600

# ANOVA OMNIBUS: SIGNIFICANT F-RATIO

**Factor (IV) had effect on DV**

- **Groups are not from same population**

**Which levels of factor differ?**

**Must compare and contrast means from different levels**

**Indicates  $\geq 1$  significant difference among all POSSIBLE comparisons**

## **Simple vs. complex comparisons**

- **Simple comparisons**
  - **Comparing 2 means, pairwise**
  - **Possible for no 'pair' of group means to significantly differ**
- **Complex comparisons**
  - **Comparing combinations of  $> 2$  means**

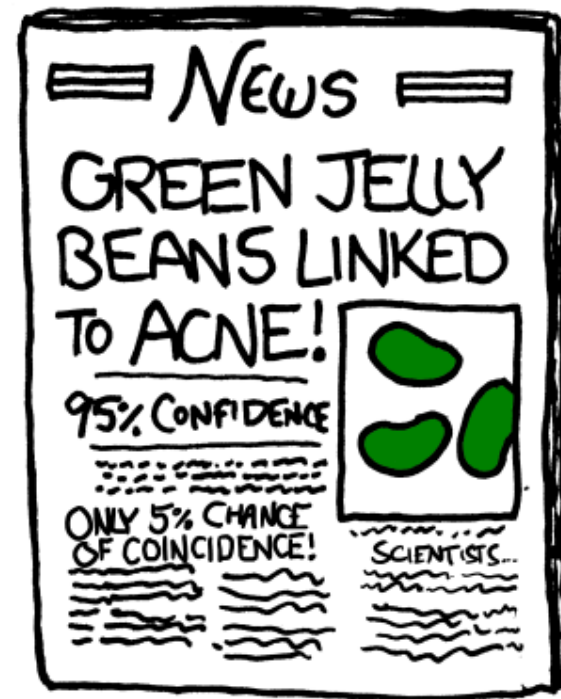
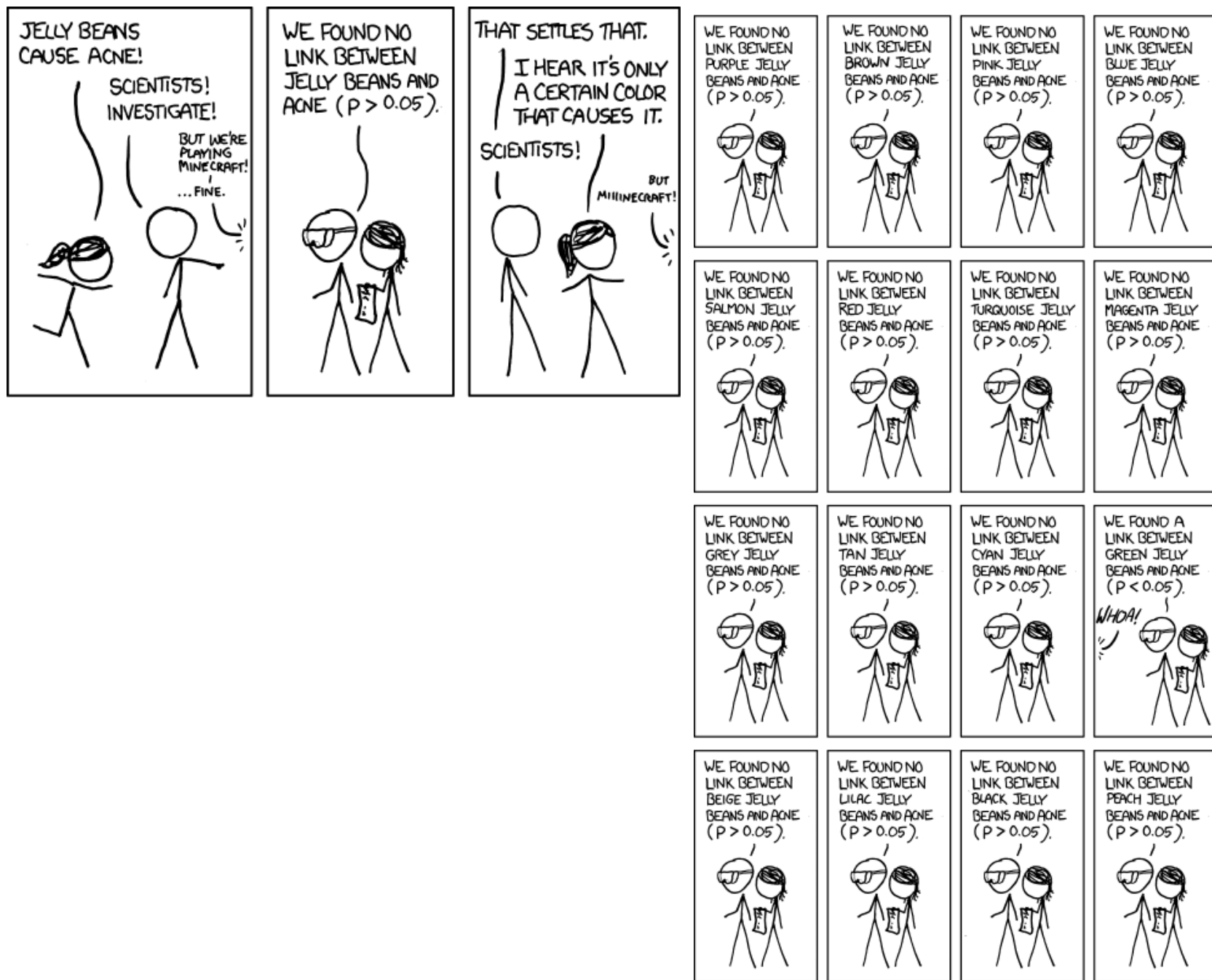
# MULTIPLE COMPARISON PROCEDURE

