

Economics 471: Introductory Econometrics

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Midterm I – Answers

$$\begin{aligned}
 1. \quad (a) \quad E(\tilde{\beta}) &= E\left[\frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] = E\left[\frac{\sum_{i=1}^n y_i(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] = E\left[\frac{\sum_{i=1}^n (\beta x_i + u_i)(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] = E\left[\frac{\beta \sum_{i=1}^n x_i(x_i - \bar{x}) + \sum_{i=1}^n u_i(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] \\
 &= E\left[\frac{\beta \sum_{i=1}^n (x_i - \bar{x})(x_i - \bar{x}) + \sum_{i=1}^n u_i(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] = \beta + E\left[\frac{\sum_{i=1}^n u_i(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] = \beta + E\left[\frac{\sum_{i=1}^n u_i(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}\right] \\
 &= \beta + \frac{1}{n} \sum_{i=1}^n E[u_i(x_i - \bar{x})] = \beta
 \end{aligned}$$

(b) Given that $E(\tilde{\beta}) = \beta$, the bias is zero.

(c) This estimator is the same as that from class and hence the variance is $\frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2}$.

2. (a) $\ln(y_i) = \ln \alpha + \beta \ln x_i + u_i$

(b) $\ln \alpha$, β , and u_i , respectively

(c) $\hat{\alpha} = \overline{\ln(y)} - \hat{\beta} \overline{\ln(x)}$

(d)
$$\hat{\beta} = \frac{\sum_{i=1}^n (\ln y_i - \overline{\ln y})(\ln x_i - \overline{\ln x})}{\sum_{i=1}^n (\ln x_i - \overline{\ln x})^2}$$

(e) Given that we have logs on each side of the equation, the slope coefficient is interpreted as an elasticity.

3. (a) This is our left-hand-side variable, in this case test scores

(b) We estimate the model via ordinary least-squares

(c) Our sample size is $n = 3733$

(d) Our intercept parameter estimate $\hat{\alpha} = \bar{y} - \hat{\beta} \bar{x}$

(e) Our slope parameter estimate
$$\hat{\beta} = \frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sum_{i=1}^n (x_i - \bar{x})^2}$$

$$(f) \text{ se}(\hat{\alpha}) = \sqrt{\frac{\frac{\hat{\sigma}^2}{n} \sum_{i=1}^n x_i^2}{\sum_{i=1}^n (x_i - \bar{x})^2}}$$

$$(g) \text{ se}(\hat{\beta}) = \sqrt{\frac{\frac{\hat{\sigma}^2}{n}}{\sum_{i=1}^n (x_i - \bar{x})^2}}$$

$$(h) R^2 = \frac{SSE}{SST} = 1 - \frac{SSR}{SST}$$

$$(i) \hat{\sigma} = \sqrt{\hat{\sigma}^2} = \sqrt{\frac{1}{n-2} SSR} = \sqrt{\frac{1}{n-2} \sum_{i=1}^n \hat{u}_i^2}$$

$$(j) SSR = \sum_{i=1}^n \hat{u}_i^2$$

$$(k) \hat{\sigma}_y = \sqrt{\frac{1}{(n-1)} \sum_{i=1}^n (y_i - \bar{y})^2}$$