

1. Use the data set “Sealing Strength of Wax-Polyethylene Blends” by Brown, Turner and Smith (*Tappi*,1958).

- (a) `seal <- read.table("hw06.txt", header=T)`  
`full <- lm(y~x1+x2+x3+I(x1^2)+I(x2^2)+I(x3^2)+x1*x2+x1*x3+x2*x3,data=seal)`  
`library(MASS)`  
`qqnorm(stdres(full),main="Normal Probability Plot",ylab="standardized residuals")`  
`qqline(stdres(full))`
- (b) `reduced <- lm(y~x1+x2+x3,data=seal)`  
`anova(reduced,full)`

Analysis of Variance Table

Model 1:  $y \sim x_1 + x_2 + x_3$

Model 2:  $y \sim x_1 + x_2 + x_3 + I(x_1^2) + I(x_2^2) + I(x_3^2) + x_1:x_2 + x_1:x_3 + x_2:x_3$

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	16	51.209				
2	10	11.859	6	39.349	5.5301	0.009126 **

F-ratio = 5.53 with p-value = 0.009. There is evidence that the response is not “planar” as a function of the  $x$ 's.

- (c) `new.x <- data.frame(-1.01,.26,.68)`  
`names(new.x) <- c("x1","x2","x3")`  
`predict(full,new.x,interval="confidence", level=0.90)`  

	fit	lwr	upr
[1,]	11.08156	10.05503	12.10809

`predict(full,new.x,interval="prediction", level=0.90)`  

	fit	lwr	upr
[1,]	11.08156	8.856805	13.30632

Or you could have done:

```
x0 <- c(1,new.x$x1,new.x$x2,new.x$x3,new.x$x1^2,new.x$x2^2,new.x$x3^2,
        new.x$x1*new.x$x2,new.x$x1*new.x$x3,new.x$x2*new.x$x3)
```

```
y0 <- sum(full$coef*x0)
```

```
X <- cbind(rep(1,dim(seal)[1]),seal$x1,seal$x2,seal$x3,seal$x1^2,seal$x2^2,seal$x3^2,
           seal$x1*seal$x2,seal$x1*seal$x3,seal$x2*seal$x3)
```

```
MSE <- anova(full)$Mean[10]
```

```
cXXc <- x0*%ginv(t(X)*%X)*%x0
```

```
# lower and upper confidence limits
```

```
ll <- y0 - qt(0.95,10)*sqrt(MSE*cXXc)
```

```
ul <- y0 + qt(0.95,10)*sqrt(MSE*cXXc)
```

```
# lower and upper prediction limits
```

```
ll <- y0 - qt(0.95,10)*sqrt(MSE*(1+cXXc))
```

```
ul <- y0 + qt(0.95,10)*sqrt(MSE*(1+cXXc))
```

2. Testing “Lack of Fit”.

```
project <- function(X) {X*%ginv(t(X)*%X)*%t(X)}
```

```
Xstar <- matrix(0,20,15)
```

```
Xstar[1:8,1:8] <- diag(rep(1,8))
```

```
Xstar[9:14,9] <- 1
```

```
Xstar[15:20,10:15] <- diag(rep(1,6))
```

```
F.ratio <- ((t(seal$y)*%(project(Xstar)-project(X))*%seal$y)/(15-10))/
```

```
          ((t(seal$y)*%(diag(rep(1,20))-project(Xstar))*%seal$y)/(20-15))
```

```
p.value <- 1-pf(F.ratio,5,5)
```

F-ratio = 1.39 with p-value = 0.363. There is no evidence of lack-of fit.

3. Use dogs data set.

```
lm.out <- lm(y~drug*disease,data=d)
anova(lm.out)          # Type I SS

Analysis of Variance Table
Response: y
      Df Sum Sq Mean Sq F value    Pr(>F)
drug     3  2992.8   997.6   9.0513 8.047e-05 ***
disease  2   365.7   182.9   1.6591  0.2015
drug:disease 6   737.9   123.0   1.1158  0.3680
Residuals 46  5070.0   110.2
```