**‘Multiple comparison procedures’ used to detect simple or complex differences**

**Significant omnibus test NOT always necessary**

- Inaccurate when assumptions violated
- Type II error

**OKAY to conduct multiple comparisons when  $p$ -value CLOSE to significance**



# ERROR RATES

$\alpha = p(\text{Type I error})$

- Determined in study design
- Generally,  $\alpha = .01, .05, \text{ or } .10$

$\alpha$  also per comparison error rate ( $\alpha_{PC}$ )

$$\alpha = \alpha_{PC}$$

$\alpha_{PC}$  = Error rate for any 1 comparison

Experimentwise ( $\alpha_{EW}$ )

$p(\geq 1 \text{ Type I error for } \underline{\text{all}} \text{ comparisons})$

Relationship between  $\alpha_{PC}$  and  $\alpha_{EW}$

$$\alpha_{EW} = 1 - (1 - \alpha_{PC})^c$$

$c$  = Number of comparisons

$(1 - \alpha_{PC})^c = p(\text{NOT making Type I error over } c)$

# ERROR RATES

## ANOVA with 4 groups

- $F$ -statistic is significant
- Comparing each group with one another
  - $c = 6$
  - $\alpha_{PC} = .05$
  - $\alpha_{EW} = ???$
  - $\alpha_{EW}$  when  $c = 10$ ?

## 3 Options...

- Ignore  $\alpha_{PC}$  or  $\alpha_{EW}$
- Modify  $\alpha_{PC}$
- Modify  $\alpha_{EW}$

$$\bar{X}_1 \text{ vs. } \bar{X}_2$$

$$\bar{X}_1 \text{ vs. } \bar{X}_3$$

$$\bar{X}_1 \text{ vs. } \bar{X}_4$$

$$\bar{X}_2 \text{ vs. } \bar{X}_3$$

$$\bar{X}_2 \text{ vs. } \bar{X}_4$$

$$\bar{X}_3 \text{ vs. } \bar{X}_4$$

# COMPARISONS

## Post hoc *(a posteriori)*

Selected **after** data collection and analysis

Used in **exploratory** research

Larger set of or all possible comparisons

Inflated  $\alpha_{EW}$ : Increased  $p$ (Type I error)

## Pre Planned *(a priori)*

Selected **before data collection**

Follow hypotheses and theory

**Smaller set** of comparisons

Not all possible combinations

$\alpha_{EW}$  is much smaller than alternatives

$\alpha_{EW}$  can slightly exceed  $\alpha$  when planned

Adjust when  $c$  is large or includes all possible comparisons?

