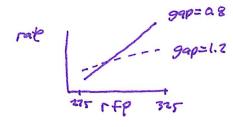
Stat 581, Problem Set #5 Solutions

- ("0.8","1.2") B= RFp setting, response = etch rate
- (a) An interaction effect occurs when the effect of one factor depends on the level of the other factor.

estimates:
$$\hat{\mu} = 776.0625$$
, $\hat{z}_1 = 50.8$, $\hat{z}_2 = -50.8$
 $\hat{\beta}_1 = -153$, $\hat{\beta}_2 = 153$

- (c) $F_{AB} = 54.31$, p = .000The experiment finds an interaction effect (between gap and RFp) on etch rate.
- (d) see interaction plot



- (e) see output for pairwise comparison p-values

 rfp= 275 | A rfp= 325 | B gap= a8) C

 grouping information gap= 0.8, 1.2 | gap=1.2 | B gap= a8) C
- (f) The experiment finds that gap has no effect on etch rate when RFp setting is 275, but that gap=0.8 leads to greater etch rates than gap=1.2 when RFp setting is 325.

```
> library("multcomp")
> library("readxl")
> setwd("C:/Users/aneath/iCloudDrive/Lexar/stat581 fall2021")
> hw5.data = read_excel("handout5data.xlsx")
  str(hw5.data)
Classes 'tbl_df',
                    'tbl' and 'data.frame':
                                                    36 obs. of 15 variables:
                       1 1 1 1 2 2 2 2 3 3
                : num
 $ plate
   temp
                        15 15 15 15 15 15 15 15 15 15
                : num
   life
                : num
                        130 155 74 180 150 188 159 126 138 110 ...
                        150 150 160 160 170 170 150 150 160 160 ...
200 200 200 200 200 200 215 215 215 215 ...
   temperature: num
   pressure
                  num
                        90.4 90.2 90.1 90.3 90.5 90.7 90.6 90.5 90.6 ... "low" "low" "low" "low" ...
   yield
                  num
                 chr
   pour.temp
                        "low" "low" "high" "high" .
   titanium
                : chr
   brk.strn
                : num
                        1.55 1.67 1.42 1.48 1.35 1.5 1.71 1.91 NA NA ...
                        1 1 2 2 3 3 1 1 2 2
   operator
                : num
   machine
                        1 1 1 1 1 1 2 2 2 2
                : num
                        109 110 110 112 116 114 110 115 110 111 ...
   strength
                : num
                        0.8 0.8 0.8 0.8 0.8 0.8 0.8 0.8 1.2 1.2 ...
275 275 275 275 325 325 325 325 275 275 ...
550 633 604 601 1052 ...
   gap
                : num
   rfp
                 num
 $
   rate
                : num
> gap = as.factor(na.omit(hw5.data$gap))
 rfp = as.factor(na.omit(hw5.data$rfp))
> rate = na.omit(hw5.data$rate)
> contrasts(gap)=contr.sum
  contrasts(rfp)=contr.sum
> two.way.mod = aov(rate ~ gap + rfp + gap:rfp)
> dummy.coef(two.way.mod)
Full coefficients are
(Intercept):
                   776.0625
gap:
                         0.8
                                    1.2
                    50.8125
                              -50.8125
rfp:
                         275
                                    325
                  -153.0625
                              153.0625
                    0.8:275
gap:rfp:
                                1.2:275
                                          0.8:325
                                                   1.2:325
                   -76.8125
                               76.8125
                                         76.8125 -76.8125
> summary(two.way.mod)
             Df Sum Sq Mean Sq F value
                                             Pr(>F)
                                    23.77 0.000382 ***
gap
                           41311
                  41311
rfp
                374850
                          374850
                                   215.66 4.95e-09 ***
gap:rfp
              1
                  94403
                           94403
                                    54.31 8.62e-06 ***
Residuals
             12
                  20858
                            1738
                  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
  interaction.plot(x.factor = rfp,trace.factor = gap,response = rate)
  interaction.plot(x.factor = gap,trace.factor = rfp,response = rate)
> combined = interaction(gap,rfp)
> levels(combined)
[1] "0.8.275" "1.2.275" "0.8.325" "1.2.325"
> comb.mod = aov(rate~combined)
> comparisons = glht(comb.mod, linfct = mcp( combined = "Tukey"))
> summary(comparisons,test=univariate())
```

