

## **Project Overview**

### **Target Audience**

Anyone who plays Fantasy Football can utilize our app to help guide them in their decision making process through data analysis.

### **Motiviations**

To create a model that accurately predicts the performance of chosen NFL players based on past performance, weather conditions, stadium, years playing, and average yearly fantasy league points scored.

#### **Tools Used:**

Linear Regression, Flask, Javascript, Pandas, CSS

## Data Overview

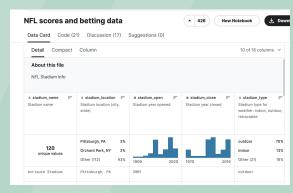
### **Primary Data Source:**

NFL scores and betting data

### NFL Stats 2012-2023

- 73 different columns that reference player stats
- 5453 rows
- Stadium, weather, & field type data

### **Site Home Pages**





3/0

4/0

50

40

3/0

## **Processing Pipeline**

### Goal

- Ensure data for prior seasons are available, accurate, and consistent across players and seasons including changing team names, stadium names, games per season, etc.
- Create a merged dataset from multiple CSV's and spreadsheets



### **Process**

- Downloaded required data from kaggle and updated fields for consistency
- Merged stadium and player data
- Output new CSV's for model building

## **Linear Regression Model**

- Used many NFL offensive stats to help predict fantasy scores (None that near perfect correlation to fantasy scores)
- Back tested model with 2023 model predictions and actual 2023 fantasy results
- Predicted 2024 fantasy football scores for upcoming season

```
no_2023_lr_model = LinearRegression()

# Training model without 2023 season data
no_2023_lr_model.fit(X_no_2023, y_no_2023)

* LinearRegression
LinearRegression()
```



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## **Model Optimization**

Added new columns that helped increase performance of model

RMSE increased by 18.59%

R2 increased by 1.37%

opt\_model\_all\_seasons\_data['target\_per\_game'] = round((opt\_model\_all\_seasons\_data['targets']/opt\_model\_all\_seasons\_data['games']),2)
opt\_model\_all\_seasons\_data['carries\_per\_game'] = round((opt\_model\_all\_seasons\_data['carries']/opt\_model\_all\_seasons\_data['games']),2)
opt\_model\_all\_seasons\_data['team\_off\_snaps\_per\_game'] = round((opt\_model\_all\_seasons\_data['teams\_offense\_snaps']/opt\_model\_all\_seasons\_data['games']
opt\_model\_all\_seasons\_data['off\_snaps\_per\_game'] = round((opt\_model\_all\_seasons\_data['offense\_snaps']/opt\_model\_all\_seasons\_data['games']),2)
opt\_model\_all\_seasons\_data['attempts\_per\_game'] = round((opt\_model\_all\_seasons\_data['attempts']/opt\_model\_all\_seasons\_data['games']),2)

### **PRE-Optimization**

# [32]: import math mse\_2023 = mean\_squared\_error(fa rmse\_2023 = math.sqrt(mse\_2023) rmse\_2023 [32]: 17.481317677683556 [33]: r2\_2023 = r2\_score(fantasy\_2023\_ r2\_2023 [33]: 0.9607669583952572

RMSE: 17.48

R2: 96.08



-3.25

### **POST-Optimization**

RMSE: 14.23

R2: 97.40

30 40 50

### **Model Results**

- RMSE of 71.00 for 2022 vs 2023 actual fantasy scores proved that our model was not just assuming a players last year performance but including their whole career performance
- RMSE => Predictions for 2024 should be expected to be +-14.23 points different on average
- R2 => 97.40% of the variation in predicted fantasy football scores can be explained by the included features or statistics

### Official 2024 Season Predictions (Top 10 Players)

	name	team	position	fantasy_2024_score_prediction	fantasy_2024_per_week_score_prediction
0	Josh Allen	BUF	QB	318.60	18.74
1	Puka Nacua	LA	WR	318.36	18.73
2	Justin Herbert	LAC	QB	308.04	18.12
3	Justin Jefferson	MIN	WR	297.64	17.51
4	Lamar Jackson	BAL	QB	293.24	17.25
5	Patrick Mahomes	KC	QB	290.89	17.11
6	Trevor Lawrence	JAX	QB	284.14	16.71
7	CeeDee Lamb	DAL	WR	277.58	16.33
8	Alvin Kamara	NO	RB	274.51	16.15
9	Tyreek Hill	MIA	WR	273.10	16.06

### 2022 VS 2023 Actual

mse\_assume = mean\_squared\_error(fa
rmse\_assume = math.sqrt(mse\_assume
rmse\_assume
71.00374769451727

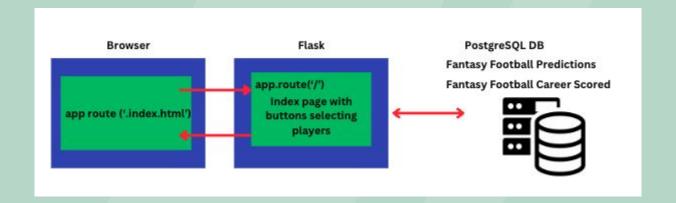
### **Final Model Evaluation**

[37]: # Calculating RMSE for model bas
mse\_2023 = mean\_squared\_error(fit
rmse\_2023 = math.sqrt(mse\_2023)
rmse\_2023

[37]: 14.229830865136156



## **Application Architecture**



30 40 50 40 30

## Demo



## Challenges & Next Steps

### **Challenges**

- Weekly data did not match yearly data.
- Could not locate and incorporate datasets including weather or date information.

### **Next Steps**

- Integrate with the ESPN API to create a live application that accounts for player injuries.
- Develop a model to predict week-over-week outcomes.
- Add in respective data for a WRs/TEs/RBs QB stats and vice versa.

# THANK YOU!

