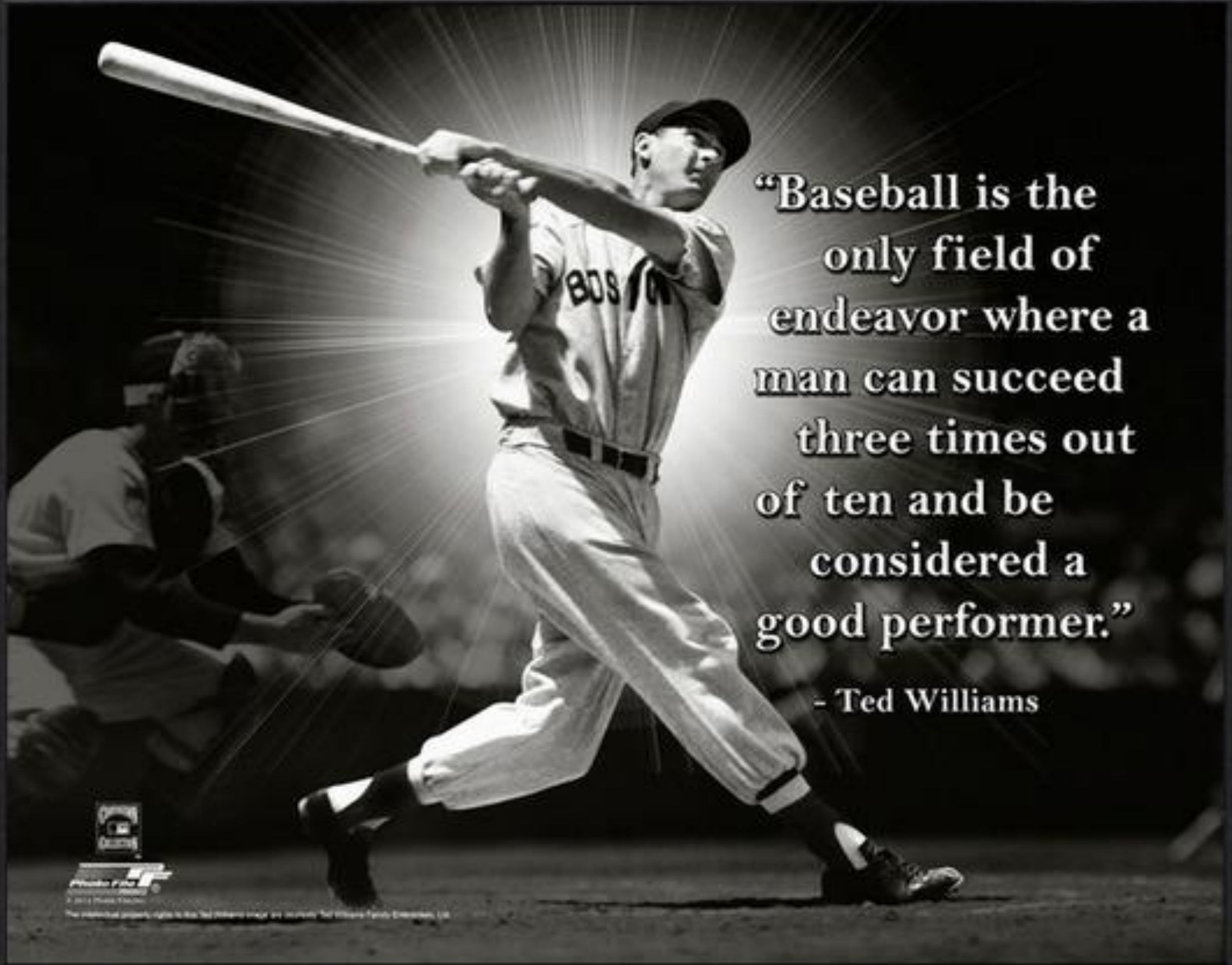


Predicting the Next Pitch



Alex Bell
Metis Data Science



“Baseball is the
only field of
endeavor where a
man can succeed
three times out
of ten and be
considered a
good performer.”

- Ted Williams



Photo by
A. J. ...

The individual property rights in this Ted Williams image are owned by Ted Williams Family Enterprises, LLC

