

Squared error loss.

Training error: $\bar{err} = \frac{1}{N} \sum_{i=1}^N (y_i - \hat{f}(x_i))^2$

In sample prediction error $Err_{in} = \frac{1}{N} \sum_{i=1}^N E_{Y^0}[(Y_i^0 - \hat{f}(x_i))^2]$

↑
observations in training set, x 's kept fixed. E only over new Y^0 's

TASK: establish that the average optimism is

$$E_y(op) = w = \frac{2}{N} \sum_{i=1}^N \text{Cov}(\hat{y}_i, y_i) \quad (7.21)$$

$$E_y(op) = E_y(Err_{in} - \bar{err}) \quad \text{"definition"}$$

↑
expected value over
the responses in the
training set

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

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

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

↑
not a
function of
 Y^0

→ $E_{Y^0}(\hat{f}(x_i)) = \hat{f}(x_i)$ and $E_{Y^0}(\hat{f}(x_i)^2) = \hat{f}(x_i)^2$

$$= \frac{1}{N} \sum_{i=1}^N \left(E_y \underbrace{E_{Y_0}(Y_i^2)}_{E_y(Y_i^2) \leftarrow \text{same distribution}} - 2 E_y(\hat{f}(x_i) \cdot \underbrace{E_{Y_0}(Y_i)}_{E_y(Y_i)}) - \cancel{E_y(Y_i^2)} + 2 E_y(Y_i \cdot \hat{f}(x_i)) \right)$$

~~$E_y(Y_i^2)$~~

Also $E_{Y_0}(Y_i) = E_y(Y_i)$ and

$$E_y(E_{Y_0}(Y_i)) = E_y(Y_i)$$

$$= \frac{1}{N} \sum_{i=1}^N \left(-2 E_y(\hat{f}(x_i)) \cdot E_y(Y_i) + 2 E_y(Y_i \cdot \hat{f}(x_i)) \right)$$

$$\text{Cov}(Y_i, \hat{Y}_i) = \text{Cov}(Y_i, \hat{f}(x_i)) = E_y(Y_i \cdot \hat{f}(x_i)) - E_y(Y_i) \cdot E_y(\hat{f}(x_i))$$

$$\text{Cov}(V, W) = E(V \cdot W) - E(V) \cdot E(W)$$

↑
def of covariance

$$= \frac{2}{N} \sum_{i=1}^N \text{Cov}(Y_i, \hat{f}(x_i))$$

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