

Simple Linear Regression:

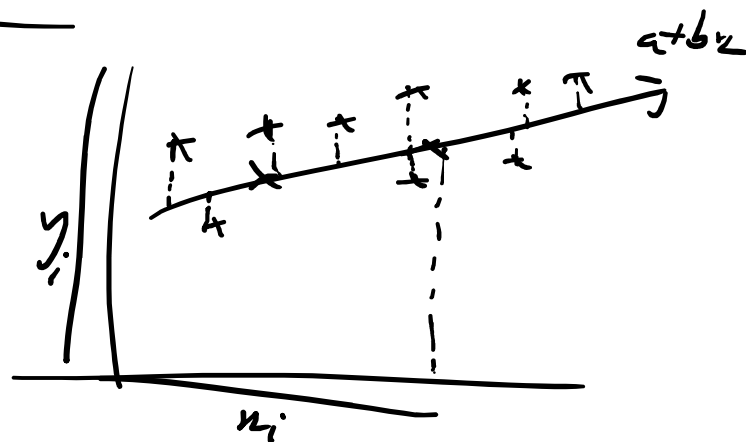
$$y = a + bx$$

Where a is intercept & b is slope of line
 x is independent & y is dependent.

→ Least Squares Estimation:

Approximation is done by

$$\hat{y}^1 = a + bx$$



Exact representation

$$y = a + bx + \text{error}$$

$$\rightarrow \hat{a} = \bar{y} - \hat{b} \bar{x}$$

$$\hat{b} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sum_{i=1}^n (x_i - \bar{x})^2} = \frac{SS_{xy}}{SS_x}$$

$$\sum_{i=1}^n (x_i - \bar{x})^2$$

SSX

→ Correlation coefficient b/w x & y

$$r = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{\sum_{i=1}^n (x_i - \bar{x})^2} \sqrt{\sum_{i=1}^n (y_i - \bar{y})^2}} = \frac{\sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})}{\sqrt{SSX} \sqrt{SSY}}$$

x	y	$(x_i - \bar{x})$	$(y_i - \bar{y})$	$(x_i - \bar{x})^2$	$(x_i - \bar{x})(y_i - \bar{y})$
42	18	10.533	12.267	110.9511	129.2889
34	6	2.533	0.267	6.4178	0.6756
25	0	-6.467	-5.733	41.8178	27.0756
35	-1	3.533	-6.733	12.4844	-23.7511
37	13	5.533	7.267	30.6178	40.2089
38	14	6.533	8.267	42.6844	54.0089
31	7	-0.467	1.267	0.2178	-0.5911
33	7	1.533	1.267	2.3511	1.9422
19	-9	-12.467	-14.733	155.4178	163.6756
29	8	-2.467	2.267	6.0844	-5.5911
38	8	6.533	2.267	42.6844	14.8089
28	5	-3.467	-0.733	12.0178	2.5422
29	3	-2.467	-2.733	6.0844	6.7422
36	14	4.533	8.267	20.5511	37.4756
18	-7	-13.467	-12.733	181.3511	171.4756

$$\begin{array}{rclcl}
 \bar{x} & = & 31.4667 & \bar{y} & = & 5.733 \\
 SS_x & = & 671.733 & SS_{xy} & = & 649.8667
 \end{array}$$

$$b^1 = \frac{SS_{xy}}{SS_x}$$

$$= \frac{649.8667}{671.733} = \underline{\underline{0.9674}}$$

$$\begin{aligned}
 a^1 &= \bar{y} - b^1 \bar{x} = 5.733 - (0.9674 \times 31.4667) \\
 &= \underline{\underline{-24.709}}
 \end{aligned}$$

$$\hat{y} = \underline{\underline{a + b_1x}}$$

$$\begin{aligned}
 x = 30 \quad \hat{y} &= a + b_1x \\
 &= -24.709 + (0.9674 \times 30) \\
 &= \underline{\underline{4.313}}
 \end{aligned}$$

→ Metric - Accuracy ~ Goodness of Fit
(R² - Score)

$$\rightarrow \sum_{i=1}^n (y_i - \hat{y})^2 \rightarrow \text{Sum of Squares of Error (SSE)}$$

$$\rightarrow \sum_{i=1}^n (\hat{y}_i - \bar{y})^2 \rightarrow \text{Sum of Squares due to Regression (SSR)}$$

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

$$SST = SSE + SSR$$

$$\rightarrow R^2 = \frac{SSR}{SST} = 1 - \left(\frac{SSE}{SST} \right)$$

$$SST = 818.9333$$

$$SSE = 190.2216$$

$$R^2 = 1 - \left(\frac{SSE}{SST} \right) = 1 - \left(\frac{190.2216}{818.9333} \right) = \underline{\underline{0.7677}}$$