

Homework 2

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Problem 1.0:

$$Q = \begin{pmatrix} \hat{\beta} \\ \hat{Y} \end{pmatrix} = \begin{pmatrix} (X'X)^{-1}X'Y \\ (I-H)Y \end{pmatrix} \quad (1.1)$$

$$Q = LY \quad (1.2)$$

$$Var(Q) = Var(LY) = LVar(Y)L' \quad (1.3)$$

$$= \sigma^2 LL' \quad (1.4)$$

$$= \sigma^2 \begin{pmatrix} (X'X)^{-1}X' \\ (I-H) \end{pmatrix} \begin{pmatrix} (X'X)^{-1}X' & (I-H) \end{pmatrix} \quad (1.5)$$

$$= \sigma^2 \begin{pmatrix} (X'X)^{-1} & (X'X)^{-1}X'(I-H) \\ (I-H)X(X'X)^{-1} & (I-H) \end{pmatrix} \quad (1.6)$$

$$Var(Q) = \sigma^2 \begin{pmatrix} (X'X)^{-1} & 0 \\ 0 & (I-H) \end{pmatrix} \quad (1.7)$$

Therefore \hat{Y} and e are independent.

Problem 2.0:

$$\hat{\beta}_c = \hat{\beta} - (X'X)^{-1}c'[c(X'X)^{-1}c']^{-1}(c\hat{\beta} - \gamma) \quad (2.8)$$

$$(2.9)$$

let $A = (X'X)^{-1}c'[c(X'X)^{-1}c']^{-1}$

$$Var(\hat{\beta}_c) = Var(\hat{\beta}) - Var(A(c\hat{\beta} - \gamma)) \quad (2.10)$$

$$Var(\hat{\beta}_c) = Var(\hat{\beta}) - AVar(c\hat{\beta} - \gamma)A' \quad (2.11)$$

$$Var(\hat{\beta}_c) = Var(\hat{\beta}) - A[cVar(\hat{\beta})c' - Var(\gamma)]A' \quad (2.12)$$

$$Var(\hat{\beta}_c) = Var(\hat{\beta}) - AcVar(\hat{\beta})c'A' + AVar(\gamma)A' \quad (2.13)$$

$$Var(\hat{\beta}_c) = Var(\hat{\beta}) - AcVar(\hat{\beta})c'A' + AVar(c\beta)A' \quad (2.14)$$

$$Var(\hat{\beta}_c) = \sigma^2(X'X)^{-1} - Ac[\sigma^2(X'X)^{-1}]c'A' \quad (2.15)$$

$$Var(\hat{\beta}_c) = \sigma^2(X'X)^{-1} - \sigma^2(X'X)^{-1}c'[c'(X'X)^{-1}c]^{-1}c(X'X)^{-1} \quad (2.16)$$

Problem 3.0:

We have that $S^2 = \frac{\sum_{i=1}^n (y_i - \bar{y})^2}{n-1}$ and $S_e^2 = \frac{e'e}{n-k-1} = \frac{\sum_{i=1}^n (y_i - \bar{y})^2 - \sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{n-k-1}$

$$R_a^2 = \frac{S^2 - s_e^2}{S^2} \quad (3.17)$$

$$= \frac{n-1}{\sum_{i=1}^n (y_i - \bar{y})^2} \left[\frac{\sum_{i=1}^n (y_i - \bar{y})^2}{n-1} - \frac{\sum_{i=1}^n (y_i - \bar{y})^2}{n-k-1} + \frac{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{n-k-1} \right] \quad (3.18)$$

$$= \frac{n-1}{\sum_{i=1}^n (y_i - \bar{y})^2} \left[\frac{(n-k-1) \sum_{i=1}^n (y_i - \bar{y})^2 - (n-1) \sum_{i=1}^n (y_i - \bar{y})^2 + (n-1) \sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{(n-1)(n-k-1)} \right] \quad (3.19)$$

$$= \left[\frac{(n-k-1) - (n-1) + (n-1) \frac{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{\sum_{i=1}^n (y_i - \bar{y})^2}}{n-k-1} \right] \quad (3.20)$$

