

# A6

$$1. \text{SSE}(b_0, b_1, b_2) = \sum_{i=1}^N (y_i - b_0 - b_1 z_{i,1} - b_2 z_{i,2})^2 = \sum_{i=1}^N e_i^2$$

$$2. \frac{\partial e_i}{\partial b_0} = -1, \frac{\partial e_i}{\partial b_1} = -z_{i,1}, \frac{\partial e_i}{\partial b_2} = -z_{i,2} \quad \frac{\partial \text{SSE}}{\partial b_0} = -2 \sum_{i=1}^N e_i \stackrel{!}{=} 0 \Rightarrow \sum_{i=1}^N e_i = 0$$

$$\frac{\partial \text{SSE}}{\partial b_1} = -2 \sum_{i=1}^N e_i z_{i,1} \stackrel{!}{=} 0 \Rightarrow \sum_{i=1}^N e_i z_{i,1} = 0 \quad \frac{\partial \text{SSE}}{\partial b_2} = -2 \sum_{i=1}^N e_i z_{i,2} \stackrel{!}{=} 0 \Rightarrow \sum_{i=1}^N e_i z_{i,2} = 0$$

$$3. \sum e_i = 0 \Rightarrow \frac{1}{N} \sum e_i = \bar{e} = 0, \text{ avg error is } 0$$

$$\sum e_i z_{i,1} = 0, \sum e_i z_{i,2} = 0, \text{ residuals are ortho. to each predictor}$$

$$4. 0 = \sum_{i=1}^N (y_i - b_0 - b_1 z_{i,1} - b_2 z_{i,2}) = \sum y_i - N b_0 - b_1 \sum z_{i,1} - b_2 \sum z_{i,2}$$

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

$$\sum (y_i - \bar{y}) z_{i,1} - b_1 \sum z_{i,1}^2 - b_2 \sum z_{i,1} z_{i,2} = 0, \sum (y_i - \bar{y}) z_{i,2} - b_1 \sum z_{i,1} z_{i,2} - b_2 \sum z_{i,2}^2 = 0$$

$$5. b = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}, A = \begin{bmatrix} \sum z_{i,1}^2 & \sum z_{i,1} z_{i,2} \\ \sum z_{i,1} z_{i,2} & \sum z_{i,2}^2 \end{bmatrix},$$

$$\begin{bmatrix} \sum z_{i,1}^2 & \sum z_{i,1} z_{i,2} \\ \sum z_{i,1} z_{i,2} & \sum z_{i,2}^2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} \sum (y_i - \bar{y}) z_{i,1} \\ \sum (y_i - \bar{y}) z_{i,2} \end{bmatrix} = \begin{bmatrix} \sum y_i z_{i,1} \\ \sum y_i z_{i,2} \end{bmatrix} = C$$

$$6. \begin{bmatrix} \frac{1}{N} \sum z_{i,1}^2 & \frac{1}{N} \sum z_{i,1} z_{i,2} \\ \frac{1}{N} \sum z_{i,1} z_{i,2} & \frac{1}{N} \sum z_{i,2}^2 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{N} \sum (y_i - \bar{y}) z_{i,1} \\ \frac{1}{N} \sum (y_i - \bar{y}) z_{i,2} \end{bmatrix}$$

$$\frac{1}{N} \sum z_{i,1}^2 = \text{Var}(X_1), \frac{1}{N} \sum z_{i,2}^2 = \text{Var}(X_2), \frac{1}{N} \sum z_{i,1} z_{i,2} = \text{Cov}(X_1, X_2), \frac{1}{N} \sum (y_i - \bar{y}) z_{i,1} = \text{Cov}(X_1, Y),$$

$$\frac{1}{N} \sum (y_i - \bar{y}) z_{i,2} = \text{Cov}(X_2, Y), \begin{bmatrix} \text{Var}(X_1) & \text{Cov}(X_1, X_2) \\ \text{Cov}(X_1, X_2) & \text{Var}(X_2) \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} = \begin{bmatrix} \text{Cov}(X_1, Y) \\ \text{Cov}(X_2, Y) \end{bmatrix} \text{ w/ } b_0^* = \bar{y}$$

$A$  is the covariance matrix of the predictors  $\hat{\Sigma}_{xx}$ .  $C$  is the vector of covariances between each predictor &  $Y$ . So after scaling,  $A = \hat{\Sigma}_{xx}$  &  $C = \hat{\Sigma}_{xy}$  so  $b = (\hat{\Sigma}_{xx})^{-1} \hat{\Sigma}_{xy}$  &  $b_0 = \bar{y}$