```
set.seed(48109711)
n=200; x=rnorm(2*n); x1=x[1:n]; x2=x[(n+1):(2*n)]; x3=.7*x1+.9*x2+rnorm(n)*.1
beta=c(1, 1, 1, 2)
y1=beta[1]+beta[2]*x1+beta[3]*x2+rnorm(n)*.3
y2=beta[1]+beta[2]*x1+beta[3]*x2+beta[4]*x3+rnorm(n)*.3
dat=data.frame(y1=y1,y2=y2,x1=x1,x2=x2,x3=x3)
outx12 = lm(y1~x1+x2, data=dat)
summary(outx12)
outx123 = lm(y1^x1+x2+x3, data=dat)
library(ellipse)
## Plot the confidence region
par(mfrow=c(1,2))
plot(ellipse(outx12, c("x1", "x2")), type="l")
plot(ellipse(outx123, c("x1", "x2")), type="1")
## Add the estimates to the plot
D1= model.matrix(outx12);D2= model.matrix(outx123);
cov(x1,x2);
V1=solve(t(D1)%*%D1); V2=solve(t(D2)%*%D2);
round(V1[2:3,2:3],4); round(V2[2:3,2:3],4);
```

1. In the codes given above, we generate $n=200\ x1,\ x2$ and x3, and $x1,\ x2$ are independent of each other. In model A, we generate y1 with the predictors x1 and x2. In model B, we generate y2 with all three predictors. In the two plots we generate, we obtain the 95% confidence regions for the coefficients, $\beta_{x1},\ \beta_{x2}$ from each of the two model. Note that even you generate $x1\ x2$ independently, the sample correlation still may not be 0, but it tends to be small. Consequently, the major axis and minor axis of the ellipse are roughly parallel with the x-axis and y-axis corresponding to $\hat{\beta}_{x1},\ \hat{\beta}_{x2}$. In model B, you are using the same x1 and x2 and you still consider the 95% confidence ellipse for $\beta_{x1},\ \beta_{x2}$, do you expect the major axis and minor axis of the ellipse are roughly parallel with the x-axis and y-axis corresponding to $\hat{\beta}_{x1},\ \hat{\beta}_{x2}$? Please briefly explain your answer using the output you obtain. Basically, in a multiple regression, unless the predictors are completely orthogonal to each other, the estimated coefficients and the

corresponding inference are adjusted to the existence of other predictors.

- 2. Using the sat data as in the previous homework and fit a model with total sat score as the response and takers, ratio and salary as predictors. Let $\alpha = .05$. Re-test H_{A1} : $\beta_{ratio} \neq 0$ and H_{A2} : $\beta_{salary} \neq 0$ by using a permutation test.
- 3. As above, but now use the permutation test to test H_0 : $\beta_{ratio} = \beta_{salary} = 0$ vs the H_A that at least one of the two coefficients are not 0 by a 0.05 permutation test.