

twylyqeih

March 16, 2025

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[23]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.cluster import KMeans
from sklearn.preprocessing import StandardScaler
from tabulate import tabulate

df = pd.read_csv("EPL.csv")
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[24]: df.fillna(df.mean(numeric_only=True), inplace=True)

df["TotalGoals"] = df["FullTimeHomeTeamGoals"] + df["FullTimeAwayTeamGoals"]
df["GoalDifference"] = df["FullTimeHomeTeamGoals"] - df["FullTimeAwayTeamGoals"]
df["WinRate"] = df["HomeTeamPoints"] / (df["HomeTeamPoints"] +
    ↪df["AwayTeamPoints"])

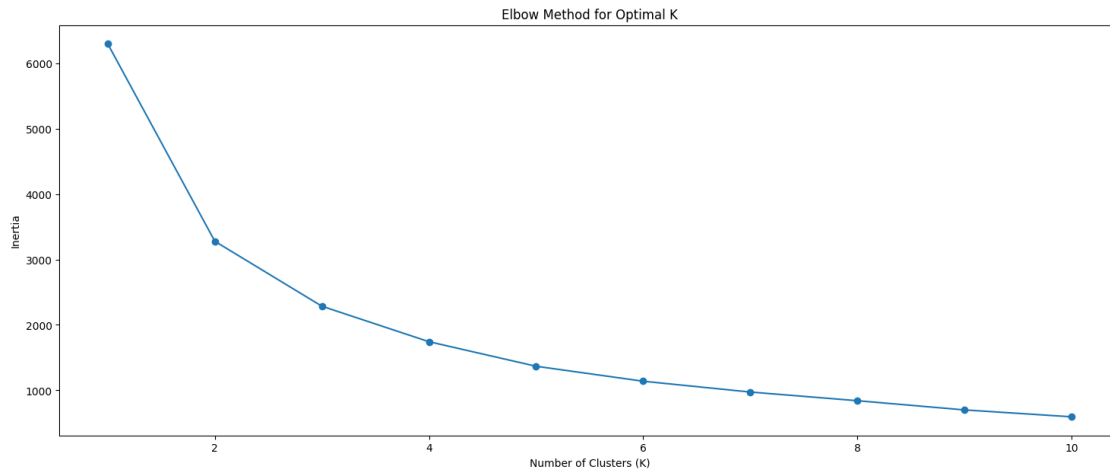
features = df[["TotalGoals", "GoalDifference", "WinRate"]]
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[25]: scaler = StandardScaler()
scaled_features = scaler.fit_transform(features)
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[26]: inertia = []
K_range = range(1, 11)

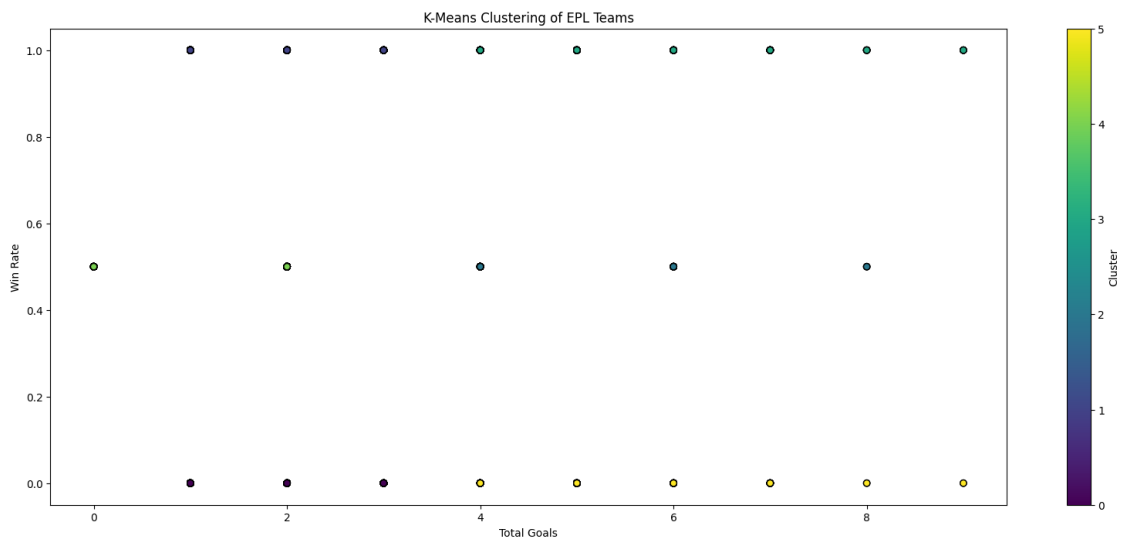
for k in K_range:
    kmeans = KMeans(n_clusters=k, random_state=42, n_init=10)
    kmeans.fit(scaled_features)
    inertia.append(kmeans.inertia_)
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[27]: plt.figure(figsize=(18, 7))
plt.plot(K_range, inertia, marker='o')
plt.xlabel('Number of Clusters (K)')
plt.ylabel('Inertia')
plt.title('Elbow Method for Optimal K')
plt.show()
```



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[28]: optimal_k = 6
kmeans = KMeans(n_clusters=optimal_k, random_state=42, n_init=10)
df["Cluster"] = kmeans.fit_predict(scaled_features)
```

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[29]: plt.figure(figsize=(20, 8))
plt.scatter(df["TotalGoals"], df["WinRate"], c=df["Cluster"], cmap="viridis",
            edgecolors='k')
plt.xlabel("Total Goals")
plt.ylabel("Win Rate")
plt.title("K-Means Clustering of EPL Teams")
plt.colorbar(label="Cluster")
plt.show()
```



```
[30]: print(tabulate(df[["HomeTeam", "TotalGoals", "GoalDifference", "WinRate",
↪ "Cluster"]].head(10), headers="keys", tablefmt="github", showindex=False))
```

HomeTeam	TotalGoals	GoalDifference	WinRate	Cluster
Liverpool	5	3	1	3
West Ham	5	-5	0	5
Bournemouth	2	0	0.5	4
Burnley	3	3	1	1
Crystal Palace	0	0	0.5	4
Watford	3	-3	0	0
Tottenham	4	2	1	1
Leicester	0	0	0.5	4
Newcastle	1	-1	0	0
Man United	4	4	1	3