

Q2.

(a) $E(\varepsilon) = 0$.

$$E(\varepsilon) = E(Y) - E(\beta_0) - E(\beta^T X) = E(Y) - (\beta_0 + \beta^T E(X))$$

$$\beta_0 = E(Y) - \beta^T E(X) \quad \text{so} \quad E(\varepsilon) = E(Y) - [E(Y) - \beta^T E(X)] - \beta^T E(X)$$

b. $\text{cov}(\varepsilon, X) = 0 \quad \varepsilon = Y - \beta_0 - \beta^T X$

$$= \text{cov}(Y, X) - \text{cov}(\beta_0, X) - \beta^T \text{cov}(X, X) \quad \beta_0 \text{ constant.}$$

$$= \text{cov}(Y, X) - \beta^T \text{var}(X) \quad \beta = \text{var}(X)^{-1} \text{cov}(X, Y)$$

$$= \text{cov}(Y, X) - [\text{var}(X)^{-1} \text{cov}(X, Y)]^T \text{var}(X)$$

$$= \text{cov}(Y, X) - \text{cov}(X, Y)^T = 0$$

(c) No. $\text{cov}(\varepsilon, X) = 0$ does not mean they are independent.

$$\text{var}(\varepsilon | X) = \text{var}(Y - \beta_0 - \beta^T X | X)$$

$$= \text{var}(Y) - 0 - 0 = \text{var}(Y)$$

Since we don't have ε, X independent, so that $\text{var}(\varepsilon | X)$ is not constant with X .

(d) $E(\hat{\beta} | X^n)$

$$= (X^T X)^{-1} X^T E(\bar{Y} | X^n) \quad \bar{Y} = \bar{\varepsilon} + X\beta$$

$$= (X^T X)^{-1} X^T X\beta + (X^T X)^{-1} X^T E(\bar{\varepsilon} | X^n)$$

$$= \beta + (X^T X)^{-1} X^T E(\bar{\varepsilon} | X^n)$$

So we can see $E(\hat{\beta} | X^n) = \beta$ only when $E(\bar{\varepsilon} | X^n) = 0$.

(e) $\text{var}(\hat{\beta} | X^n)$

$$\text{var}(\bar{\varepsilon} | X^n) = \Sigma$$

$$= (X^T X)^{-1} X^T \text{var}(\bar{Y} | X^n) X (X^T X)^{-1}$$

$$= (X^T X)^{-1} X^T \text{var}(\bar{Y} | X^n) X (X^T X)^{-1}$$

$$= (X^T X)^{-1} X^T \cdot (\text{var}(\bar{\varepsilon} | X^n) + \text{var}(X\beta | X^n) + 2 \cdot \text{cov}(\bar{\varepsilon}, X\beta | X^n)) X (X^T X)^{-1}$$

$$= (X^T X)^{-1} X^T \cdot \Sigma X (X^T X)^{-1}$$

So when $\Sigma = \sigma^2 I$,

$$\text{var}(\hat{\beta} | X^n) = (X^T X)^{-1} X^T \sigma^2 I X (X^T X)^{-1}$$

$$= \sigma^2 (X^T X)^{-1}$$