4. Use table of fake 2-way factorial data.

```
(a) Y <- c(12,14,10,12,9,11,12,6,7,10,11,7)
A <- c(rep(1,4),rep(2,5),rep(3,3))
B <- c(1,2,3,3,1,2,2,3,3,1,2,3)
d <- data.frame(Y,A,B)
d$A <- as.factor(d$A)
d$B <- as.factor(d$B)
lm.out <- lm(Y~A*B,data=d)
means <- tapply(d$Y,list(d$A,d$B),mean)
> means
  1  2  3
1 12 14.0 11.0
2  9 11.5  6.5
3 10 11.0  7.0

x.axis <- unique(d$A)
matplot(c(1,3),c(0,20),type="n",xlab="A",ylab="Mean Response")
matlines(x.axis,means,type="b",lty=c(1,3,7))
lm.out <- lm(Y~A*B,data=d)
```

```
> anova(lm.out)          # Type I SS
Analysis of Variance Table
Response: Y
      Df Sum Sq Mean Sq F value    Pr(>F)
A         2  22.250   11.125  11.1250 0.04095 *
B         2  37.039   18.520  18.5195 0.02051 *
A:B        4   2.628    0.657   0.6569 0.66182
Residuals  3   3.000    1.000
```

```
(b) > Xeffects
      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9]
[1,]    1    1    0    1    0    1    0    0    0
[2,]    1    1    0    0    1    0    1    0    0
[3,]    1    1    0   -1   -1   -1   -1    0    0
[4,]    1    1    0   -1   -1   -1   -1    0    0
[5,]    1    0    1    1    0    0    0    1    0
[6,]    1    0    1    0    1    0    0    0    1
[7,]    1    0    1    0    1    0    0    0    1
[8,]    1    0    1   -1   -1    0    0   -1   -1
[9,]    1    0    1   -1   -1    0    0   -1   -1
[10,]   1   -1   -1    1    0   -1    0   -1    0
[11,]   1   -1   -1    0    1    0   -1    0   -1
[12,]   1   -1   -1   -1   -1    1    1    1    1
```

```

ssA <- t(d$Y)%*(project(Xeffects[,1:3])-project(Xeffects[,1]))%*d$Y # R(alpha|mu)
ssB <- t(d$Y)%*(project(Xeffects[,1:5])-project(Xeffects[,1:3]))%*d$Y # R(beta|mu,alpha)
ssAB <- t(d$Y)%*(project(Xeffects)-project(Xeffects[,1:5]))%*d$Y # R(alphabeta|mu,alpha,beta)
Type I SS:  $R(\alpha|\mu) = 22.250$ ,  $R(\beta|\mu, \alpha) = 37.039$ ,  $R(\alpha\beta|\mu, \alpha, \beta) = 2.628$ .

```

```

> Xcells
  [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9]
[1,]  1   0   0   0   0   0   0   0   0
[2,]  0   1   0   0   0   0   0   0   0
[3,]  0   0   1   0   0   0   0   0   0
[4,]  0   0   1   0   0   0   0   0   0
[5,]  0   0   0   1   0   0   0   0   0
[6,]  0   0   0   0   1   0   0   0   0
[7,]  0   0   0   0   1   0   0   0   0
[8,]  0   0   0   0   0   1   0   0   0
[9,]  0   0   0   0   0   1   0   0   0
[10,] 0   0   0   0   0   0   1   0   0
[11,] 0   0   0   0   0   0   0   1   0
[12,] 0   0   0   0   0   0   0   0   1

```

Type III SS for testing main effect A:  $H_0 : \bar{\mu}_1 = \bar{\mu}_2 = \bar{\mu}_3$ . That is,  $H_0 : \bar{\mu}_1 - \bar{\mu}_3 = 0$  and  $\bar{\mu}_2 - \bar{\mu}_3 = 0$ .

```
C <- matrix(c(1/3,1/3,1/3,0,0,0,-1/3,-1/3,-1/3,0,0,0,1/3,1/3,1/3,-1/3,-1/3,-1/3),2,9,byrow=T)
```

Type III SS for testing main effect B:  $H_0 : \bar{\mu}_1 = \bar{\mu}_2 = \bar{\mu}_3$ . That is,  $H_0 : \bar{\mu}_1 - \bar{\mu}_3 = 0$  and  $\bar{\mu}_2 - \bar{\mu}_3 = 0$ .

```
C <- matrix(c(rep(c(1/3,0,-1/3),3),rep(c(0,1/3,-1/3),3)),2,9,byrow=T)
```

Then compute:

```

CXXC <- C%*%ginv(t(Xcells)%*%Xcells)%*%t(C)
Cb <- C%*%ginv(t(Xcells)%*%Xcells)%*%t(Xcells)%*%d$Y
SSH0 <- t(Cb)%*%solve(CXXC)%*%Cb

```

Type III SS A:  $SS_{H0} = 25.108$ . Type III SS B:  $SS_{H0} = 32.331$ .

```

(c) Xincomp.full <- Xcells[-c(6,7),-5]
Xincomp.red <- Xeffects[-c(6,7),1:5]
Y.incomp <- d$Y[-c(6,7)]
full <- lm(Y.incomp~Xincomp.full-1)
reduced <- lm(Y.incomp~Xincomp.red-1)
anova(reduced,full)

```

Analysis of Variance Table

Model 1: Y.incomp ~ Xincomp.red - 1

Model 2: Y.incomp ~ Xincomp.full - 1

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
1	5	3.8298				
2	2	2.5000	3	1.3298	0.3546	0.7954

F-ratio = 0.35 with p-value= 0.7954. There is not enough evidence that there is interaction between factors A and B. (Note that in 4(a) the interaction term is not significant either.)

```

(d) X <- Xeffects[-c(6,7),1:5]
C <- c(1,0,1,0,1)
mu22 <- C%*%ginv(t(X)%*%X)%*%t(X)%*%Y.incomp
> mu22
  [,1]
[1,] 10.08511

```