# Analysis of Factor Level Means

#### Question:

What happens when the F-test leads to the conclusion that the factor level means  $\mu_i$  differ?

- Analysis of the factor level means of interest using *estimation* techniques.
- Statistical tests concerning the factor level means of interest.

### Inference for Factor Level Means

- A single factor level mean.
- A difference between two factor level means.
- A contrast among factor level means.
- A linear combination of factor level means.

## Single Factor Level Mean

- Estimation of  $\mu_i$ :  $\hat{\mu}_i = \bar{y}_i$ .
- Distribution of  $\hat{\mu}_i$ :  $E(\hat{\mu}_i) = \mu_i$ ,  $Var(\hat{\mu}_i) = \frac{\sigma^2}{n_i}$ . The estimated variance of  $\bar{y}_i$  is  $s_{\bar{y}_i}^2 = \frac{1}{n_i} \cdot \frac{RSS}{n-r}$ .
- Under the  $H_0$

$$rac{ar{y}_{i\cdot} - \mu_i}{s_{ar{y}_{i\cdot}}}$$
 is distributed as  $T_{n-r}.$ 

• Confidence Interval for  $\mu_i$ :

$$\mu_i \in \bar{y}_{i\cdot} \pm T_{n-r}(\alpha/2) \ s_{\bar{y}_{i\cdot}}$$

• In order to obtain confidence intervals for the factor level means in R, we fit the *means model* and we use the confint command:

```
# CI for single Factor Level Mean
gl=lm(coag-diet-1)
confint(gl)
```

```
## 2.5 % 97.5 %
## dietA 58.53185 63.46815
## dietB 63.98477 68.01523
## dietC 65.98477 70.01523
## dietD 59.25476 62.74524
```

#### Difference between Two Factor Level Means

The difference between two factor level means (pairwise comparison) is defined as

$$D = \mu_i - \mu_{i'}$$

- Estimation of D:  $\hat{D} = \bar{y}_{i\cdot} \bar{y}_{i'\cdot}$
- Distribution of  $\hat{D}$ :  $E(\hat{D}) = \mu_i \mu_{i'}$ ,  $Var(\hat{D}) = \sigma^2 \left(\frac{1}{n_i} + \frac{1}{n_{i'}}\right)$ .

The estimated variance of  $\hat{D}$  is

$$s_{\hat{D}}^2 = \frac{RSS}{n-r} \cdot \left(\frac{1}{n_i} + \frac{1}{n_{i'}}\right).$$

• Under the  $H_0$ 

$$\frac{\hat{D}-D}{s_{\hat{D}}}$$
 is distributed as  $T_{n-r}$ 

- Confidence Interval for  $D: D \in \hat{D} + T_{n-r}(\alpha/2) s_{\hat{D}}$
- Hypothesis Test for *D*:

$$\begin{cases} H_0: \mu_i = \mu_{i'} \\ H_\alpha: \mu_i \neq \mu_{i'} \end{cases} \Leftrightarrow \begin{cases} H_0: \mu_i - \mu_{i'} = D = 0 \\ H_\alpha: \mu_i - \mu_{i'} \neq 0 \end{cases}$$

The test statistic is  $t = \frac{D}{s_n} \sim T_{n-r}$ .

#### Contrast of Factor Level Means

A contrast is a comparison involving two or more level means:

$$L = \sum_{i=1}^{r} c_i \mu_i$$
, where  $\sum_{i=1}^{r} c_i = 0$ .

- Estimation of L:  $\hat{L} = \sum_{i=1}^{r} c_i \bar{y}_i$ .
- Distribution of  $\hat{L}$ :  $E(\hat{L}) = \sum_{i=1}^{r} c_i \mu_i$ ,  $Var(\hat{L}) = \sigma^2 \sum_{i=1}^{r} \frac{c_i^2}{n_i}$ The estimated variance of  $\hat{L}$  is

$$RSS = \frac{r}{r}$$

$$s_{\hat{L}}^2 = \frac{RSS}{n-r} \cdot \sum_{i=1}^r \frac{c_i^2}{n_i}$$

• Under the  $H_0$ 

$$\frac{\hat{L}-L}{s_{\hat{x}}}$$
 is distributed as  $T_{n-r}$ 

- Confidence Interval for L:  $L \in \hat{L} + T_{n-r}(\alpha/2)$   $s_{\hat{L}}$
- Hypothesis Testing for *L*:

$$\begin{cases} H_0: L = 0 \\ H_\alpha: L \neq 0 \end{cases}$$

The test statistic is  $t = \frac{\hat{L}}{s_{\hat{r}}} \sim T_{n-r}$ .

#### Linear Combination of Factor Level Means

$$L = \sum_{i=1}^{r} c_i \mu_i$$
, no restrictions on  $c_i's$ 

- Point estimator and estimated variance same as before.
- Single Degree of Freedom Tests

$$\begin{cases} H_0: L = c \\ H_\alpha: L \neq c \end{cases}$$

Under the  $H_0$  the test statistic here is

$$F = t^2 = \left(\frac{\hat{L} - c}{s_{\hat{L}}}\right)^2 \sim F_{1,n-r}$$

#### Limitations of Inference Procedures

- ullet The confidence coefficient 1-lpha for the estimation procedures described is a statement confidence coefficient and applies only to a particular estimate, not to a series of estimates.
- Similarly the specified Type I error rate  $\alpha$  applies only to a particular test and not to a series of tests.

