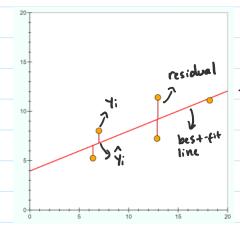
Finding m and c equation in least squared method to get the best fit line

$$Y = \underline{m} \times + C \longrightarrow m = \sum_{i=1}^{n} (x_i - \overline{x})(y_i - \overline{y})$$

$$(x_i - \overline{x})^2$$

$$C = \overline{y} - m\overline{x}$$



→ Y:mx+C

residual =  $\gamma_i - \hat{\gamma}_i$   $\longrightarrow$  sum of squared residuals  $= \xi (\gamma_i - \hat{\gamma}_i)^2$   $= \xi (\gamma_i - (c + m \chi_i))^2$ minimize!

Steps to get the equation for m and c:

- 1. Find the partial derivatives of sum of squared equation with respect to m and c
- 2.—Set the partial derivatives equal to  $\boldsymbol{0}$
- 3. Solve for m and c

## 1. Find the equation

a. c

i. Partial derivative

$$\frac{\partial}{\partial c} \, \mathcal{E} \left( Y_{i} - (c + m \times i) \right)^{2} = \mathcal{E} \, \frac{\partial}{\partial c} \, \left( Y_{i} - (c + m \times i) \right)^{2}$$

$$= \mathcal{E} \, 2 \, \left( Y_{i} - (c + m \times i) \right)^{2} \, (-1)$$

$$= -2 \, \mathcal{E} \left( Y_{i} - (c + m \times i) \right)$$

ii. Set to 0

iii. Solve for c

a. m

i. Partial derivative

$$\frac{\partial}{\partial m} \not \geq (\forall i - ((+m \times i))^2) = \frac{\partial}{\partial m} \not \geq (\forall i - (m \times i + (\bar{\gamma} - m\bar{\chi})))^2$$

$$= \frac{\partial}{\partial m} \not \geq (\forall i - (\bar{\gamma} + m(x_i - \bar{\chi})))^2$$

$$= \frac{\partial}{\partial m} \not \geq (\forall i - \bar{\gamma} - m(x_i - \bar{\chi}))^2$$

$$= 2 \not \leq (\forall i - \bar{\gamma} - m(x_i - \bar{\chi})) \cdot -(x_i - \bar{\chi})$$

$$= -2 \not \leq (\forall i - \bar{\gamma} - m(x_i - \bar{\chi})) \cdot (x_i - \bar{\chi})$$

i. Set to 0

-2 
$$\Xi(Y_i - \bar{Y} - m(x_i - \bar{x}))(x_i - \bar{x}) = 0$$
  
 $\Xi(Y_i - \bar{Y} - m(x_i - \bar{x}))(x_i - \bar{x}) = 0$ 

i. Solve for m

$$\begin{split} & \{ (Y_{i} - \bar{Y} - m(X_{i} - \bar{X}))(X_{i} - \bar{X}) = 0 \\ & \{ (X_{i} - \bar{X}) - \bar{Y}(X_{i} - \bar{X}) - m(X_{i} - \bar{X})^{2} = 0 \\ & \{ (X_{i} - \bar{X}) - \bar{Y}(X_{i} - \bar{X}) - m(X_{i} - \bar{X})^{2} = 0 \\ & \{ (Y_{i} - \bar{Y})(X_{i} - \bar{X}) = M \{ (X_{i} - \bar{X})^{2} \\ & \{ (Y_{i} - \bar{Y})(X_{i} - \bar{X}) = m \{ (X_{i} - \bar{X})^{2} \\ & \} \\ & m = \{ (Y_{i} - \bar{Y})(X_{i} - \bar{X}) - m \} \end{split}$$