Justified conducting ANY planned comparison

ANOVA need NOT be statistically significant

# PROBLEMS WITH COMPARISONS

Decision to statistically test certain post hoc comparisons made after examining data

- When only 'most-promising' comparisons are selected, need to correct for inflated  $p$ (Type I error)
- Biased sample data often deviates from population

When all possible pairwise comparisons are conducted,  $p$ (Type I error) or  $\alpha_{EW}$  is same for *a priori* and *post hoc* comparisons

For example, a significant  $F$ -statistic is obtained:

Assume 20 pairwise comparisons are possible

- But, in population, no significant differences exist
- Made a Type I error obtaining significant  $F$ -statistic
- However, a *post hoc* comparison using sample data suggests largest and smallest means differ  $\bar{X}_L - \bar{X}_S$

If we had conducted 1 planned comparison

- 1 in 20 chance ( $\alpha = .05$ ) of conducting this comparison and making a type I error

If we had conducted all possible comparisons

- 100% chance ( $\alpha = 1.00$ ) of conducting this comparison and making a type I error
- If researcher decides to make only 1 comparison after looking at data, between largest and smallest means, chance of type I error is still 100%
  - All other comparisons have been made 'in head' and this is only one of all possible comparisons
  - Testing largest vs. smallest means is probabilistically similar to testing all possible comparisons



# COMMON TECHNIQUES

## *a priori* tests

- Multiple *t*-tests
- Bonferroni (Dunn)
- *Dunn-Šidák*\*
- *Holm*\*
- Linear contrasts

## Complex comparison

\*adjusts  $\alpha_{PC}$

\*\**Italicized*: not covered

## *post hoc* tests

- Fisher LSD
- Tukey HSD
- Student-Newman-Keuls (SNK)
- Tukey-b
- Tukey-Kramer
- Games-Howell
- Duncan's
- Dunnett's
- REGWQ
- Scheffé

Many more comparison techniques available

Most statistical packages make no *a priori* / *post hoc* distinction  
All called *post hoc* (SPSS) or multiple comparisons (R)

In practice, most *a priori* comparison techniques can be used as *post hoc* procedures

Most refer to these techniques collectively as post hoc, not because they were planned after doing the study per se, but because they are conducted **after an omnibus test**

# A PRIORI PROCEDURES: MULTIPLE T-TESTS

## Homogeneity of variance

- $MS_W$  (estimated pooled variance) and  $df_W$  (both from ANOVA) for critical value (smaller  $F_{crit}$ )

$$t = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{MS_W}{n_1} + \frac{MS_W}{n_2}}} = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{2MS_W}{n_j}}}$$

## Heterogeneity of variance and equal $n$

- Above equation: Replace  $MS_W$  with  $s_j^2$  and  $df_W$  with  $df = 2(n_j - 1)$  for  $t_{crit}$

## Heterogeneity of variance and unequal $n$

- Above equation: Replace  $MS_W$  with  $s_j^2$  and  $df_W$  with Welch-Satterwaite  $df$  for  $t_{crit}$

# A PRIORI PROCEDURES: BONFERRONI (DUNN) T-TEST

Developed by both Fisher ('Splitting') and Dunn

Bonferroni inequality

- $p(\text{occurrence for set of events (additive)}) \leq \sum \text{ of probabilities for each event}$

Adjusting  $\alpha_{PC}$

- Each comparison has  $p(\text{Type I error}) = \alpha_{PC} = .05$
- $\alpha_{EW} = .05$
- $\alpha_{EW} \leq c * \alpha_{PC}$ 
  - $p(\geq 1 \text{ Type I error})$  can never exceed  $c * \alpha_{PC}$

**Example for 6 comparisons:**

$$\alpha_{PC} = .05/6 = .0083$$

Conduct standard independent-samples  $t$ -tests per pair

**$t$ -tables lack Bonferroni-corrected critical values**

- Software: Exact  $p$ -values
- Is exact  $p$ -value  $\leq$  Bonferroni-corrected  $\alpha$ -level?

