# STAT 222 Spring 2022 HW8

#### Matthew Zhao

```
warpbreaks = read.table("http://www.stat.uchicago.edu/~yibi/s222/warpbreaks.txt", h=T)
warpbreaks$wool = as.factor(warpbreaks$wool)
warpbreaks$tension = factor(warpbreaks$tension, labels=c("L","M","H"))
lm1 = lm(breaks ~ wool*tension, data=warpbreaks)
```

### Q1 - 5 points

We can use Tukey's HSD to control the FWER, where the HSD is given as  $\frac{q_{g,dfE,\alpha}}{\sqrt{2}} \times \sqrt{\text{MSE}(\frac{1}{r} + \frac{1}{r})}$ 

```
anova(lm1)
## Analysis of Variance Table
##
## Response: breaks
##
             Df Sum Sq Mean Sq F value Pr(>F)
## wool
              1
                  451 450.7 3.765 0.058213 .
## tension
             2
                  2034 1017.1 8.498 0.000693 ***
## wool:tension 2 1003 501.4 4.189 0.021044 *
## Residuals 48 5745 119.7
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Calculating HSD:

```
mse<-119.7
qval<-qtukey(1-0.02,6,48)/sqrt(2)
qval * sqrt(mse*(2/9))
## [1] 17.137</pre>
```

```
sort(mean(breaks~wool+tension,data=warpbreaks))
## 2.H 1.M 1.H 2.L 2.M 1.L
## 18.7778 24.0000 24.5556 28.2222 28.7778 44.5556
```

```
2.H 1.M 1.H 2.L 2.M 1.L
18.7778 24.0000 24.5556 28.2222 28.7778 44.5556
```

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#### Q2 — 5 points

We can again use Tukey's HSD

Calculating HSD:

```
mse<-119.7
qval<-qtukey(1-0.02,3,48)/sqrt(2)
qval * sqrt(mse*(2/(9*2)))
## [1] 10.192</pre>
```

```
H M L
21.6667 26.3889 36.3889
```

# Q3 — 5 points

We again use Tukey.

```
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
lm1emmean <- emmeans(lm1,~tension:wool)</pre>
summary(contrast(lm1emmean, method="pairwise", adjust="tukey"),
infer=c(T,F), level=0.98)
   contrast estimate
##
                         SE df lower.CL upper.CL
   L 1 - M 1
                20.556 5.16 48
                                             37.7
##
                                   3.419
##
   L 1 - H 1
                20.000 5.16 48
                                   2.864
                                             37.1
                16.333 5.16 48
##
   L 1 - L 2
                                  -0.803
                                             33.5
##
   L 1 - M 2
                15.778 5.16 48
                                  -1.358
                                             32.9
   L 1 - H 2
                25.778 5.16 48
                                             42.9
##
                                   8.642
   M 1 - H 1
                -0.556 5.16 48
##
                                 -17.692
                                             16.6
   M 1 - L 2
                -4.222 5.16 48
                                 -21.358
                                             12.9
##
   M 1 - M 2
                -4.778 5.16 48
                                 -21.914
                                             12.4
##
   M 1 - H 2
                 5.222 5.16 48
                                             22.4
##
                                 -11.914
##
   H 1 - L 2
                -3.667 5.16 48
                                -20.803
                                             13.5
   H 1 - M 2
                -4.222 5.16 48
##
                                -21.358
                                             12.9
##
   H 1 - H 2
                 5.778 5.16 48
                                 -11.358
                                             22.9
   L 2 - M 2
                -0.556 5.16 48
                                -17.692
##
                                             16.6
   L 2 - H 2
                 9.444 5.16 48
                                  -7.692
                                             26.6
##
   M 2 - H 2
                10.000 5.16 48
                                  -7.136
                                             27.1
##
## Confidence level used: 0.98
## Conf-level adjustment: tukey method for comparing a family of 6 estimates
```

```
\mu_{1L} - \mu_{2L}: (-0.803,33.5)

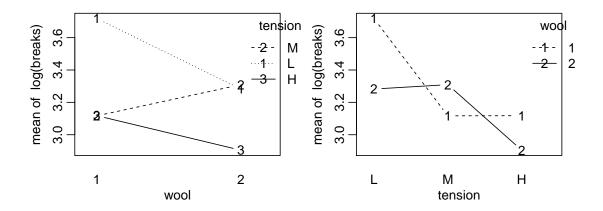
\mu_{1M} - \mu_{2M}: (-21.914,12.4)

\mu_{1H} - \mu_{2H}: (-11.358,22.9)
```

We can conclude that there could be an interaction between wool type and level of tension since the CIs change with the level of tension. However, since the CIs all overlap, we cannot be sure that the interaction term is not insignificantly different than zero i.e. we cannot conclude that there is an interaction.

# Q4 — 6 points

