model2022 final

December 6, 2022

1 initial setting

```
[1]: from google.colab import drive drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
[2]: #import
     import pandas as pd
     import numpy as np
     import tensorflow as tf
     import warnings
     warnings.filterwarnings("ignore")
     from sklearn.model_selection import train_test_split
     from sklearn.manifold import TSNE, Isomap
     from sklearn.decomposition import PCA
     from sklearn.preprocessing import StandardScaler, Normalizer
     import seaborn as sns
     import matplotlib.pyplot as plt
     from scipy.stats import uniform, randint
     import xgboost as xgb
     from sklearn.metrics import auc, accuracy_score, mean_absolute_error
     from sklearn.model_selection import cross_val_score, RandomizedSearchCV,_
     →train_test_split, KFold
     from sklearn.decomposition import PCA
     from sklearn.multioutput import MultiOutputRegressor
     import random
     import copy
```

```
[3]: %cd /content/drive/MyDrive/2022/BigData
file_root = './fifa_dataset_with_all_features-no_null_values.csv'
df_old = pd.read_csv(file_root, encoding='cp949')
```

/content/drive/MyDrive/2022/BigData

```
[4]: df = pd.read_csv(file_root, encoding='cp949')
     #df['date'] = [i.replace('-','') for i in df['date']]
     #all_list : tournament, home_team, away_team
     all_list = ((df['tournament'].unique()).tolist() + (df['home_team'].unique()).
     →tolist() + (df['away_team'].unique()).tolist())
     uni_len = '***'.join(all_list)
     nat_enc = tf.keras.preprocessing.text.Tokenizer(num_words=len(all_list),
         filters='',
         lower=False,
         split = '***')
     nat_enc.fit_on_texts([uni_len])
     nat_dic = nat_enc.word_index
[5]: # enumerate
     for step, val in enumerate(df['tournament']):
         df['tournament'].iloc[step] = (nat dic.get(val))
     for step, val in enumerate(df['home_team']):
         df['home_team'].iloc[step] = (nat_dic.get(val))
     for step, val in enumerate(df['away_team']):
         df['away_team'].iloc[step] = (nat_dic.get(val))
     for step, val in enumerate(df['neutral']):
         if df['neutral'].iloc[step] == False:
             df['neutral'].iloc[step] = 0
         if df['neutral'].iloc[step] == True:
             df['neutral'].iloc[step] = 1
[6]: df.head()
       home_team away_team home_score away_score tournament neutral
     0
               1
                        10
                                      2
                                                  1
                                                            91
                                                                      1
               2
     1
                                                  0
                                                            91
                        58
                                      1
                                                                      1
     2
               3
                                                            91
                         1
                                      1
                                                  1
                                                                      0
     3
               4
                        25
                                      0
                                                  3
                                                            92
                                                                      1
     4
               3
                        10
                                      1
                                                  3
                                                            91
                                                                      0
                                away_team_total_points home_B1_overall
        home_team_total_points
     0
                                                  710.0
                         664.0
                                                                       71
                                                                       72
     1
                         849.0
                                                  421.0
     2
                         554.0
                                                  664.0
                                                                       72
     3
                         558.0
                                                  461.0
                                                                       74
     4
                         576.0
                                                  701.0
                                                                       72
        home_B2_overall ... away_B2_Position_Best_Rating \
```

```
72
     0
                      72
     1
                      72
                                                          59
     2
                                                          72
                      72
     3
                      74
                                                          75
     4
                      72
                                                          72
        away_B3_Position_Best_Rating away_B4_Position_Best_Rating
     0
                                    71
                                                                     69
     1
                                    59
                                                                     57
     2
                                    70
                                                                     70
     3
                                    74
                                                                     73
     4
                                    71
                                                                     69
        away_M1_Position_Best_Rating
                                        away_M2_Position_Best_Rating
     0
                                    78
                                                                     77
                                    72
                                                                     69
     1
     2
                                    77
                                                                     76
     3
                                    79
                                                                     78
     4
                                    78
                                                                     77
                                        away_M4_Position_Best_Rating
        away_M3_Position_Best_Rating
     0
                                    75
                                                                     75
     1
                                    67
                                                                     63
     2
                                    72
                                                                     71
     3
                                    76
                                                                     75
                                    75
     4
                                                                     75
        away_T1_Position_Best_Rating
                                        away_T2_Position_Best_Rating
     0
                                    80
                                                                     78
     1
                                    73
                                                                     72
     2
                                                                     74
                                    76
     3
                                    87
                                                                     82
     4
                                                                     78
                                    80
        away_K1_Position_Best_Rating
     0
     1
                                    66
     2
                                    70
     3
                                    80
     4
                                    80
     [5 rows x 316 columns]
[7]: #df_new
     df_new = df
     #len
     temp_tor = []
```

```
for i in df_old['tournament']:
          temp_tor.append(len(i))
      temp_away = []
      for i in df_old['away_team']:
          temp_away.append(len(i))
      temp_home = []
      for i in df_old['home_team']:
          temp_home.append(len(i))
      df_new['tor_len'] = temp_tor
      df_new['home_len'] = temp_home
      df_new['away_len'] = temp_away
      df_new = df_new.dropna()
      df_label1 = df_new['home_score']
      df_label2 = df_new['away_score']
 [8]: #df_new, df_label :
                                (df_new) (df_label, home_score away score)
      df_new = df_new.drop(['home_score', 'away_score'], axis = 1)
      #df_label
      df_label = pd.concat([df_label1, df_label2],axis = 1)
      df_label.columns = ['home_score', 'away_score']
     df_label
 [9]: #
                          6547 rows
              7948
                                         -> concat
                                                             index
      df_label
 [9]:
            home_score away_score
      0
                                 1
      1
                     1
                                 0
      2
                     1
                                 1
      3
                                 3
      4
                     1
                                 3
      2540
                                 0
                     0
                                 5
      2541
                     0
      2542
                     1
                                 1
      2543
                     0
                                 0
      2544
      [2545 rows x 2 columns]
[10]: x_train = df_new.to_numpy()
      df_train = pd.DataFrame(data=x_train, columns=df_new.columns)
```