$$= 1 - \frac{(n-1) - (n-1)R^2}{n-k-1} \quad (3.21)$$

$$\boxed{R_a^2 = 1 - \frac{n-1}{n-k-1}(1-R^2)} \quad (3.22)$$

Problem 4.0:

a

$$\text{var} \left(\frac{(n-k-1)S_e^2}{\sigma^2} \right) = 2(n-k-1) \quad (4.23)$$

$$\frac{(n-k-1)^2}{\sigma^4} \text{Var}(S_e^2) = 2(n-k-1) \quad (4.24)$$

$$\frac{n-k-1}{\sigma^4} \text{Var}(S_e^2) = 2 \quad (4.25)$$

$$\text{Var}(S_e^2) = \frac{2\sigma^4}{n-k-1} \quad (4.26)$$

b

c

Problem 5.0:

$$(Y - Xc)'(Y - Xc) - (Y - X\hat{\beta})'(Y - X\hat{\beta}) \quad (5.27)$$

$$Y'Y - Y'Xc - c'X'Y + c'X'Xc - Y'Y - Y'Y + Y'X\hat{\beta} + \hat{\beta}'X'Y - \hat{\beta}'X'X\hat{\beta} \quad (5.28)$$

$$(5.29)$$

We use the normal equations to substitute for $Y'X$ and $X'Y$.

$$-\hat{\beta}'X'Xc - c'X'X\hat{\beta} + c'X'Xc + \hat{\beta}X'X\hat{\beta} + \hat{\beta}'X'X\hat{\beta} - \hat{\beta}X'X\hat{\beta} \quad (5.30)$$

$$c'X'Xc - \hat{\beta}'X'Xc - c'X'X\hat{\beta} + \hat{\beta}'X'X\hat{\beta} \quad (5.31)$$

$$(c'X'X - \hat{\beta}'X'X)(c - \hat{\beta}) \quad (5.32)$$

$$(c' - \hat{\beta}')X'X(c - \hat{\beta}) \quad (5.33)$$

$$\boxed{(c - \hat{\beta})'X'X(c - \hat{\beta})} \quad (5.34)$$

Problem 6.0:

a

1

$$\hat{\beta} = \begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix} = (X'X)^{-1}X'Y \quad (6.35)$$

$$= \begin{pmatrix} 2 & 0 & 0 \\ 0 & 3 & -1 \\ 0 & -1 & 1 \end{pmatrix} \begin{pmatrix} 4 \\ -2 \\ 5 \end{pmatrix} \quad (6.36)$$

$$\begin{pmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \\ \hat{\beta}_2 \end{pmatrix} = \begin{pmatrix} 8 \\ -11 \\ 7 \end{pmatrix} \quad (6.37)$$

2

$$\sigma^2(X'X)^{-1} = \begin{pmatrix} 2\sigma^2 & 0 & 0 \\ 0 & 3\sigma^2 & -\sigma^2 \\ 0 & -\sigma^2 & \sigma^2 \end{pmatrix} \quad (6.38)$$

Therefore $Cov(\hat{\beta}_1, \hat{\beta}_2) = 0$

b

$$X = \begin{pmatrix} 1 & 3 & 1 & -1 & 1 \\ 1 & 4 & 1 & 1 & -1 \\ 1 & 5 & -1 & 1 & 1 \\ 1 & 6 & 0.5 & 0.2 & 0.3 \\ 1 & 8 & 0.8 & 0.1 & 0.1 \\ 1 & 9 & 0.3 & 0.5 & 0.2 \\ 1 & 10 & 0.2 & 0.3 & 0.5 \\ 1 & 13 & 0.1 & 0.6 & 0.3 \end{pmatrix} \quad (6.39)$$

$x_2 + x_3 + x_4 = 1 \implies$ matrix is not full rank and we cannot invert $X'X$.