#### Bonferroni Correction

When? The family of interest is a particular set of pairwise comparisons, contrasts, or linear combinations that is specified by the user.

- ullet Suppose m is the number of statements in the family.
- In order to control the family wise error rate to be  $\alpha$ , we need to reduce the error rate for each individual comparison to be  $\alpha/m$ .
- That is we need to increase the significance level from  $(1 \alpha)$  to  $(1 \alpha/m)$ .
- ullet Not applicable when m is large, since the CIs would be too wide due to the increase of the significant level.

• In R, we can obtain the *p*-values for the Bonferroni corrections for pairwise differences using the pairwise.t.test command.

```
# Bonferroni correction: this test outputs the p-values for the corresponding differences.
pairwise.t.test(coag, diet, p.adjust.method = "bonferroni")

##
## Pairwise comparisons using t tests with pooled SD
##
## data: coag and diet
##
## A B C
## B 0.02282 - -
## C 0.00108 0.95266 -
## D 1.00000 0.00518 0.00014
##
## P value adjustment method: bonferroni
```

## Tukey's Paired Comparison Procedures

When? the family of interest is a set of all pairwise comparisons of factor level means, i.e. it consists of estimates of all pairs  $D = \mu_i - \mu_{i'}$ .

A confidence interval is given by

$$D \in \hat{D} + \frac{q(\alpha/2; r, n-r)}{\sqrt{2}} s(\hat{D}),$$

where  $q(\alpha/2; r, n-r)$  refers to the  $\alpha/2$  upper quantile of the **studentized range** for r means and n-r degrees of freedom.

The coverage probability is exact when the sample sizes in each group are identical and is approximate otherwise.

Remark: The studentized range refers to the distribution of

$$max_{i\neq j}\sqrt{n}(\bar{y}_i-\bar{y}_j)/\hat{\sigma}$$

where  $\bar{y}_i$  and  $\bar{y}_j$  are sample means from independent samples of size n from normal distributions with common means and variance  $\sigma^2$ .

• To obtain Tukey family CIs for all pairwise comparisons in R, we use the TukeyHSD command.

```
# Tukey Simultaneous 95% CI for all mean differences
TukeyHSD(aov(coag~diet), data=coagulation)
```

```
Tukey multiple comparisons of means
##
      95% family-wise confidence level
##
##
## Fit: aov(formula = coag ~ diet)
##
## $diet
##
      diff
                 lwr
                           upr p adj
## B-A 5 0.5932529 9.406747 0.0228300
## C-A 7 2.5932529 11.406747 0.0013858
## D-A 0 -4.2789880 4.278988 1.0000000
## C-B 2 -1.9415144 5.941514 0.4988550
## D-B -5 -8.7981383 -1.201862 0.0075558
## D-C -7 -10.7981383 -3.201862 0.0002854
```

#### Scheffé's Method for Contrasts

**When?** The family of interest is the set of **contrasts** among the factor level means:

$$L = \sum c_i \mu_i$$
, where  $\sum c_i = 0$ 

An confidence interval is given by

$$L \in \hat{L} + (r-1) F_{r-1,n-r}(\alpha) s_{\hat{L}}$$

• To obtain Scheffé family Cls for all pairwise comparisons in R, we use the ScheffeTest in the *DescTools* library.

```
q2aov=aov(coaq~diet)
# If you want all the pairwise comparisons with Scheffe's method:
ScheffeTest(q2aov)
##
    Posthoc multiple comparisons of means: Scheffe Test
##
      95% family-wise confidence level
##
## $diet
      diff lwr.ci upr.ci pval
## B-A 5 0.342883 9.657117 0.03233 *
## C-A 7 2.342883 11.657117 0.00210 **
## D-A 0 -4.418129 4.418129 1.00000
## C-B 2 -2.165452 6.165452 0.55494
## D-B -5 -8.896424 -1.103576 0.00876 **
## D-C -7 -10.896424 -3.103576 0.00031 ***
##
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

• If we want to obtain intervals for specific contrasts, such as

$$L_1=\mu_A-\frac{1}{2}\mu_B-\frac{1}{2}\mu_C$$
 and  $L_2=\mu_B-\frac{1}{2}\mu_C-\frac{1}{2}\mu_D$ 

then we can specify this in the contrasts argument as follows:

```
ScheffeTest(g2aov, contrasts=matrix(c(1,-0.5,-0.5,0,
0,1,-0.5,-0.5), ncol=2))
```

```
##
## Posthoc multiple comparisons of means: Scheffe Test
## 95% family-wise confidence level
##
## $diet
## diff lwr.ci upr.ci pval
## A-B,C -6.0 -10.165452 -1.834548 0.0032 **
## B-C,D 1.5 -2.031434 5.031434 0.6482
##
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```