

PREDICTING OUTCOMES OF NFL MATCHUPS

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PROBLEM CONTEXT

- ▶ NFL has a huge market – 8 Billion in revenue in 2017
- ▶ Top experts ~67% accurate

Predicting NFL Outcomes

- ▶ Sports gambling platforms
- ▶ NFL Analysts and other analysis platforms

Understanding important components of a winning team

- ▶ Owners, GMs, Coaches, Talent Scouts, Agents

DATA COLLECTION

Online API - <https://profootballapi.com/>

- ▶ Flexible data collection and feature engineering
- ▶ Leverages domain knowledge
- ▶ Aggregate statistics at a yearly level for each team
- ▶ Compare statistics for each team matchup
- ▶ Predict winner based upon matchup

FEATURES

- ▶ Data from 2011 to 2017 season
- ▶ Features should cover many aspects of gameplay
- ▶ Overall ~40 features were chosen (~20 for each team per matchup)
- ▶ Focus on features which can influence team decisions, examples below

Feature	Value
QB Pass/Run ratio	Effectiveness of QB playstyle
Average Pass Length	Effectiveness of play call styles
Turnovers	Impact of turnovers influence risky play calling

EXPLORATORY ANALYSIS

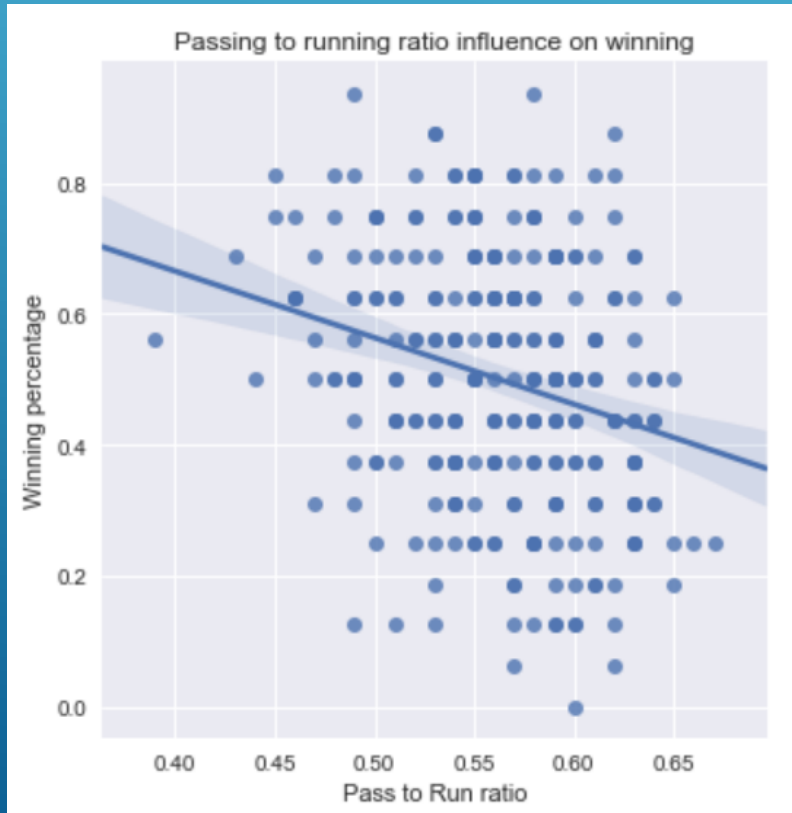
Average statistics of winning teams

- ▶ Turnovers are most significant
- ▶ Features are more extreme for away teams

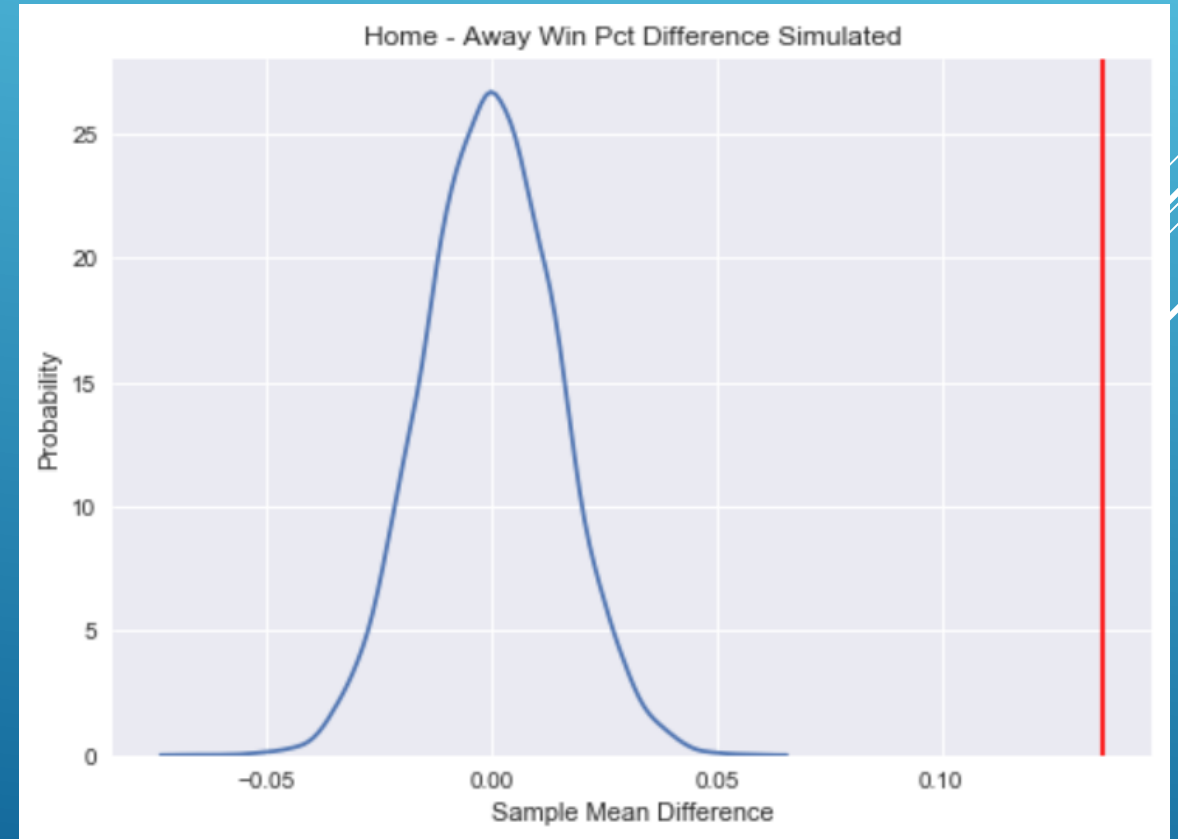


STATISTICAL ANALYSIS

- ▶ Statistically significant correlation between low passing ratio and winning percentage
- ▶ P-value < .01



- ▶ Hypothesis test statistically validates home field advantage
- ▶ P-value < .01



BUILDING THE MODEL

Explore and tune a list of applicable models

- ▶ Linear Regression
- ▶ Logistic Regression
- ▶ SVM
- ▶ Random Forrest

Choose top model and explore for insights

Several white lines of varying lengths and slopes are positioned in the bottom right corner of the slide, creating a modern, abstract graphic element.

