

1. SSE

$$\sum_{i=1}^N (y_i - (b_0 + b_1 z_{i1} + b_2 z_{i2}))^2$$

2.  $\frac{\partial SSE}{\partial b_0} = -2 \sum_{i=1}^N (y_i - b_0 - b_1 z_{i1} - b_2 z_{i2})$

$$\frac{\partial SSE}{\partial b_1} = -2 \sum_{i=1}^N z_{i1} (y_i - b_0 - b_1 z_{i1} - b_2 z_{i2})$$

$$\frac{\partial SSE}{\partial b_2} = -2 \sum_{i=1}^N z_{i2} (y_i - b_0 - b_1 z_{i1} - b_2 z_{i2})$$

3.  $\frac{\partial SSE}{\partial b_0} = -2 \sum_{i=1}^N (y_i - b_0 - b_1 z_{i1} - b_2 z_{i2}) = 0$

$$-2 \sum_{i=1}^N e_i = 0 \Rightarrow \text{Avg err} = 0$$

$$\sum_{i=1}^N z_{ij} e_i = 0$$

$\Rightarrow e \cdot z = 0$  at optimum

4.  $b_0^* = \bar{y}$

$$\sum_{i=1}^N (y_i - b_0 - b_1 z_{i1} - b_2 z_{i2}) = 0 \quad (\sum z_{ij} = 0)$$

$\Downarrow$

$$\sum_{i=1}^N y_i - N b_0 = 0 \Rightarrow b_0 = \frac{1}{N} \sum y_i = \bar{y}$$

$$b_0 = \bar{y}$$

$$5. \hat{y}_i = \bar{y} + b_1 z_{i1} + b_2 z_{i2}$$

Assuming  $\frac{1}{N} \sum_{i=1}^N z_{ij} = 0$  for 1 and 2

$$Z = \begin{bmatrix} z_{11} & z_{12} \\ z_{21} & z_{22} \\ \vdots & \vdots \\ z_{N1} & z_{N2} \end{bmatrix} \quad A$$

$$b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \quad b$$

$$y_i - \bar{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_N \end{bmatrix} \quad C = \|y - Zb\|^2$$

$$Z^T Z b = \|y - Zb\|^2$$

$$6. \frac{1}{N} Z^T Z b = \frac{1}{N} \|y - Zb\|^2$$

$$\text{Where } A = \frac{1}{N} Z^T Z \text{ and } C = \frac{1}{N} \|y - Zb\|^2$$

$$\text{So } Ab = C$$

$$z_{ij} = x_{ij} - m_j$$

A is a matrix of Var and Cov between preds  
C is a vector of cov between pred. + outcome

So A gives how predictors relate to each other  
and C gives how <sup>each</sup> pred. relates to y