# Untitled

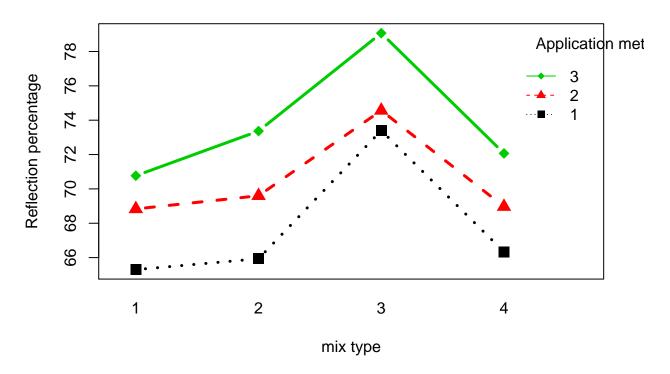
#### Andrew Liu

November 25, 2018

#### **4a**

```
day <- as.factor(c(rep(1,12),rep(2,12),rep(3,12)))</pre>
application <- as.factor(rep(c(rep(1,4),rep(2,4),rep(3,4)),3))
mix \leftarrow as.factor(rep(c(1,2,3,4),9))
reflect <- c(64.5,66.3,74.1,66.5,
             68.3, 69.5, 73.8, 70.0,
             70.3, 73.1, 78.0, 72.3,
             65.2, 65.0, 73.8, 64.8,
             69.2, 70.3, 74.5, 68.3,
             71.2, 72.8, 79.1, 71.5,
             66.2, 66.5, 72.3, 67.7,
             69.0, 69.0, 75.4, 68.6,
             70.8, 74.2, 80.1, 72.4)
df.4 <- data.frame(day,application,mix,reflect)</pre>
interaction.plot(x.factor=df.4$mix,trace.factor=df.4$application,
    response=df.4$reflect,trace.label="Application method",
    xlab="mix type",ylab="Reflection percentage",
    col=1:4,lwd=3,type="b",pch=c(15,17,18:19),
    main="Factor Plot for Grass Experiment",cex=1.5)
```

## **Factor Plot for Grass Experiment**



Because the lines don't intersect, we don't suspect an interaction effect present.

## **4**b

```
out.4b <- aov(reflect~day+mix*application,data=df.4)</pre>
summary(out.4b)
##
                    Df Sum Sq Mean Sq F value
                                                 Pr(>F)
                                 1.02
## day
                         2.04
                                         1.470
                                                 0.2517
                     3 307.48
                               102.49 147.584 1.01e-14 ***
## mix
                               111.05 159.903 7.85e-14 ***
## application
                     2 222.09
## mix:application
                        10.04
                                 1.67
                                         2.409
                                                 0.0609 .
                    6
## Residuals
                    22
                       15.28
                                 0.69
##
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

the p-value for mix:application is .0609 which is higher than 0.05, so we conclude that there is no interaction effect present. This agrees with the statment from 4a.

### **4c**

The application's pvalue is  $1.01 * 10^{-14}$  and the mix's pvalue is  $7.85 * 10^{-14}$  Both application and mix type have p-values under 0.05, so we conclude that both application and mix type are significant effects.

## **4**d

## **5**a

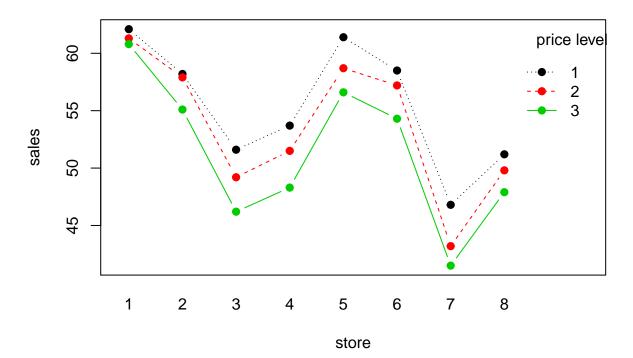
```
sales <- c(62.1, 61.3, 60.8,
58.2, 57.9, 55.1,
51.6, 49.2, 46.2,
53.7, 51.5, 48.3,
61.4, 58.7, 56.6,
58.5, 57.2, 54.3,
46.8, 43.2, 41.5,
51.2, 49.8, 47.9)

store<- as.factor(sort(rep(c(1,2,3,4,5,6,7,8),3)))

price<- as.factor(rep(c(1,2,3),8))

df.5<-data.frame(sales,store,price)
interaction.plot(x.factor=df.5$store,trace.factor=df.5$price,
    response=df.5$sales,fun=mean,type="b",col=1:6,trace.label="price level",
    xlab="store",ylab="sales",main="Grapefruit sales by store",fixed=T,pch=19)</pre>
```

## **Grapefruit sales by store**



Yes each of the lines look roughly parallel to the others.

## **5**b