LINEAR MODEL

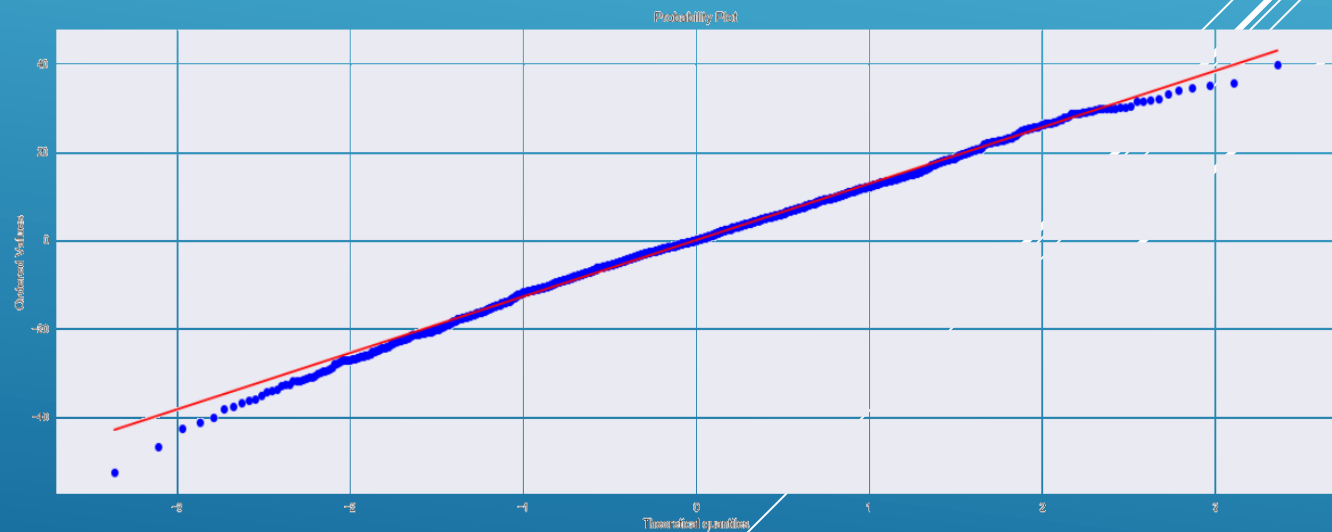
- ▶ Train model on the score difference (an integer outcome)
- ▶ Convert to binary classification to tune using accuracy as the performance metric

Cross Validation
Accuracy

69.2%

Assumption – predictions are normally distributed around correct value

QQPlot shows non-normal distribution weakening the validity of the model



CLASSIFICATION MODELS

Logistic Regression

- ▶ Top performance among models
- ▶ Feature importance visibility

Random Forest/Gradient Boosted Forest

- ▶ Poor model performance
- ▶ Feature importance visibility
- ▶ Gradient Boosted improved performance but model is still weak

SVM

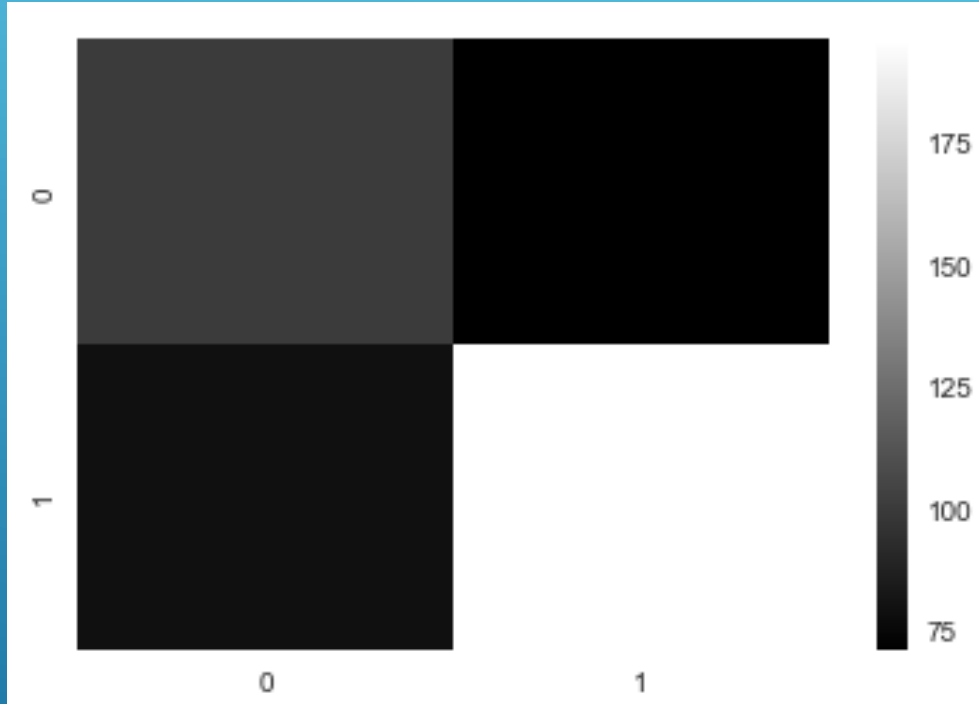
- ▶ Top performance among models
- ▶ No feature visibility