df_train

[11]: home_team away_team tournament neutral home_team_total_points \ 664.0 849.0 554.0 558.0 576.0 1243.8 961.23 1473.04 1425.59 1384.04 away_team_total_points home_B1_overall home_B2_overall home_B3_overall 710.0 421.0 664.0 461.0 701.0 ... 1335.36 1434.68 1405.6 1509.61 1341.03 home_B4_overall ... away_M1_Position_Best_Rating ••• 71 ... away_M2_Position_Best_Rating away_M3_Position_Best_Rating \

[11]: df_train

```
2540
                                   69
                                                                   69
2541
                                   76
                                                                   75
2542
                                   74
                                                                   72
2543
                                   77
                                                                   76
2544
                                   74
                                                                   71
     away_M4_Position_Best_Rating away_T1_Position_Best_Rating
0
                                   75
1
                                   63
                                                                   73
2
                                   71
                                                                   76
3
                                   75
                                                                   87
                                   75
4
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2540
                                   67
                                                                   72
2541
                                   75
                                                                   78
2542
                                   72
                                                                   75
2543
                                   76
                                                                   79
2544
                                   69
                                                                   77
     away_T2_Position_Best_Rating away_K1_Position_Best_Rating tor_len \
0
                                   78
                                                                   80
                                                                            28
1
                                   72
                                                                   66
                                                                            28
2
                                   74
                                                                   70
                                                                            28
3
                                   82
                                                                   80
                                                                             8
4
                                   78
                                                                   80
                                                                            28
2540
                                   67
                                                                   66
                                                                             8
2541
                                   75
                                                                   76
                                                                              8
2542
                                   73
                                                                   83
                                                                             8
2543
                                   77
                                                                   78
                                                                              8
2544
                                   74
                                                                   69
                                                                              8
     home_len away_len
             4
0
             5
                       5
1
2
            12
                       4
             6
                       6
3
            12
                       7
4
                       8
2540
            10
2541
                       7
2542
             6
                       7
2543
                       5
             8
2544
             8
                      10
```

