Sample title

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Overview

1 Q1: Bias-variance tradeoff

Q2: Hastie and Tibshirani

Bias Error

Also called 'Overfitting'

Variance

Bias Error

Also called 'Overfitting'

Variance

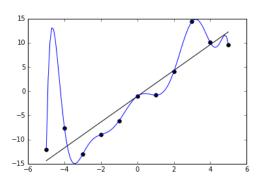
Also called 'Underfitting'

Bias Error

Also called 'Overfitting'

Variance

Also called 'Underfitting'

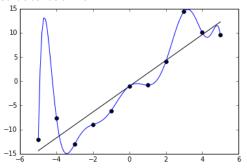


Bias Error

- Also called 'Overfitting'
- Predicts test data too well

Variance

Also called 'Underfitting'

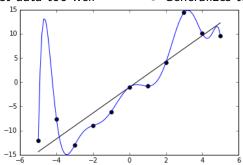


Bias Error

- Also called 'Overfitting'
- Predicts test data too well

Variance

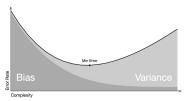
- Also called 'Underfitting'
- Generalizes too well



- Aiming for the lowest total error typically means finding a "middle-ground"
- A common technique for this is determining the minimum mean squared error.

$$MSE = \left(E\left[\hat{f}(x)\right] - f(x)\right)^{2} + E\left[\left(\hat{f}(x) - E\left[\hat{f}(x)\right]\right)^{2}\right] + \sigma_{e}^{2}$$

$$= Bias^{2} + Variance + Irreducible Error$$



Hastie Lectures

- Statistical Learning and Regression
- Dimensionality and Parametric Models
- Assessing Model Accuracy and Bias-Variance Trade-Off
- Classification Problems and K-Nearest Neighbors

Statistical Learning and Regression

Predicting values through various means The simplest method of regression may be to use $\hat{f}(x_n) = \bar{x}_n$ A model may be improved by extending the values contributing to the average

Dimensionality and Parametric Models

Assessing Model Accuracy and Bias-Variance Trade-Off

Classification Problems and K-Nearest Neighbors