

Premier League Dashboard

Kiran Seetharam, Kaushik Nandan
Prayag Ganesh Prabhu, Tejas Pratap, Pratik Sabne

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1 Introduction

1.1 Problem Statement

We aim to develop a user-friendly Premier League dashboard that provides real-time match predictions, updating after each match and resetting at the end of the season. This dashboard will feature a dynamic table predictor, ensuring users stay informed with the latest match outcomes and predictions. Additionally, users will have access to various interactive graphs and charts representing key attributes utilized in our prediction model. Our goal is to offer an intuitive platform for football enthusiasts to analyze and engage with Premier League data effectively.

1.2 Inspiration

Due to the growing interest in data-driven decision-making across various industries, including sports, we have decided to propose this idea. Predicting the outcome of sports events has always captivated enthusiasts, and the Premier League, being one of the most popular and competitive football leagues worldwide, offers an exciting challenge for data scientists and has a potential for lots of learning.

1.3 Background

This project builds upon existing work and background research in sports analytics and predictive modeling. Several studies have explored predictive models for sports outcomes, including football. Researchers have investigated various factors such as team statistics, player performance, playing style, and historical results to develop predictive models. However, given the dynamic nature of the Premier League and the interplay of multiple variables, there is a constant need for improved models that capture the complexity of the competition.

2 Method

Our methodological approach is delineated below, encompassing a systematic series of steps:

- **Data Acquisition:** To ensure the integrity and reliability of our dataset, we meticulously gathered comprehensive Premier League standings, statistics, and match results spanning from the 2000-2001 season to the present campaign. This process entailed leveraging web scraping techniques on the *fbref.com* platform [1], utilizing the powerful Python libraries *Beautiful Soup*, *NumPy*, and *Pandas*.
- **Dataset Composition:** The resultant dataset encapsulates matchwise data from the 2000-01 season onwards, comprising a diverse array of meticulously collected attributes and calculated metrics. Key components of this dataset include team names, manager names, full-time results, match-specific details such as dates, full-time and half-time goals, as well as auxiliary metrics such as teams' head-to-head records, managers' head-to-head records, past 5 match form, and expected goals (xG).
- **Model Development and Prediction:** Leveraging the acquired xG data, we employed the robust *xG Boost Random Forest Regressor Model* [2] to facilitate predictive analysis. This model assimilates historical xG data and employs sophisticated algorithms to generate accurate match outcome predictions, thereby enhancing the dashboard's prognostic capabilities.
- **Dashboard Creation:** In a bid to furnish users with an immersive and intuitive platform for data exploration, all scraped and computed data, in conjunction with the prediction model, were seamlessly integrated into *Power BI*. Leveraging the platform's extensive visualization capabilities, we crafted an

interactive dashboard that offers users a rich and dynamic visual experience. The incorporation of drill-through features enables seamless navigation between different pages, facilitating access to nuanced and meaningful insights within the dataset.

3 Result

Our prediction model yields results with an accuracy rate of 54.5%, a commendable achievement given the dual task of predicting match outcomes and the number of goals scored by teams. This multifaceted approach underscores the complexity of our model and its ability to provide nuanced insights into Premier League matches. The accompanying figures offer a glimpse into the comprehensive nature of our project.

The first figure showcases a page from our Premier League Dashboard, providing users with a visually appealing and intuitive interface for accessing match predictions and insights. Subsequent figures provide glimpses into segments of our extensive dataset, demonstrating the breadth and depth of information incorporated into our analysis. Lastly, the xG dataset, displayed in the final figure, serves as a foundational component for predicting match results and offers valuable insights into the expected goals scored by teams, enriching the predictive capabilities of our model.