**More conservative:** Reduced  $p(\text{Type I error})$

**Less powerful:** Increased  $p(\text{Type II error})$

# A PRIORI PROCEDURES: LINEAR CONTRASTS - IDEA

Linear combination of means:

$$L = c_1\bar{X}_1 + c_2\bar{X}_2 + \cdots + c_k\bar{X}_k = \sum_{i=1}^k c_j\bar{X}_j$$

- Each group mean weighted by constant ( $c$ )
- Products summed together

Weights selected so means of interest are compared

Sum of weights = 0

Example 1: 4 means

Compare  $M_1$  to  $M_2$ , ignore others

$c_1 = 1, c_2 = -1, c_3 = 0, c_4 = 0$

$$L = (1)\bar{X}_1 + (-1)\bar{X}_2 + (0)\bar{X}_3 + (0)\bar{X}_4 = \bar{X}_1 - \bar{X}_2$$

Example 2: Same 4 means

Compare  $M_1, M_2$ , and  $M_3$  to  $M_4$

$c_1 = 1/3, c_2 = 1/3, c_3 = 1/3, c_4 = -1$

$$L = (1/3)\bar{X}_1 + (1/3)\bar{X}_2 + (1/3)\bar{X}_3 + (-1)\bar{X}_4 = \frac{(\bar{X}_1 + \bar{X}_2 + \bar{X}_3)}{3} - \bar{X}_4$$

# A PRIORI PROCEDURES: LINEAR CONTRASTS - SS

Each linear combination:  $SS_{Contrast}$

Equal ns:

$$SS_{Contrast} = \frac{n_j L^2}{\sum_{j=1}^k c_j^2} = \frac{n_j (\sum_{j=1}^k c_j \bar{X}_j)^2}{\sum_{j=1}^k c_j^2}$$

Unequal ns:

$$SS_{Contrast} = \frac{L^2}{\sum_{j=1}^k \left( \frac{c_j^2}{n_j} \right)} = \frac{(\sum_{j=1}^k c_j \bar{X}_j)^2}{\sum_{j=1}^k \left( \frac{c_j^2}{n_j} \right)}$$

$SS_{Between}$  partitioned into  $k$   $SS_{Contrasts}$

▪  $SS_{Between} = SS_{Contrast 1} + SS_{Contrast 2} + \dots + SS_{Contrast k}$

$$F = \frac{MS_{Contrast}}{MS_W} = \frac{nL^2 / \sum c_j^2}{MS_W} = \frac{nL^2}{\sum c_j^2 * MS_W} \text{ or } \frac{L^2}{\sum_{j=1}^k \left( \frac{c_j^2}{n_j} \right) * MS_W}$$

$df$  for  $SS_B = k - 1$

$df$  for  $SS_{Contrast}$  = Number of 'groups/sets' included in contrast minus 1

$$F = MS_{Contrast} / MS_W$$

$$MS_{Contrast} = SS_{Contrast} / df_{Contrast}$$

As  $df = 1$ ,  $MS_{Contrast} = SS_{Contrast}$

$MS_W$  from omnibus ANOVA results

**Max # 'legal' contrasts =  $df_B$**   
Do not need to consume all available  $df$   
Use smaller  $\alpha_{EW}$  if # contrasts >  $df_B$

# A PRIORI PROCEDURES: LINEAR CONTRASTS - EXAMPLE

3 Ms: 9.2, 6.6, 6.2;  $N_t = 15$ ,  $n_i = 5$

Test each Contrast (ANOVA:  $SS_{Between} = 26.53$ ,  $SS_{Within} = 22.8$ )

