Topic4. Bias-Variance decomposition

Simply the note from http://www.cs.cornell.edu/courses/cs4780/2018fa/lectures/lecturenote12.html

$$\begin{split} &E(x_0) = E\left[\left(f(x_0) - \hat{f}(x_0) \right)^2 | X = x_0 \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) + \left(\hat{f}(x_0) - \hat{f}(x_0) \right)^2 \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) + \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right]^2 \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) + \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) + \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \right] \times E\left[\left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \right] \times \left(E\left[\hat{f}(x_0) \right] - E\left[f(x_0) \right] \right) \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \right] \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \right] \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \right] \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) + \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[\left(f(x_0) - \hat{f}(x_0) \right) \times \left(\hat{f}(x_0) - \hat{f}(x_0) \right) \right] \\ &= E\left[f(x_0) - \hat{f}(x_0) \right] \times E\left[\hat{f}(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[\hat{f}(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[\hat{f}(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[\hat{f}(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[\hat{f}(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[\hat{f}(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right] \\ &= (E\left[f(x_0) - \hat{f}(x_0) \right) \times E\left[f(x_0) - \hat{f}(x_0) \right]$$

Combining (3) & (4), we get

$$E\left[\left(f(x_0) - \overline{\hat{f}(x_0)}\right)^2\right] = E\left[\left(f(x_0) - \overline{f(x_0)}\right)^2\right] + E\left[\left(\overline{f(x_0)} - \overline{\hat{f}(x_0)}\right)^2\right] \cdots (5)$$

Combining (1), (2) & (5), we get

$$E(x_0) = E\left[\left(f(x_0) - \hat{f}(x_0) \right)^2 | X = x_0 \right]$$

$$= E\left[\left(f(x_0) - \overline{f(x_0)} \right)^2 \right] + E\left[\left(\overline{f(x_0)} - \overline{\hat{f}(x_0)} \right)^2 \right] + E\left[\left(\overline{\hat{f}(x_0)} - \hat{f}(x_0) \right)^2 \right]$$

= Noise + Bias 2 + Variance