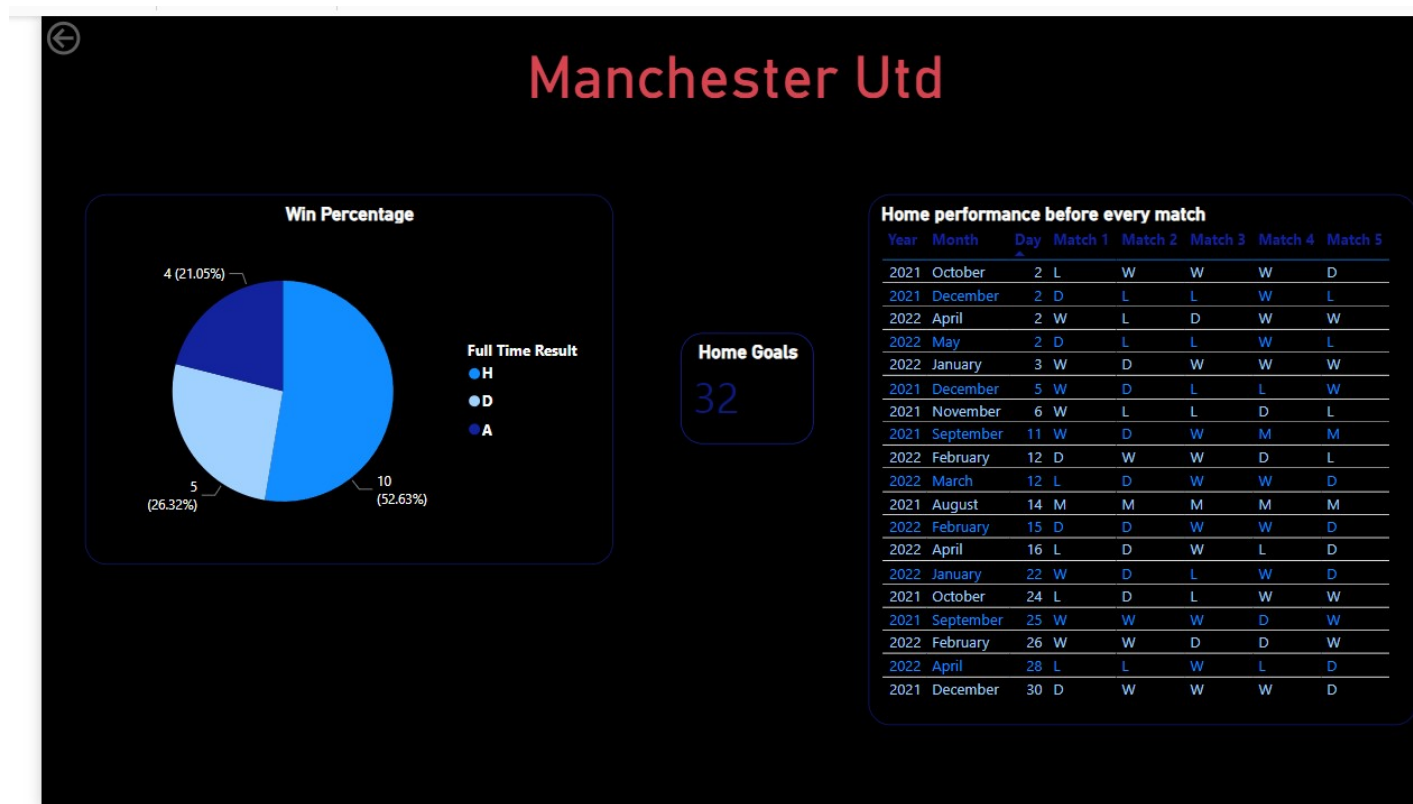


Figure 1: Power Bi Dashboard Page

| Date | HomeTeam | AwayTeam | Home Team Manager | Away Team Manager | FTHG | FTAG | FTR | HTH_hom | HTH_away | HTGS |
|------------------|-----------------|-----------------|-------------------|-------------------|------|------|-----|---------|----------|------|
| 19-08-2000 00:00 | Charlton Ath | Manchester City | Alan Curbishley | Joe Royle | 4 | 0 | H | 0 | 0 | |
| 19-08-2000 00:00 | Chelsea | West Ham | Gianluca Vialli | Harry Redknapp | 4 | 2 | H | 0 | 0 | |
| 19-08-2000 00:00 | Coventry City | Middlesbrough | Gordon Strachan | Bryan Robson | 1 | 3 | A | 0 | 0 | |
| 19-08-2000 00:00 | Derby County | Southampton | Jim Smith | Glenn Hoddle | 2 | 2 | D | 0 | 0 | |
| 19-08-2000 00:00 | Leeds United | Everton | David O'Leary | Walter Smith | 2 | 0 | H | 0 | 0 | |
| 19-08-2000 00:00 | Leicester City | Aston Villa | Peter Taylor | John Gregory | 0 | 0 | D | 0 | 0 | |
| 19-08-2000 00:00 | Liverpool | Bradford City | G  rard Houllier | Chris Hutchings | 1 | 0 | H | 0 | 0 | |
| 19-08-2000 00:00 | Sunderland | Arsenal | Peter Reid | Ars  ne Wenger | 1 | 0 | H | 0 | 0 | |
| 19-08-2000 00:00 | Tottenham | Ipswich Town | George Graham | George Burley | 3 | 1 | H | 0 | 0 | |
| 20-08-2000 00:00 | Manchester Utd | Newcastle Utd | Alex Ferguson | Bobby Robson | 2 | 0 | H | 0 | 0 | |
| 21-08-2000 00:00 | Arsenal | Liverpool | Ars  ne Wenger | G  rard Houllier | 2 | 0 | H | 0 | 0 | |
| 22-08-2000 00:00 | Bradford City | Chelsea | Chris Hutchings | Gianluca Vialli | 2 | 0 | H | 0 | 0 | |
| 22-08-2000 00:00 | Ipswich Town | Manchester Utd | George Burley | Alex Ferguson | 1 | 1 | D | 0 | 0 | |
