

Assignment 1 - Câu 2 Report: Analysis of Premier League Statistics

Nguyễn Đức Hà Hùng

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Abstract

This report provides a detailed explanation and results for Question 2 of Assignment 1, focusing on analyzing player statistics from the Premier League season stored in `results.csv`. The analysis includes identifying top performers, calculating statistical summaries, plotting histograms, and determining the best-performing team, with a focus on offensive metrics. An improved approach addressing the latest requirement of analyzing 3 offensive and 3 defensive stats with bar charts is also proposed.

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1 Objective

The objective of Question 2 is to analyze the player statistics collected in `results.csv` from Question 1. The tasks include:

- Identifying the top 3 highest and lowest performers for selected statistics.
- Calculating median, mean, and standard deviation for each statistic across all players and per team.
- Plotting histograms to visualize the distribution of statistics.
- Identifying the team with the highest average scores for each statistic.
- Analyzing the best-performing team based on overall performance.

Additionally, based on the latest requirement, the focus is adjusted to select 3 offensive statistics, 3 defensive statistics, and create bar charts for visualization.

2 Code Explanation

This section explains the provided Python code, breaking it into key components.

2.1 Importing Libraries

```
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import os
```

- `pandas`: Handles data manipulation and CSV operations.
- `numpy`: Supports numerical computations.
- `matplotlib.pyplot`: Creates visualizations (histograms and potential bar charts).
- `os`: Manages file and directory operations.

2.2 Defining Paths and Loading Data

```
1 results_csv_path = r'C:\Users\nguye\Downloads\results.csv'
2 top_3_path = r'C:\Users\nguye\Downloads\top_3.txt'
3 results2_csv_path = r'C:\Users\nguye\Downloads\results2.csv'
4 plots_dir = r'C:\Users\nguye\Downloads\plots'
5 os.makedirs(plots_dir, exist_ok=True)
6 df = pd.read_csv(results_csv_path)
```

- Defines paths for input (`results.csv`), output files (`top_3.txt`, `results2.csv`), and `plotsdir`.
- Loads the dataset from `results.csv`.

2.3 Selecting Statistics and Data Preparation

```
1 stats = ['Gls', 'Ast', 'xG', 'xAG']
2 for stat in stats:
3     df[stat] = pd.to_numeric(df[stat], errors='coerce')
4 df = df.dropna(subset=stats, how='all')
```

- Selects four offensive statistics: GlS (Goals), Ast (Assists), xG (Expected Goals), xAG (Expected Assisted Goals).
- Converts statistic columns to numeric values, replacing non-numeric entries (e.g., "N/a") with NaN.
- Drops rows where all selected stats are NaN, ensuring only valid player data is analyzed.

2.4 Task 1: Identifying Top 3 Performers

```
1 top_3_content = []
2 for stat in stats:
3     stat_df = df[['Player', stat]].dropna()
4     top_high = stat_df.nlargest(3, stat)
5     top_3_content.append(f"Top 3 highest for {stat}:\n")
6     for _, row in top_high.iterrows():
7         top_3_content.append(f"{row['Player']}: {row[stat]}\n")
8     stat_df_nonzero = stat_df[stat_df[stat] > 0]
9     if len(stat_df_nonzero) >= 3:
10         top_low = stat_df_nonzero.nsmallest(3, stat)
11     else:
12         top_low = stat_df.nsmallest(3, stat)
13     top_3_content.append(f"Top 3 lowest for {stat}:\n")
14     for _, row in top_low.iterrows():
15         top_3_content.append(f"{row['Player']}: {row[stat]}\n")
16     top_3_content.append("\n")
17 with open(top_3_path, 'w', encoding='utf-8') as f:
18     f.writelines(top_3_content)
```

- Identifies the top 3 highest and lowest performers for each statistic.
- Excludes zero values for lowest performers to ensure meaningful results.
- Saves results to `top3.txt`.

2.5 Task 2: Calculating Statistical Summary

```
1 stats_summary = []
2 for stat in stats:
3     row = {'Team': 'all', f'Median of {stat}': df[stat].median(),
4           f'Mean of {stat}': df[stat].mean(), f'Std of {stat}': df[
5           stat].std()}
6     stats_summary.append(row)
```

```

6 teams = df['Squad'].unique()
7 for team in teams:
8     team_df = df[df['Squad'] == team]
9     row = {'Team': team}
10    for stat in stats:
11        row[f'Median of {stat}'] = team_df[stat].median()
12        row[f'Mean of {stat}'] = team_df[stat].mean()
13        row[f'Std of {stat}'] = team_df[stat].std()
14    stats_summary.append(row)
15 stats_df = pd.DataFrame(stats_summary)
16 stats_df = stats_df.fillna('N/a')
17 stats_df.to_csv(results2_csv_path, index=False)

```

- Calculates median, mean, and standard deviation for each statistic across all players and per team.
- Stores results in a DataFrame and saves to results2.csv.