```
library(car)
## Warning: package 'car' was built under R version 3.4.4
## Loading required package: carData
## Warning: package 'carData' was built under R version 3.4.4
df5.2 <- with(df.5,cbind(sales[price=="1"],sales[price=="2"],</pre>
    sales[price=="3"]))
df5.mlm <- lm(df5.2~1)
price <- as.factor(1:3)</pre>
options(contrasts=c("contr.sum", "contr.poly"))
df5.aov <- Anova(df5.mlm,idata=data.frame(price),</pre>
    idesign=~price,type="III")
summary(df5.aov,multivariate=F)
##
## Univariate Type III Repeated-Measures ANOVA Assuming Sphericity
##
##
               Sum Sq num Df Error SS den Df F value
                                                         Pr(>F)
## (Intercept)
               68587
                           1
                               745.18
                                            7 644.282 3.762e-08 ***
## price
                   67
                           2
                                  9.57
                                           14 49.346 4.567e-07 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
## Mauchly Tests for Sphericity
##
##
         Test statistic p-value
                0.67868 0.3126
## price
##
##
## Greenhouse-Geisser and Huynh-Feldt Corrections
##
   for Departure from Sphericity
##
##
          GG eps Pr(>F[GG])
## price 0.75682 8.802e-06 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
##
            HF eps
                     Pr(>F[HF])
## price 0.9212925 1.186949e-06
```

The p=value for the Mauchly test is .3126, so the test is not significant against the Huynh-Feldt condition and sphericity is an appropriate assumption.

### 5c

The p-vlue for this test is  $4.567 * 10^{-7}$ . This pvalue is less than alpha, so we conclude there is sufficient evidence that price level has a significant effect.

#### 6a

```
truth < - c(3, 1, 2, 5, 4,
4, 2, 1, 3, 5,
 4, 2, 3, 1, 5,
3, 1, 2, 5, 4,
 4, 1, 2, 5, 3,
 4, 2, 1, 3, 5,
 4, 1, 2, 3, 5,
 5, 1, 3, 2, 4,
 4, 2, 3, 1, 5,
 5, 1, 2, 3, 4)
subject<- as.factor(sort(rep(c(1,2,3,4,5,6,7,8,9,10),5)))
ad<- as.factor(rep(c(1,2,3,4,5),10))
df6<-data.frame(truth, subject, ad)</pre>
out6 <- anova(lm(truth~ad,data=df6))</pre>
n <- length(unique(df6$subject))</pre>
r <- length(unique(df6$ad))
out6
## Analysis of Variance Table
##
## Response: truth
##
                                           Pr(>F)
             Df Sum Sq Mean Sq F value
                  63.4 15.8500 19.488 2.302e-09 ***
                  36.6 0.8133
## Residuals 45
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
tmp <- out6$"Sum Sq"/c((r-1),(n-1)*(r-1))
F.star <- tmp[1]/tmp[2]
```

```
pf(F.star,r-1,(n-1)*(r-1),lower=F)
## [1] 1.723857e-07
```

So  $F_R^* = \frac{MSTR}{MSRM} = \frac{63.4/(10-1)}{36.6/((10-1)(5-1))} = 15.59$  which is distributed F with degrees 9,36

which has a pvalue of  $1.72 * 10^{-7}$  which indicates highly significant evidence of advertisment affecting how participants perceive truthfulness.

#### **6**b

```
library(xtable)
## Warning: package 'xtable' was built under R version 3.4.3
g <- r*(r-1)/2
truth.means \leftarrow c(by(df6[,1],df6[,3],mean))
CIs \leftarrow rbind((apply(combn(truth.means, 2), 2, diff)) - qnorm(1-(.2/(2*g))) * sqrt(r*(r+1)/(6*n)),
(apply(combn(truth.means,2), 2, diff))+qnorm(1-(.2/(2*g)))*sqrt(r*(r+1)/(6*n)))
comparison.first<-c("A","A","A","A","B","B","B","C","C","D")</pre>
comparison.second<- c("B","C","D","E","C","D","E","D","E","E")
CI.lower<-CIs[1,]</pre>
CI.upper<-CIs[2,]</pre>
differ<-c("yes","yes","no","no",</pre>
           "no", "yes", "yes",
           "no", "yes",
           "no")
df.6b<-data.frame(comparison.first,comparison.second,CI.lower,CI.upper,differ)
xtable(df.6b)
```

% latex table generated in R 3.4.0 by xtable 1.8-2 package % Mon Nov 26 00:08:26 2018

	comparison.first	comparison.second	CI.lower	CI.upper	differ
1	A	В	-4.24	-0.96	yes
2	A	$\mathbf{C}$	-3.54	-0.26	yes
3	A	D	-2.54	0.74	no
4	A	E	-1.24	2.04	no
5	В	C	-0.94	2.34	no
6	В	D	0.06	3.34	yes
7	В	E	1.36	4.64	yes
8	C	D	-0.64	2.64	no
9	C	E	0.66	3.94	yes
_10	D	Е	-0.34	2.94	no