| 22-08-2000 00:00 | Middlesbrough | Tottenham | Bryan Robson | George Graham | 1 | 1 | D | 0 | 0 | |
| 23-08-2000 00:00 | Everton | Charlton Ath | Walter Smith | Alan Curbishley | 3 | 0 | H | 0 | 0 | |
| 23-08-2000 00:00 | Manchester City | Sunderland | Joe Royle | Peter Reid | 4 | 2 | H | 0 | 0 | |
| 23-08-2000 00:00 | Newcastle Utd | Derby County | Bobby Robson | Jim Smith | 3 | 2 | H | 0 | 0 | |
| 23-08-2000 00:00 | Southampton | Coventry City | Glenn Hoddle | Gordon Strachan | 1 | 2 | A | 0 | 0 | |
| 23-08-2000 00:00 | West Ham | Leicester City | Harry Redknapp | Peter Taylor | 0 | 1 | A | 0 | 0 | |
| 26-08-2000 00:00 | Arsenal | Charlton Ath | Ars  ne Wenger | Alan Curbishley | 5 | 3 | H | 0 | 0 | |
| 26-08-2000 00:00 | Bradford City | Leicester City | Chris Hutchings | Peter Taylor | 0 | 0 | D | 0 | 0 | |
| 26-08-2000 00:00 | Everton | Derby County | Walter Smith | Jim Smith | 2 | 2 | D | 0 | 0 | |
| 26-08-2000 00:00 | Ipswich Town | Sunderland | George Burley | Peter Reid | 1 | 0 | H | 0 | 0 | |
| 26-08-2000 00:00 | Manchester City | Coventry City | Joe Royle | Gordon Strachan | 1 | 2 | A | 0 | 0 | |
| 26-08-2000 00:00 | Middlesbrough | Leeds United | Bryan Robson | David O'Leary | 1 | 2 | A | 0 | 0 | |
| 26-08-2000 00:00 | Newcastle Utd | Tottenham | Bobby Robson | George Graham | 2 | 0 | H | 0 | 0 | |
| 26-08-2000 00:00 | Southampton | Liverpool | Glenn Hoddle | G  rard Houllier | 3 | 3 | D | 0 | 0 | |
| 26-08-2000 00:00 | West Ham | Manchester Utd | Harry Redknapp | Alex Ferguson | 2 | 2 | D | 0 | 0 | |
| 27-08-2000 00:00 | Aston Villa | Chelsea | John Gregory | Gianluca Vialli | 1 | 1 | D | 0 | 0 | |
| 05-09-2000 00:00 | Leeds United | Manchester City | David O'Leary | Joe Royle | 1 | 2 | A | 0 | 0 | |

