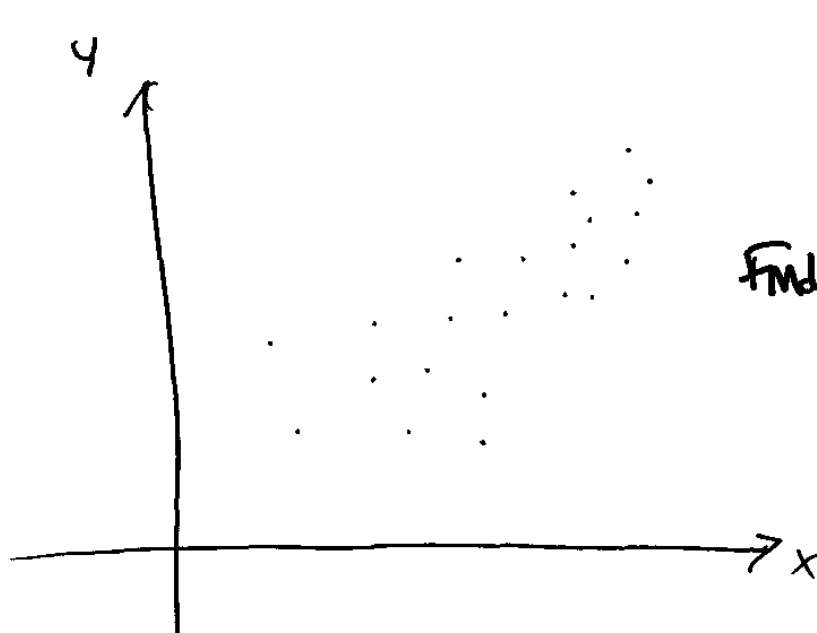




Stat 451
11-17-16
①

Find the best line

Model : $i=1, \dots, n$ $y_i = \beta_0 + \beta_1 x_i + \varepsilon_i$

Assume that $\varepsilon_1, \dots, \varepsilon_n$ are
independent, identically distributed
 $N(0, \sigma^2)$

②

Then $y_1, \dots, y_n$ are independent,
with $y_i \sim N(\beta_0 + \beta_1 x_i, \sigma^2)$

$$\begin{aligned} \text{lik}(\beta_0, \beta_1, \sigma) &= \prod_{i=1}^n f(y_i) \\ &= \prod_{i=1}^n \frac{1}{\sigma \sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{y_i - \beta_0 - \beta_1 x_i}{\sigma} \right)^2} \end{aligned}$$

$$\frac{\partial \mathcal{L}}{\partial \beta_1} = -\frac{1}{2\sigma^2} \sum (y_i - \beta_0 - \beta_1 x_i) x_i \stackrel{\text{set}}{=} 0 \quad (3)$$

$$\sum y_i x_i - \beta_0 \sum x_i - \beta_1 \sum x_i^2 = 0$$

↑
substitute

$$\sum x_i y_i - (\bar{y} - \beta_1 \bar{x}) \sum x_i - \beta_1 \sum x_i^2 = 0$$

$$\sum x_i y_i - \bar{y} \sum x_i + \beta_1 \bar{x} \sum x_i - \beta_1 \sum x_i^2 = 0$$

$$\beta_1 (\sum x_i^2 - \bar{x} \sum x_i) = \sum x_i y_i - \bar{y} \sum x_i$$

$$= \sigma^{-n} (2\pi)^{-\frac{n}{2}} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2} \quad (4)$$

$$\mathcal{L}(\beta_0, \beta_1, \sigma) = -n \ln \sigma - \frac{n}{2} \ln(2\pi) - \frac{1}{2\sigma^2} \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2$$

$$\frac{\partial \mathcal{L}}{\partial \beta_0} = -\frac{1}{2\sigma^2} \sum (y_i - \beta_0 - \beta_1 x_i) (-1) \stackrel{\text{set}}{=} 0$$

$$\sum y_i - n\beta_0 - \beta_1 \sum x_i = 0$$

$$\hat{\beta}_0 = \frac{\sum y_i}{n} - \beta_1 \frac{\sum x_i}{n}$$

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

$$\hat{\beta}_1 = \frac{\sum x_i y_i - \frac{(\sum x_i)(\sum y_i)}{n}}{\sum x_i^2 - \frac{(\sum x_i)^2}{n}} \left\{ \begin{array}{l} SS_{xy} \\ SS_{xx} \end{array} \right. \quad (5)$$

$$\hat{\beta}_1 = \frac{SS_{xy}}{SS_{xx}}, \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$$

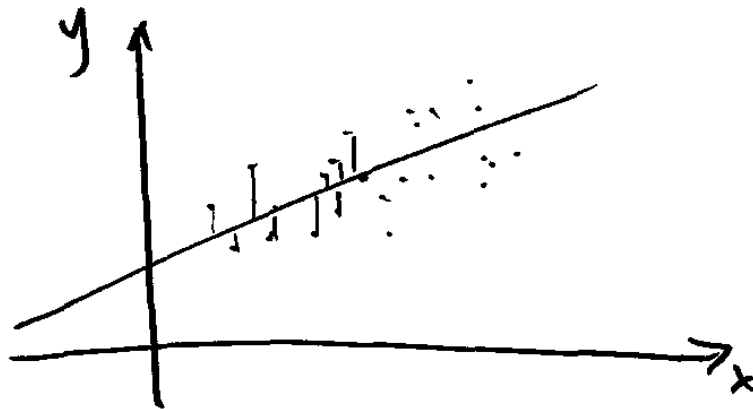
$$\frac{\partial \mathcal{L}}{\partial \sigma} = -\frac{1}{\sigma} - \frac{1}{2}(-2\sigma^{-3}) \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2 \stackrel{\text{set}}{=} 0$$

$$-n\sigma^2 + \sum (y_i - \beta_0 - \beta_1 x_i)^2 = 0$$

$$\hat{\sigma}^2 = \frac{\sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2}{n} \quad (6)$$

(7)

Least-squares solution:



Find the line that minimized the sum of squares of the vertical distances from the points to the line $\hookrightarrow$ residuals

(8)

$$\text{Model: } y_i = \beta_0 + \beta_1 x_i + \varepsilon_i$$

$\uparrow$ No further assumptions

$$\varepsilon_i = y_i - \beta_0 - \beta_1 x_i$$

$$\sum_{i=1}^n \varepsilon_i^2 = \sum_{i=1}^n (y_i - \beta_0 - \beta_1 x_i)^2$$

"
SSE

We want to minimize SSE

$$\frac{\partial SSE}{\partial \beta_0} = \sum 2(y_i - \beta_0 - \beta_1 x_i)(-1) \stackrel{\text{set}}{=} 0 \quad (9)$$

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

$$\frac{\partial SSE}{\partial \beta_1} = \sum 2(y_i - \beta_0 - \beta_1 x_i)(-x_i) \stackrel{\text{set}}{=} 0$$

$$\hat{\beta}_1 = \frac{SS_{xy}}{SS_{xx}}$$

Same solutions as we got using MLE !!