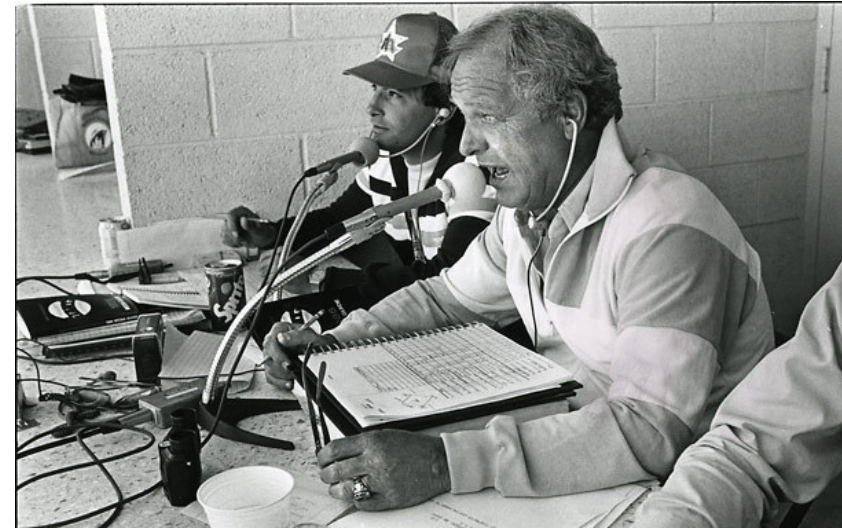
“Guessing what the pitcher is going to throw is 80 percent of being a successful hitter. The other 20 percent is just execution.”

Hank Aaron

- **Predictions can help improve hitter performance**
- **Predictions can help pitchers decrease their predictability**
- **Predictions can help teams increase their win totals, win championships, and rake in the cash!**



The Data:

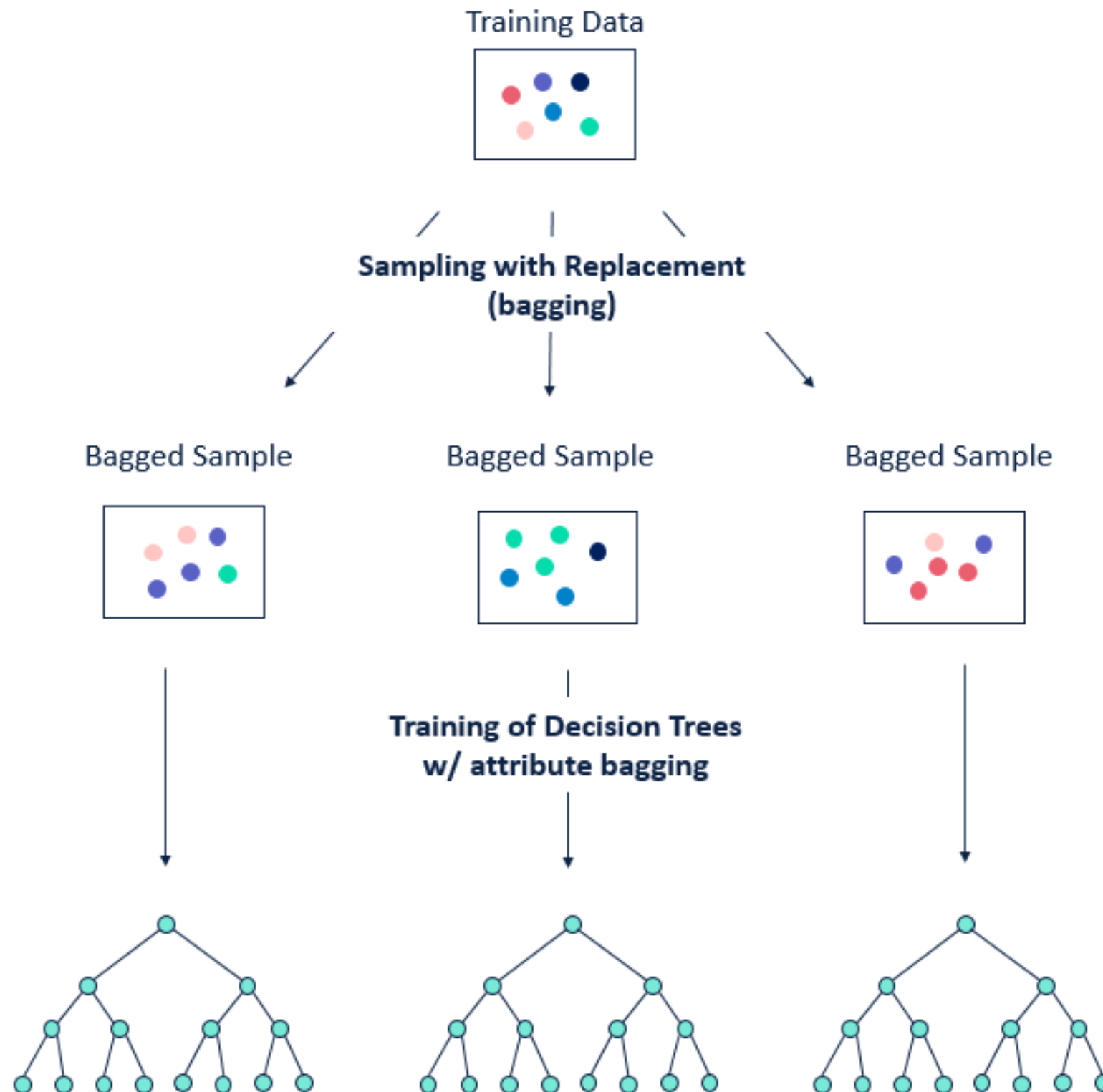


- Characteristics of 720,000+ Major League Baseball Pitches 2016 - 2018 from baseballsavant.com
- Generated Data Tables for Individual Pitchers
- Tables Contain 88 Features for Classifying Pitch Types
- Pitches Grouped into Multiple Classes:
Fastball Breaking Ball Off-Speed