Figure 2: Dataset

| ATLossStr | HTGD | ATGD | DiffPts | DiffFormP | HS | AS | HST | AST | HHW | AHW | Prev_hth | HTH_hom | HTH_away | season | HM1_map | H |
|-----------|----------|----------|----------|-----------|----|----|-----|-----|-----|-----|----------|---------|----------|--------|---------|----|
| 0 | 0 | 0 | 0 | 0 | 0 | 17 | 8 | 14 | 4 | 2 | 1 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 17 | 12 | 10 | 5 | 1 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 6 | 16 | 3 | 9 | 0 | 1 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 6 | 13 | 4 | 6 | 0 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 17 | 12 | 8 | 6 | 0 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 5 | 5 | 4 | 3 | 0 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 16 | 3 | 10 | 2 | 0 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 8 | 14 | 2 | 7 | 0 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 20 | 15 | 6 | 5 | 2 | 1 | 0 | 0 | 0 | 2000 | 0 |
| 0 | 0 | 0 | 0 | 0 | 0 | 19 | 9 | 9 | 6 | 2 | 0 | 0 | 0 | 0 | 2000 | 0 |
| 0 | -0.5 | 0.5 | -1.5 | -1.5 | | 17 | 7 | 12 | 4 | 0 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | -0.5 | 1 | -1.5 | -1.5 | | 12 | 14 | 3 | 6 | 0 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | -1 | 1 | -1.5 | -1.5 | | 13 | 15 | 8 | 6 | 0 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | 1 | 1 | 0 | 0 | | 12 | 11 | 6 | 4 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | -1 | 2 | -1.5 | -1.5 | | 13 | 8 | 8 | 4 | 0 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | -2 | 0.5 | -1.5 | -1.5 | | 15 | 9 | 10 | 4 | 0 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | -1 | 0 | -0.5 | -0.5 | | 9 | 10 | 4 | 5 | 0 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | 0 | -1 | 0.5 | 0.5 | | 12 | 7 | 4 | 5 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | -1 | 0 | -0.5 | -0.5 | | 17 | 4 | 12 | 2 | 1 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | -0.5 | 2 | -1.5 | -1.5 | | 18 | 7 | 9 | 4 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | 0.333333 | 0.333333 | -0.33333 | -0.33333 | | 8 | 13 | 4 | 8 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | 0.333333 | -0.33333 | 0.666667 | 0.666667 | | 12 | 7 | 9 | 4 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | -0.66667 | -0.33333 | -0.66667 | -0.66667 | | 14 | 9 | 5 | 3 | 1 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | -0.66667 | -0.33333 | 0 | 0 | | 14 | 9 | 5 | 8 | 1 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | 0.666667 | 1 | -0.66667 | -0.66667 | | 15 | 16 | 8 | 11 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | -0.33333 | 0.666667 | -0.33333 | -0.33333 | | 15 | 10 | 6 | 2 | 1 | 1 | 0 | 0 | 0 | 2000 | 1 |
| 0 | -0.33333 | -0.33333 | -0.66667 | -0.66667 | | 14 | 9 | 7 | 4 | 2 | 0 | 0 | 0 | 0 | 2000 | -1 |
| 0 | -1 | 0.666667 | -1.33333 | -1.33333 | | 17 | 8 | 8 | 5 | 0 | 2 | 0 | 0 | 0 | 2000 | -1 |
| 0 | 0 | 0 | -0.33333 | -0.33333 | | 12 | 11 | 9 | 7 | 0 | 0 | 0 | 0 | 0 | 2000 | 1 |
| 0 | 1 | -0.66667 | 1 | 1 | | 6 | 8 | 1 | 3 | 1 | 1 | 0 | 0 | 0 | 2000 | 1 |

Figure 3: Dataset

| Date | Home | xG(Home) | xG(Away) | Away |
|------------|-----------------|----------|----------|-----------------|
| 11-08-2017 | Arsenal | 2.5 | 1.5 | Leicester City |
| 12-08-2017 | Brighton | 0.3 | 1.9 | Manchester City |
| 12-08-2017 | Southampton | 2 | 0.3 | Swansea City |
| 12-08-2017 | Everton | 0.6 | 0.4 | Stoke City |
| 12-08-2017 | Chelsea | 1.5 | 0.6 | Burnley |
| 12-08-2017 | West Brom | 1.3 | 0.5 | Bournemouth |
| 12-08-2017 | Crystal Palace | 1.1 | 1.5 | Huddersfield |
| 12-08-2017 | Watford | 2.1 | 2.6 | Liverpool |
| 13-08-2017 | Manchester Utd | 2.1 | 0.5 | West Ham |
| 13-08-2017 | Newcastle Utd | 0.8 | 2.5 | Tottenham |
| 19-08-2017 | Stoke City | 0.7 | 1.5 | Arsenal |
| 19-08-2017 | Liverpool | 2.5 | 0.7 | Crystal Palace |
| 19-08-2017 | Burnley | 1.3 | 0.9 | West Brom |
| 19-08-2017 | Leicester City | 2 | 0.2 | Brighton |
| 19-08-2017 | Southampton | 2.1 | 2 | West Ham |
| 19-08-2017 | Bournemouth | 1 | 2.4 | Watford |
| 19-08-2017 | Swansea City | 0.4 | 3 | Manchester Utd |
| 20-08-2017 | Tottenham | 0.7 | 0.7 | Chelsea |
| 20-08-2017 | Huddersfield | 0.3 | 0.7 | Newcastle Utd |
| 21-08-2017 | Manchester City | 1.1 | 0.6 | Everton |
| 26-08-2017 | Manchester Utd | 2.8 | 0.9 | Leicester City |
| 26-08-2017 | Newcastle Utd | 2.3 | 0.6 | West Ham |
| 26-08-2017 | Crystal Palace | 0.9 | 0.7 | Swansea City |
| 26-08-2017 | Watford | 0.3 | 1.1 | Brighton |
| 26-08-2017 | Huddersfield | 1.4 | 0.7 | Southampton |
| 26-08-2017 | Bournemouth | 0.5 | 1.4 | Manchester City |
| 27-08-2017 | Liverpool | 3.1 | 0.6 | Arsenal |
| 27-08-2017 | Tottenham | 2.3 | 0.9 | Burnley |
| 27-08-2017 | West Brom | 0.8 | 1.3 | Stoke City |

Figure 4: xG data used for prediction

4 Conclusion

In conclusion, our project aimed to develop a user-friendly Premier League dashboard that provides real-time match predictions, updating after each match and resetting at the end of the season. This dashboard features a dynamic table predictor and various interactive graphs and charts representing key attributes utilized in our prediction model.

Drawing inspiration from the growing interest in data-driven decision-making, particularly in sports, our project focused on predicting Premier League match outcomes, leveraging existing research in sports analytics and predictive modeling.

Our method involved gathering valid and reliable data from fbref.com using web scraping techniques and Python libraries such as BeautifulSoup, NumPy, and Pandas. The dataset included matchwise data from the 2000-01 season onwards, covering various aspects such as team and manager names, match results, head-to-head statistics, past form, and expected goals (xG).

We utilized the xG data in a Random Forest Regressor Model [2] for match outcome prediction. Additionally, we incorporated all scraped and calculated data, along with the prediction model, into Power BI for creating a visually appealing dashboard. Drill-through features were utilized for easy navigation between different pages, facilitating access to meaningful data.

Our prediction model achieved a decent accuracy rate, considering its ability to predict both match outcomes and the number of goals scored by teams. The Power BI dashboard provides users with a comprehensive tool for analyzing and engaging with Premier League data effectively.

Overall, our project contributes to the advancement of predictive modeling in sports analytics and provides football enthusiasts with an intuitive platform for analyzing Premier League data.

5 Future Work

For future development, the Premier League dashboard and prediction model could undergo significant improvements. Firstly, refining the prediction model is essential, involving the continuous integration of additional features such as player injuries, weather conditions, and team tactics to enhance accuracy and robustness. Secondly, advanced data visualization techniques could be employed to enrich the dashboard's user experience. This could include the incorporation of heat-maps, network graphs for visualizing player connections, and in-

teractive maps displaying match locations and statistics, offering users deeper insights into match dynamics. Thirdly, enhancing user interaction and customization options could be prioritized, enabling users to personalize their dashboard experience by selecting favorite teams, comparing player or team performance, and setting up notifications for match updates and predictions. These improvements would not only elevate the dashboard's functionality but also provide users with a more engaging and tailored platform for analyzing Premier League data.

References

- [1] 2023-2024 Premier League Stats, <https://fbref.com/en/comps/9/Premier-League-Stats>, [Accessed: 21-3-2024].
- [2] Random forests(tm) in xgboost, <https://xgboost.readthedocs.io/en/stable/tutorials/rf.html>, [Accessed: 21-3-2024].