Model	CV Accuracy
Logistic	70.28%
Random Forest	63.2%
Boosted RF	67.5%
SVM	70.34%

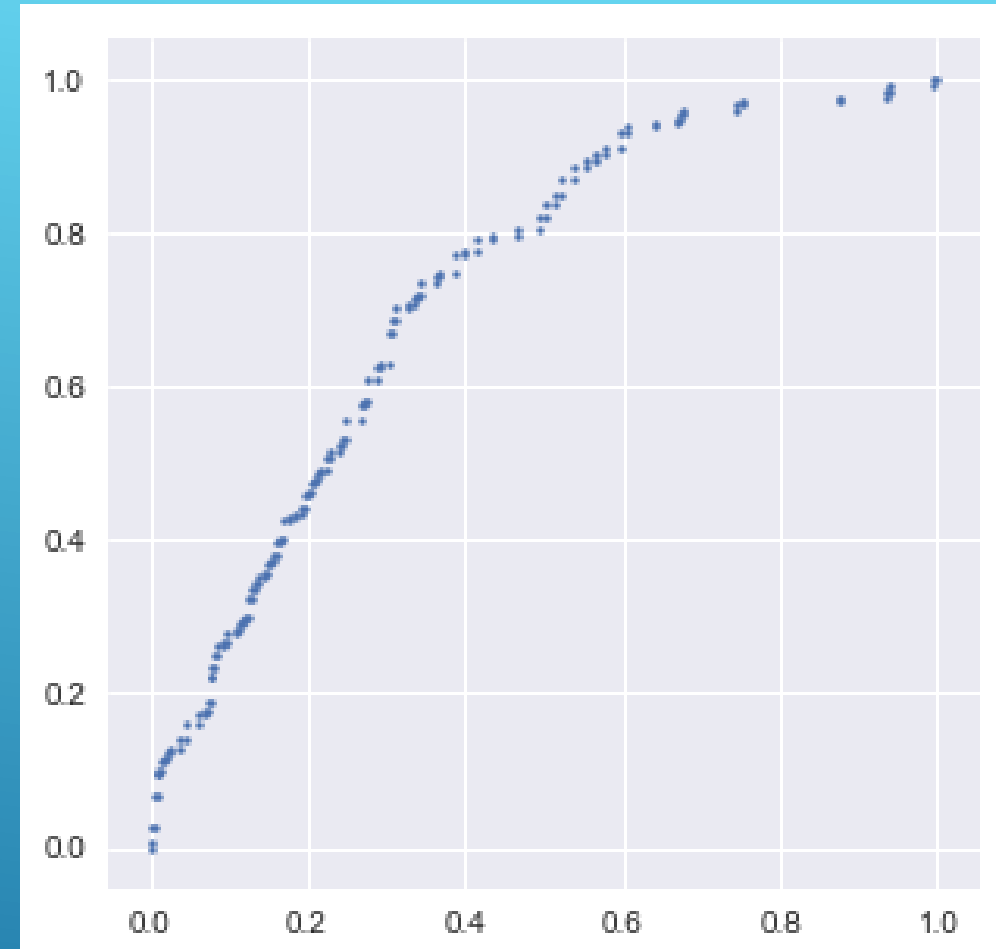
Logistic Regression selected as final model