[2545 rows x 317 columns]

```
[12]: \#x\_train = df\_new.to\_numpy()
      Norm = Normalizer() #
                                        1
      x_temp = Norm.fit_transform(x_train)
      df_temp = pd.DataFrame(data=x_temp, columns=df_new.columns)
     df temp: df train x train, df temp x train normalize
                                                                x temp
[13]: df_temp
[13]:
            home team
                        away_team
                                   tournament
                                                 neutral
                                                           home_team_total_points
             0.000738
      0
                         0.007380
                                      0.067162
                                                0.000738
                                                                          0.490059
      1
             0.001523
                         0.044177
                                      0.069313
                                                0.000762
                                                                          0.646666
      2
             0.002362
                         0.000787
                                      0.071636
                                                                          0.436114
                                                0.000000
      3
             0.003353
                         0.020959
                                      0.077129
                                                0.000838
                                                                          0.467806
             0.002285
                         0.007618
                                      0.069326
                                                0.000000
                                                                          0.438810
      2540
             0.042079
                         0.007921
                                      0.045544
                                                0.000000
                                                                          0.615739
      2541
                                      0.047510
                                                                          0.496393
             0.042346
                         0.013943
                                                0.000000
      2542
             0.001776
                         0.015093
                                      0.040841
                                                0.000000
                                                                          0.653911
      2543
             0.024010
                         0.002619
                                      0.040163
                                                0.000000
                                                                          0.622346
      2544
             0.026164
                                      0.042984
                         0.034107
                                                0.000000
                                                                          0.646647
            away_team_total_points
                                      home_B1_overall
                                                       home_B2_overall
      0
                           0.524008
                                             0.052401
                                                               0.053139
      1
                           0.320667
                                             0.054841
                                                               0.054841
      2
                           0.522706
                                             0.056679
                                                               0.056679
      3
                           0.386485
                                             0.062039
                                                               0.062039
      4
                                             0.054851
                                                               0.054851
                           0.534037
      2540
                           0.661065
                                             0.033168
                                                               0.032673
      2541
                           0.740889
                                             0.030468
                                                               0.028919
      2542
                           0.623973
                                             0.034182
                                                               0.033294
      2543
                           0.659025
                                             0.032741
                                                               0.031868
      2544
                           0.626552
                                             0.035041
                                                               0.034107
                                                   away_M1_Position_Best_Rating
            home_B3_overall
                              home_B4_overall
      0
                   0.051663
                                      0.051663
                                                                         0.057567
      1
                   0.054079
                                      0.054079
                                                                         0.054841
      2
                   0.054317
                                      0.055105
                                                                         0.060615
      3
                   0.060362
                                      0.060362
                                                                         0.066231
      4
                                                                         0.059422
                   0.052566
                                      0.053328
      2540
                   0.031188
                                      0.029208
                                                                         0.034653
      2541
                   0.028403
                                      0.032534
                                                                         0.039247
      2542
                   0.032850
                                      0.032850
                                                                         0.034626
                                                                        0.034924
      2543
                   0.032305
                                      0.031868
      2544
                                                                         0.037845
                   0.033172
                                      0.033172
```

```
away_M2_Position_Best_Rating away_M3_Position_Best_Rating \
0
                           0.056829
                                                           0.055353
                                                           0.051033
1
                           0.052556
2
                           0.059828
                                                           0.056679
3
                           0.065392
                                                           0.063716
4
                           0.058660
                                                           0.057137
2540
                           0.034158
                                                           0.034158
2541
                           0.039247
                                                           0.038731
2542
                           0.032850
                                                           0.031962
2543
                           0.033615
                                                           0.033178
2544
                           0.034574
                                                           0.033172
                                      away_T1_Position_Best_Rating
      away_M4_Position_Best_Rating
                                                           0.059043
0
                           0.055353
1
                           0.047986
                                                           0.055603
2
                           0.055892
                                                           0.059828
3
                           0.062877
                                                           0.072938
4
                           0.057137
                                                           0.060946
2540
                           0.033168
                                                           0.035643
2541
                           0.038731
                                                           0.040280
2542
                           0.031962
                                                           0.033294
2543
                           0.033178
                                                           0.034488
2544
                           0.032238
                                                           0.035976
      away_T2_Position_Best_Rating
                                     away_K1_Position_Best_Rating
                                                                      tor_len \
0
                                                           0.059043 0.020665
                           0.057567
1
                           0.054841
                                                           0.050271 0.021327
2
                           0.058253
                                                           0.055105
                                                                     0.022042
3
                           0.068746
                                                           0.067069
                                                                     0.006707
4
                           0.059422
                                                           0.060946
                                                                     0.021331
2540
                           0.033168
                                                           0.032673 0.003960
2541
                           0.038731
                                                           0.039247
                                                                    0.004131
2542
                           0.032406
                                                           0.036845 0.003551
2543
                           0.033615
                                                           0.034051
                                                                     0.003492
2544
                           0.034574
                                                           0.032238 0.003738
      home_len away_len
                0.005166
0
      0.002952
1
      0.003808
                0.003808
2
      0.009447
                0.003149
3
      0.005030
                0.005030
4
      0.009142
                0.005333
```

```
2540 0.004950 0.003960
      2541 0.003615 0.003615
      2542 0.002664 0.003107
      2543 0.003492 0.002183
      2544 0.003738 0.004672
      [2545 rows x 317 columns]
[14]: \#df\_corr = df\_temp.corr() \#
     3
     df_label df_temp(normalize )
[15]: y_es = df_label.to_numpy()
      x_es = df_temp.to_numpy()
      x_t, x_val , y_t, y_val = train_test_split(x_es, y_es, random_state=28)
[16]: params = {
          "estimator__colsample_bylevel": uniform(0.7, 0.3),
          "estimator__gamma": uniform(0, 0.5),
          "estimator_learning_rate": uniform(0.003, 0.3),
          "estimator__max_depth": randint(2, 6),
          "estimator n estimators": randint(100, 500),
          "estimator__subsample": uniform(0.6, 0.4)
      }
      : XGBRegressor
[17]: #Multioutput regression are regression problems that involve predicting two on
      →more numerical values given an input example.
      xgb model = MultiOutputRegressor(xgb.XGBRegressor(objective='reg:squarederror',__
       →eval_metric=['mae'], random_state = 1, use_label_encoder=False))
[18]: #search :
      search = RandomizedSearchCV(xgb_model, param_distributions=params, random_state_
      \rightarrow=1, n_iter=5, cv=3,
                                  verbose=0, n_jobs=1, return_train_score=True)
      search.fit(x_t, y_t)
      best_param = search.best_params_
      print('-')
      print(search.best_score_)
      print(best param)
      print('-')
```

```
0.08699591513110606
     {'estimator_colsample_bylevel': 0.9147911547905092, 'estimator_gamma':
     0.40137875196868245, 'estimator_learning_rate': 0.030840242592221373,
     'estimator max depth': 3, 'estimator n estimators': 101,
     'estimator_subsample': 0.9460081007915934}
[19]: #RandomizedSearchCV best param
     xgb_model = MultiOutputRegressor(xgb.XGBRegressor(objective='reg:
      →squarederror', eval_metric=['mae'], random_state = 1,
                                   use_label_encoder=False, **best_param)).fit(x_t,_
      \rightarrowy_t)
[20]: #MAE : MSE loss function
           loss function
     print('MAE: ',mean_absolute_error(y_val, xgb_model.predict(x_val)))
     MAE: 0.8795515228431303
[21]: #K-fold :
           Κ ,
     #K-fold
     folds = KFold(n_splits = 5, shuffle = True, random_state = 100)
     scores = cross val score(xgb model, x val, y val,
      ⇒scoring='neg_mean_absolute_error', cv=folds, verbose=0)
     print('KFold_Neg_MAE: ',np.mean(scores))
     KFold_Neg_MAE: -0.9389189271674322
     4 predict score
[22]: file_WC_team = './input_test_with_14_features.csv'
     df_teams = pd.read_csv(file_WC_team, encoding='cp949')
[23]: #inp_list: home_team, away team
     def predict_score(inp_list, autofill, Norm, seed, model):
         # ind_list = []
         # for i in inp_list:
               ind_list.append(nat_dic.get(i))
         # seed[:,1] = ind_list[0]
         # seed[:,2] = ind list[1]
         x_test = []
         qatar_flag = False
```

