

Bias and variance

$$Err(\mathbf{x}) = \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}))^2 \right] + (\bar{f}(\mathbf{x}) - y)^2 + \mathbb{E}_D \left[(y_D - y)^2 \right]$$

► variance



► bias²

► noise

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► bias²

► noise

$$\begin{aligned}
E(f; D) &= \mathbb{E}_D \left[(f(\mathbf{x}; D) - y_D)^2 \right] \\
&= \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}) + \bar{f}(\mathbf{x}) - y_D)^2 \right] \\
&= \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}))^2 \right] + \mathbb{E}_D \left[(\bar{f}(\mathbf{x}) - y_D)^2 \right] \\
&\quad + \mathbb{E}_D \left[2 (f(\mathbf{x}; D) - \bar{f}(\mathbf{x})) (\bar{f}(\mathbf{x}) - y_D) \right] \\
&= \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}))^2 \right] + \mathbb{E}_D \left[(\bar{f}(\mathbf{x}) - y_D)^2 \right] \\
&= \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}))^2 \right] + \mathbb{E}_D \left[(\bar{f}(\mathbf{x}) - y + y - y_D)^2 \right] \\
&= \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}))^2 \right] + \mathbb{E}_D \left[(\bar{f}(\mathbf{x}) - y)^2 \right] + \mathbb{E}_D \left[(y - y_D)^2 \right] \\
&\quad + 2\mathbb{E}_D \left[(\bar{f}(\mathbf{x}) - y) (y - y_D) \right] \\
&= \mathbb{E}_D \left[(f(\mathbf{x}; D) - \bar{f}(\mathbf{x}))^2 \right] + (\bar{f}(\mathbf{x}) - y)^2 + \mathbb{E}_D \left[(y - y_D)^2 \right]
\end{aligned}$$