

hw1

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Question 2

(a)

(b)

The estimates are given by the following formula:

$$\hat{\alpha}_1 = \frac{\sum_{i=1}^n \tilde{x}_i (y_i - \bar{y})}{\sum_{i=1}^n \tilde{x}_i^2}$$

and

$$\hat{\alpha}_0 = \bar{y}$$

because $\frac{1}{n} \sum_{i=1}^n \tilde{x}_i = 0$

(c)

We can write:

$$\begin{aligned} \hat{\alpha}_1 &= \frac{\sum_{i=1}^n \tilde{x}_i (y_i - \bar{y})}{\sum_{i=1}^n \tilde{x}_i^2} \\ &= \frac{(1/s_X) \times \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{(1/s_X)^2 \times \sum_{i=1}^n (x_i - \bar{x})^2} \\ &= s_X \times \frac{n \times s_{XY} \times s_Y}{n \times s_X^2 \times s_Y} \end{aligned}$$

Finally :

$$\hat{\alpha}_1 = s_Y \times \widehat{\rho(X, Y)}$$

(d)

By noting that $var(Y) = \sigma^2$, the sampling variance of these estimates is given by:

$$\begin{aligned} var(\hat{\alpha}_1) &= s_X^2 \times \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{\left(\sum_{i=1}^n (x_i - \bar{x})^2\right)^2} \times \sigma^2 \\ &= s_X^2 \times \frac{\sigma^2}{\sum_{i=1}^n (x_i - \bar{x})^2} \end{aligned}$$

Finally :

$$\boxed{var(\hat{\alpha}_1) = \frac{\sigma^2}{n}}$$

Then :

$$var(\hat{\alpha}_0) = var(\bar{y})$$

So then

$$\boxed{var(\hat{\alpha}_0) = \frac{\sigma^2}{n}}$$

Furthermore,

$$cov(\hat{\alpha}_0, \hat{\alpha}_1) = \sum \frac{\tilde{x}_i}{\tilde{x}_i^2} cov(y_i - \bar{y}, \bar{y}) \text{ with } cov(y_i, \bar{y}) = \frac{1}{n} \sigma^2 \text{ as } cov(y_i, y_k) = 0 \text{ when } i \neq k \text{ because samples are iid}$$

and also $cov(\bar{y}, \bar{y}) = \frac{1}{n} \sigma^2$.

So finally:

$$cov(\hat{\alpha}_0, \hat{\alpha}_1) = \sigma^2 \times \frac{\sum \tilde{x}_i}{\sum \tilde{x}_i^2} \times \left(\frac{1}{n} - \frac{1}{n}\right) \implies \boxed{cov(\hat{\alpha}_0, \hat{\alpha}_1) = 0}$$

(e)

(f)