

$$\hat{y}_i = \beta_0 + \beta_1 x_i \quad \text{advertising}$$

$$\text{error } e_i = y_i - \hat{y}_i$$

$$= y_i - (\beta_0 + \beta_1 x_i)$$

Find  $\beta_0$  and  $\beta_1$  such that the square of error is minimum

$$\min \sum e_i^2 = \min \left\{ \left( y_i - (\beta_0 + \beta_1 x_i) \right)^2 \right\}$$

$$\frac{\partial}{\partial \beta_0} = 0 \quad \frac{\partial}{\partial \beta_1} = 0$$

$$\hat{\beta}_1 \text{ (or } b_1) = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} - \text{slope}$$

$\bar{x}$  = Mean of  $x$

$\bar{y}$  = Mean of  $y$

$$\hat{\beta}_0 = b_0 = \bar{y} - \hat{\beta}_1 \bar{x} \rightarrow \text{intercept.}$$

(OR)

$$\bar{y} - \hat{\beta}_1 \bar{x}$$

$$\hat{\beta}_1 = \frac{\Delta \bar{y}}{\Delta \bar{x}} = \text{change in } y \text{ for one unit change in } x$$

## Example

Advertising and Sales.

X (advertising)

5

2

3

1

4

Y (Sales)

50

15

25

10

40

$$\bar{X} = \frac{5+2+3+1+4}{5} = \frac{15}{5} = 3$$

$$\bar{Y} = \frac{50+15+25+10+40}{5} = 28$$

$X$	$y$	$X_i - \bar{X}$	$y_i - \bar{Y}$	$(X_i - \bar{X})(y_i - \bar{Y})$	$(X_i - \bar{X})^2$
5	50	$5 - 3 = 2$	$50 - 28 = 22$	$2 \times 22 = 44$	4
2	15	$2 - 3 = -1$	$15 - 28 = -13$	$-1 \times -13 = 13$	-
3	25	$3 - 3 = 0$	$25 - 28 = -3$	$0 \times -3 = 0$	0
1	10	$1 - 3 = -2$	$10 - 28 = -18$	$-2 \times -18 = 36$	4
$\sum$		$4 - 3 = 1$	$40 - 28 = 12$	$1 \times 12 = 12$	-1
$\sum$		40	0	10	10
$\bar{X} = 3$		$\hat{b}_1 = b_1 = \frac{\sum (X_i - \bar{X})(y_i - \bar{Y})}{\sum (X_i - \bar{X})^2}$			10
$\bar{Y} = 28$					10
$b_0 = 10 \cdot 5 = 50$					Slope:

$$\begin{aligned} b_0 - \hat{\beta}_0 &= \bar{Y} - \hat{\beta}_1 \bar{X} \\ &= 28 - 10.5(3) \\ &= -3.5 \end{aligned}$$

If you do not advertise you expect to have -3.5 in sales.

Best fit linear regression line is

$$\hat{Y} = -3.5 + 10.5(X)$$

OLS estimation

↳ ordinary Least Squares.

## Algebraic Properties of OLS Estimates.

1. Sum of residuals = 0

Residuals = observed - fitted.

$$\begin{aligned} &= \sum_{i=1}^n y_i - (\hat{\beta}_0 + \hat{\beta}_1 x_i) \\ &= 0 \end{aligned}$$

2. Covariance between  $x$  and residuals is zero

$$\sum_{i=1}^n x_i e_i = 0$$

3.  $(\bar{x}, \bar{y})$  always fall on the estimated regression line

$$\bar{y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x}$$

$$\bar{y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{x}$$

$$y = -3.5 + 10.5(x)$$

$$\bar{x} = 3$$

$$y = -3.5 + 10.5(3)$$

$$= -3.5 + 31.5$$

$$= 28$$

$$= \bar{y}$$

4. Covariance between fitted values and residuals is zero.

$$e_i = y_i - \hat{y}_i$$

Sum of Square total

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

total variation from the mean to be explained. This is a measure of total sample variation.

Sum of Square Residual

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

total variation explained by X

Sum of Squared Error

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

total variation left unexplained

$$SST = SSR + SSE$$

## Goodness -of- fit

Measures how well X explains Y

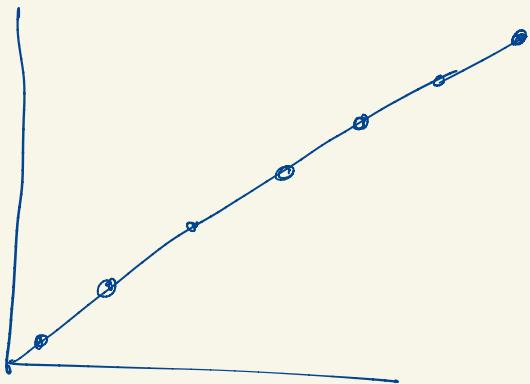
$R^2$  = Fraction of Sample Variation  
in y explained by x  
(or)

Ratio of Variation explained  
to the total Variation.

$$R^2 = \frac{SSE}{SST}$$

$$= 1 - \frac{SSR}{SST}$$

$R^2$  is also called Coefficient of determination



$$R^2 = 1$$

$R^2 = 0$  poor fit