Bias Variance Tradeoff

AntonMu

Definitions

Definition (Average Hypothesis at x). For each point $x \in X$ define the average hypothesis at x as the expectation of $g^{(D)}(x)$ with respect to \mathcal{D} as

$$\bar{g}(x) = \mathbb{E}_{\mathcal{D}}[g^{(D)}(x)].$$

Definition (Prediction Error at x). For each point $x \in X$ define the prediction error at x as the expected squared error between the ground truth f(x) at x and the learned hypothesis $g^{(D)}(x)$ at x as

$$\mathbb{E}_{\mathcal{D}}[(f(x) - g^{(D)}(x))^2],$$

where the expectation is with respect to the set of data sets \mathcal{D} .

Derivation

To derive the bias variance decomposition at x, we "insert" the average hypothesis $\bar{g}(x)$ in the prediction error at x.

$$\mathbb{E}_{\mathcal{D}} \left[\left(f(x) - g^{(D)}(x) \right)^{2} \right] = \mathbb{E}_{\mathcal{D}} \left[\left(f(x) - \bar{g}(x) + \bar{g}(x) - g^{(D)}(x) \right)^{2} \right]$$

$$= \mathbb{E}_{\mathcal{D}} \left[\left(f(x) - \bar{g}(x) \right)^{2} + \left(\bar{g}(x) - g^{(D)}(x) \right)^{2} \right]$$

$$= \mathbb{E}_{\mathcal{D}} \left[\left(f(x) - \bar{g}(x) \right)^{2} \right] + \mathbb{E}_{\mathcal{D}} \left[\left(\bar{g}(x) - g^{(D)}(x) \right)^{2} \right]$$

$$= \left(f(x) - \bar{g}(x) \right)^{2} + \mathbb{E}_{\mathcal{D}} \left[\left(\bar{g}(x) - g^{(D)}(x) \right)^{2} \right]$$

$$= \text{bias}(x) + \text{variance}(x),$$
(II)

where in (I) we used that $2\mathbb{E}_{\mathcal{D}}[(g(x) - \bar{g}(x)] = 0$ and in (II) we used that neither f nor \bar{g} depend on $D \in \mathcal{D}$. Taking the expectation with respect to the domain X yields:

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$$E_{\mathcal{D},X} \left[\left(f(x) - g^{(D)}(x) \right)^2 \right] = \mathbb{E}_X \left[\left(f(x) - \bar{g}(x) \right)^2 \right] + \mathbb{E}_{\mathcal{D},X} \left[\left(\bar{g}(x) - g^{(D)}(x) \right)^2 \right]$$

$$= \text{bias} + \text{variance}$$