## Specific Comparisons

• If any of the F-tests reveal that the factor(s) have significant effects on the response, we can perform:

- Preplanned comparisons (contrasts)
- Post-hoc multiple comparisons (Fisher LSD or Tukey)

in order to determine which factor levels produce significantly different mean responses.

• This is straightforward when there is <u>no significant</u> <u>interaction</u> between factors.

• We may then treat each factor separately, and use contrasts or multiple comparisons to compare mean responses among the levels of each factor.

• Basically just like in previous chapter, except we do it for two factors separately.

**Example:** 

• If we do have significant interaction (as we actually did in the gas mileage example), we must investigate contrasts about one factor given a specific level of the <u>other factor</u>.

**Example 1**: Do the mean mileages of 4-cylinder and 6cylinder engines differ significantly, when the oil type is "Gasmiser"?

**Relevant contrast:** 

We test:

**Example 2:** Do the mean mileages for the cheap oil ("standard") and the expensive oils differ significantly, when the engine is "4-cylinder"?

**Relevant contrast:** 

We test:

**Conclusions based on computer output:** 

## **Post-Hoc Comparisons**

• If there is significant interaction, we test for significant differences in mean response for <u>each pair</u> of <u>factor level combinations</u>.

We test:

• Again, Fisher LSD procedure has P{Type I error} = α for each comparison.

• Tukey procedure has  $P{at least one Type I error} = \alpha$ for the entire set of comparisons.

• For Tukey procedure, we conclude a difference in mean response is significant, at level α, if:

(for  $i' \neq i'', j' \neq j''$ )

### Example (Gas mileage data):

#### **Additional Considerations**

• What if we have no replication (i.e.,  $n = 1 \rightarrow$  one observation for each cell)?

• We then have no estimate of  $\sigma^2$  (the variation among responses in the same cell).

• Solution: Assume there is no interaction. The interaction MS will then serve as an estimate of  $\sigma^2$ .

• If we do this, and interaction <u>does exist</u>, then our F-tests will be biased (conservative  $\rightarrow$  less likely to reject H<sub>0</sub>).

#### **Three or More Factors**

# • If we have three or more factors, we have the possibility of <u>higher-order interactions</u>.

**Example:** Factors A, B, and C:

• If the 3-way interaction is significant, this implies, for example, that the A×B interaction is not consistent across the levels of C.

• Having 3 or more factors means having lots of "cells".

• If resources are limited, the number of replicates could be small (n = 1? n = 2?)

• It may be better to assume higher-order interactions do not exist (often they are of no practical interest anyway).

• Thus we could devote more degrees of freedom to estimating  $\sigma^2$ .

• Analysis of three-factor studies can be done with software in a similar way.

Example: (Table 9.27 data, p. 515)

**<u>Response</u>:** Rice yield

<u>Factors</u>: Location (4 levels) Variety (3 levels) Nitrogen (4 levels)

• We have *n* = 1 observation for each factor level combination.

Analysis: