DS3001 HW1 Q1

February 11, 2021

[1]: import pandas as pd

```
import numpy as np
    import warnings
    import matplotlib.pyplot as plt
    warnings.filterwarnings('ignore')
[2]: df = pd.read csv('soccer17.csv')
    df = df.loc[(df.Y == 17)]
    flds = ['Div','Date','Y','HomeTeam','AwayTeam','FTHG','FTAG']
    df = df.loc[:,flds].copy()
    df = df.rename({'HomeTeam':'Team_Home','AwayTeam':'Team_Away','FTHG':
     df.loc[:,'GameID'] = df.index
    df['pd_Home'] = df.G_Home - df.G_Away
    df['pd_Away'] = df.G_Away - df.G_Home
[3]: df_melt = pd.melt(df, id_vars = ['GameID', 'Div'], value_vars =
     var_name = 'isHome', value_name = 'Team')
    df_melt.loc[:,'isHome'] = 1*(df_melt.isHome == 'Team_Home')
[4]: df_melt2 = pd.melt(df, id_vars=['GameID', 'Div'], value_vars=['pd_Home',__
     →'pd_Away'], var_name = 'isHome', value_name = 'pd')
    df_melt2.loc[:,'isHome'] = 1*(df_melt2.isHome == 'pd_Home')
[5]: df_merge = df_melt.merge(df_melt2, on=['GameID', 'Div', 'isHome'])
    df_merge['Win'] = 1*(df_merge.pd > 0)
    df_merge['Draw'] = 1*(df_merge.pd == 0)
    df_merge['Loss'] = 1*(df_merge.pd < 0)</pre>
[6]: avg_pd = df_merge.groupby(['Team']).mean()['pd']
    gp = df_merge.groupby(['Team']).size()
    wins = df merge.groupby(['Team']).agg({'Win': sum})['Win']
    draws = df_merge.groupby(['Team']).agg({'Draw': sum})['Draw']
    losses = df_merge.groupby(['Team']).agg({'Loss': sum})['Loss']
    team = avg_pd.index
```

1 1a

```
[7]: df1 = output.loc[output.Div == 'EPL'].drop('Div', 1)
df1.sort_values(by='Avg PD', ascending=False).reset_index().drop('index', 1)
```

| [7]: | Team | Avg PD | Games Played | Wins | Losses | Draws | |
|------|----------------|-----------|--------------|------|--------|-------|--|
| 0 | Man City | 2.078947 | 38 | 32 | 2 | 4 | |
| 1 | Liverpool | 1.210526 | 38 | 21 | 5 | 12 | |
| 2 | Man United | 1.052632 | 38 | 25 | 7 | 6 | |
| 3 | Tottenham | 1.000000 | 38 | 23 | 7 | 8 | |
| 4 | Chelsea | 0.631579 | 38 | 21 | 10 | 7 | |
| 5 | Arsenal | 0.605263 | 38 | 19 | 13 | 6 | |
| 6 | Burnley | -0.078947 | 38 | 14 | 12 | 12 | |
| 7 | Leicester | -0.105263 | 38 | 12 | 15 | 11 | |
| 8 | Newcastle | -0.210526 | 38 | 12 | 18 | 8 | |
| 9 | Crystal Palace | -0.263158 | 38 | 11 | 16 | 11 | |
| 10 | Everton | -0.368421 | 38 | 13 | 15 | 10 | |
| 11 | Bournemouth | -0.421053 | 38 | 11 | 16 | 11 | |
| 12 | Southampton | -0.500000 | 38 | 7 | 16 | 15 | |
| 13 | Brighton | -0.526316 | 38 | 9 | 16 | 13 | |
| 14 | Watford | -0.526316 | 38 | 11 | 19 | 8 | |
| 15 | West Ham | -0.526316 | 38 | 10 | 16 | 12 | |
| 16 | West Brom | -0.657895 | 38 | 6 | 19 | 13 | |
| 17 | Swansea | -0.736842 | 38 | 8 | 21 | 9 | |
| 18 | Huddersfield | -0.789474 | 38 | 9 | 19 | 10 | |
| 19 | Stoke | -0.868421 | 38 | 7 | 19 | 12 | |

2 1b

```
[8]: df1['Points'] = df1.Wins*3 + df1.Draws
df1.sort_values(by='Points', ascending=False).reset_index().drop('index', 1)
```

```
[8]:
                   Team
                            Avg PD
                                    Games Played Wins
                                                         Losses
                                                                  Draws
                                                                         Points
     0
               Man City 2.078947
                                               38
                                                     32
                                                               2
                                                                      4
                                                                             100
             Man United 1.052632
                                                     25
                                                               7
     1
                                                                      6
                                                                              81
                                               38
     2
              Tottenham 1.000000
                                               38
                                                     23
                                                               7
                                                                      8
                                                                              77
     3
              Liverpool
                         1.210526
                                               38
                                                     21
                                                               5
                                                                     12
                                                                              75
```

| 4 | Chelsea | 0.631579 | 38 | 21 | 10 | 7 | 70 |
|----|----------------|-----------|----|----|----|----|----|
| 5 | Arsenal | 0.605263 | 38 | 19 | 13 | 6 | 63 |
| 6 | Burnley | -0.078947 | 38 | 14 | 12 | 12 | 54 |
| 7 | Everton | -0.368421 | 38 | 13 | 15 | 10 | 49 |
| 8 | Leicester | -0.105263 | 38 | 12 | 15 | 11 | 47 |
| 9 | Crystal Palace | -0.263158 | 38 | 11 | 16 | 11 | 44 |
| 10 | Bournemouth | -0.421053 | 38 | 11 | 16 | 11 | 44 |
| 11 | Newcastle | -0.210526 | 38 | 12 | 18 | 8 | 44 |
| 12 | West Ham | -0.526316 | 38 | 10 | 16 | 12 | 42 |
| 13 | Watford | -0.526316 | 38 | 11 | 19 | 8 | 41 |
| 14 | Brighton | -0.526316 | 38 | 9 | 16 | 13 | 40 |
| 15 | Huddersfield | -0.789474 | 38 | 9 | 19 | 10 | 37 |
| 16 | Southampton | -0.500000 | 38 | 7 | 16 | 15 | 36 |
| 17 | Stoke | -0.868421 | 38 | 7 | 19 | 12 | 33 |
| 18 | Swansea | -0.736842 | 38 | 8 | 21 | 9 | 33 |
| 19 | West Brom | -0.657895 | 38 | 6 | 19 | 13 | 31 |

3 1c

```
[9]: df2 = output.sort_values(['Div', 'Avg PD'], ascending=[True, False])
df2.groupby('Div').head(3).reset_index(drop=1)
```

| [9]: | Div | Team | Avg PD | Games Played | Wins | Losses | Draws |
|------|------------|---------------|----------|--------------|------|--------|-------|
| 0 | Bundesliga | Bayern Munich | 1.882353 | 34 | 27 | 4 | 3 |
| 1 | Bundesliga | Hoffenheim | 0.529412 | 34 | 15 | 9 | 10 |
| 2 | Bundesliga | Dortmund | 0.500000 | 34 | 15 | 9 | 10 |
| 3 | EPL | Man City | 2.078947 | 38 | 32 | 2 | 4 |
| 4 | EPL | Liverpool | 1.210526 | 38 | 21 | 5 | 12 |
| 5 | EPL | Man United | 1.052632 | 38 | 25 | 7 | 6 |
| 6 | La_Liga | Barcelona | 1.842105 | 38 | 28 | 1 | 9 |
| 7 | La_Liga | Real Madrid | 1.315789 | 38 | 22 | 6 | 10 |
| 8 | La_Liga | Ath Madrid | 0.947368 | 38 | 23 | 5 | 10 |
| 9 | Ligue_1 | Paris SG | 2.078947 | 38 | 29 | 3 | 6 |
| 10 | Ligue_1 | Lyon | 1.157895 | 38 | 23 | 6 | 9 |
| 11 | Ligue_1 | Monaco | 1.052632 | 38 | 24 | 6 | 8 |
| 12 | Serie_A | Juventus | 1.631579 | 38 | 30 | 3 | 5 |
| 13 | Serie_A | Napoli | 1.263158 | 38 | 28 | 3 | 7 |
| 14 | Serie_A | Lazio | 1.052632 | 38 | 21 | 8 | 9 |

DS3001_HW1_Q2

February 11, 2021

```
[1]: import pandas as pd
  import numpy as np
  import matplotlib.pyplot as plt
  from scipy.stats import norm
  import math

[2]: df = pd.read_csv('soccer17.csv')

[3]: df['All_Goals'] = df.FTHG + df.FTAG
  df['H1_Goals'] = df.HTHG + df.HTAG
  df['H2_Goals'] = df['All_Goals'] - df['H1_Goals']
  X = np.linspace(0, 10)
```

1 2a

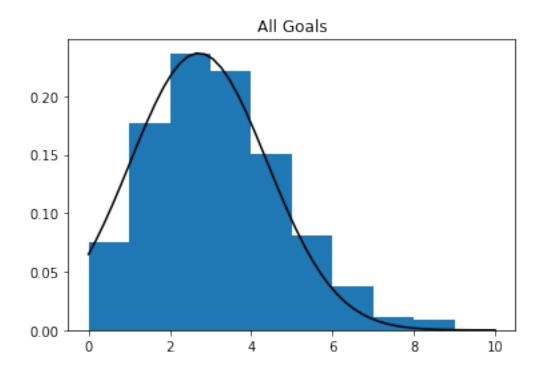
1.1 i. Model Selection

 $A_1,...,A_{7304} \stackrel{iid}{\sim} \mathcal{N}(\mu_A,\sigma_A^2)$ where A_i represents all goals scored in a single game i

1.2 ii. Histogram

```
mu, sigma = norm.fit(df['All_Goals'])
plt.hist(df['All_Goals'], bins=np.arange(0, 10), density=True)
plt.plot(X, norm.pdf(X, mu, sigma), color='black')
plt.title('All Goals')
```

[4]: Text(0.5, 1.0, 'All Goals')



2 2b

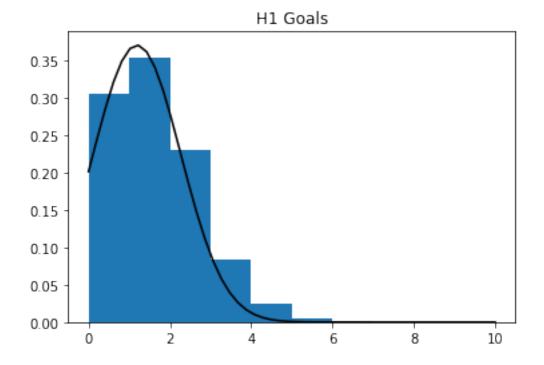
2.1 i. Model Selection

 $H_1,...,H_{7304} \stackrel{iid}{\sim} \mathcal{N}(\mu_H,\sigma_H^2)$ where H_i represents all first half goals scored in a single game i

2.2 ii. Histogram

```
[5]: mu, sigma = norm.fit(df['H1_Goals'])
   plt.hist(df['H1_Goals'], bins=np.arange(0, 10), density=True)
   plt.plot(X, norm.pdf(X, mu, sigma), color='black')
   plt.title('H1 Goals')
```

[5]: Text(0.5, 1.0, 'H1 Goals')



3 2c

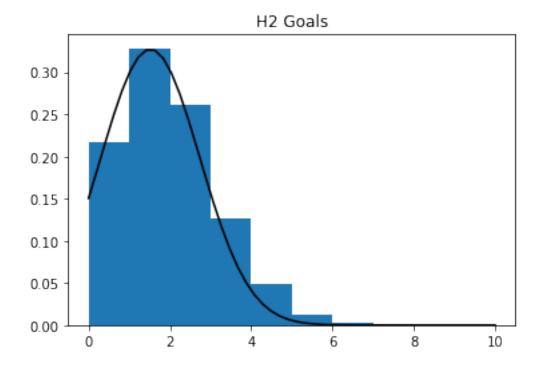
3.1 i. Model Selection

 $I_1,...,I_{7304} \stackrel{iid}{\sim} \mathcal{N}(\mu_I,\sigma_I^2)$ where I_i represents all second half goals scored in a single game i

3.2 ii. Histogram

```
[6]: mu, sigma = norm.fit(df['H2_Goals'])
   plt.hist(df.H2_Goals, bins=np.arange(0, 10), density=True)
   plt.plot(X, norm.pdf(X, mu, sigma), color='black')
   plt.title('H2 Goals')
```

[6]: Text(0.5, 1.0, 'H2 Goals')



4 2d

```
[7]: df2 = pd.DataFrame()
  df2['Num_Games'] = df.groupby('Div').count()['All_Goals']
  df2['Avg_Goals'] = df.groupby('Div').mean()['All_Goals']

df2.reset_index().sort_values(by='Avg_Goals', ascending=False)
```

```
[7]:
               Div Num_Games Avg_Goals
        Bundesliga
                          1224
                                 2.811275
     2
           La_Liga
                          1520
                                 2.759211
     4
           Serie_A
                          1520
                                 2.725658
               EPL
     1
                          1520
                                 2.686184
     3
           Ligue_1
                          1520
                                 2.588158
```

5 2e

```
[8]: epsilon = 0.04
df4 = df.loc[df.All_Goals == 4]
fgg = df4.loc[(np.abs(df4.pH - df4.pA) < epsilon)]</pre>
```

```
[9]: prob_draw = math.factorial(4)/(math.factorial(2) * math.factorial(2)) * (0.

→5)**2 * (0.5)**2

exp_draws = len(fgg) * prob_draw
draws = len(fgg.loc[fgg.FTHG == fgg.FTAG])
sigma = np.sqrt(0.375*0.625*len(fgg))

t = (draws-exp_draws)/sigma
p = 1 - norm.cdf(t)
print('p-value = {}'.format(p))
```

p-value = 0.007348710885011767

Our null hypothesis states that there is no "comeback tendency" within the dataset provided

Using an epsilon of $\epsilon = 0.04$, we were able to obtain a statistically significant p-value meaning that we reject the null hypothesis. This means that there is empirical evidence that suggests that a "comeback tendency" exists.