



JOHNS HOPKINS

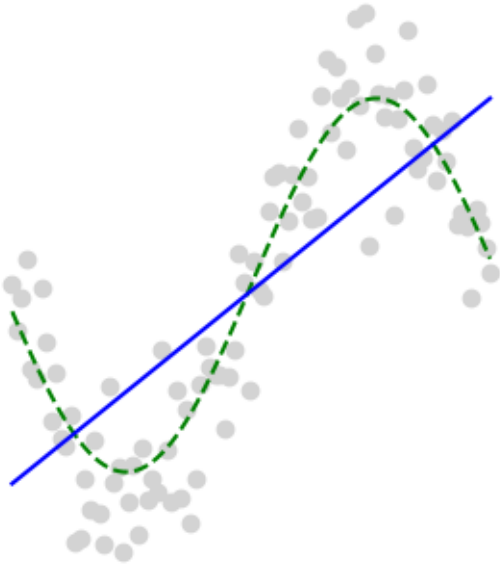
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685.621 Algorithms for Data Science

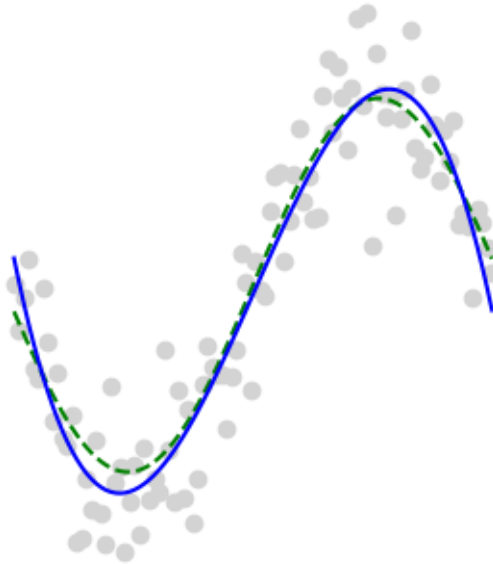
Supervised Learning: Model Optimization

The Bias-Variance Tradeoff

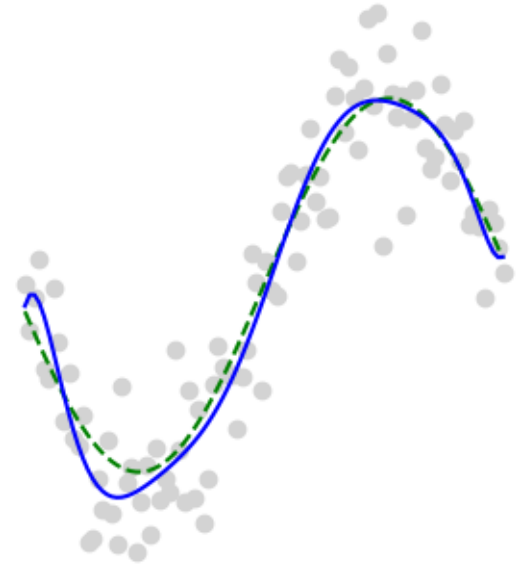
Underfitting (High Bias)



Optimal Fit (Bias-Variance Tradeoff)



Overfitting (High Variance)



Feature Selection and Imbalanced Data

- **Why Feature Selection?**

- Reduces overfitting by removing redundant/noisy features
- Improves model interpretability
- Speeds up training time

- **Methods:**

- Fisher's Linear Discriminant Ratio
- Decision Tree

- **Problem:** When one class dominates, models may favor the majority class.

- **Solutions:**

- **Resampling**

- **Oversampling** the minority class (SMOTE)
 - **Under-sampling** the majority class

- **Class Weighting:** Assign higher penalty to minority misclassifications

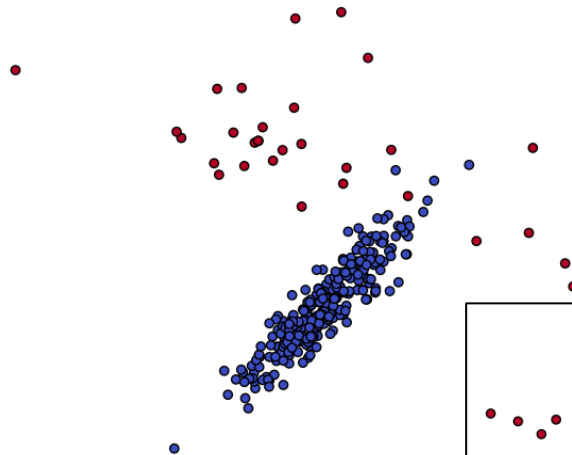
- **Synthetic Data Generation:** Generate new examples for the minority class

SMOTE: Synthetic Minority Over-sampling Technique

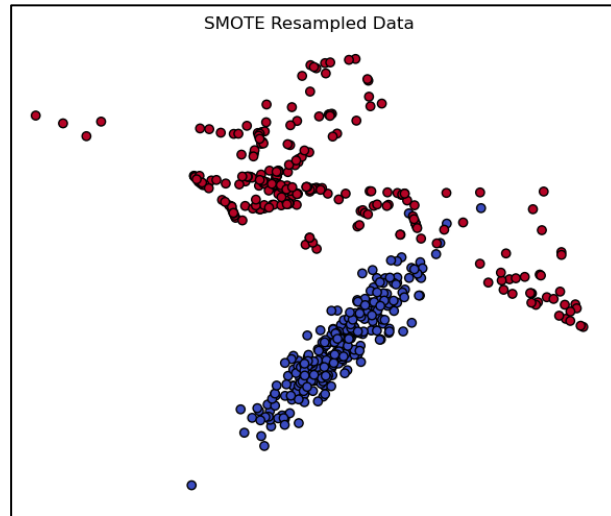
- **How SMOTE Works:**

- Identify k-nearest neighbors of a minority class sample.
- Generate synthetic points between the sample and its neighbors
- Repeat until class distribution is balanced

Original Imbalanced Data



SMOTE Resampled Data



Why Data Leakage Can Ruin Your Model

- **What is data leakage?**
 - Occurs when a model has access to information that wouldn't be available during real-world predictions
- **Types of Data Leakage:**
 - **Target Leakage:** Features contain information about the label that wouldn't be available at prediction time.
 - **Cross-Validation Leakage:** Training data is inadvertently exposed to test data.
- **Example of Target Leakage:**
 - **Bad Feature:** A credit risk model includes "Number of late payments in the next 3 months"
 - **Why It's Bad:** The model would "cheat" by using future information

How to Avoid Data Leakage

- **Best Practices to Prevent Leakage:**

- Remove future-dependent features
- Perform preprocessing (e.g., scaling, encoding) only on training data.
- Use proper cross-validation techniques
 - **Mistake** – splitting after preprocessing (introduces leakage)
 - **Mistake** – not stratifying data in imbalanced classification
 - **Fix** – Use Stratified K-Fold Cross-Validation for class balance
 - **Fix** – Apply transformations inside the cross-validation loop.



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