Distribution of Pitch Classes



Random Forest Classification Method

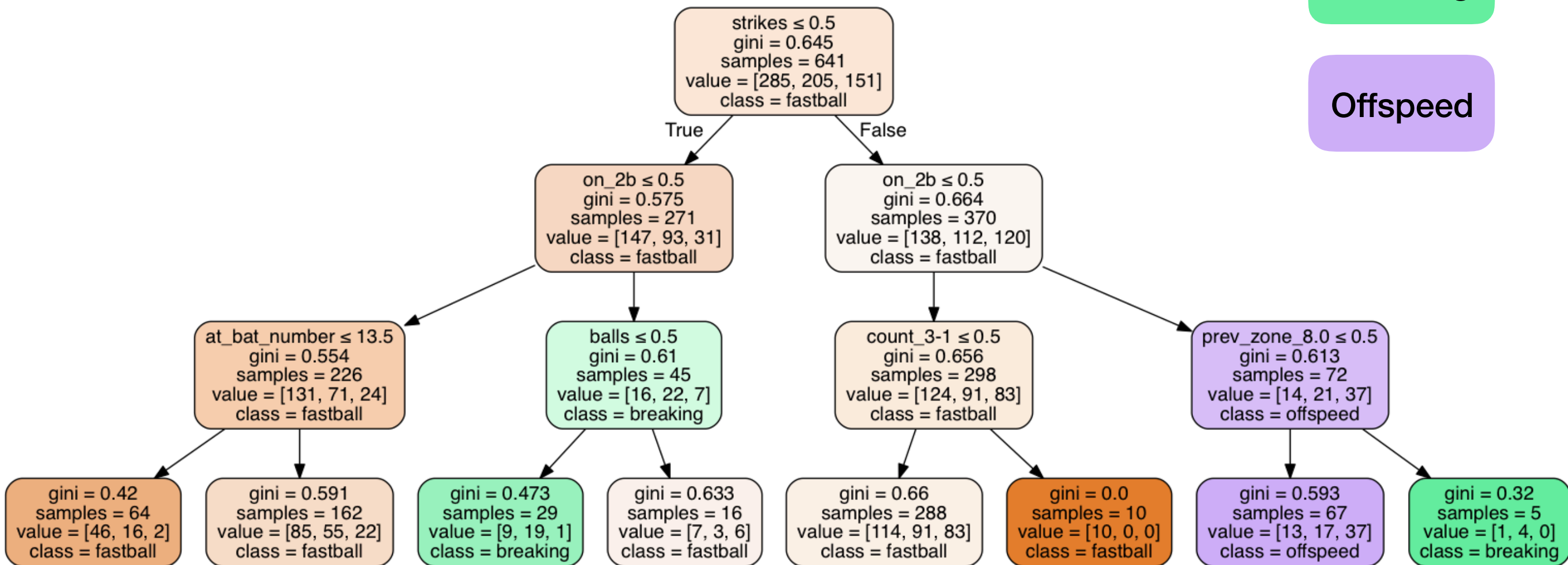


Sample Decision Tree From Random Forest Model

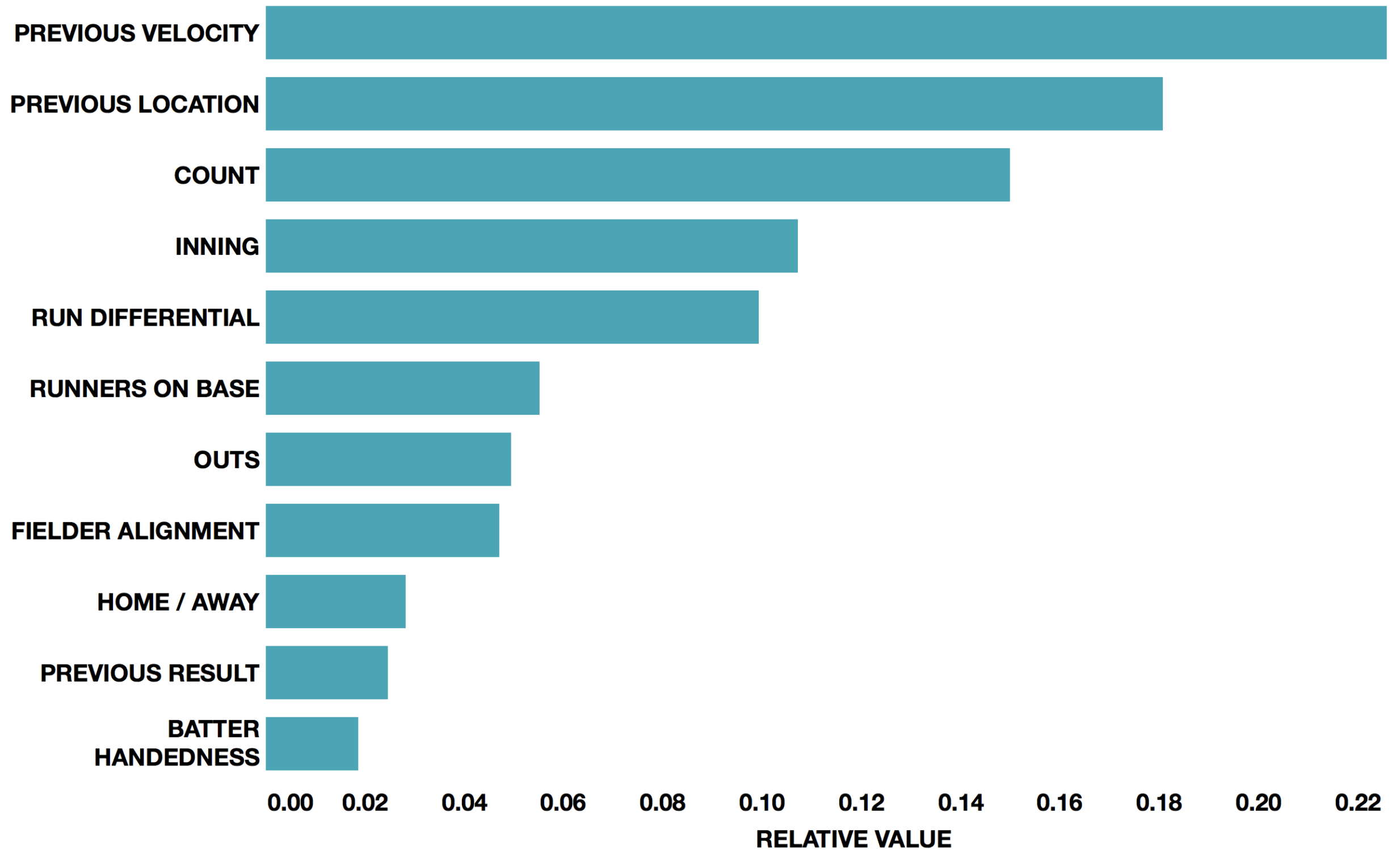
Fastball

Breaking

Offspeed



Feature Importance



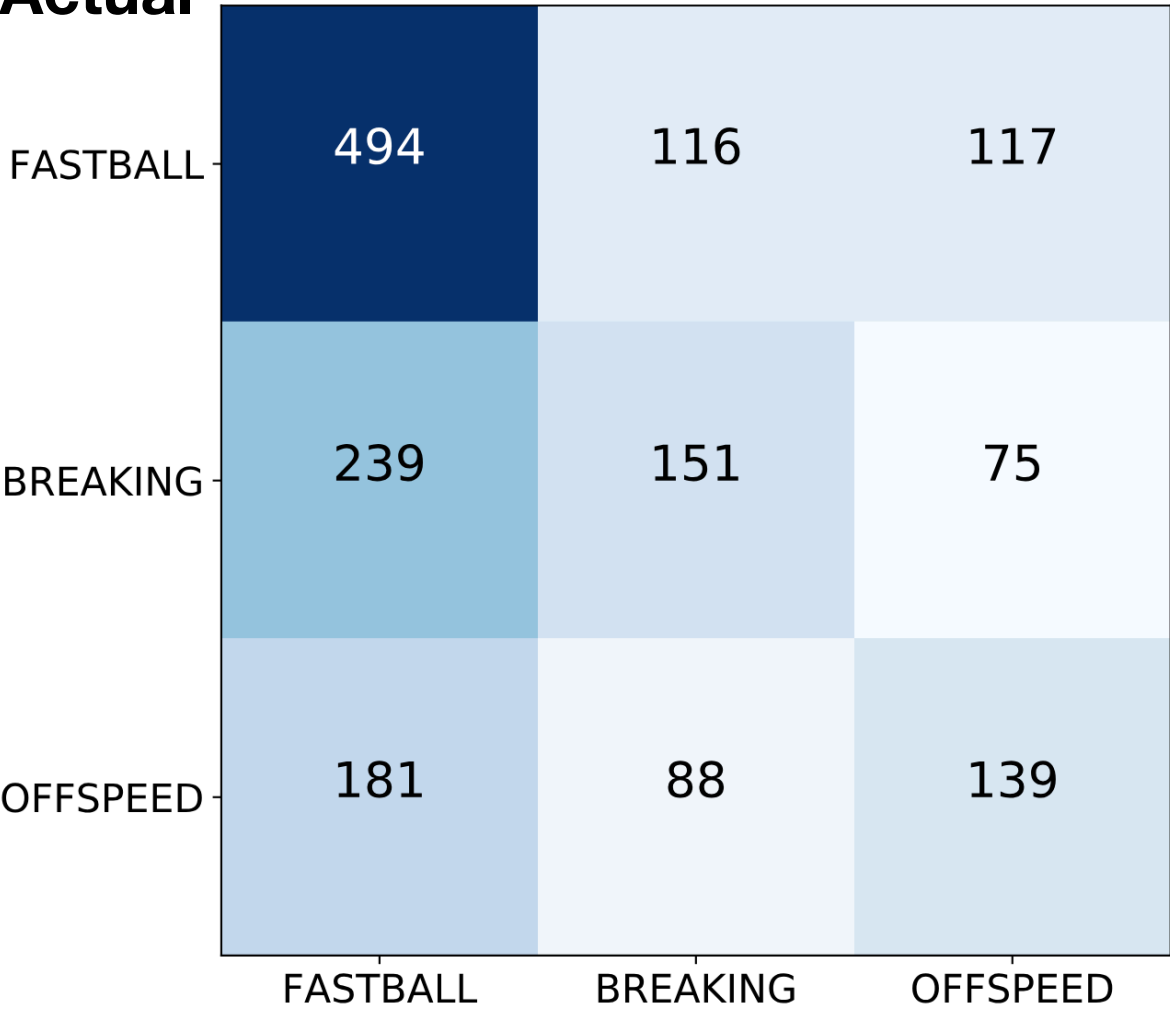
Felix Hernandez Pitch Types

Fastball Accuracy 68%

Overall Accuracy 49%