Simultaneous Tests for General Linear Hypotheses

Multiple Comparisons of Means: Tukey Contrasts

Fit: aov(formula = rate ~ combined)

```
Linear Hypotheses:
```

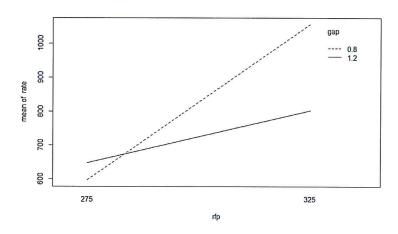
```
Estimate Std. Error t value Pr(>|t|)
1.2.275 - 0.8.275 == 0
                                            52.00
                                                              29.48
                                                                            1.764 0.103166
0.8.325 - 0.8.275 == 0

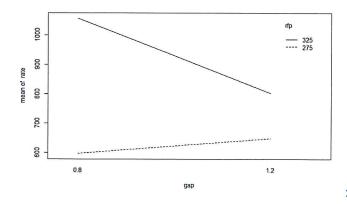
1.2.325 - 0.8.275 == 0

0.8.325 - 1.2.275 == 0

1.2.325 - 1.2.275 == 0
                                          459.75
204.50
407.75
                                                                          15.595 2.49e-09 ***
6.937 1.57e-05 ***
13.831 9.79e-09 ***
5.173 0.000232 ***
                                                               29.48
                                                              29.48
                                                               29.48
                                          152.50
                                                              29.48
1.2.325 - 0.8.325 == 0
                                         -255.25
                                                              29.48
                                                                          -8.658 1.66e-06 ***
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 (Univariate p values reported)





(2.) Factor
$$A = machine$$
, factor $B = operator$, response = breaking strength (b=3)

(a)
$$F_A = 1.095$$
, $F_B = 21.143$, $F_{AB} = 1.963$
 $(P_A = .389)$ $(P_B = .000)$ $(P_{AB} = .15)$

The experiment finds that operator has an effect on strength, but that machine has no effect on strength. Furthermore, the experiment finds that there is no interaction effect.

(b) see output for pairwise comparison p-values grouping information operator
$$\frac{1}{A}$$
 $\frac{3}{B}$

(c) confidence intervals for between operator differences
CI for
$$\mu_2 - \mu_1 = [-1.1, 3.6]$$

CI for $\mu_3 - \mu_1 = [3.65, 8.35]$
CI for $\mu_3 - \mu_2 = [2.4, 7.1]$

(d) see interaction plots; interaction model, additive model
The model smooths over the randomness in the data,
Simplifying the analysis.

```
> operator = as.factor(na.omit(hw5.data$operator))
> machine = as.factor(na.omit(hw5.data$machine))
 strength = na.omit(hw5.data$strength)
> x.mod = aov(strength ~ operator*machine)
> summary(x.mod)
                 Df Sum Sq Mean Sq F value 2 160.33 80.17 21.143
                                              Pr(>F)
                                     21.143 0.000117 ***
operator
                               4.15
                                      1.095 0.388753
                     12.46
machine
operator:machine
                  6
                     44.67
                               7.44
                                      1.963 0.150681
Residuals
                  12
                      45.50
                               3.79
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
> a.mod = aov(strength ~ operator + machine)
> compare = glht(a.mod, linfct = mcp( operator = "Tukey"))
> summary(compare, test=univariate())
         Simultaneous Tests for General Linear Hypotheses
Multiple Comparisons of Means: Tukey Contrasts
Fit: aov(formula = strength ~ operator + machine)
Linear Hypotheses:
           Estimate Std. Error t value Pr(>|t|)
2 - 1 == 0
              1.250
                          1.119
                                  1.117 0.278683
3 - 1 == 0
                                  5.362 4.27e-05 ***
              6.000
                          1.119
              4.750
                          1.119
                                  4.245 0.000487 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
(Univariate p values reported)
> cld(summary(compare, test=univariate()))
"a" "a" "b"
> confint(compare,calpha = univariate_calpha())
         Simultaneous Confidence Intervals
Multiple Comparisons of Means: Tukey Contrasts
Fit: aov(formula = strength ~ operator + machine)
Quantile = 2.1009
95% confidence level
Linear Hypotheses:
           Estimate lwr
 - 1 == 0
           1.2500 -1.1011
                             3.6011
 - 1 == 0
           6.0000
                     3.6489
                             8.3511
3 - 2 == 0
                      2.3989
           4.7500
                              7.1011
 plot(cld(summary(compare, test=univariate())))
> a.means = predict(a.mod)
> interaction.plot(operator, machine, strength)
> interaction.plot(operator, machine, a.means)
```

