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In a simple regression problem, we have $n=16$, $\bar{x}=10$, $S_x = \frac{1}{\sqrt{8}}$, and $S_e = 4$. At $x=11$, what is the value of T_0 such that $\text{prob}(\text{prediction error} > T_0) = 0.01$?

(Hint: how do we standardize prediction error?)

$$\text{prob}(\text{pred. err.} > T_0) = .01 \quad \left. \vphantom{\text{prob}(\text{pred. err.} > T_0) = .01} \right\} \text{standardize}$$

$$\text{prob}\left(\frac{\text{pred err}}{S_{\text{pred err}}} > \frac{T_0}{\sqrt{S_{\hat{y}}^2 + S_e^2}}\right) = .01$$

$$\begin{aligned} S_{\hat{y}} &= S_e \sqrt{\frac{1}{n} + \frac{(x-\bar{x})^2}{S_{xx}}} = S_e \sqrt{\frac{1}{n} + \frac{(x-\bar{x})^2}{(n-1)S_x^2}} = 4 \sqrt{\frac{1}{16} + \frac{(11-10)^2}{15(\frac{1}{8})}} \\ &\approx 4 \sqrt{\frac{1}{16} + \frac{8}{15}} \approx 4 \sqrt{\frac{9}{16}} = 3 \\ &\quad \rightarrow \approx \frac{8}{16} \end{aligned}$$

$$\text{prob}\left(t > \frac{T_0}{\sqrt{9+16}}\right) = .01 \Rightarrow \underline{T_0 = 2.6(5)}$$

$\underbrace{\quad}_{2.6} \leftarrow df = n-2 = 14, \text{Table VI}$

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Here is the regression version of a problem we have seen many times before. Show that, at a given x , the prob that a random y would fall into the obs CI for $y(x)$ is

$$pr(t_{obs} - t^* \sqrt{\frac{1}{n} + \frac{(x-\bar{x})^2}{S_{xx}}} < t < t_{obs} + t^* \sqrt{\frac{1}{n} + \frac{(x-\bar{x})^2}{S_{xx}}})$$

$$\text{where } t_{obs} = \frac{\hat{y}_{obs}(x) - y(x)}{S_e}$$

(Hint: How do you standardize observation error?)

$$\text{Obs CI for } y(x): \hat{y}_{obs}(x) \pm t^* S_{est.err}$$

$$pr(\hat{y}_{obs}(x) - t^* S_{est.err} < y < \hat{y}_{obs}(x) + t^* S_{est.err}) \quad \text{standardize}$$

$$= pr\left(\frac{\hat{y}_{obs}(x) - y(x) - t^* S_{est.err}}{S_e} < \frac{y - y(x)}{S_e} < \frac{\hat{y}_{obs}(x) - y(x) + t^* S_{est.err}}{S_e}\right)$$

$$\frac{t_{obs} - t^* \frac{S_e \sqrt{\frac{1}{n} + \frac{(x-\bar{x})^2}{S_{xx}}}}{S_e}}{S_e}$$

$$= pr(t_{obs} - t^* \sqrt{\frac{1}{n} + \frac{(x-\bar{x})^2}{S_{xx}}} < t < \dots)$$

b) Show that, at a given x , the prob that a random $\hat{y}(x)$ would fall into the obs CI for $y(x)$ is $pr(t_{obs} - t^* < t < t_{obs} + t^*)$

$$\text{where } t_{obs} = \frac{\hat{y}_{obs}(x) - y(x)}{S_{est.err}}$$

(Hint: how do you standardize estimation error?)

$$pr(\hat{y}_{obs}(x) - t^* S_{est.err} < \hat{y}(x) < \hat{y}_{obs}(x) + t^* S_{est.err})$$

$$= pr\left(\frac{\hat{y}_{obs}(x) - y(x) - t^* S_{est.err}}{S_{est.err}} < \frac{\hat{y}(x) - y(x)}{S_{est.err}} < \frac{\hat{y}_{obs}(x) - y(x) + t^* S_{est.err}}{S_{est.err}}\right)$$

$$= pr(t_{obs} - t^* < t < t_{obs} + t^*)$$