

(A)

$$\begin{aligned} &= \left(E(x_{1i}x'_{1i}) \right)^{-1} E(x_{1i}y_i) \\ &= \left(E(x_{1i}x'_{1i}) \right)^{-1} E(x_{1i}(x'_{1i}\beta_1 + x'_{2i}\beta_2 + e_i)) \\ &= \beta_1 + \left(E(x_{1i}x'_{1i}) \right)^{-1} E(x_{1i}x'_{2i}) \beta_2 \end{aligned}$$

$$E(b_R) = \beta_1 + P\beta_2.$$

(B)

$$\begin{aligned} &= E[(b_R - \beta)(b_R - \beta)'] = E[(X'X)^{-1}X'\varepsilon\varepsilon'X(X'X)^{-1}] \\ &= (X'X)^{-1}X'E[\varepsilon\varepsilon']X(X'X)^{-1} = (X'X)^{-1}X'(\sigma^2 I)X(X'X)^{-1} \end{aligned}$$

$$\text{var}(b_R) = \sigma^2(X'_1X_1)^{-1}$$

(C)

In the restricted model (where y is regressed on $k - g$ regressors) the $(k - g) \times 1$ vector of least squares estimates is given by $b_R = (X'_1X_1)^{-1}X'_1y$. In the unrestricted model (where y is regressed on $k - g$ regressors and g additional regressors) and $k \times 1$ vector of least squares estimates is given by $b = (X'_1X_1)^{-1}X'_1y$.

Let b be decomposed in two parts as $b = (b'_1, b'_2)'$, where the $(k - g) \times 1$ vector b_1 corresponds to the regressors of the restricted model and b_2 to the g added regressors. Then, the relation between b_R and b_1 is given by $b_R = b_1 + Pb_2$

$$b_R = b_1 + Pb_2$$

(D)

Dependent: Logwage		
Variable	Coefficients	Std. Error
(Constant)	4.734	.024
Female	-.251	.040
R ²	.073	

Dependent: Logwage	
Variable	Coefficients
(Constant)	3.866
Female	-.248
Age	.022
R ²	.344

Dependent: Logwage	
Variable	Coefficients
(Constant)	3.866
Female	-.137
Edu	.339
R ²	.350

Dependent: Logwage	
Variable	Coefficients
(Constant)	3.866
Female	-.277
Partime	.106
R ²	.083

(E)

P = 4.734

(F)

4.734	==	3.866	+	40.051	X	0.022
0.024		- 0.248		- 0.110		