Contrast 1:  $M_{No\ Noise}$  versus  $M_{Moderate}$  and  $M_{loud}$ ,

Contrast 2:  $M_{Moderate}$  versus  $M_{loud}$

# A PRIORI PROCEDURES: LINEAR CONTRASTS - ORTHOGONAL

## Independent (orthogonal) contrasts

- If  $M_1$  is larger than average of  $M_2$  and  $M_3$
- Tells us nothing about  $M_4$  and  $M_5$

## Dependent (non-orthogonal) contrasts

- If  $M_1$  is larger than average of  $M_2$  and  $M_3$
- Increased probability that  $M_1 > M_2$  or  $M_1 > M_3$

**Can conduct non-orthogonal contrasts, but...**

**Dependency in data**

**Inefficiency in analysis**

**Contain redundant information**

**Increased  $p$ (Type I error)**

# A PRIORI PROCEDURES: LINEAR CONTRASTS - ORTHOGONAL

Orthogonality indicates  $SS_{Contrasts}$  are independent partitions of  $SS_B$

## Orthogonality obtained when

- $\Sigma$  of  $SS_{Contrasts} = SS_{Between}$

- Two rules are met:

- Rule 1:  $\sum_{j=1}^k c_j = 0$

Rule 2:  $\sum_{j=1}^k c_{1j}c_{2j}c_{Lj} = 0$

where  $c_{Lj}$  = Contrast weights from additional linear combinations

From example...Orthogonal!

- Rule 1:  $L_1 = (1)+(1)+(-2) = 0$ ;  $L_2 = 1+(-1)+(0) = 0$

- Rule 2:  $-2*0 + 1*1 + 1*-1 = 1 + -1 + 0 = 0$



# A PRIORI PROCEDURES: RECOMENDATIONS

## 1 pairwise comparison of interest

- Standard  $t$ -test

## Several pairwise comparisons

- Bonferroni, Multiple  $t$ -tests
- Bonferroni is most widely used (varies by field), and can be used for multiple statistical testing situations

## 1 complex comparison

- Linear contrast

## Several complex comparisons

- Orthogonal linear contrasts – no adjustment
- Non-orthogonal contrasts – Bonferroni correction or more conservative  $\alpha_{PC}$

# POST HOC PROCEDURES: FISHER'S LSD TEST

Fisher does it again (1951)

**Aka: Fisher's Protected  $t$ -test = Multiple  $t$ -test**

- Conduct as described previously: 'multiple  $t$ -tests'
- 'Fisher's LSD test': Only after significant  $F_{stat}$
- 'Multiple  $t$ -test': Planned *a priori*

One advantage is that equal  $ns$  are not required

## Logic

If  $H_0$  true and all means equal one another, significant overall  $F$ -statistic ensures  $\alpha_{EW}$  is fixed at  $\alpha_{PC}$

## Powerful: No adjustment to $\alpha_{PC}$

Most liberal *post hoc* comparison

Highest  $p$ (Type I error)

Not recommended in most cases

Only use when  $k = 3$

# POST HOC PROCEDURES: STUDENTIZED RANGE Q

$t$ -distribution derived under assumption of comparing only 2 sample means

- With  $>2$  means, sampling distribution of  $t$  is NOT appropriate as  $p(\text{Type I error}) > \alpha$

Need sampling distributions based on comparing multiple means

- **Studentized range  $q$ -distribution**

- ☐  $k$  random samples (equal  $n$ ) from population
- ☐ Difference between high and low means
- ☐ Differences divided by  $\sqrt{\frac{MS_W}{n_j}}$
- ☐ Obtain probability of multiple mean differences
- ☐ Critical value varies to control  $\alpha_{EW}$

**Rank order group means (low to high)**

- $r = \underline{\text{Range}}$  or distance between groups being compared
  - 4 means: Comparing  $M_1$  to  $M_4$ ,  $r = 4$ ; comparing  $M_3$  to  $M_4$ ,  $r = 2$
- **Not part of calculations, used to find critical value**

