

PS 7 Hangyu Huang

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Problem 1

The 1000 estimated beta creates a empirical distribution for beta. So calculate the standard deviation, $\text{sd}(\beta)$ from the 1000 estimators and compare it with the mean of the standard errors.

Problem 2

$$\text{set } y = P^T Z \cdot \beta \quad \|y\|_2 = 1$$

$$\|y\|_2 = \sqrt{y^T y} = \sqrt{(P^T \beta)^T (P^T \beta)} = \sqrt{\beta^T (P^T P) \beta}$$

$$= \sqrt{\beta^T \beta} \quad (\text{As } P^T P = I)$$

$$\text{For a diagonal matrix } D = \begin{bmatrix} \sigma_1 & & \\ & \ddots & 0 \\ 0 & \cdots & \sigma_n \end{bmatrix} \quad y = \begin{bmatrix} y_1 \\ \vdots \\ y_n \end{bmatrix} \quad y_1 + y_2 + \dots + y_n = 1$$

$$\text{we have } y^T D y = \sigma_1^2 y_1^2 + \sigma_2^2 y_2^2 + \dots + \sigma_n^2 y_n^2 \leq 1,$$

$$\text{As we know } A = A^T = P \Lambda P^T$$

$$\begin{aligned} \|A\|_2 &= \sqrt{(A^T)(A)} = \sqrt{z^T (A^T A) z} \\ &= \sqrt{z^T (P \Lambda P^T P \Lambda P^T) z} \\ &= \sqrt{z^T P \Lambda^2 P^T z} \end{aligned}$$

As Λ is a diagonal matrix with eigenvectors of A

$$\text{we have } \|A\|_2 = \sqrt{q^T \Lambda^2 q} \quad q = z^T P$$

$$\begin{aligned} &= \sqrt{\sigma_1^2 q_1^2 + \sigma_2^2 q_2^2 + \dots + \sigma_n^2 q_n^2} \\ &\leq \sqrt{\sigma_1^2} = |\sigma_1| \end{aligned}$$

Problem 3

For a rectangular matrix $X (n \times p)$

$$X = UDV^T \quad U: n \times n \quad D: n \times p \text{ diagonal matrix}$$
$$V: p \times p$$

$$\begin{aligned} X^T X &= (UDV^T)^T (UDV^T) \\ &= VD^T U^T U D V^T \\ &= VD^T D V^T \quad (U^T U = I) \end{aligned}$$

As D is a diagonal matrix,

$D^T D$ is a diagonal matrix, whose first k diagonal entries are σ_i^2 .
 $X^T X = V(D^T D)V^T$ is the EVD of $X^T X$, V is the eigenvectors of the matrix $X^T X$ and the eigenvalue of $X^T X$ is the square of the singular value of X , which is σ_i^2 .

let $S = X^T X = VD^T D V^T$ V is the eigenvectors of S .

We can guarantee their orthogonality. so that

$$V_j^T V_i = 0 \text{ for } j \neq i$$
$$(X^T V_j)^T (X^T V_i) = V_j^T X^T X V_i = V_j^T \sigma_i^2 V_i = \begin{cases} \sigma_i^2 & \text{if } j=i \\ 0 & \text{if } j \neq i \end{cases}$$

Thus, $X^T X$ is positive semidefinite.

b) $\Sigma = P \Lambda P^T$
 $C I = C P \Lambda P^T = P C \Lambda P^T \quad C = \begin{bmatrix} c & & & \\ & c & & \\ & & \ddots & \\ 0 & & \cdots & c \end{bmatrix} \quad \Lambda = \begin{bmatrix} \sigma_1^2 & & & \\ & \ddots & & \\ & & \ddots & \\ 0 & & \cdots & \sigma_n^2 \end{bmatrix}$

$$Z = \Sigma + C I = P \Lambda P^T + P C \Lambda P^T = P (\Lambda + C) P^T$$

so the eigenvalue of Z is $\sigma_i + c$
the eigenvectors of Z is P

It is $O(n)$ operation as shown.

Problem 4

a) $X = QR$ (using QR decomposition)

$$X^T = R^T Q^T$$

$$C = X^T X = R^T Q^T Q R \quad (Q^T Q = I) \quad [p \times p]$$

$$= R^T R$$

As R is a upper triangular matrix,

$C = R^T R$ is a upper triangular matrix as well.

① Calculate $d = X^T f$

② Calculate $e = C^{-1} d$ (using backsolve (C, d))

$$\begin{aligned}\hat{\beta} &= e + C^{-1} A^T (A C^{-1} A^T)^{-1} (-Ae + b) \\ &= e + (C^{-1} A^T)(A^T)^{-1} C A^{-1} (-Ae + b)\end{aligned}$$

③ Calculate $f = -Ae + b$

④ calculate $g = A^{-1} f$

⑤ Calculate $D = (C^{-1} A^T)^T$

calculate $E = \text{crossprod}(D)$ calculate Eg
Calculate from right to left to get $\hat{\beta}$

b) $X = \text{matrix}(\text{rnorm}(np), \text{nrow}=n)$

$$X.qr = qr(X)$$

$$Q = qr.Q(X.qr)$$

$$R = qr.R(X.qr)$$

$$C \leftarrow \text{crossprod}(R)$$

$$d \leftarrow \text{crossprod}(X, Y)$$

$$e \leftarrow \text{backsolve}(C, d)$$

$$f \leftarrow b - A \%*\% e$$

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e ← backsolve(C, a)
f ← b - A %*% e
g ← ihsolve(A) %*% f
h ← C %*% g
D ← backsolve(C, A, transpose = TRUE)
E ← crossprod(D)
j ← E %*% h
Beta-hat ← e + j

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Problem 5

$Z(Z^T Z)^{-1}Z$ is a 60 million by 60 million matrix. The computer ran out of memory, thus we can not do the calculation in two stages as given.

b) ① Calculate $A = (Z^T Z)^{-1} \Rightarrow A$ is a 630×630 matrix
 $\beta_0 \times e^7 \text{ by } e^7 \times \beta_0$

② Calculate $\hat{x}^T = (Z \Lambda Z^T X)^T = X^T Z \Lambda^{-1} Z^T \Rightarrow \hat{x}^T$ is $600 \times 6e^7$

③ Calculate $B = \hat{x}^T \hat{x} = \hat{x}^T Z \Lambda^{-1} (Z^T Z) A Z^T X$
 $= \hat{x}^T Z \Lambda^{-1} (A^{-1} A) Z^T X$

$$= \hat{x}^T Z \Lambda^{-1} Z^T X$$

$B \hat{x}^T$ is 600×600 matrix

④ calculate $C = \hat{x}^T X$ is 600×1 matrix

⑤ calculate $\beta - BC$ is 600×1 matrix