2.6 Task 3: Plotting Histograms

```

1 for stat in stats:
2     plt.figure(figsize=(10, 6))
3     plt.hist(df[stat].dropna(), bins=30, edgecolor='black')
4     plt.title(f'Distribution of {stat} for All Players')
5     plt.xlabel(stat)
6     plt.ylabel('Frequency')
7     plt.savefig(os.path.join(plots_dir, f'{stat}_all_players.png'))
8     plt.close()
9 for team in teams:
10    team_df = df[df['Squad'] == team]
11    for stat in stats:
12        plt.figure(figsize=(10, 6))
13        plt.hist(team_df[stat].dropna(), bins=30, edgecolor='black')
14        plt.title(f'Distribution of {stat} for {team}')
15        plt.xlabel(stat)
16        plt.ylabel('Frequency')
17        plt.savefig(os.path.join(plots_dir, f'{stat}_{team.replace(" ", "_")}.png'))
18    plt.close()

```

- Creates histograms for each statistic, both for all players and per team.
- Saves plots as PNG files in the plots directory.

2.7 Task 4: Identifying Top Teams

```

1 team_means = df.groupby('Squad')[stats].mean()
2 top_teams = {}
3 for stat in stats:
4     top_team = team_means[stat].idxmax()

```

```

5     top_score = team_means[stat].max()
6     top_teams[stat] = (top_team, top_score)
7 print("Teams with highest average scores for each statistic:")
8 for stat, (team, score) in top_teams.items():
9     print(f"{stat}: {team} with average {score:.2f}")

```

- Calculates the mean of each statistic per team and identifies the team with the highest average.
- Prints the results.

2.8 Task 5: Analyzing the Best-Performing Team

```

1 team_scores = team_means.mean(axis=1)
2 best_team = team_scores.idxmax()
3 best_team_score = team_scores.max()
4 print(f"\nAnalysis of Best-Performing Team:")
5 print(f"The team with the highest overall average across all
   statistics is {best_team} with an average score of {
   best_team_score:.2f}.")
6 print("Reasoning:")
7 print(f"- {best_team} shows strong performance across key offensive
   metrics (Gls, Ast, xG, xAG), indicating a balanced and effective
   attacking strategy.")
8 print(f"- High xG and xAG suggest they create high-quality chances,
   while GlS and Ast show they convert these chances effectively.")
9 print(f"- In the Premier League season, this aligns with teams that
   have top players and tactical consistency.")

```

- Identifies the team with the highest overall average across all statistics.
- Provides reasoning based on offensive performance.

3 Results

The code executed the following tasks and produced the specified outputs.

3.1 Output Files

- **top₃.txt** : Contains the top 3 highest and lowest performers for each statistic (e.g., GlS, Ast, xG, xAG). Contains median, mean, and standard deviation for each statistic across all players and per team.
- **plots Directory**: Contains histogram images for each statistic, both for all players (e.g., GlS_{all_players.png}) and per team (e.g.,

3.2 Sample Output

- **top₃.txt Sample** :

Top 3 highest for Gls:

Erling Haaland: 15

Harry Kane: 12

Mohamed Salah: 10

Top 3 lowest for Gls:

Joe Gomez: 0.1

Ben Chilwell: 0.2

Trent Alexander-Arnold: 0.3

	Team	Median of Gls	Mean of Gls	Std of Gls
results2.csv Sample:	all	2.5	3.1	2.0
	Arsenal	2.0	2.8	1.5
	Chelsea	3.0	3.5	2.2

(Note: Values are illustrative; actual results depend on results.csv.)

3.3 Challenges and Solutions

- **NaN Values:** Handled by converting to numeric and dropping rows with all NaN stats.
- **Zero Values in Lowest:** Excluded zeros for meaningful lowest performer identification.
- **File Naming:** Replaced spaces in team names with underscores for file names.

4 Analysis

- **Statistical Coverage:** The analysis focuses on offensive metrics (Gls, Ast, xG, xAG), providing a solid foundation for attacking performance evaluation.
- **Visualization:** Histograms effectively show data distribution, aiding in identifying trends.
- **Limitations:** Lacks defensive statistics and bar charts, which are required by the latest instruction to analyze 3 offensive and 3 defensive stats.

5 Improved Approach

Based on the latest requirement to analyze 3 offensive stats (e.g., Gls_{per90}), This approach better aligns with the requirement, providing clear comparisons of team performance across offensive and defensive metrics.

6 Conclusion

Question 2 was partially completed with the provided code, effectively analyzing offensive statistics and producing histograms, top 3 lists, and team performance insights. However, it does not fully meet the latest requirement of including 3 defensive stats and using bar charts. The suggested improvement addresses these gaps, offering a streamlined solution for future analysis. The dataset and visualizations are ready for further exploration in subsequent questions.