home_away_chagned_flag = False

```
home_team = inp_list[0]
away_team = inp_list[1]
# Netural == False
if (home_team == 'Qatar' or away_team == 'Qatar'):
    qatar_flag = True
   if(away_team == 'Qatar'):
        home_away_chagned_flag = True
        home_team, away_team = away_team, home_team
home_data = df_teams.loc[df_teams['team'] == home_team]
away_data = df_teams.loc[df_teams['team'] == away_team]
# Home_team norm
x_test.append(nat_dic.get(home_team))
# Away_team norm
x_test.append(nat_dic.get(away_team))
# Tournament norm
x_test.append(nat_dic.get('FIFA World Cup'))
# Neutral
if (qatar_flag):
   x_test.append(0)
else:
   x_test.append(1)
# Rank
x_test.append(float(home_data.iloc[0]['total_points']))
x_test.append(float(away_data.iloc[0]['total_points']))
feature_num = 14
                                   # TODO: Feature
for i in range(0, feature_num):
 x_test += home_data.iloc[0][2+(i*11):13+(i*11)].to_list()
 x_test += away_data.iloc[0][2+(i*11):13+(i*11)].to_list()
# Tournament
               len
x_test.append(len('FIFA World Cup'))
# Home_team
              len
x_test.append(len(home_team))
# Away_team len
x_test.append(len(away_team))
```

```
# Norm
    x_test_pd = pd.DataFrame(data=np.array(x_test).reshape(-1, len(x_test)),
columns=df_new.columns)
    x_test_norm = Norm.fit_transform(x_test_pd)
    pred = model.predict(x_test_norm)

output = np.ravel(pred).tolist()

if (home_away_chagned_flag):
    output[0], output[1] = output[1], output[0]

return np.ravel(pred).tolist()
```

5 def

```
[24]: import math
[25]: def create_params():
        seed = x_train[random.randint(0,x_train.shape[0])].reshape(1,-1)
        params = {
          'autofill': True,
          'Norm': Norm,
          'seed': seed,
          'model': xgb_model
        return params
[26]: def create_input(home_team, away_team):
        # home_team, away_team
                                                  2
        ind_list = [home_team, away_team]
        return ind_list#input_vector
[27]: def select_winning_team(ans, qual): #probability_array
          pred_ans = copy.deepcopy(ans)
          \#: x,
          if qual == True :
            pred_ans[0] = round(pred_ans[0])
            pred_ans[1] = round(pred_ans[1])
          # #
          11 11 11
          else:
            pred_ans[0] = round(pred_ans[0], 1)
```

```
pred_ans[1] = round(pred_ans[1], 1)
"""

if (pred_ans[0]>pred_ans[1]):
    out=0 # home

elif (pred_ans[0]<pred_ans[1]):
    out=1 # away

elif (pred_ans[0]==pred_ans[1]):
    out=2 #

return out, pred_ans</pre>
```

6

```
[29]: ##Group stage Matches
print("======Qualifying Games=====")
Group_standings={}
game_id=0
game_win_predicted={}
game_win_predicted_team={}
for grp_name in list(Groups.keys()):
    print(f"[{grp_name} Matches]")
    #
    probable_countries=Groups[grp_name]

    team_wins_dct={}
    goal_scored_dct={}
    goal_against_dct={}
```

```
win_dct={}
   draw_dct={}
   lost_dct={}
   for i in range(len(probable_countries)):
       #team i: team1, team j: team2
       j=i+1
       team_1=probable_countries[i]
       team wins=0
       while j<len((probable_countries)):</pre>
           team_2=probable_countries[j]
           team_lst=[team_1,team_2]
               home_team, away_team
           input=create_input(team_1, team_2) #np.
\rightarrow array([[year, stadium_num, team_1_num, team_2_num]])
           params = create_params()
           #ans : ex 1.3 : 1.5
           ans = predict_score(input, **params)
           #qual:
                       true,
                                false
           qual = True
           win, prob_lst=select_winning_team(ans, qual)
           #print("applied_score : ", prob_lst,", predicted_score: ", ans)
           goal_scored_dct[team_1] = goal_scored_dct.get(team_1,0)+prob_lst[0]
           goal_scored_dct[team_2] = goal_scored_dct.get(team_2,0)+prob_lst[1]
           goal_against_dct[team_1] = goal_against_dct.

    get(team_1,0)+prob_lst[1]
           goal_against_dct[team_2] = goal_against_dct.
\rightarrowget(team_2,0)+prob_lst[0]
           game_id+=1
           try:
               print(game_id," [win]: [", win, "]")
               game_win_predicted[game_id]=win
               if(win==2):
                    game_win_predicted_team[game_id]='draw'
               else:
                    game_win_predicted_team[game_id]=team_lst[win]
               print(f" {team_1} vs {team_2} \n {team_lst[win]} wins \n")
```

```
# print(str(prob_lst[0]) + " : " + str(prob_lst[1]))
    # print()
    #team1
    if (win) == 0:
        team_wins_dct[team_1] = team_wins_dct.get(team_1,0)+2
        team_wins_dct[team_2] = team_wins_dct.get(team_2,0)
        #team1 1
        win_dct[team_1] = win_dct.get(team_1,0)+1
        win_dct[team_2] = win_dct.get(team_2,0)
        #team2 1
        lost_dct[team_2] = lost_dct.get(team_2,0)+1
        lost_dct[team_1] = lost_dct.get(team_1,0)
        draw_dct[team_2] = draw_dct.get(team_2,0)
        draw_dct[team_1] = draw_dct.get(team_1,0)
    elif (win)==1:
        team_wins_dct[team_2] = team_wins_dct.get(team_2,0)+2
        team_wins_dct[team_1] = team_wins_dct.get(team_1,0)
        win_dct[team_2] = win_dct.get(team_2,0)+1
        win_dct[team_1] = win_dct.get(team_1,0)
        #team1 1
        lost_dct[team_1] = lost_dct.get(team_1,0)+1
        lost_dct[team_2] = lost_dct.get(team_2,0)
        draw_dct[team_1] = draw_dct.get(team_1,0)
        draw_dct[team_2] = draw_dct.get(team_2,0)
except IndexError:
    print(f"{team_1} vs {team_2} \n Match Draw \n")
    team_wins_dct[team_1] = team_wins_dct.get(team_1,0)+1
    team_wins_dct[team_2] = team_wins_dct.get(team_2,0)+1
    draw_dct[team_1] = draw_dct.get(team_1,0)+1
    draw_dct[team_2] = draw_dct.get(team_2,0)+1
    win_dct[team_1] = win_dct.get(team_1,0)
    lost_dct[team_1] = lost_dct.get(team_1,0)
    win_dct[team_2] = win_dct.get(team_2,0)
    lost_dct[team_2] = lost_dct.get(team_2,0)
```

```
j=j+1
 →group_results=[win_dct,draw_dct,lost_dct,team_wins_dct,goal_scored_dct,goal_against_dct]
    Group_standings[grp_name] = group_results
======Qualifying Games======
[Group A Matches]
1 [win]: [0]
Netherlands vs Senegal
  Netherlands wins
2 [win]: [0]
 Netherlands vs Ecuador
 Netherlands wins
3 [win]: [ 0 ]
Netherlands vs Qatar
  Netherlands wins
4 [win]: [0]
 Senegal vs Ecuador
  Senegal wins
5 [win]: [0]
 Senegal vs Qatar
  Senegal wins
6 [win]: [ 0 ]
 Ecuador vs Qatar
 Ecuador wins
[Group B Matches]
7 [win]: [0]
 England vs United States
  England wins
8 [win]: [ 0 ]
England vs Iran
 England wins
9 [win]: [ 0 ]
England vs Wales
  England wins
10 [win]: [ 0 ]
```

United States vs Iran

United States wins

11 [win]: [0]
United States vs Wales
United States wins

12 [win]: [2]
Iran vs Wales
Match Draw

[Group C Matches]
13 [win]: [0]
Argentina vs Poland
Argentina wins

14 [win]: [0]
Argentina vs Mexico
Argentina wins

15 [win]: [0]
Argentina vs Saudi Arabia
Argentina wins

16 [win]: [0]
Poland vs Mexico
Poland wins

17 [win]: [0]
Poland vs Saudi Arabia
Poland wins

18 [win]: [0]
Mexico vs Saudi Arabia
Mexico wins

[Group D Matches]
19 [win]: [0]
France vs Australia
France wins

20 [win]: [0]
France vs Tunisia
France wins

21 [win]: [0]
France vs Denmark
France wins

- 22 [win]: [0]
 Australia vs Tunisia
 Australia wins
- 23 [win]: [2]
 Australia vs Denmark
 Match Draw
- 24 [win]: [0]
 Tunisia vs Denmark
 Tunisia wins

[Group E Matches]
25 [win]: [2]
Japan vs Spain
Match Draw

- 26 [win]: [0]
 Japan vs Germany
 Japan wins
- 27 [win]: [0]
 Japan vs Costa Rica
 Japan wins
- 28 [win]: [0] Spain vs Germany Spain wins
- 29 [win]: [0]
 Spain vs Costa Rica
 Spain wins
- 30 [win]: [0]
 Germany vs Costa Rica
 Germany wins

[Group F Matches]
31 [win]: [2]
Morocco vs Croatia
Match Draw

- 32 [win]: [2]
 Morocco vs Belgium
 Match Draw
- 33 [win]: [0]
 Morocco vs Canada

Morocco wins

34 [win]: [2]
Croatia vs Belgium
Match Draw

35 [win]: [0]
Croatia vs Canada
Croatia wins

36 [win]: [0]
Belgium vs Canada
Belgium wins

[Group G Matches]
37 [win]: [0]
Brazil vs Switzerland
Brazil wins

38 [win]: [0]
Brazil vs Cameroon
Brazil wins

39 [win]: [0]
Brazil vs Serbia
Brazil wins

40 [win]: [0]
Switzerland vs Cameroon
Switzerland wins

41 [win]: [0]
Switzerland vs Serbia
Switzerland wins

42 [win]: [0]
Cameroon vs Serbia
Cameroon wins

[Group H Matches]
43 [win]: [0]
Portugal vs South Korea
Portugal wins

44 [win]: [0]
Portugal vs Uruguay
Portugal wins

```
45 [win]: [ 0 ]
Portugal vs Ghana
Portugal wins

46 [win]: [ 0 ]
South Korea vs Uruguay
South Korea wins

47 [win]: [ 0 ]
South Korea vs Ghana
South Korea wins

48 [win]: [ 0 ]
Uruguay vs Ghana
Uruguay wins
```

8 ACCURACY

```
[30]: file_qualifying_label='./qualifying_label.csv'
df_qualifying_label = pd.read_csv(file_qualifying_label)

list_label_win=df_qualifying_label['label_win'].tolist()
game_win_label = {}
game_id_label=1
for i in range(48):
    game_win_label[game_id_label]=list_label_win[i]
    game_id_label +=1

#game_win_label
#game_win_label
```

```
15: 1,
       16: 2,
       17: 0,
       18: 0,
       19: 0,
       20: 1,
       21: 0,
       22: 0,
       23: 0,
       24: 2,
       25: 0,
       26: 0,
       27: 1,
       28: 2,
       29: 0,
       30: 0,
       31: 2,
       32: 0,
       33: 0,
       34: 2,
       35: 0,
       36: 0,
       37: 0,
       38: 1,
       39: 0,
       40: 0,
       41: 0,
       42: 2,
       43: 1,
       44: 0,
       45: 0,
       46: 2,
       47: 1,
       48: 0}
[32]: for i in range(len(game_win_predicted)):
          df_qualifying_label['predicted_win'][i]=int(game_win_predicted[i+1])
          df_qualifying_label['predicted_win_team'][i]=game_win_predicted_team[i+1]
[33]: num_correct=0
      for i in range(len(game_win_predicted)):
        if(game_win_predicted[i+1] == game_win_label[i+1]):
          df_qualifying_label['accuracy'][i]=True
          num_correct+=1
        else:
          df_qualifying_label['accuracy'][i]=False
```

14: 0,

accuracy_rate = num_correct/len(game_win_predicted)*100

[34]: df_qualifying_label

[34]:	game_id	group	team_1	team_2	label_win	predicted_win	\
0	0	Α	Netherlands	Senegal	0	0.0	
1	1	Α	Netherlands	Ecuador	2	0.0	
2	2	Α	Netherlands	Qatar	0	0.0	
3	3	Α	Senegal	Ecuador	0	0.0	
4	4	Α	Senegal	Qatar	0	0.0	
5	5	Α	Ecuador	Qatar	0	0.0	
6	6	В	England	United States	2	0.0	
7	7	В	England	Iran	0	0.0	
8	8	В	England	Wales	0	0.0	
9	9	В	United States	Iran	0	0.0	
10	10	В	United States	Wales	2	0.0	
11	11	В	Iran	Wales	0	2.0	
12	12	C	Argentina	Poland	0	0.0	
13	13	C	Argentina	Mexico	0	0.0	
14	14	C	Argentina	Saudi Arabia	1	0.0	
15	15	C	Poland	Mexico	2	0.0	
16	16	C	Poland	Saudi Arabia	0	0.0	
17	17	C	Mexico	Saudi Arabia	0	0.0	
18	18	D	France	Australia	0	0.0	
19	19	D	France	Tunisia	1	0.0	
20	20	D	France	Denmark	0	0.0	
21	21	D	Australia	Tunisia	0	0.0	
22	22	D	Australia	Denmark	0	2.0	
23	23	D	Tunisia	Denmark	2	0.0	
24	24	Ε	Japan	Spain	0	2.0	
25	25	E	Japan	${\tt Germany}$	0	0.0	
26	26	E	Japan	Costa Rica	1	0.0	
27	27	E	Spain	${\tt Germany}$	2	0.0	
28	28	E	Spain	Costa Rica	0	0.0	
29	29	E	Germany	Costa Rica	0	0.0	
30	30	F	Morocco	Croatia	2	2.0	
31	31	F	Morocco	Belgium	0	2.0	
32	32	F	Morocco	Canada	0	0.0	
33	33	F	Croatia	Belgium	2	2.0	
34	34	F	Croatia	Canada	0	0.0	
35	35	F	Belgium	Canada	0	0.0	
36	36	G	Brazil	Switzerland	0	0.0	
37	37	G	Brazil	Cameroon	1	0.0	
38	38	G	Brazil	Serbia	0	0.0	
39	39	G	Switzerland	Cameroon	0	0.0	
40	40	G	Switzerland	Serbia	0	0.0	

41	41	G	Cameroon	Serbia	2	0.0
42	42	Н	Portugal	South Korea	1	0.0
43	43	Н	Portugal	Uruguay	0	0.0
44	44	Н	Portugal	Ghana	0	0.0
45	45	Н	South Korea	Uruguay	2	0.0
46	46	Н	South Korea	Ghana	1	0.0
47	47	Н	Uruguav	Ghana	0	0.0

	accuracy	labe win team	predicted_win_team
0	True	Netherlands	Netherlands
1	False	Draw	Netherlands
2	True	Netherlands	Netherlands
3	True	Senegal	Senegal
4	True	Senegal	Senegal
5	True	Ecuador	Ecuador
6	False	Draw	England
7	True	England	England
8	True	England	England
9	True	United States	United States
10	False	Draw	United States
11	False	Iran	draw
12	True		
13	True	Argentina Argentina	Argentina
14	False	Saudi Arabia	Argentina
15	False	_	Argentina Poland
		Draw	
16	True	Poland	Poland Mexico
17	True	Mexico	
18	True	France	France
19	False	Tunisia	France
20	True	France	France
21	True	Australia	Australia
22	False	Australia	draw -
23	False	Draw	Tunisia
24	False	Japan	draw
25	True	Japan	Japan
26	False	Costa Rica	Japan
27	False	Draw	Spain
28	True	Spain	Spain
29	True	Germany	Germany
30	True	Draw	draw
31	False	Morocco	draw
32	True	Morocco	Morocco
33	True	Draw	draw
34	True	Croatia	Croatia
35	True	Belgium	Belgium
36	True	Brazil	Brazil
37	False	Cameroon	Brazil

Brazil	Brazil
Switzerland	Switzerland
Switzerland	Switzerland
e Draw	Cameroon
South Korea	Portugal
Portugal	Portugal
Portugal	Portugal
e Draw	South Korea
Ghana	South Korea
uruguay	Uruguay
	Switzerland Switzerland Draw South Korea Portugal Portugal Draw Ghana

[35]: accuracy_rate

[35]: 62.5

[36]: df_qualifying_label

[36]:	game_id	group	team_1	team_2	label_win	<pre>predicted_win \</pre>
0	0	Α	Netherlands	Senegal	0	0.0
1	1	Α	Netherlands	Ecuador	2	0.0
2	2	Α	Netherlands	Qatar	0	0.0
3	3	Α	Senegal	Ecuador	0	0.0
4	4	Α	Senegal	Qatar	0	0.0
5	5	Α	Ecuador	Qatar	0	0.0
6	6	В	England	United States	2	0.0
7	7	В	England	Iran	0	0.0
8	8	В	England	Wales	0	0.0
9	9	В	United States	Iran	0	0.0
10	10	В	United States	Wales	2	0.0
11	11	В	Iran	Wales	0	2.0
12	12	C	Argentina	Poland	0	0.0
13	13	C	Argentina	Mexico	0	0.0
14	14	C	Argentina	Saudi Arabia	1	0.0
15	15	C	Poland	Mexico	2	0.0
16	16	C	Poland	Saudi Arabia	0	0.0
17	17	C	Mexico	Saudi Arabia	0	0.0
18	18	D	France	Australia	0	0.0
19	19	D	France	Tunisia	1	0.0
20	20	D	France	Denmark	0	0.0
21	21	D	Australia	Tunisia	0	0.0
22	22	D	Australia	Denmark	0	2.0
23	23	D	Tunisia	Denmark	2	0.0
24	24	Ε	Japan	Spain	0	2.0
25	25	Ε	Japan	${\tt Germany}$	0	0.0
26	26	E	Japan	Costa Rica	1	0.0
27	27	E	Spain	Germany	2	0.0
28	28	E	Spain	Costa Rica	0	0.0

29	29	Ε	${\tt Germany}$	Costa Rica	0	0.0
30	30	F	Morocco	Croatia	2	2.0
31	31	F	Morocco	Belgium	0	2.0
32	32	F	Morocco	Canada	0	0.0
33	33	F	Croatia	Belgium	2	2.0
34	34	F	Croatia	Canada	0	0.0
35	35	F	Belgium	Canada	0	0.0
36	36	G	Brazil	Switzerland	0	0.0
37	37	G	Brazil	Cameroon	1	0.0
38	38	G	Brazil	Serbia	0	0.0
39	39	G	Switzerland	Cameroon	0	0.0
40	40	G	Switzerland	Serbia	0	0.0
41	41	G	Cameroon	Serbia	2	0.0
42	42	Н	Portugal	South Korea	1	0.0
43	43	Н	Portugal	Uruguay	0	0.0
44	44	Н	Portugal	Ghana	0	0.0
45	45	Н	South Korea	Uruguay	2	0.0
46	46	Н	South Korea	Ghana	1	0.0
47	47	Н	Uruguay	Ghana	0	0.0

	accuracy	labe_win_team	<pre>predicted_win_team</pre>
0	True	Netherlands	Netherlands
1	False	Draw	Netherlands
2	True	Netherlands	Netherlands
3	True	Senegal	Senegal
4	True	Senegal	Senegal
5	True	Ecuador	Ecuador
6	False	Draw	England
7	True	England	England
8	True	England	England
9	True	United States	United States
10	False	Draw	United States
11	False	Iran	draw
12	True	Argentina	Argentina
13	True	Argentina	Argentina
14	False	Saudi Arabia	Argentina
15	False	Draw	Poland
16	True	Poland	Poland
17	True	Mexico	Mexico
18	True	France	France
19	False	Tunisia	France
20	True	France	France
21	True	Australia	Australia
22	False	Australia	draw
23	False	Draw	Tunisia
24	False	Japan	draw
25	True	Japan	Japan

```
26
      False
                 Costa Rica
                                           Japan
27
                                           Spain
      False
                        Draw
28
       True
                       Spain
                                           Spain
29
       True
                    Germany
                                         Germany
30
       True
                       Draw
                                             draw
      False
31
                    Morocco
                                             draw
32
       True
                    Morocco
                                         Morocco
33
       True
                       Draw
                                             draw
34
                                         Croatia
       True
                    Croatia
35
       True
                    Belgium
                                         Belgium
36
       True
                     Brazil
                                          Brazil
37
      False
                   Cameroon
                                          Brazil
38
       True
                     Brazil
                                          Brazil
39
       True
                Switzerland
                                     Switzerland
40
                Switzerland
                                     Switzerland
       True
41
      False
                       Draw
                                        Cameroon
42
      False
                South Korea
                                        Portugal
43
       True
                   Portugal
                                        Portugal
44
       True
                   Portugal
                                        Portugal
45
      False
                        Draw
                                     South Korea
46
      False
                                     South Korea
                       Ghana
47
       True
                    Uruguay
                                         Uruguay
```

```
[37]: #game_win_label game_win_predicted
num_correct=0
correct_game_id=[]
num_wrong=0
wrong_game_id=[]
for i in range(1, len(game_win_label)+1):
    if(game_win_label[i] == game_win_predicted[i]):
        num_correct+=1
        correct_game_id.append(i)
    else:
        num_wrong+=1
        wrong_game_id.append(i)
```

```
[45]: print("[PREDICTION]")

for grp_name in list(Group_standings.keys()):

    team_wins_dct= dict(sorted(Group_standings[grp_name][3].items()))
    goal_scored_dct=dict(sorted(Group_standings[grp_name][4].items()))
    goal_against_dct=dict(sorted(Group_standings[grp_name][5].items()))
```

```
win_dct=dict(sorted(Group_standings[grp_name][0].items()))
   draw_dct=dict(sorted(Group_standings[grp_name][1].items()))
   lost_dct=dict(sorted(Group_standings[grp_name][2].items()))
   lst_teams=list(team_wins_dct.keys())
   win_lst=list(win_dct.values())
   draw_lst=list(draw_dct.values())
   lost_lst=list(lost_dct.values())
   lst_win_count=list(team_wins_dct.values())
   goal_scored=list(goal_scored_dct.values())
   goal_against=list(goal_against_dct.values())
   goal_differance=[goal_scored[i]-goal_against[i] for i in range_
→(len(goal_scored))]
   ranking_table=pd.
→DataFrame(list(zip(lst_teams,win_lst,draw_lst,lost_lst,goal_scored,goal_against,goal_differ

→Scored", "Goal Against", "Goal Differance", "Points"])
   ranking_table=ranking_table.sort_values("Points", ascending=False).
→reset_index(drop=True)
   ranking_table.index = ranking_table.index + 1
   print(f"\n\n{grp_name} Final Rankings")
   print(ranking_table.to_markdown())
```

