## Week2 Assignment Econometrics

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## Questions

(a) Prove:

$$E(b_R) = \beta_1 + P\beta_2$$

First, it is important to express  $b_R$  in terms of  $\epsilon$ :

$$b_{R} = (X_{1}^{'}X_{1})^{-1}X_{1}^{'}y = (X_{1}^{'}X_{1})^{-1}X_{1}^{'}(X_{1}\beta_{1} + X_{2}\beta_{2} + \epsilon) = (X_{1}^{'}X_{1})^{-1}X_{1}^{'}X_{2}\beta_{2} + (X_{1}^{'}X_{1})^{-1}X_{1}^{'}\epsilon = \beta_{1} + P\beta_{2} + (X_{1}^{'}X_{1})^{-1}X_{1}^{'}\epsilon so, E(b_{1}^{'}X_{1})^{-1}X_{1}^{'}(a_{1}^{'}X_{1}) = (A_{1}^{'}X_{1})^{-1}X_{1}^{'}(a_{1}^{'}X_{1}) + (A_{1}^{'}X_{1})^{-1}X_{1}^{'}(a_{1}^{'}X_{1}$$

(b) Prove:

$$var(b_R) = \sigma^2(X_1'X_1)^{-1}$$

First, it is important to express  $b_R$  in terms of  $\epsilon$ :

$$b_R - E(b_R) = \beta_1 + P\beta_2 + (X_1^{'}X_1)^{-1}X_1^{'}\epsilon - \beta_1 + P\beta_2 = (X_1^{'}X_1)^{-1}X_1^{'}\epsilon var(b_R) = E[((X_1^{'}X_1)^{-1}X_1^{'}\epsilon)((X_1^{'}X_1)^{-1}X_1^{'}\epsilon)^{'}] = E[((X_1^{'}X_1)^{-$$

 $b_{R} = (X_{1}^{'}X_{1})^{-1}X_{1}^{'}y = (X_{1}^{'}X_{1})^{-1}X_{1}^{'}(X_{1}b_{1} + X_{2}b_{2} + res) = b_{1} + Pb_{2} + (X_{1}^{'}X_{1})^{-1}X_{1}^{'}res = b_{1} + Pb_{2} \text{as}X_{1}^{'}res = 0 \text{because of orthogonery}$ 

(c) Prove:

$$b_R = b_1 + Pb_2$$

(d) Argue that the columns of the  $(2\times3)$  matrix P are obtained by regressing each of the variables 'Age', 'Educ', and 'Parttime' on a constant term and the variable 'Female'.

So,

$$P = (X_1'X_1)^{-1}X_1'(X_2)$$

is a  $(2\times3)$  matrix, with columns as such:

Column 1:

$$(X_1^{'}X_1)^{-1}X_1^{'}(Age)$$
the OLS formula for regressing 'Age' on X1.

Column 2:

$$(X_{1}^{'}X_{1})^{-1}X_{1}^{'}(Educ) {\rm the~OLS}$$
 formula for regressing 'Educ' on X1.

Column 3:

$$(X_{1}^{'}X_{1})^{-1}X_{1}^{'}(Parttime)$$
 the OLS formula for regressing 'Parttime' on X1.

(e) P matrix value.

```
file25 = "C:\\Users\\Gannon_chef\\Documents\\Pradeepta\\Coursera\\Econometrics\\TrainExer21.txt"
new_logwage<-read.table(file25, header = TRUE, sep = "", dec = ".")</pre>
str(new_logwage)
## 'data.frame':
                    500 obs. of 7 variables:
##
   $ Observ. : int 1 2 3 4 5 6 7 8 9 10 ...
            : int 66 34 70 47 107 188 123 57 42 200 ...
## $ LogWage : num 4.19 3.53 4.25 3.85 4.67 ...
## $ Female : int 0 1 1 0 1 1 1 1 1 0 ...
## $ Age
              : int 49 42 42 38 54 54 47 39 25 59 ...
## $ Educ
              : int 1 1 1 1 1 1 1 1 1 1 ...
## $ Parttime: int 1 1 1 0 1 0 0 1 0 0 ...
X1 = new_logwage['Female']
X1['Constant'] <-rep(1, length(new_logwage['Female']))</pre>
X1 \leftarrow X1[,c(2,1)]
X2 = new_logwage[c('Age', 'Educ', 'Parttime')]
str(X1)
## 'data.frame':
                    500 obs. of 2 variables:
   $ Constant: num 1 1 1 1 1 1 1 1 1 1 ...
## $ Female : int 0 1 1 0 1 1 1 1 1 0 ...
str(X2)
                    500 obs. of 3 variables:
## 'data.frame':
              : int 49 42 42 38 54 54 47 39 25 59 ...
## $ Age
              : int 1 1 1 1 1 1 1 1 1 ...
   $ Educ
## $ Parttime: int 1 1 1 0 1 0 0 1 0 0 ...
Now, value of P matrix can be computed as:
x1_mat = as.matrix(sapply(X1, as.numeric))
x2_mat = as.matrix(sapply(X2, as.numeric))
P<-solve(t(x1_mat)%*%(x1_mat))%*%t(x1_mat)%*%x2_mat)
##
                             Educ Parttime
                   Age
## Constant 40.0506329 2.2594937 0.1962025
            -0.1104155 -0.4931893 0.2494496
## Female
(f) Check the numerical validity of the result in part (c). Note: This equation will not hold
exactly because the coefficients have been rounded to two or three decimals; preciser results
would have been obtained for higher precision coefficients.
model <- lm(LogWage~Female+Age+Educ+Parttime, data=new_logwage)</pre>
summary(model)
##
## Call:
## lm(formula = LogWage ~ Female + Age + Educ + Parttime, data = new_logwage)
##
## Residuals:
##
        Min
                  1Q
                       Median
                                     3Q
## -0.73799 -0.15705 -0.00388 0.16495 0.77827
##
```

```
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 3.052697 0.055333 55.170
                                              <2e-16 ***
## Female
              -0.041073
                          0.024710 -1.662
                                              0.0971 .
## Age
               0.030606
                          0.001273 24.041
                                              <2e-16 ***
## Educ
               0.233163
                           0.010660 21.873
                                              <2e-16 ***
## Parttime
              -0.365407
                           0.031570 -11.575
                                              <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.2452 on 495 degrees of freedom
## Multiple R-squared: 0.704, Adjusted R-squared: 0.7016
## F-statistic: 294.3 on 4 and 495 DF, p-value: < 2.2e-16
b1=as.matrix(model$coeff[1:2])
b2=as.matrix(model$coeff[3:5])
b1+P%*%b2
##
                     [,1]
## (Intercept) 4.7336234
## Female
              -0.2505962
Now, we should be able to replicate this value if we regress only using the Female and intercept! If we do
```

x1\_model <- lm(LogWage~Female, data=new\_logwage)
x1\_model\$coeff</pre>

```
## (Intercept) Female
## 4.7336234 -0.2505962
```

We can see that the values match exactly!