# Homework 1

April 19, 2017

**Earle Aguilar** (804501476)

## Problem 1.0:

$$SSE = \sum_{i=1}^{n} e_i^2 = (Y - X\beta)'(Y - X\beta)$$
 (1.1)

$$= Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'X\hat{\beta}$$
(1.2)

(1.3)

Substitude  $\hat{\beta} = (X'X)^{-1}X'Y$  to the last term

$$SSE = Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'X((X'X)^{-1}X'Y)$$
(1.4)

$$= Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'Y \tag{1.5}$$

$$SSE = Y'Y - \hat{\beta}'X'Y$$
 (1.6)

$$SSE = Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'X((X'X)^{-1}X'Y)$$
(1.7)

$$= Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'HY \tag{1.8}$$

$$SSE = Y'Y - \hat{\beta}'X'\hat{Y}$$
(1.9)

$$SSE = Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'X((X'X)^{-1}X'Y)$$
(1.10)

$$= Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'Y \tag{1.11}$$

$$SSE = Y'Y - \hat{\beta}'X'Y \tag{1.12}$$

Substitude normal equations  $X'Y = X'X\hat{\beta}$ 

$$SSE = Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'X((X'X)^{-1}X'Y)$$
(1.13)

$$= Y'Y - Y'X\hat{\beta} - \hat{\beta}'X'Y + \hat{\beta}'X'Y$$
 (1.14)

$$SSE = Y'Y - \hat{\beta}'X'X\hat{\beta}$$
(1.15)

#### Problem 2.0:

$$X = \begin{pmatrix} 1 & x_{11} & x_{12} & \cdots & .001x_{1i} & \cdots & x_{1k} \\ 1 & x_{21} & x_{22} & \cdots & .001x_{2i} & \cdots & x_{2k} \\ \vdots & \vdots & \vdots & \ddots & \vdots & & \vdots \\ 1 & x_{i1} & x_{i2} & \cdots & .001x_{ii} & \cdots & x_{ik} \\ \vdots & \vdots & \vdots & & \vdots & \ddots & \vdots \\ 1 & x_{n1} & x_{n2} & \cdots & .001x_{ni} & \cdots & x_{nk} \end{pmatrix}$$

$$(2.16)$$

$$X'X = \begin{pmatrix} n & \sum x_{i1} & \sum x_{i2} & \cdots & .001 \sum x_{ji} & \cdots & \sum x_{ik} \\ \sum x_{i1} & \sum x_{i1}^{2} & \sum x_{i1}x_{i2} & \cdots & .001 \sum x_{i1} \sum x_{ji} & \cdots & \sum x_{i1}x_{ik} \\ \sum x_{i2} & \sum x_{i1}x_{i2} & \sum x_{i2}^{2} & \cdots & .001 \sum x_{i2} \sum x_{ji} & \sum x_{i2}x_{ik} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ .001 \sum x_{ji} & .001 \sum x_{i1} \sum x_{ji} & .001 \sum x_{i2} \sum x_{ji} & \cdots & (.001)^{2} \sum x_{ii}^{2} & \cdots & .001 \sum x_{ik}x_{ji} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \sum x_{ik} & \sum x_{i1}x_{ik} & \sum x_{i1}x_{ik} & \cdots & .001 \sum x_{ik}x_{ji} & \cdots & \sum x_{ik}^{2} \end{pmatrix}$$

$$\hat{\beta}_{i_{new}} = ((.001)^2 X'_{ii} X_{ii})^{-1} (.001 X'_{ii}) Y$$
(2.18)

$$\hat{\beta}_{i_{new}} = 1000(X'_{ii}X_{ii})^{-1}X'_{ii}Y$$
(2.19)

$$\hat{\beta}_{i_{new}} = 1000\hat{\beta}_i \tag{2.20}$$

$$Var(\hat{\beta}_{i_{new}}) = Var(1000\hat{\beta}_{i})$$

$$= (1000)^{2} (\sigma(X'_{ii}X_{ii})^{-1})$$
(2.21)

$$= (1000)^{2} (\sigma(X'_{ii}X_{ii})^{-1}) \tag{2.22}$$

### Problem 3.0:

```
library(MASS)
a <- read.table("http://www.stat.ucla.edu/~nchristo/statistics100c/restaurant.txt",
                header = T)
x \leftarrow matrix(c(rep(1,100),a\$food,a\$decor, a\$ser), nrow=100, ncol=4)
x_transpose_x <- (t(x) %*% x) ;x_transpose_x</pre>
        [,1] [,2] [,3] [,4]
## [1,] 100 1943 1609 1764
## [2,] 1943 38445 31764 34722
## [3,] 1609 31764 27671 29108
## [4,] 1764 34722 29108 31748
y <- matrix(a$cost)</pre>
beta_hat = ginv(x_transpose_x) %*% t(x) %*% y;beta_hat
##
               [,1]
## [1,] -35.3098383
## [2,] 1.2838518
        1.7773844
## [3,]
## [4,] 0.2804128
lm(cost ~ ., data= a)
##
## Call:
## lm(formula = cost ~ ., data = a)
##
## Coefficients:
## (Intercept)
                       food
                                   decor
                                                   ser
      -35.3098
                     1.2839
                                   1.7774
                                                0.2804
H <- x %*% ginv(x_transpose_x) %*% t(x)</pre>
H[1:5,1:5]
              [,1]
                          [,2]
                                       [,3]
                                                   [,4]
## [1,] 0.02333445 0.017090141 0.010247706 0.030288494 0.008920920
## [2,] 0.01709014 0.030828287 0.009747574 0.007121384 0.012518985
## [3,] 0.01024771 0.009747574 0.013348148 0.008131717 0.015062149
## [4,] 0.03028849 0.007121384 0.008131717 0.053772247 0.001936585
## [5,] 0.00892092 0.012518985 0.015062149 0.001936585 0.018606756
```