Problem 7.0:

a

$E[\hat{y}_{n+1}]$:

$$E[\hat{y}_{n+1}] = E[x'_{n+1}\hat{\beta}] \quad (7.40)$$

$$= x'_{n+1}E[\hat{\beta}] \quad (7.41)$$

$$E[\hat{y}_{n+1}] = x'_{n+1}\beta \quad (7.42)$$

$E[y_{n+1}]$:

$$E[y_{n+1}] = E[x'_{n+1}\beta + \varepsilon_{n+1}] = x'_{n+1}\beta \quad (7.43)$$

b

$$E(\tilde{y}_{n+1} - y_{n+1}) = 0 \quad (7.44)$$

$$E[\tilde{y}_{n+1}] - E[y_{n+1}] = 0 \quad (7.45)$$

$$a' E[Y] - x'_{n+1}\beta = 0 \quad (7.46)$$

$$a' X\beta - x'_{n+1}\beta = 0 \quad (7.47)$$

$$(a' X - x'_{n+1})\beta = 0 \implies a' X = x'_{n+1} \quad (7.48)$$

c

$$Var(\hat{y}_{n+1}) = Var(x'_{n+1}\hat{\beta}) \quad (7.49)$$

$$= x'_{n+1}Var(\hat{\beta})x_{n+1} \quad (7.50)$$

$$= x'_{n+1}[\sigma^2(X'X)^{-1}]x_{n+1} \quad (7.51)$$

$$\boxed{Var(\hat{y}_{n+1}) = \sigma^2 x'_{n+1}(X'X)^{-1}x_{n+1}} \quad (7.52)$$

$$\boxed{Var(\tilde{y}_{n+1}) = Var(a'Y) = \sigma^2 a'a} \quad (7.53)$$

d

$$Var(\tilde{y}_{n+1}) - Var(\hat{y}_{n+1}) = 0 \quad (7.54)$$

$$\sigma^2 a'a - \sigma^2 x'_{n+1}(X'X)^{-1}x_{n+1} = \sigma^2 a'a - \sigma^2 a'x(X'X)^{-1}x'a \quad (7.55)$$

$$\sigma^2 a'(I - X(X'X)^{-1}X')a = \sigma^2 a'(I - H)a \quad (7.56)$$

$$\sigma^2 a'(I - H)(I - H)'a = \sigma^2 ((I - H)'a)'(I - H)'a \quad (7.57)$$

We notice that $((I - H)'a)'(I - H)'a$ is of the form $(M'l)'(M'l)$ which implies that $(I - H)$ is at least postive semidefinite so then

$$\sigma^2 ((I - H)'a)'(I - H)'a \geq 0 \quad (7.58)$$

which implies that

$$Var(\tilde{y}_{n+1}) - Var(\hat{y}_{n+1}) \geq 0 \quad (7.59)$$

$$Var(\tilde{y}_{n+1}) \geq Var(\hat{y}_{n+1}) \quad (7.60)$$

Problem 8.0:

a

$$e'e = ((I - H)Y)'((I - H)Y) \quad (8.61)$$

$$= Y'(I - H)'(I - H)Y \quad (8.62)$$

$$\boxed{e'e = Y'(I - H)Y} \quad (8.63)$$

b

$$tr[(I - H)E[Y'Y]] = tr[(I - H)(\sigma^2 I + X\beta\beta'X')] \quad (8.64)$$

$$= tr[\sigma^2 I + X\beta\beta'X' - \sigma^2 H - HX\beta\beta'X'] \quad (8.65)$$

$$= \sigma^2 tr(I) - \sigma^2 tr(H) \quad (8.66)$$

$$\boxed{tr[(I - H)E[Y'Y]] = \sigma^2(n - k - 1)} \quad (8.67)$$

c

$$Var(e) = Var((I - H)Y) \quad (8.68)$$

$$= \sigma^2(I - H)(I - H)' \quad (8.69)$$

$$= \sigma^2(I - H) \quad (8.70)$$

$$Var(e) = \sigma^2(I - X(X'X)^{-1}X') \quad (8.71)$$

Therefore $Var(e_i) = \sigma^2(1 - x_i'(X'X)^{-1}x_i)$

d

$$Var(\hat{Y}) = Var(HY) \quad (8.72)$$

$$= HVar(Y)H' \quad (8.73)$$

$$= H(\sigma^2 I)H' \quad (8.74)$$

$$= \sigma^2 H \quad (8.75)$$

$$Var(\hat{Y}) = \sigma^2 X(X'X)^{-1}X' \quad (8.76)$$

Therefore $Var(e_i) = \sigma^2 - Var(\hat{Y})$