

Review:

Notation:  $S_{XX} = \sum_{i=1}^n (x_i - \bar{x})^2 \stackrel{(1)}{=} \sum_{i=1}^n x_i^2 - n\bar{x}^2$

$S_{XY} = \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y}) \stackrel{(2)}{=} \sum_{i=1}^n x_i y_i - n\bar{x}\bar{y}$

$S_{YY} = \sum_{i=1}^n (y_i - \bar{y})^2 \stackrel{(3)}{=} \sum_{i=1}^n y_i^2 - n\bar{y}^2$

$(x_1, y_1), (x_2, y_2) \dots (x_n, y_n)$

$\Rightarrow$  fit a least square regression line  
simple

$$\hat{y} = b_0 + b_1 x$$

$b_0$  and  $b_1$  are the estimators

of  $\beta_0$  &  $\beta_1$  where  $y = \beta_0 + \beta_1 x + \varepsilon$   
is the true regression model

$$b_1 = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = \frac{S_{XY}}{S_{XX}}$$

$$b_0 = \bar{y} - b_1 \bar{x}$$

why (2) is true?

$$\sum (x_i - \bar{x})(y_i - \bar{y}) = \sum (x_i y_i - x_i \bar{y} - y_i \bar{x} + \bar{x} \bar{y})$$

$$\begin{aligned}
 &= \sum x_i y_i - \underbrace{\sum x_i \bar{y}}_{\bar{y} \cdot n\bar{x}} - \underbrace{\sum y_i \bar{x}}_{\bar{x} \cdot n\bar{y}} + \sum \bar{x} \bar{y} \\
 &\quad - \bar{y} \cdot n\bar{x} - \bar{x} \cdot n\bar{y} + n\bar{x}\bar{y}
 \end{aligned}$$

For any inference on slope  $\beta_1$ :

$$T = \frac{b_1 - \beta_1}{s/\sqrt{S_{xx}}} \sim t(n-2) \quad \star$$

$$\text{where } s = \sqrt{MSE} = \sqrt{\frac{SSE}{n-2}} = \sqrt{\frac{S_{yy} - b_1 S_{xy}}{n-2}}$$

mean square error

sum of squared error

① ex: Find a 95% CI for regression slope  $\beta_1$ .

$$b_1 \pm t_{\frac{\alpha}{2}}^{(n-2)} \cdot \frac{s}{\sqrt{S_{xx}}}$$

②  $H_0: \beta_1 = 0$  under  $H_0: \frac{b_1}{s/\sqrt{S_{xx}}} \sim t(n-2)$   
 $H_1: \beta_1 \neq 0$

Inference on ~~the~~ mean response  $\mu_{Y|X_0}$  and on a single response  $Y_0|X_0$ .

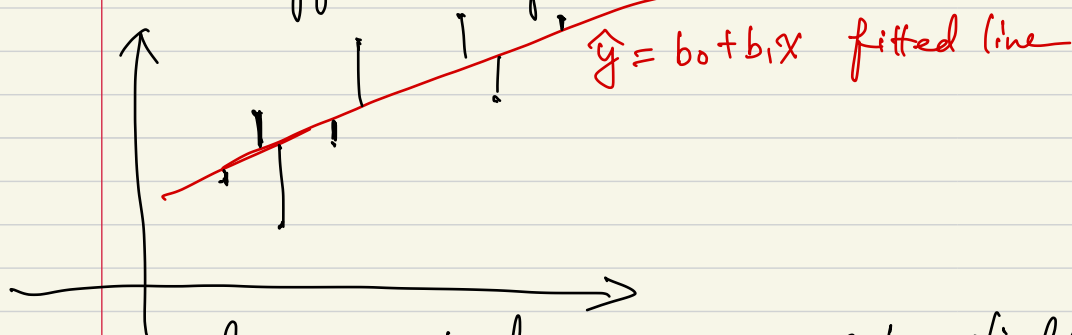
$$T = \frac{\hat{Y}_0 - \mu_{Y|X_0}}{S \cdot \sqrt{\frac{1}{n} + \frac{(X_0 - \bar{X})^2}{S_{XX}}}} \sim T(n-2)$$

and:

$$T = \frac{\hat{Y}_0 - Y_0}{S \sqrt{\left(1 + \frac{1}{n} + \frac{(X_0 - \bar{X})^2}{S_{XX}}\right)}} \sim T(n-2)$$

A measure of quality of fit:

Coefficient of Determination,



Before we introduce  $X$  as an input variable, we see a lot variation in  $y$ .

$$SST = \sum_{i=1}^n (y_i - \bar{y})^2 (= S_{yy})$$

(total sum of squares)

$$SSE = \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

↑  
sum of square error

error after  
introducing the  
regression model.

$$R^2 = \text{Coefficient of Determination} = 1 - \frac{SSE}{SST}$$

↑  
total error.

$$R^2 = r^2$$

↑  
 $r$ : correlation coefficient of  
 $x_i$  &  $y_i$

$$-1 \leq r \leq 1.$$

§10.13.

Ex: Survey 200 Dem. 150 Rep. 150 Indep

Law	Dem	Rep	Indep	
For	85??	70	62	214
Against	92	62	67	222
undecided	25	18	21	64
	200	150	150	500

$H_0$ : For each opinion (for, against, or decided)  
the 3 group of people have the same  
proportion.

$H_1$ : Not

under  $H_0$ :  $P_{D,for} = P_{rep,fa} = P_{ind,for}$

$$\begin{array}{ccc} & \text{against} & \\ & \text{outside} & \\ \hat{p}_{for} = \frac{214}{500} & \hat{p}_{again} = \frac{222}{500} & \hat{p}_{outside} = \frac{64}{500} \end{array}$$

$$e_{Dem,for} = 200 * \frac{214}{500} =$$

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$$\chi^2 = \sum_{i=1}^n \frac{(e_i - o_i)^2}{e_i} \sim \chi^2(n-1)$$

Goodness-of-fit

$$e_i \geq 5$$