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#### Problem 4.0:

$$(Y - X\hat{\beta} + X\hat{\beta} - X\beta)'(Y - X\hat{\beta} + X\hat{\beta} - X\beta) \tag{4.23}$$

$$((Y - X\hat{\beta}) + X(\hat{\beta} - \beta))'((Y - X\hat{\beta}) + X(\hat{\beta} - \beta)) \tag{4.24}$$

$$(Y - X\hat{\beta})'(Y - X\hat{\beta}) + (Y - X\hat{\beta})'X(\hat{\beta} - \beta) + (\hat{\beta} - \beta)'X'(Y - X\hat{\beta}) + (\hat{\beta} - \beta)'X'X(\hat{\beta} - \beta)$$
(4.25)

$$(Y - X\hat{\beta})'(Y - X\hat{\beta}) + (Y - \hat{Y})'X(\hat{\beta} - \beta) + (\hat{\beta} - \beta)'X'(Y - \hat{Y}) + (\hat{\beta} - \beta)'X'X(\hat{\beta} - \beta)$$
(4.26)

$$(Y - X\hat{\beta})'(Y - X\hat{\beta}) + (\hat{\beta} - \beta)'X'X(\hat{\beta} - \beta)$$

$$(4.27)$$

#### Problem 5.0:

$$E[\hat{Y}'\hat{Y}] = E[(HY)'HY] = E[Y'H'HY]$$
(5.28)

$$= E[tr(Y'HY)] = tr(HE[Y'Y])$$
(5.29)

$$= tr(H(\sigma^2 I + X\beta\beta' X')) = tr(\sigma^2 H + X\beta\beta' X')$$
(5.30)

$$= tr(\sigma^2 H) + tr(X\beta\beta' X') \tag{5.31}$$

$$E[\hat{Y}'\hat{Y}] = \sigma(K+1) + \beta'X'X\beta \tag{5.32}$$

### Problem 6.0:

$$Q = \begin{pmatrix} \hat{\beta}_1 - 2\hat{\beta}_2 + 3\hat{\beta}_4 \\ \hat{\beta}_0 + \hat{\beta}_4 + 3\hat{\beta}_5 \end{pmatrix}$$
 (6.33)

$$= \begin{pmatrix} 0 & 1 & -2 & 0 & 3 & 0 \\ 1 & 0 & 0 & 0 & 1 & 3 \end{pmatrix} \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \\ \hat{\beta}_3 \\ \hat{\beta}_4 \\ \hat{\beta}_5 \end{pmatrix}$$
(6.34)

$$Q = c\hat{\beta} \tag{6.35}$$

We know that  $\hat{\beta} \sim N_6(\beta, \sigma(X'X)^{-1})$  then,

$$E[Q] = E[c\hat{\beta}] = cE[\hat{\beta}] \tag{6.36}$$

$$E[Q] = c\beta \tag{6.37}$$

$$Var(Q) = Var(c\hat{\beta}) \tag{6.38}$$

$$Var(Q) = c(\sigma(X'X)^{-1})c'$$
 (6.39)

$$Q \sim N_2(c\beta, \sigma c(X'X)^{-1}c')$$
(6.40)

### Problem 7.0:

 $\mathbf{a}$ 

$$\hat{\beta}_1 = (X_1' X_1)^{-1} X_1' Y_1 \tag{7.41}$$

$$\hat{\beta}_2 = (X_2' X_2)^{-1} X_2' Y_2 \tag{7.42}$$

b

$$\hat{\beta} = \left( \begin{pmatrix} X_1' & X_2' \end{pmatrix} \begin{pmatrix} X_1 \\ X_2 \end{pmatrix} \right)^{-1} \begin{pmatrix} X_1' & X_2' \end{pmatrix} \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix}$$
 (7.43)

$$= \left(X_1'X_1 + X_2'X_2\right)^{-1} \left(X_1' \quad X_2'\right) \begin{pmatrix} Y_1 \\ Y_2 \end{pmatrix} \tag{7.44}$$

$$\hat{\beta} = \left(X_1' X_1 + X_2' X_2\right)^{-1} \left(X_1' Y_1 + X_2' Y_2\right) \tag{7.45}$$

 $\mathbf{c}$ 

$$\hat{\beta} = (2X_1'X_1)^{-1}(X_1'Y_1 + X_2'Y_2) \tag{7.46}$$

$$= \frac{1}{2} (X_1' X_1)^{-1} (X_1' Y_1 + X_2' Y_2)$$
 (7.47)

$$= \frac{1}{2} [(X_1'X_1)^{-1}X_1'Y_1 + (X_1'X_1)^{-1}X_2'Y_2]$$
(7.48)

$$= \frac{1}{2} [(X_1'X_1)^{-1}X_1'Y_1 + (X_2'X_2)^{-1}X_2'Y_2]$$
(7.49)

$$\hat{\beta} = \frac{1}{2}(\hat{\beta}_1 + \hat{\beta}_2) \tag{7.50}$$