

# Machine Learning Football Predictor: English Premier League Match Outcome Prediction

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This project implements a machine learning system to predict match outcomes in the English Premier League. Combining web scraping techniques with supervised learning algorithms, the system analyzes historical match data to forecast home wins, away wins, or draws. The project demonstrates data science applications in sports analytics while addressing football's unique prediction challenges.

## Introduction

### Motivation

Football has long been a domain where statistical analysis and intuition intersect. As a lifelong football enthusiast and data scientist, I developed this predictive model to bridge these worlds. The English Premier League's competitiveness and data availability make it ideal for analysis.

### Project Objectives

- Collect and process historical EPL match data
- Engineer relevant predictive features
- Implement and evaluate machine learning models
- Develop an accurate prediction system

## Data Collection and Preparation

### Data Source

Selected [FBref](#) for:

- Comprehensive historical coverage
- Consistent data formatting
- Advanced metrics availability
- Free educational access

### Data Cleaning

Key preprocessing steps:

- Handling missing values
- Standardizing team names
- Creating derived features
- Normalizing numerical features

## Feature Engineering

### Predictor Sets

- **Baseline Predictors:** Venue, opponent, time, day
- **Rolling Averages:** Goals, shots, possession metrics
- **Full Feature Set:** Baseline + rolling + FIFA rankings

### Target Variable

Match outcomes encoded as:

- 0: Away win
- 1: Draw
- 2: Home win

## Model Selection

### Candidate Models

- **Basic Models:** KNN, LDA, QDA, Logistic Regression
- **Tree-Based Models:** Decision Trees, Random Forest, Gradient Boosting

### Evaluation Metrics

- Accuracy
- Precision
- Training vs testing performance

## Results

### Optimal Models

- **Primary Recommendation:** Random Forest (Full Feature Set)
  - Testing Accuracy: 65.93%

- Testing Precision: 64.64%
  - Best generalization capability
- **Secondary Recommendation:** K-Nearest Neighbors (Rolling Features)
  - Training Accuracy: 74.17%
  - Training Precision: 76.39%
  - Best training performance

### Key Findings

- Strong home advantage prediction ( 70% accuracy)
- Draws most challenging to predict ( 50%)
- Recent form more predictive than long-term history

## Challenges and Future Work

### Limitations

- Missing data in early seasons
- Football's inherent randomness
- Team strength fluctuations

### Future Improvements

- Incorporate player-level data
- Add betting odds benchmarks
- Implement temporal pattern recognition

## Conclusion

This project demonstrates machine learning's potential for football outcome prediction, with Random Forest emerging as the most reliable approach , yet future enhancements could always be done.