

# Bias Variance Tradeoff

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## 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  $x$ .

$$\begin{aligned} \mathbb{E}_{\mathcal{D}} \left[ \left( f(x) - g^{(D)}(x) \right)^2 \right] &= \mathbb{E}_{\mathcal{D}} \left[ \left( f(x) - \bar{g}(x) + \bar{g}(x) - g^{(D)}(x) \right)^2 \right] \\ &= \mathbb{E}_{\mathcal{D}} \left[ (f(x) - \bar{g}(x))^2 + (\bar{g}(x) - g^{(D)}(x))^2 \right] \quad (\text{I}) \\ &= \mathbb{E}_{\mathcal{D}} \left[ (f(x) - \bar{g}(x))^2 \right] + \mathbb{E}_{\mathcal{D}} \left[ (\bar{g}(x) - g^{(D)}(x))^2 \right] \\ &= (f(x) - \bar{g}(x))^2 + \mathbb{E}_{\mathcal{D}} \left[ (\bar{g}(x) - g^{(D)}(x))^2 \right] \quad (\text{II}) \\ &= \text{bias}(x) + \text{variance}(x), \end{aligned}$$

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  $\bar{g}$  depends on  $D \in \mathcal{D}$ . Taking the expectation with respect to the domain  $X$  yields:

### Bias Variance Tradeoff

$$\begin{aligned} \mathbb{E}_{\mathcal{D},X} \left[ \left( f(x) - g^{(D)}(x) \right)^2 \right] &= \mathbb{E}_X \left[ (f(x) - \bar{g}(x))^2 \right] + \mathbb{E}_{\mathcal{D},X} \left[ (\bar{g}(x) - g^{(D)}(x))^2 \right] \\ &= \text{bias} + \text{variance} \end{aligned}$$