$q_{crit}$ : Use  $r$ ,  $df_W$  from ANOVA, and  $\alpha$

- $q_{crit}$  always positive

**Most tests of form:**

$$q = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{MS_W}{n_j}}}$$

# POST HOC PROCEDURES: STUDENTIZED RANGE Q

**Table A.11**  
Critical Values of the Studentized Range Statistic ( $q$ ) for  $\alpha = .05$

NUMBER OF GROUPS (OR NUMBER OF STEPS BETWEEN ORDERED MEANS)

df for Error Term	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
1	17.97	26.98	32.82	37.08	40.41	43.12	45.40	47.36	49.07	50.59	51.96	53.20	54.33	55.36	56.32	57.22	58.04	58.83	59.56
2	6.08	8.33	9.80	10.88	11.74	12.44	13.03	13.54	13.99	14.39	14.75	15.08	15.38	15.65	15.91	16.14	16.37	16.57	16.77
3	4.50	5.91	6.82	7.50	8.04	8.48	8.85	9.18	9.46	9.72	9.95	10.15	10.35	10.52	10.69	10.84	10.98	11.11	11.24
4	3.93	5.04	5.76	6.29	6.71	7.05	7.35	7.60	7.83	8.03	8.21	8.37	8.52	8.66	8.79	8.91	9.03	9.13	9.23
5	3.64	4.60	5.22	5.67	6.03	6.33	6.58	6.80	6.99	7.17	7.32	7.47	7.60	7.72	7.83	7.93	8.03	8.12	8.21
6	3.46	4.34	4.90	5.30	5.63	5.90	6.12	6.32	6.49	6.65	6.79	6.92	7.03	7.14	7.24	7.34	7.43	7.51	7.59
7	3.34	4.16	4.68	5.06	5.36	5.61	5.82	6.00	6.16	6.30	6.43	6.55	6.66	6.76	6.85	6.94	7.02	7.10	7.17
8	3.26	4.04	4.53	4.89	5.17	5.40	5.60	5.77	5.92	6.05	6.18	6.29	6.39	6.48	6.57	6.65	6.73	6.80	6.87
9	3.20	3.95	4.41	4.76	5.02	5.24	5.43	5.59	5.74	5.87	5.98	6.09	6.19	6.28	6.36	6.44	6.51	6.58	6.64
10	3.15	3.88	4.33	4.65	4.91	5.12	5.30	5.46	5.60	5.72	5.83	5.93	6.03	6.11	6.19	6.27	6.34	6.40	6.47
11	3.11	3.82	4.26	4.57	4.82	5.03	5.20	5.35	5.49	5.61	5.71	5.81	5.90	5.98	6.06	6.13	6.20	6.27	6.33
12	3.08	3.77	4.20	4.51	4.75	4.95	5.12	5.27	5.39	5.51	5.61	5.71	5.80	5.88	5.95	6.02	6.09	6.15	6.21
13	3.06	3.73	4.15	4.45	4.69	4.88	5.05	5.19	5.32	5.43	5.53	5.63	5.71	5.79	5.86	5.93	5.99	6.05	6.11
14	3.03	3.70	4.11	4.41	4.64	4.83	4.99	5.13	5.25	5.36	5.46	5.55	5.64	5.71	5.79	5.85	5.91	5.97	6.03
15	3.01	3.67	4.08	4.37	4.59	4.78	4.94	5.08	5.20	5.31	5.40	5.49	5.57	5.65	5.72	5.78	5.85	5.90	5.96
16	3.00	3.65	4.05	4.33	4.56	4.74	4.90	5.03	5.15	5.26	5.35	5.44	5.52	5.59	5.66	5.73	5.79	5.84	5.90
17	2.98	3.63	4.02	4.30	4.52	4.70	4.86	4.99	5.11	5.21	5.31	5.39	5.47	5.54	5.61	5.67	5.73	5.79	5.84
18	2.97	3.61	4.00	4.28	4.49	4.67	4.82	4.96	5.07	5.17	5.27	5.35	5.43	5.50	5.57	5.63	5.69	5.74	5.79
19	2.96	3.59	3.98	4.25	4.47	4.65	4.79	4.92	5.04	5.14	5.23	5.31	5.39	5.46	5.53	5.59	5.65	5.70	5.75
20	2.95	3.58	3.96	4.23	4.45	4.62	4.77	4.90	5.01	5.11	5.20	5.28	5.36	5.43	5.49	5.55	5.61	5.66	5.71
24	2.92	3.53	3.90	4.17	4.37	4.54	4.68	4.81	4.92	5.01	5.10	5.18	5.25	5.32	5.38	5.44	5.49	5.55	5.59
30	2.89	3.49	3.85	4.10	4.30	4.46	4.60	4.72	4.82	4.92	5.00	5.08	5.15	5.21	5.27	5.33	5.38	5.43	5.47
40	2.86	3.44	3.79	4.04	4.23	4.39	4.52	4.63	4.73	4.82	4.90	4.98	5.04	5.11	5.16	5.22	5.27	5.31	5.36
60	2.83	3.40	3.74	3.98	4.16	4.31	4.44	4.55	4.65	4.73	4.81	4.88	4.94	5.00	5.06	5.11	5.15	5.20	5.24
120	2.80	3.36	3.68	3.92	4.10	4.24	4.36	4.47	4.56	4.64	4.71	4.78	4.84	4.90	4.95	5.00	5.04	5.09	5.13
$\infty$	2.77	3.31	3.63	3.86	4.03	4.17	4.29	4.39	4.47	4.55	4.62	4.68	4.74	4.80	4.85	4.89	4.93	4.97	5.01