Random Forest Model

Actual



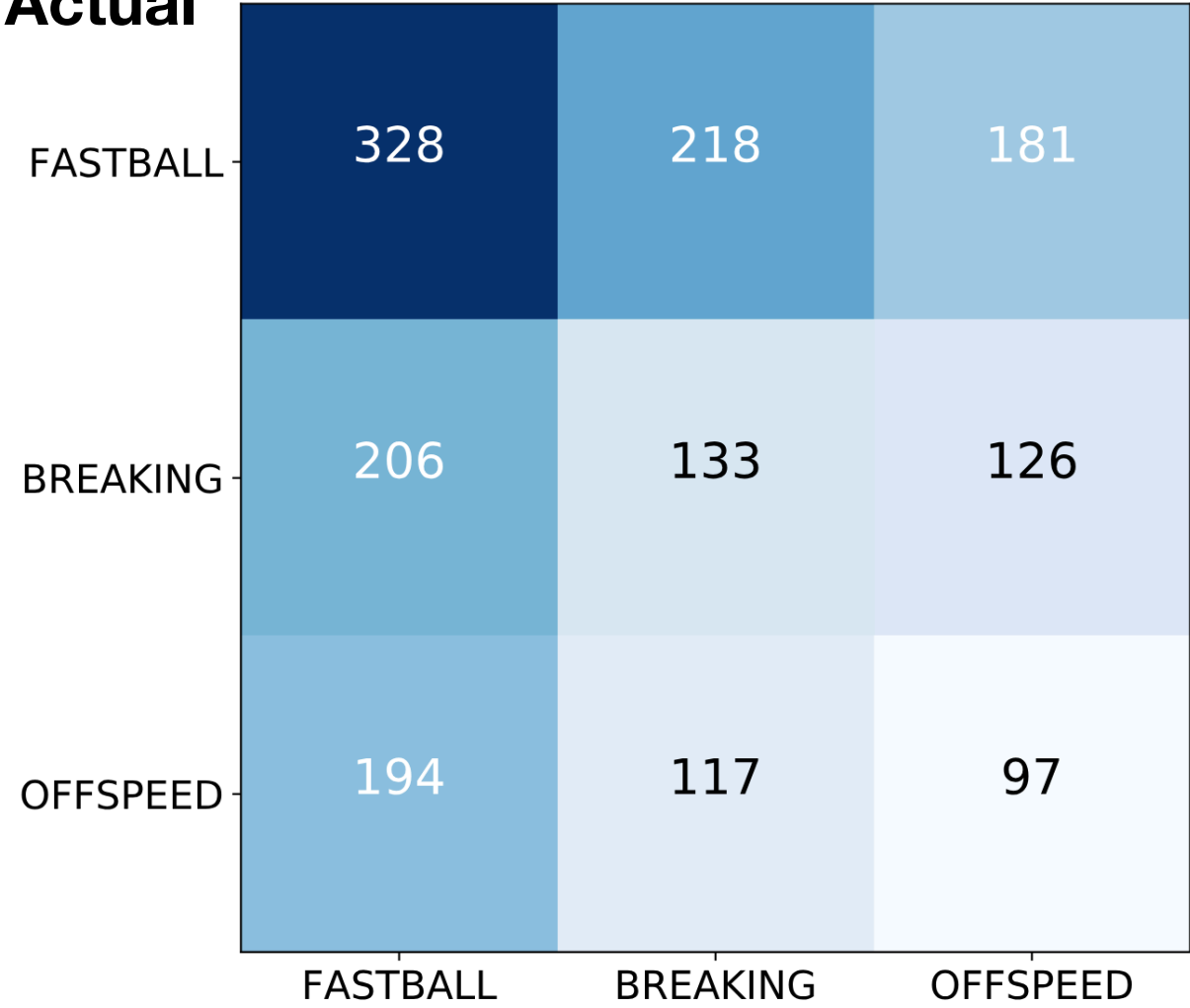
Predicted

Fastball Accuracy 45%

Overall Accuracy 35%

Stratified Naive Model

Actual



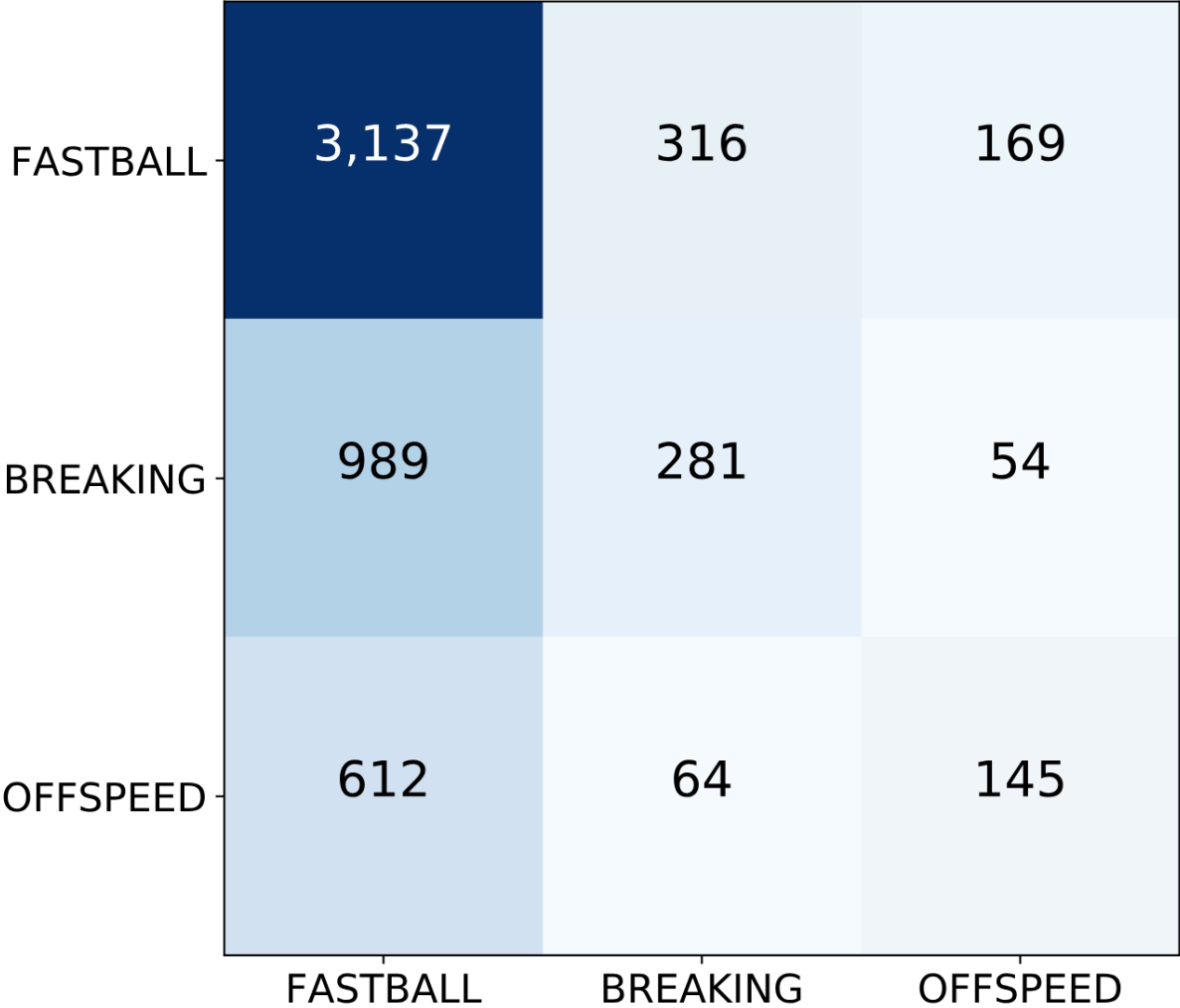
Predicted

Seattle Mariners 2018

Fastball Accuracy 87%

Overall Accuracy 62%

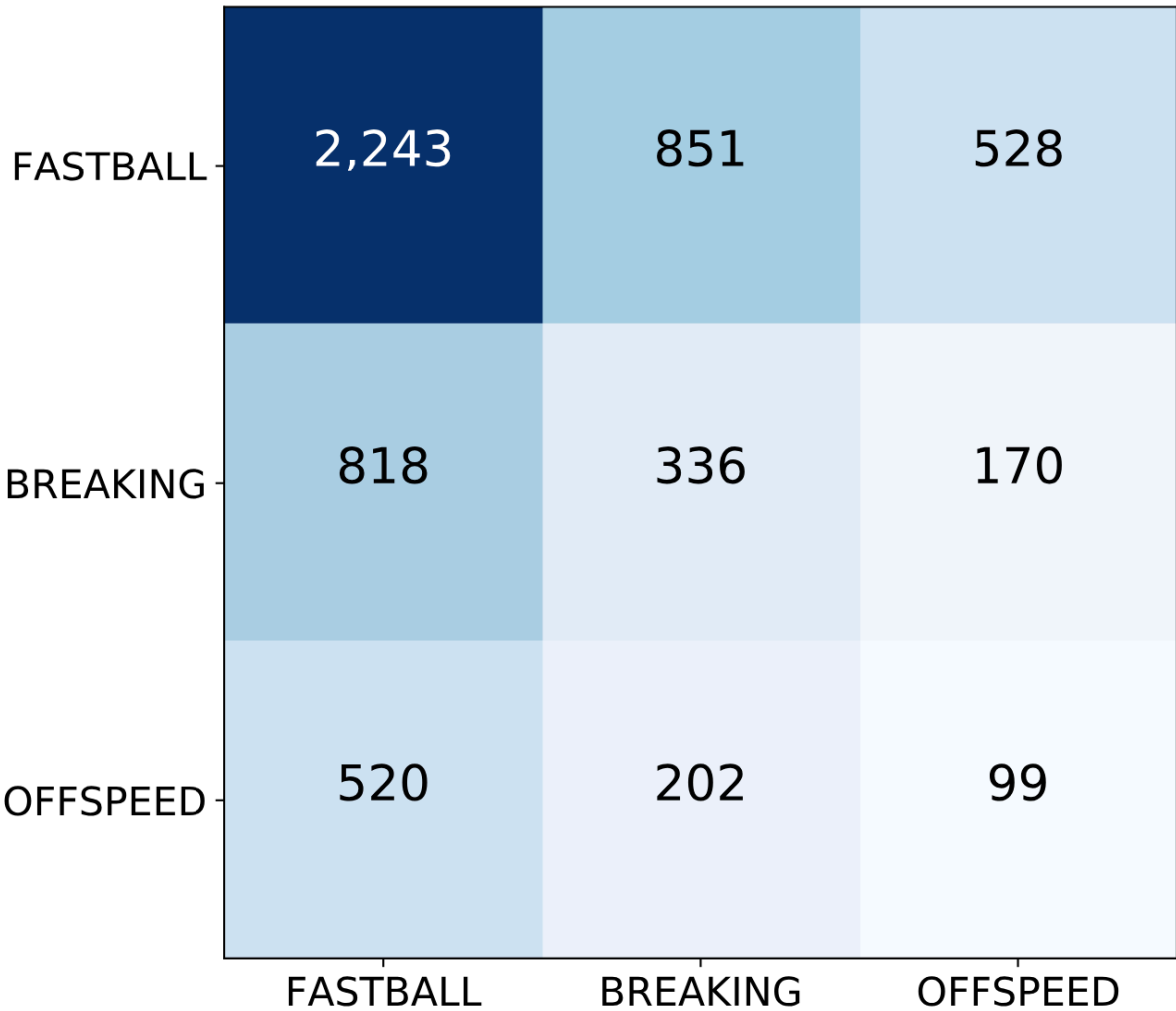
Random Forest Model



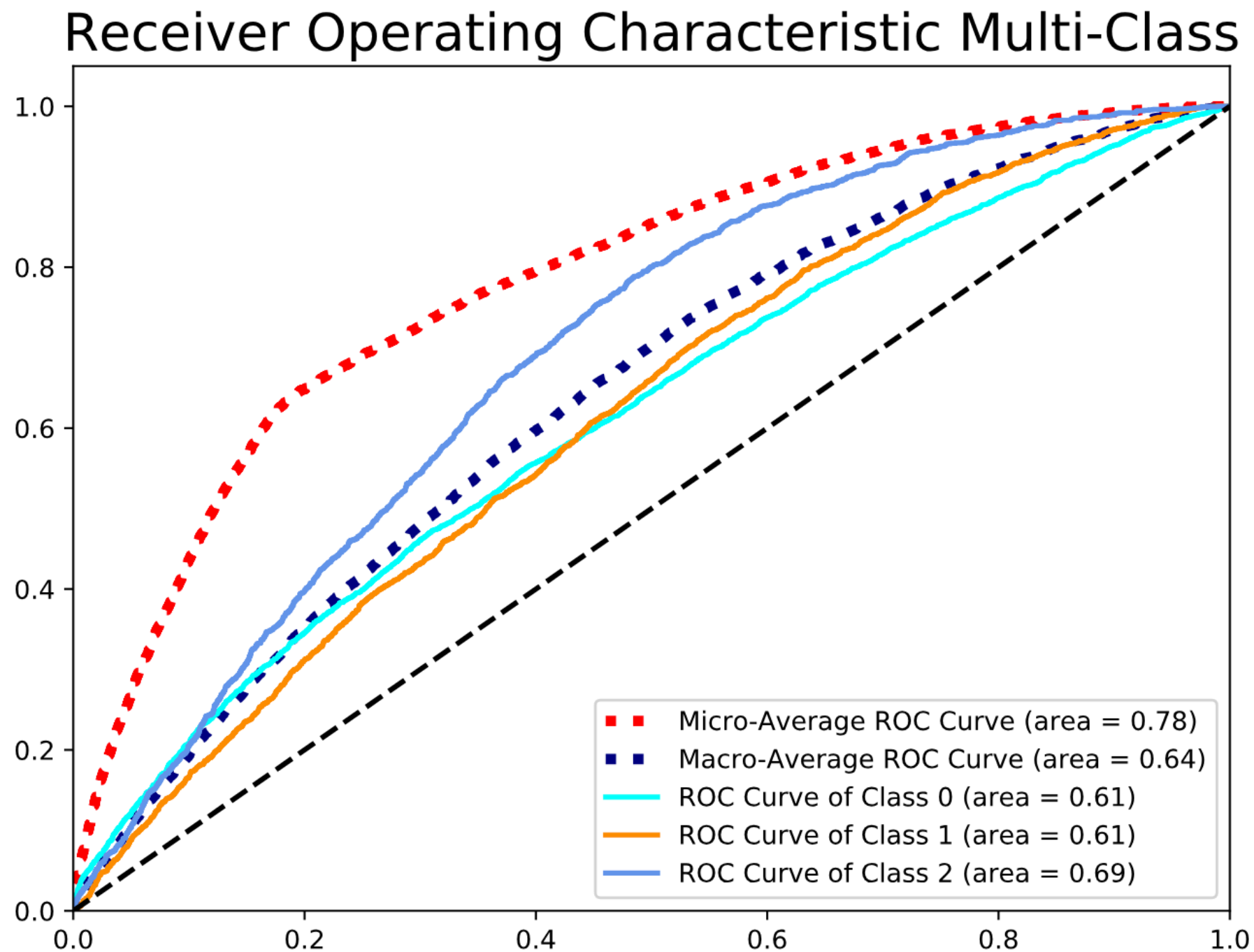
