

proof for matrix form of $y_i = \beta_0 + \beta_1 x_i + e_i$

$$\hat{\beta} = (X^T X)^{-1} X^T Y, \quad X = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_n \end{pmatrix} \quad Y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$E(\hat{\beta}) = E[(X^T X)^{-1} X^T Y] \quad Y = X\beta + e$$

$$= (X^T X)^{-1} X^T E(Y)$$

$$= \beta$$

$$= \begin{bmatrix} \frac{\sum x_i y_i}{S_{xx}} - \frac{\bar{x} \sum y_i}{n} \\ \frac{-n \bar{x} \bar{y}}{S_{xx}} + \frac{\sum x_i y_i}{S_{xx}} \end{bmatrix}$$

$$\text{Var}(\hat{\beta}) = \text{Var}[(X^T X)^{-1} X^T Y]$$

$$= (X^T X)^{-1} X^T \text{Var}(Y) X (X^T X)^{-1}$$

$$= \sigma^2 (X^T X)^{-1} X^T X (X^T X)^{-1}$$

$$= \sigma^2 (X^T X)^{-1}$$

$2 \times n \quad n \times 2 = 2 \times 2 \text{ dim}$

$$\text{Cov}(\hat{\beta}) = \sigma^2 \begin{bmatrix} \frac{\sum x_i^2}{n S_{xx}} & -\frac{\bar{x}}{S_{xx}} \\ -\frac{\bar{x}}{S_{xx}} & \frac{1}{S_{xx}} \end{bmatrix}$$

RHS
Show = $\text{Cov}(\hat{\beta}_1, \hat{\beta}_0) \quad \text{Var}(\hat{\beta}_1)$

$$\text{LHS} = \sigma^2 \left(\frac{1}{n} + \frac{\bar{x}^2}{S_{xx}} \right)$$

$$\text{RHS} = \frac{\sum (x_i - \bar{x})^2}{n S_{xx}} + \frac{\sum (x_i - \bar{x})(\bar{x} - \bar{x})}{n S_{xx}}$$

$$= \frac{1}{n} + \frac{\sum (2x_i \bar{x} - 2\bar{x}^2)}{n S_{xx}} + \frac{\sum \bar{x}^2}{n S_{xx}}$$

$$= \frac{1}{n} + \frac{2\bar{x} \sum (x_i - \bar{x})}{n S_{xx}} + \frac{n \bar{x}^2}{n S_{xx}}$$

$$= \frac{1}{n} + \frac{\bar{x}^2}{S_{xx}}$$