Predicting the 2022 World Cup Winner

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Team Index: 90

Goal

Use world cup performance data and socioeconomic data to predict the 2022 World Cup winner

Motivation

5,000,000,000

worldwide viewers

227,270,000

viewers per day

Data



Data Collection

Data Needed:

- Matchup Data
 - All Matchups between two Countries in World Cup History
 - o Ex: Argentina vs France, etc.
- Socioeconomic Data
 - For each Country and World Cup Year
 - Ex: GDP per Capita, Unemployment Rate, etc.
- Soccer Data
 - Each Country's Performance in Each World Cup
 - Ex: Goals, Assists, etc.



- FbRef
 - Matchup Data and Soccer Data
- MacroTrends
 - Socioeconomic Data





Data Description

- Each Row Represents a Matchup between Team A and Team B
- For Every Feature, we will have Team A's value, Team B's value, and the Difference of the Two Values
- We will also record whether Team A recorded a Win, Loss, or Draw
- Ex:

Year	Team A	Team B	GDP	GDP	GDP Growth Difference	Result
2018	France	Croatia	1.87 %	2.9 %	-1.03 %	Win

Data Description (Cont'd):

- But wait! We also need to record the matchup between Team B and Team A
- The one row only records Team A's result, not Team B's result
 - We don't want to lose any data when creating our DataFrame
- Ex:

Year	Team A	Team B	Team A GDP Growth	GDP	GDP Growth Difference	Result
2018	France	Croatia	1.87 %	2.9 %	-1.03 %	Win
2018	Croatia	France	2.9 %	1.87 %	1.03 %	Loss

Data Description (Cont'd):

- Overall, we recorded every matchup in World Cup History between a given Team A and Team B
 - Going back to the 1st World Cup in 1930

- For predicting the 2022 World Cup, we have:
 - All matchups in the Group Stage
 - We will build the World Cup Bracket after predicting the winners of the Group Stage
 - Each team's most recently recorded socioeconomic data
 - Each team's soccer stats from 2022 World Cup Qualification

What ML Method Will We Use?

We know the categories we want to predict:

(Win, Loss, or Draw)

Therefore, we will use <u>Classification</u>

Method

- Decision Tree Classifier
- K-Nearest Neighbor Classifier:
 - Wins
 - Losses
 - Draws

- Nearest Neighbors Classifier:
- Socioeconomic vs.
 performance factors
 - All factors

- Random Forest
- **Classifier:**
 - Feature Importances
- Nearest Neighbors Classifier:
 - Top 10 factors

Method

- Decision Tree Classifier
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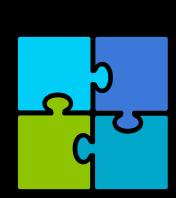
- Random Forest Classifier:
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- **Classifier:**
 - Top 10 factors

Method

- Decision Tree Classifier
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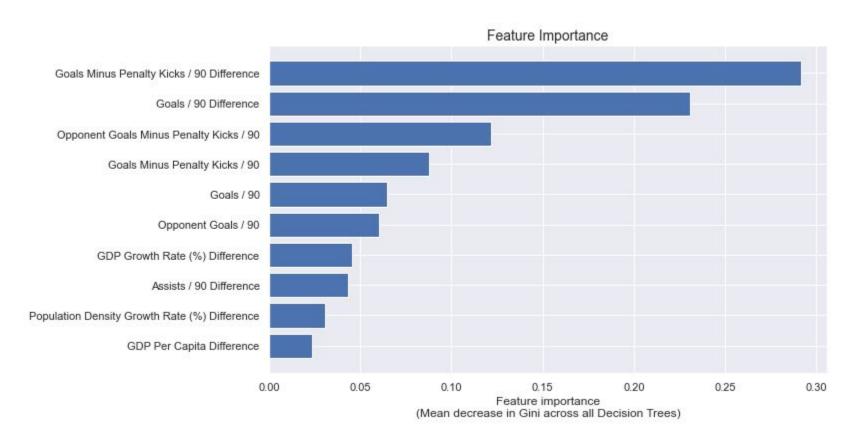
- Nearest Neighbors Classifier:
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- Random Forest Classifier:
 - Feature Importances
- Nearest Neighbors Classifier:
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Results

Random Forest Classification & Feature Importance

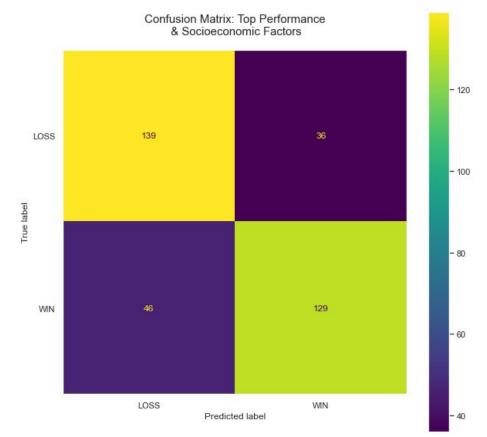


Machine Learning & KNN Classification

Accuracy Score: 0.76

Other Models

- Top Performance Factors
 - Accuracy score = 0.71
- Top Socioeconomic Factors
 - Accuracy score = 0.58



Note. Draws were excluded from model

Our Prediction: 2022 WCF Group Stage Winners

Group A: Netherlands, Qatar

Group B: England, IR Iran

Group C: Argentina, Saudi Arabia

Group D: Denmark, France

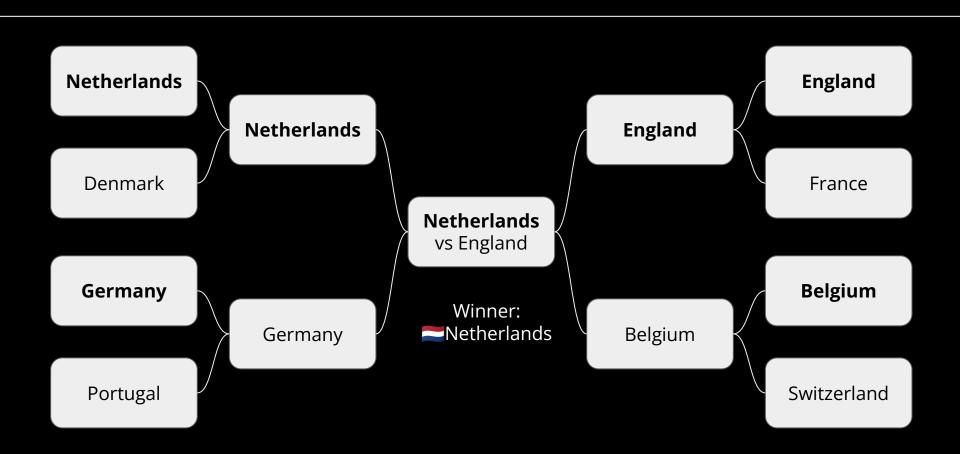
Group E: Germany, Japan

Group F: Belgium, Canada

Group G: Brazil, Switzerland

Group H: Portugal, Korea Republic

Our Prediction



Key Takeaways



Challenges

- Our initial model had extreme difficulty predicting draws
- Much more data available in 2018 that would improve performance of our model

Future Considerations

- KNN classifier built off the 10 most important factors but those factors were all given the same weight
 - When training our KNN classifier we only used historical world cup data

Questions?