

Linear regression case

Noisy target $y = \mathbf{w}^{*\top} \mathbf{x} + \text{noise}$

Data set $\mathcal{D} = \{(\mathbf{x}_1, y_1), \dots, (\mathbf{x}_N, y_N)\}$

Linear regression solution: $\mathbf{w} = (\mathbf{X}^\top \mathbf{X})^{-1} \mathbf{X}^\top \mathbf{y}$

In-sample error vector = $\mathbf{X}\mathbf{w} - \mathbf{y}$

'Out-of-sample' error vector = $\mathbf{X}\mathbf{w} - \mathbf{y}'$

Learning curves for linear regression

Best approximation error = σ^2

Expected in-sample error = $\sigma^2 \left(1 - \frac{d+1}{N}\right)$

Expected out-of-sample error = $\sigma^2 \left(1 + \frac{d+1}{N}\right)$

Expected generalization error = $2\sigma^2 \left(\frac{d+1}{N}\right)$

