

# Group 20: Kobe Bryant

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# Challenge

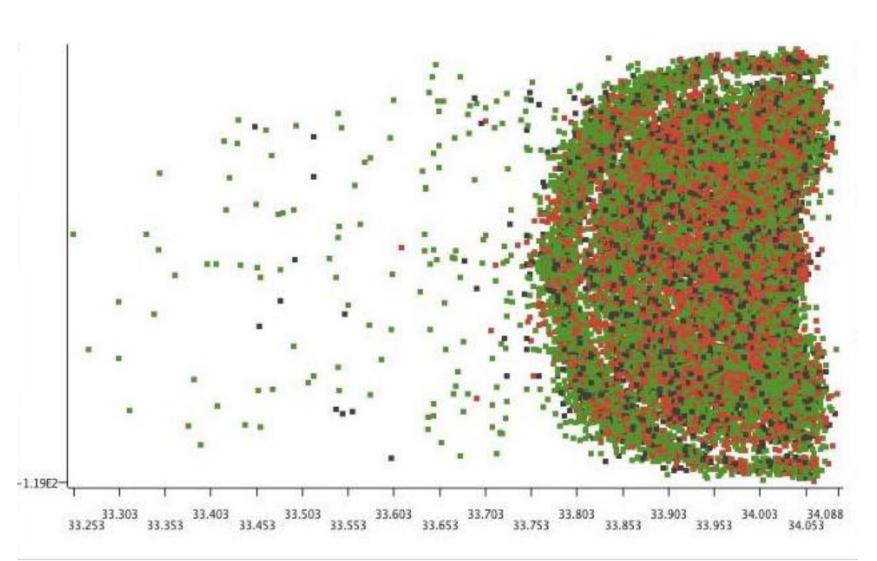
## 0 - Shots Missed

- Shots Made

Predict whether Kobe Bryant will make a shot or not.

# Dataset

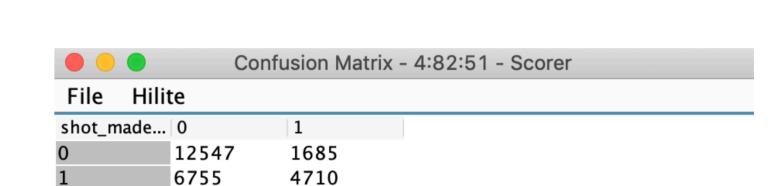
- 6 types of different shots played by Kobe in provided data set.
- 30,697 records.
- 7 categorical features, with 2-8 possible values each. There are also 11 Numerical Features and 4 nominal features in specific.



- Shots made by Kobe from different zones in the court

# Data Preparation

Fig. 2 - Types of Shots made by Kobe, Jumpshot attempts are highest



# Data Understanding

- · Values were not fully represented for · Selected Attributes: some features.
- Dataset is unbiased with respect to **Target Attribute**

- - Shot Made Flag
  - Shot Type
  - Latitude, Longitude
- Minutes remaining, seconds remaining
- Shot Zone Area
- 5,000 missing values

# Modeling

- Final Partitioning:
  - 80% Training Data
  - 20% Test Data
- Classification Models used:
  - Naive-Bayes
  - Random Forest
  - Decision Trees
  - Logistic Regression
- The quality measure is to minimise **Error rate and Increase accuracy** in prediction whether Kobe will make a shot or not.
- Final Model
  - Random Forest
  - Information Gain Split criterion

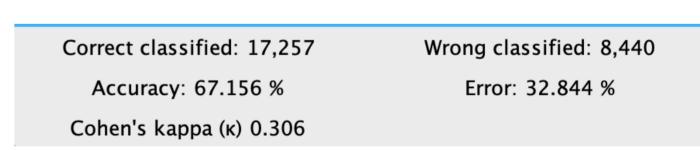


Fig. 3 - Confusion Matrix for Random Forest

# Results and Evaluation

Models	Accuracy
Decision Tree	67.23%
Random Forest	67.15%
Logistic Regression	68.11%
Naive Bayes	61.52%

 Data on left shows the accuracy rate for various models tested on dataset.

# Conclusion

Based on reason that Random Forest works good on Categorical Target attribute as compared to Logistic Regression that generally works on continuous class labels, as well as its performance is better as compared to single decision trees, we chose RF as final model.

# Knime-Workflow

