

(null hypothesis is rejected at level 5%)

Interpretation:

One additional year of studying causes increase in Earnings by 9.2 \$ (slope)

People with no studying have earnings 4.1 \$ (intercept)

P3 - $N = 90$

I would choose model 2 cause it has ~~no~~ bigger R^2 , it means that this model better describe variance of grades. Also, s.e. (H) in the second model is less, it is better (less variance and random)

~~16.4~~
~~0.366~~ Also, I don't believe that people with no practicing of studying econometrics can get more than a zero for the test (best specification is without intercept) personally

P.4. $N = 100$

$$E = \beta_0 + \beta_1 W + \beta_2 NW + u$$

\uparrow
365-W

It is impossible, cause there is a ~~strong~~ multicollinearity (NW is linear ~~exact~~ function of W), so $\beta_1 = 0$ then (It's impossible to define what effect was done, for example, for decreasing)

Of E : decrease in W or increase NW : and
 what effects W and NW have on E
 (not can not explain variation in E : are they
 the same meaning attributable to variations in W or variations in NW
 or both)

$$\hat{\beta}_2 = \frac{\sum (E_i - \bar{E}) \text{ something } (NW_i - \bar{NW})}{\sum (W_i - \bar{W}) \sum (NW_i - \bar{NW})^2 - (\sum (W_i - \bar{W}) (NW_i - \bar{NW}))^2}$$

$$= \text{something} \frac{(\sum (W_i - \bar{W})^2)^2 - (\sum (W_i - \bar{W}) (W_i + \bar{W}))^2}{\text{something}}$$

if researcher do nothing with multicollinearity
 then he couldn't fit the model.

It he will fight multicollinearity,
 for example, by dropping NW from regression,

then it $u \sim N(0, \sigma^2)$ let $\beta_0 = k$, $\beta_1 = c$.

If u now $\sim N(a, \sigma^2)$, then $\beta_1 = k + a$, β_1 remain
 the same: cause intuitively: now: $y = \beta_0 + \beta_1 W + a + \epsilon$ new random component

was: $E = \beta_0 + \beta_1 W + u \Rightarrow$ then $E = \beta_0 + \beta_1 W + a + \epsilon$
 $y^* = y + a = (\beta_0 + a) + \beta_1 W + \epsilon$ now y^* is new
 P.S. $\hat{\beta} = (X^T X)^{-1} X^T y$ or $\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = \bar{y} + a - \hat{\beta}_1 \bar{x} = \beta_0 + a$

$$E(\hat{\beta}) = E((X^T X)^{-1} X^T y) = (X^T X)^{-1} X^T E(X\beta + u) =$$

$$= (X^T X)^{-1} X^T X \beta = \beta \text{ g.e.d.}$$

$$\text{Var}(y) = \text{Var}(X\beta + u) = \text{Var}(u) = \sigma_u^2$$

$$\text{Var}(\hat{\beta}) = \text{Var}((X^T X)^{-1} X^T y) = (X^T X)^{-1} X^T \text{Var}(y) (X^T X)^{-1}$$

$$= \sigma_u^2 (X^T X)^{-1} X^T X (X^T X)^{-1} = \sigma_u^2 (X^T X)^{-1}$$

P.S. $\hat{\beta}_1 = \frac{\sum (X_i - \bar{X})(y_i + a - \bar{y} - a)}{\sum (X_i - \bar{X})^2} = \hat{\beta}_1$
 to P.S. $\hat{\beta}_1$ remain the same.