

a1q1

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**q1** (a) Let  $\beta = (\beta_1, \beta_2, \dots, \beta_n)^T$ ,  $X = (x_1, x_2, \dots, x_n)^T$  and  $\Sigma$  be variance-covariance matrix.

$$\begin{aligned} E(\tilde{\beta}_{WLS}) &= E((X^T W X)^{-1} X^T W Y) \\ &= (X^T W X)^{-1} X^T W E(Y) \\ &= (X^T W X)^{-1} X^T W E(X\beta + \epsilon) \\ &= (X^T W X)^{-1} X^T W X E(\beta) + (X^T W X)^{-1} X W E(\epsilon) \\ &= (X^T W X)^{-1} X^T W X \beta + 0 \quad (E(\epsilon) = 0) \\ &= \beta \end{aligned}$$

- (b) Let  $W = \text{diag}(\frac{1}{g(x_1)}, \frac{1}{g(x_2)}, \dots, \frac{1}{g(x_n)})$ . So  $W^T = W$ .  
 Since  $\Sigma = \sigma^2 \text{diag}(g(x_1), g(x_2), \dots, g(x_n))$ , so  $\Sigma \times W = \sigma^2$

$$\begin{aligned}
 \text{Var}(\tilde{\beta}_{WLS}) &= \text{Var}((X^T W X)^{-1} X^T W Y) \\
 &= (X^T W X)^{-1} X^T W \times \text{Var}(Y) \times ((X^T W X)^{-1} X^T W)^T \\
 &= (X^T W X)^{-1} X^T W \times \text{Var}(X\beta + \epsilon) \times ((X^T W X)^{-1} X^T W)^T \\
 &= (X^T W X)^{-1} X^T W \times \text{Var}(\epsilon) \times W^T X (X^T W^T X)^{-1} \\
 &= (X^T W X)^{-1} X^T W \times \Sigma \times W^T X (X^T W^T X)^{-1} \\
 &= \sigma^2 (X^T W X)^{-1} X^T W X (X^T W^T X)^{-1} \\
 &= \sigma^2 (X^T W X)^{-1}
 \end{aligned}$$