SOURCE: Adapted from *Biometrika Tables for Statisticians*, Vol 1, 3rd ed., by E. Pearson & H. Hartley, Table 29. Copyright © 1966 University Press. Used with the permission of the Biometrika Trustees.

# POST HOC PROCEDURES: STUDENTIZED RANGE Q

**Note square root of 2 missing from denominator**

- Each critical value ( $q_{crit}$ ) in  $q$ -distribution has already been multiplied by square root of 2

$$q = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{MS_W}{n_j}}} \quad \text{Vs.} \quad t = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{MS_W}{n_1} + \frac{MS_W}{n_2}}} = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{2MS_W}{n_j}}}$$

*Post hoc tests that rely on studentized range distribution:*

Tukey HSD

Tukey's b

S-N-K

Games-Howell

REGWQ

Duncan

**Assumes all samples are of same  $n$**

- Unequal  $n$ s can lead to inaccuracies depending on group size differences
- If  $n$ s are unequal, alternatives are:
  - Compute harmonic mean (below) of  $n$  (if  $n$ s differ slightly)
  - Equal variance: Tukey-Kramer, Gabriel, Hochberg's GT2
  - Unequal variance: Games-Howell

# POST HOC PROCEDURES: TUKEY'S HSD TEST

Based on premise that Type I error can be controlled for **comparison involving largest and smallest means**, thus controlling error for all

Significant ANOVA NOT required

$q_{crit}$  based on  $df_W$ ,  $\alpha_{EW}$  (table .05), and **largest  $r$**

- If we had 5 means, all comparisons would be evaluated using  $q_{crit}$  based on  $r = 5$

$q_{crit}$  compared to  $q_{obt}$

- $MS_W$  from ANOVA

One of most conservative *post hoc* comparisons, good control of  $\alpha_{EW}$

Compared to LSD...