Test Accuracy: **66.36%**

MODEL INSPECTION

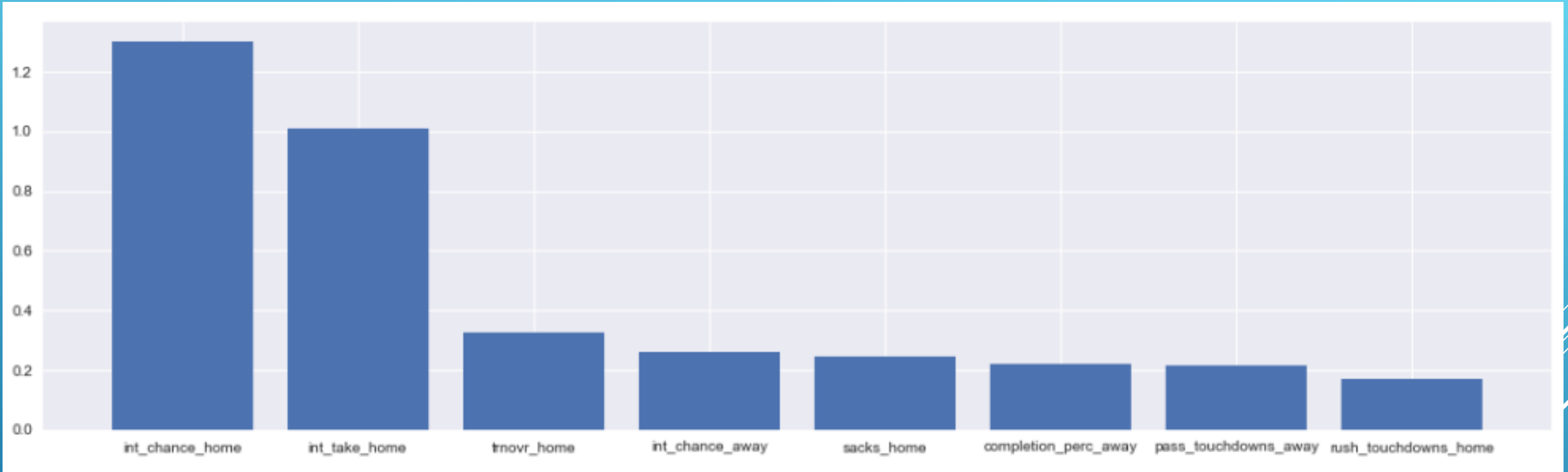


- ▶ Confusion Matrix behaving normally
- ▶ Even counts of misclassification



- ▶ ROC Curve is also normal
- ▶ Low curve shows room performance increase

FEATURE ANALYSIS



**** INT CHANCE = (PASS-RUN RATIO OF OPPONENT * INTERCEPTION TAKEAWAYS) ****

- ▶ Pass to run ratio important to both home and away teams
- ▶ Turnovers and takeaways (turnover differential) very important
- ▶ Sacks for home team
- ▶ Completion percentage for away team

IMPROVEMENTS

Very Noisy Data

- ▶ Adding features caused overfitting
- ▶ Model performed best with limited features
- ▶ Decrease noise with more relevant features
- ▶ Further explore feature Engineering

Hierarchical Models?

- ▶ Lots of correlation between features
- ▶ Build relationships between features for a structured model

SOURCES

Revenue

<https://qz.com/1383416/amid-controversy-the-nfl-is-still-thriving-financially/>

Analyst Predictions

<https://www.fantasyfootballnerd.com/nfl-picks/accuracy>