

Ex. 7.4 Consider the in-sample prediction error (7.18) and the training error $\bar{\text{err}}$ in the case of squared-error loss:

$$\begin{aligned}\text{Err}_{\text{in}} &= \frac{1}{N} \sum_{i=1}^N E_{Y^0} (Y_i^0 - \hat{f}(x_i))^2 \\ \bar{\text{err}} &= \frac{1}{N} \sum_{i=1}^N (y_i - \hat{f}(x_i))^2.\end{aligned}$$

Add and subtract $f(x_i)$ and $E\hat{f}(x_i)$ in each expression and expand. Hence establish that the average optimism in the training error is

$$\frac{2}{N} \sum_{i=1}^N \text{Cov}(\hat{y}_i, y_i),$$

as given in (7.21).

$$op = \text{Err}_{\text{in}} - \bar{\text{err}}$$

We have to show that

$$W = E_Y(op) = E_Y(\text{Err}_{\text{in}} - \bar{\text{err}}) = \frac{2}{N} \sum_{i=1}^N \text{Cov}(\hat{y}_i, y_i)$$

$$\text{Proof) } E_Y(\text{Err}_{\text{in}} - \bar{\text{err}})$$

$$\text{let } \hat{f}(x_i) = \hat{y}_i$$

$$= \frac{1}{N} \sum_{i=1}^N E_{Y^0} E_Y (Y_i^0 - \hat{f}(x_i))^2 - E_Y \left[\frac{1}{N} \sum_{i=1}^N (y_i - \hat{f}(x_i))^2 \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \left[E_{Y^0} E_Y (Y_i^0 - \hat{y}_i)^2 - E_Y (y_i - \hat{y}_i)^2 \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \left[E_{Y^0} E_Y (Y_i^0{}^2 - \hat{y}_i^2) - E_Y (y_i - \hat{y}_i)^2 \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \left[E_{Y^0} E_Y (Y_i^0{}^2 - 2Y_i^0 \hat{y}_i + \hat{y}_i^2) - E_Y (y_i^2 - 2y_i \hat{y}_i + \hat{y}_i^2) \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \left[E_{Y^0} (Y_i^0{}^2) - 2E_{Y^0} E_Y (Y_i^0 \hat{y}_i) + E_Y (\hat{y}_i^2) \right.$$

$$\left. + E_Y (\hat{y}_i^2) - E_Y (y_i^2) + 2E_Y (y_i \hat{y}_i) - E_Y (\hat{y}_i^2) \right]$$

$$= \frac{1}{N} \sum_{i=1}^N \left[E_{Y^0} (Y_i^0{}^2) - 2E_{Y^0} E_Y (Y_i^0 \hat{y}_i) - E_Y (y_i^2) + 2E_Y (y_i \hat{y}_i) \right]$$

using that $E_{y^0}(x_i^0) = E_y(y_i^2)$

$$= \frac{1}{N} \cdot \sum_{i=1}^n [2E_y(y_i \hat{y}_i) - 2E_{y^0} E_y(x_i^0 \hat{y}_i)]$$

$$= \frac{1}{N} \cdot \sum_{i=1}^n [2E_y(y_i \hat{y}_i) - 2E_{y^0}(x_i^0) \cdot E_y(\hat{y}_i)]$$

We know that $E_{y^0}(x_i^0) = E_y(x_i)$

$$= \frac{1}{N} \cdot \sum_{i=1}^n [2E_y(y_i \hat{y}_i) - 2E_y(x_i) \cdot E_y(\hat{y}_i)]$$

$$= \frac{2}{N} \cdot \sum_{i=1}^n \text{Cov}(y_i, \hat{y}_i)$$