

This time: regression

Next time: ANOVA

No HW#5

HW#4 is due Friday June 9, 2017 @ 11:59 on canvas

There will be either 1 or 2 makeup lectures in finals week, on Monday (+ Tuesday) more details soon

Take home final given out Friday June 9, 2017 during class and will be due a week later

Q<sub>1</sub>: When is a regression slope  $\hat{\beta}_1$ , large in practical terms (practsig)?

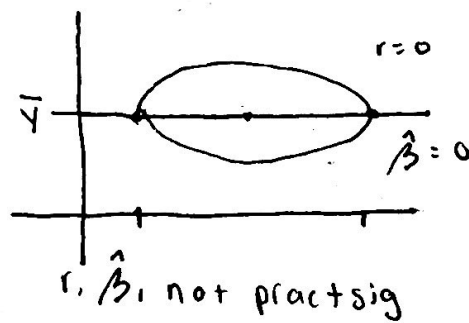
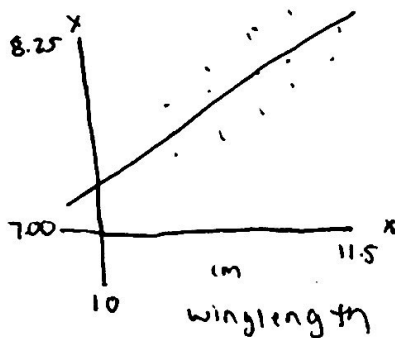
A<sub>1</sub>: Same method as in answering the question

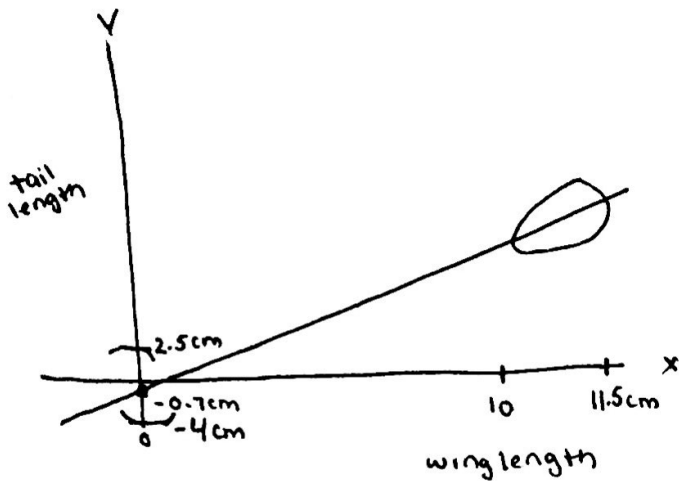
Q<sub>2</sub>: When is  $r$  a correlation coefficient in practical terms

A<sub>2</sub> = A<sub>1</sub>

Smallest birds have  $x = \text{wing length} = 10 \text{ cm}$ ,  $\hat{y} = \text{predicted tail length} = 7.0 \text{ cm}$ ; longest have  $x = 11.5$ ,  $\hat{y} = 8.25 \text{ cm}$ .

Since 8.25 cm differs from 7.0 cm by an amount that's practsig, the correlation  $r$  + slope  $\hat{\beta}_1$ , associated w/ predicting  $y$  from  $x$  are also ~~practsig~~ practsig.

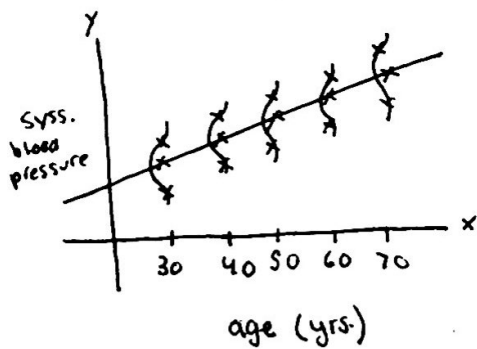




often, y-int. involves extrapolation away from data;  
it therefore may be meaningless

~~Another way~~

Another way



~~$y_i = \beta_0 + \beta_1 x_i + e_i$~~

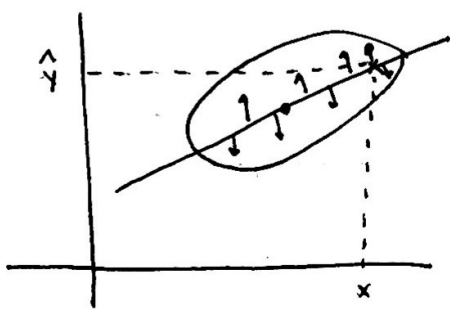
$$y_i = \beta_0 + \beta_1 x_i + e_i$$

obs = "truth" + "error"

$$y_i = (\hat{\beta}_0 + \hat{\beta}_1 x_i) + \hat{e}_i$$

obs = predicted + residual

residual  $\hat{e}_i = y_i - (\hat{y}_i)$   
 $= y_i - (\hat{\beta}_0 + \hat{\beta}_1 x_i)$



$$\hat{y} = \beta_0 + \beta_1 x$$

If available,  
 $\hat{y}_x = \hat{\beta}_0 + \hat{\beta}_1 x$   
 $SE(\hat{y}_x) = S_{y|x}$   
 residuals = RMSE =

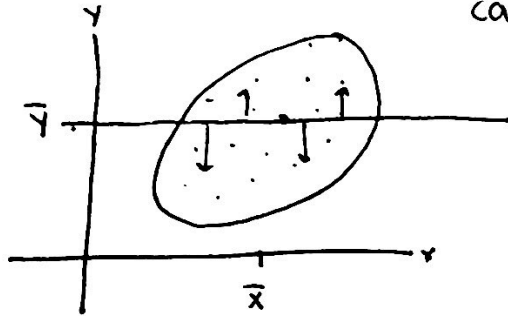
% of variance in  $y^*$  is "explained by" the regression of  $y$  on  $x$

↓ change to  
"associated with" (better)

$r^2 =$  coefficient of determination

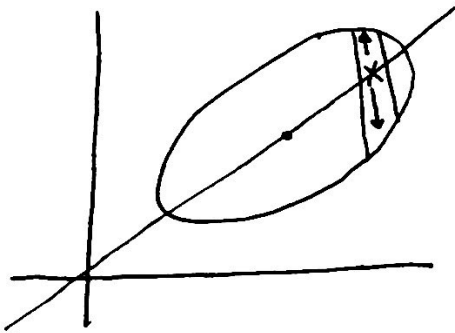
big  $r^2 =$  useful  
↳ big enough to matter

method 2 ↓



case (A) do not have  $x$  or choose to ignore  $x$

best  $\hat{y} = \bar{y} + SE(\hat{y}) = S_y$



case (B) use  $x$  to predict  $y$

now best  $\hat{y} = \hat{\beta}_0 + \beta_1 x$   
and  $SE(\hat{y}) = S_{y|x} = \sqrt{1-r^2}$

regression is successful if  $S_y \sqrt{1-r^2}$  is smaller than  $S_y$  by an amount that's ~~mean~~ meaningful

One way to quantify this is

$$\frac{\text{ignore } x \quad S_y \quad - \quad \text{use } x \quad S_y \sqrt{1-r^2}}{\text{ignore } x \quad S_y} = 1 - \sqrt{1-r^2}$$

better measure of success of regression than  $r^2$