

Ex. 3.9

First we will examine \hat{y} when we ~~add~~^{add} a single additional variable.

We know that for X_i we may find Z_0, \dots, Z_{q-1} via algorithm 3.1 e.g.
for j in $1, 2, \dots, q$ $\hat{x}_{l,j} = \frac{\langle Z_l, X_j \rangle}{\langle Z_l, Z_l \rangle}$, $l=0, 1, \dots, j-1$

$$\text{and } Z_j = X_j - \sum_{k=0}^{j-1} \hat{x}_{k,j} Z_k$$

$$\text{Then } Q = \begin{bmatrix} Z_0 & Z_1 & \dots & Z_{q-1} \end{bmatrix} \begin{bmatrix} \frac{1}{\|Z_0\|} & & 0 \\ & \ddots & \\ 0 & & \frac{1}{\|Z_{q-1}\|} \end{bmatrix}$$

$N \times (q+1) \qquad (q+1) \times (q+1)$

$$\text{Then } \hat{y} = QQ^T y \quad \text{and} \quad \text{RSS} = \sum_{i=1}^N (\hat{y}_i - y_i)^2$$

Now suppose we add a single new variable in X_2 , then we obtain an additional vector Z_{q+1} .

Notation: We will denote the N elements of a vector $Z_j = (Z_{j,1}, Z_{j,2}, \dots, Z_{j,N})$
and $\hat{Z}_{j,i} = \frac{Z_{j,i}}{\|Z_j\|}$

$$\text{Now: } Q = \begin{bmatrix} \hat{Z}_{0,1} & \dots & \hat{Z}_{0,N} \\ \vdots & & \vdots \\ \hat{Z}_{q+1,1} & \dots & \hat{Z}_{q+1,N} \end{bmatrix}$$

$$\text{and } QQ^T = \begin{bmatrix} \sum_{i=0}^{q+1} \hat{Z}_{i,1} \hat{Z}_{i,1} & \dots & \sum_{i=0}^{q+1} \hat{Z}_{i,1} \hat{Z}_{i,N} \\ \vdots & & \vdots \\ \sum_{i=0}^{q+1} \hat{Z}_{i,N} \hat{Z}_{i,1} & \dots & \sum_{i=0}^{q+1} \hat{Z}_{i,N} \hat{Z}_{i,N} \end{bmatrix}$$

$N \times N$

Now we notice that this is

$$QQ^T = QQ_{\text{original}}^T + QQ_{\text{new}}^T$$

$$\begin{array}{c} \downarrow \\ (QQ^T \text{ for first } q \text{ variables}) \end{array} \quad \begin{array}{c} \downarrow \\ \begin{bmatrix} \hat{Z}_{q+1,1} & \hat{Z}_{q+1,1} & \dots & \hat{Z}_{q+1,1} & \hat{Z}_{q+1,N} \\ \vdots & \vdots & & \vdots & \vdots \\ \hat{Z}_{q+1,N} & \hat{Z}_{q+1,1} & \dots & \hat{Z}_{q+1,N} & \hat{Z}_{q+1,N} \end{bmatrix} \end{array}$$

Now we may efficiently calculate the RSS for each of $P-q$ new variables. For new variable K we obtain RSS_K via:

$$\hat{Y}_K = QQ^T Y = \overbrace{QQ_{\text{original}}^T}^{\text{static}} Y + QQ_{\text{new}}^T Y$$
$$\Rightarrow RSS_K = \sum_{i=1}^N (\hat{Y}_{Ki} - Y_i)^2$$