Fastball Accuracy 62%

Overall Accuracy 47%

Stratified Naive Model



One vs All Methodology



Next Steps

- Combine Data from Specific Batter-Pitcher Match-ups
- Expand Predictions to Pitch Location Zones
- Develop Application to Make Live In-Game Predictions

Thank You



Questions?



| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.54 | 0.68 | 0.60 | 727 |
| 1 | 0.43 | 0.32 | 0.37 | 465 |
| 2 | 0.42 | 0.34 | 0.38 | 408 |
| micro avg | 0.49 | 0.49 | 0.49 | 1600 |
| macro avg | 0.46 | 0.45 | 0.45 | 1600 |
| weighted avg | 0.48 | 0.49 | 0.48 | 1600 |

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.45 | 0.45 | 0.45 | 727 |
| 1 | 0.33 | 0.33 | 0.33 | 465 |
| 2 | 0.24 | 0.23 | 0.23 | 408 |
| micro avg | 0.36 | 0.36 | 0.36 | 1600 |
| macro avg | 0.34 | 0.34 | 0.34 | 1600 |
| weighted avg | 0.36 | 0.36 | 0.36 | 1600 |

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.66 | 0.87 | 0.75 | 3622 |
| 1 | 0.43 | 0.21 | 0.28 | 1324 |
| 2 | 0.39 | 0.18 | 0.24 | 821 |
| micro avg | 0.62 | 0.62 | 0.62 | 5767 |
| macro avg | 0.49 | 0.42 | 0.43 | 5767 |
| weighted avg | 0.57 | 0.62 | 0.57 | 5767 |

| | precision | recall | f1-score | support |
|--------------|-----------|--------|----------|---------|
| 0 | 0.62 | 0.62 | 0.62 | 3622 |
| 1 | 0.23 | 0.25 | 0.24 | 1324 |
| 2 | 0.15 | 0.14 | 0.14 | 821 |
| micro avg | 0.47 | 0.47 | 0.47 | 5767 |
| macro avg | 0.33 | 0.33 | 0.33 | 5767 |
| weighted avg | 0.47 | 0.47 | 0.47 | 5767 |