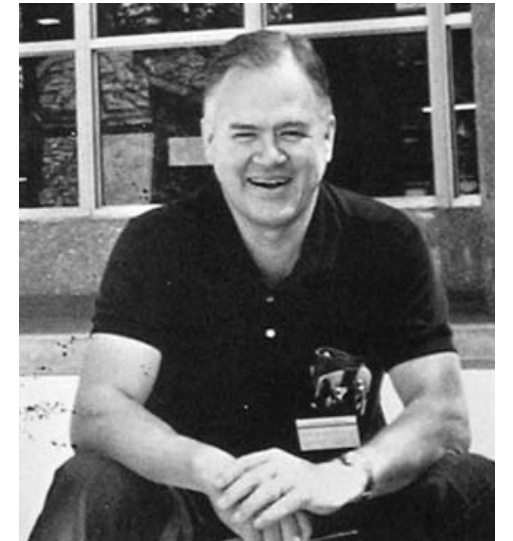
- HSD less powerful w/ 3 groups (Type II error)
- HSD more conservative; less Type I error w/ > 3 groups

Preferred with > 3 groups

**Fisher's LSD is most liberal**

**Tukey's HSD is nearly most conservative**

**Others are in-between**



# POST HOC: CONFIDENCE INTERVALS: HSD

$$q = \frac{\bar{X}_1 - \bar{X}_2}{\sqrt{\frac{MS_W}{n_j}}}$$

**Simultaneous** Confidence Intervals for all possible pairs of populations means...at the same time!

$$\mu_i - \mu_j = (\bar{X}_i - \bar{X}_j) \pm q \sqrt{\frac{MS_W}{n}} = (\bar{X}_i - \bar{X}_j) \pm HSD$$

- Interval **DOES INCLUDES** zero → fail to reject H<sub>0</sub>: means are the same...no difference
- Interval **does NOT INCLUDES** zero → **REJECT H<sub>0</sub>** → evidence there **IS a DIFFERENCE**

# POST HOC PROCEDURES: SCHEFFÉ TEST

Most conservative and least powerful

Uses  $F$ - rather than  $t$ -distribution to find critical value

- $F_{Scheffé} = (k-1) * F_{crit}(k-1, N-k)$ 
  - Scheffé recommended running his test with  $\alpha_{EW} = .10$
- $F_{Scheffé}$  is now  $F_{crit}$  used in testing

Similar to Bonferroni;  $\alpha_{PC}$  is computed by determining all possible linear contrasts AND pairwise contrasts

**Not** recommended in most situations

- Only use for **complex post-hoc** comparisons
  - Compare  $F_{contrast}$  to  $F_{Scheffé}$





# POST HOC PROCEDURES: RECOMMENDATIONS

## 1 pairwise comparison of interest

- Standard independent-samples  $t$ -test

## Several pairwise comparisons

- 3 → LSD
- > 3 → HSD or other alternatives such as Tukey-b or REGWQ
- Control vs. set of Tx groups → Dunnett's

## 1 complex comparison (linear contrast)

- No adjustment

## Several complex comparisons (linear contrasts)

- Non-orthogonal – Scheffé test
- Orthogonal – Use more conservative  $\alpha_{PC}$

# ANALYSIS OF TREND COMPONENTS

Try when the independent variable (IV) is highly ordinal or truly underlying continuous

- \* **LINEAR regression:**

- Run linear regression with the IV as predictor
- Compare the F-statistic's p-value for the source=regression to the ANOVA source=between

- \* **CURVE-a-linear regression:**

- create a new variable that is = IV variable SQUARED
- Run linear regression with BOTH the original IV & the squared-IV as predictors
- Compare the F-statistic's p-value for the source=regression

# CONCLUSION

Not all researchers agree about best approach/methods

## Method selection depends on

- Researcher preference (conservative/liberal)
- Seriousness of making Type I vs. II error
- Equal or unequal  $ns$
- Homo- or heterogeneity of variance

Can also run mixes of pairwise and complex comparisons

Adjusting  $\alpha_{PC}$  to  $\downarrow p(\text{type I error})$ ,  $\uparrow p(\text{Type II error})$

- *a priori more powerful than post hoc*
- *a priori are better choice*
  - Fewer in number; more meaningful
  - Forces thinking about analysis in advance