```
par(mai=c(.6,.6,.1,.3),mgp=c(2,.6,0))
with(warpbreaks, interaction.plot(wool, tension, log(breaks), type="b"))
with(warpbreaks, interaction.plot(tension, wool, log(breaks), type="b"))
```



i) 
$$C_1$$
:  $\sum_i c_{ij}$   
 $j=L$ :  $(1)\mu_{1L} + (-1)\mu_{2L} = 1 + (-1) = 0$   
 $j=M$ :  $(-1)\mu_{1M} + (1)\mu_{2M} = (-1) + 1 = 0$   
 $\sum_j c_{ij}$   
 $i=1$ :  $(1)\mu_{1L} + (-1)\mu_{1M} = 1 + (-1) = 0$   
 $i=2$ :  $(-1)\mu_{2L} + (1)\mu_{2M} = (-1) + 1 = 0$   
 $C_2$ :  $\sum_i c_{ij}$   
 $j=H$ :  $(1)\mu_{1H} + (-1)\mu_{2H} = 1 + (-1) = 0$   
 $j=M$ :  $(-1)\mu_{1M} + (1)\mu_{2M} = (-1) + 1 = 0$   
 $\sum_j c_{ij}$   
 $i=1$ :  $(1)\mu_{1H} + (-1)\mu_{1M} = 1 + (-1) = 0$   
 $i=2$ :  $(-1)\mu_{2H} + (1)\mu_{2M} = (-1) + 1 = 0$   
 $C_3$ :  $\sum_i c_{ij}$   
 $j=L$ :  $(0.5)\mu_{1L} + (-0.5)\mu_{2L} = 0.5 + (-0.5) = 0$   
 $j=M$ :  $(-1)\mu_{1M} + (1)\mu_{2M} = (-1) + 1 = 0$   
 $j=H$ :  $(-0.5)\mu_{1H} + (0.5)\mu_{2H} = (-0.5) + 0.5 = 0$   
 $\sum_j c_{ij}$   
 $i=1$ :  $(0.5)\mu_{1L} + (0.5)\mu_{1H}(-1)\mu_{1M} = 0.5 + 0.5 + 1 = 0$   
 $i=2$ :  $(-0.5)\mu_{2L} + (-0.5)\mu_{2H} + (1)\mu_{2M} = (-0.5) + (-0.5) + 1 = 0$ 

ii) We use Scheffe since we are performing contrasts after observing the data/based off of prior analysis.

```
anova(lm1)

## Analysis of Variance Table

##

## Response: breaks

##

Df Sum Sq Mean Sq F value Pr(>F)
```

```
## wool
                     451
                            450.7 3.765 0.058213 .
                 2
                          1017.1
                                    8.498 0.000693 ***
## tension
                     2034
## wool:tension 2
                     1003
                            501.4
                                    4.189 0.021044 *
## Residuals
                48
                     5745
                            119.7
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
lm1emmean <- emmeans(lm1,~tension:wool)</pre>
summary(contrast(lm1emmean, method=list(C_1 = c(1,-1,0,-1,1,0), C_2 = c(0,-1,1,0,1,-1), C_3 = c(0,-1,1,0,1,-1)
infer=c(F,T), level=0.96)
   contrast estimate
                        SE df t.ratio p.value
##
                 21.1 7.29 48
                                2.895 0.0210
## C 2
                 10.6 7.29 48
                                1.447 0.3588
##
  C_3
                 15.8 6.32 48
                                2.507 0.0522
##
## P value adjustment: scheffe method with rank 2
```

At FWER = 0.04, only the first contrast is significant. We can conclude that the change in effect of wool type on breaks is nonzero from low to medium.

## Q5 — 1 point

The upper bound of the FWER for the entire analysis is no more than the sum of the FWER for each family so it is 0.02 + 0.02 + 0.02 + 0.04 = 0.1.

# Q6 — 3 points

Bonferroni.

##

C\_3

15.8 6.32 48

```
anova(lm1)
## Analysis of Variance Table
##
## Response: breaks
##
                Df Sum Sq Mean Sq F value
                                             Pr(>F)
## wool
                 1
                      451
                           450.7 3.765 0.058213 .
## tension
                 2
                     2034 1017.1
                                     8.498 0.000693 ***
                     1003 501.4 4.189 0.021044 *
## wool:tension 2
                48
                             119.7
## Residuals
                     5745
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
lm1emmean <- emmeans(lm1,~tension:wool)</pre>
summary(contrast(lm1emmean, method=list(C_1 = c(1,-1,0,-1,1,0), C_2 = c(0,-1,1,0,1,-1), C_3 = c(0,-1,1,0,1,-1)
infer=c(F,T), level=0.96)
##
   contrast estimate
                        SE df t.ratio p.value
                                 2.895 0.0171
##
  C_{\perp}1
                 21.1 7.29 48
## C_2
                 10.6 7.29 48
                                 1.447 0.4630
```

2.507 0.0469

##

## P value adjustment: bonferroni method for 3 tests