[PREDICTION]

```
Group A Final Rankings
  | Team
        | Wins | Draw | Lost | Goal Scored | Goal Against |
Goal Differance | Points |
|---:|:-----:|-----:|-----:|-----:|-----:|
-----:
| 1 | Netherlands | 3 | 0 | 0 |
                                     12 |
                                                 7 |
5 l
      6 I
         | 2 | 0 | 1 |
| 2 | Senegal
                                    10 |
                                                10
      4 |
| 3 | Ecuador | 1 | 0 | 2 | 9 |
                                                11 |
-2 | 2 |
| 4 | Qatar
          | 0 | 0 | 3 |
                                9 |
                                                12 l
-3 | 0 |
Group B Final Rankings
  | Team
              Wins | Draw | Lost | Goal Scored | Goal Against
  Goal Differance | Points |
```

:	:	:			
1 England	3	0	0 1	12	1 6
_	l 6				
2 United States	1 2 1	0	1	8	1 8
1 0			•		•
3 Iran			2	7	10
-3					
4 Wales			2	7	l 10
	1			'	1
1		1			
Group C Final Rankin	or c				
-	~	Draw	Logt	Cool Scored	Goal Against
Goal Differance			TOP!	Goal Scoled	Goal Against
: :			. 1	. 1	
			: -	:	:
:			0 1	40	
1 Argentina			0	12	6
	1 6				
2 Poland			1	10	8
1 2					
3 Mexico			2	8	10
-2					
4 Saudi Arabia	0	0	3	6	12
l –6	0				
Group D Final Rankin	gs				
-	-	raw I	Lost (Goal Scored	Goal Against
-	Wins Da	raw I	Lost (Goal Scored	Goal Against
Team	Wins Da Points				
Team Goal Differance	Wins Di Points :				
Team Goal Differance : :	Wins Di Points :	:	:		
Team Goal Differance : :	Wins Di Points :	:	:	:	:
Team Goal Differance	Wins Di Points : : 3	: 0	: 0	13	7
Team Goal Differance : :	Wins Di Points : : 3	:	: 0	:	:
Team Goal Differance	Wins Di Points : : 3 1	: 0 1	0 1	: 13 9	7 11
Team Goal Differance	Wins Di Points : : 3 1	: 0 1	0 1	13	7
Team Goal Differance	Wins Di Points : : 3 1 1	: 0	0 1 2	: 13 9 8	7 11 10
Team Goal Differance	Wins Di Points : : 3 1 1	: 0	0 1 2	: 13 9	7 11
Team Goal Differance	Wins Di Points : : 3 1 1	: 0	0 1 2	: 13 9 8	7 11 10
Team Goal Differance	Wins Di Points : : 3 1 1	: 0	0 1 2	: 13 9 8	7 11 10
Team Goal Differance	Wins Di Points : 3 1 1 0	: 0	0 1 2	: 13 9 8	7 11 10
Team Goal Differance : :	Wins Di Points : : 3 1 1 0	: 0	: 0 1 2 2	13 9 8 10	7 11 10 12
Team Goal Differance : :	Wins Di Points : 3 1 1 0	: 0	: 0 1 2 2	13 9 8 10	7 11 10 12
Team Goal Differance : :	Wins Di Points :	: 0	: 0 1 2 2 Lost	: 13 9 8 10 Goal Scored	7 11 10 12 Goal Against
Team Goal Differance : :	Wins Di Points :	: 0	: 0 1 2 2 Lost	: 13 9 8 10 Goal Scored	7 11 10 12 Goal Against
Team Goal Differance : :	Wins Di Points : 3 1 1 0 gs Wins Points : :	: 0	: 0	: : :	: 7 11 10 12 Goal Against
Team Goal Differance : :	Wins Di Points :	: 0	: 0	: : :	: 7 11 10 12 Goal Against

2 Spain	2	1	0	11	7
3 Germany	1	0	2	9	10
-1 2 4 Costa Rica -6 0	0	0	3	6	12
Group F Final Ranki	-				
Team	Wins Dra	w Los	t Goa	l Scored Goa	al Against
Goal Differance					
: :		-:	-:	:	:
:		0 1	0 1	40.1	0.1
1 Belgium	1	2	0	10	8
2 4	4 1	0 1	0 1	40.	0.1
2 Croatia	T	2	0	10	8
2 4	4 1	o 1	0 1	0.1	0 1
3 Morocco	1	2	0	9	8
1 4	0 1	o 1	2	6. 1	44
4 Canada -5 0	0 1	0	3	6	11
-5 0					
Group G Final Ranki Team Goal Differance : ::	Wins Points :				
1 Brazil 5 6	•	0	0	12	7
2 Switzerland 4	2	0	1	10	8
3 Cameroon	1 1	0	2	9	10
-1 2 4 Serbia		0 1	2 I	6 1	10
-6 0	0 1	0 1	3	0	12
-0 0					
Group H Final Ranki	nøs				
Team	•	Draw	Lost	Goal Scored	Goal Against
Goal Differance		Draw ,	2000	dodi boolou i	4041 1184111111 1
: :					·I
:		• •	• •	• 1	• •
1 Portugal	•	0	0	10	6 I
4 6		- 1	- 1	1	~ I
2 South Korea	2	0	1 l	10	8
2 4	•	•	·	- 1	
3 Uruguay	1	0	2	9	9

```
0 | 2 |
| 4 | Ghana | 0 | 0 | 3 | 6 | 12 |
-6 | 0 |
```

```
[46]: ##Round of 16 Section 1
      qualified teams 1=[]
      standings=list(Group_standings.keys())
      print(f"Round of 16\n")
      while i < (len(standings)):</pre>
          A_team= sorted(Group_standings[standings[i]][3].items(), key=lambda x:__
       \rightarrowx[1], reverse=True)
          team_1=A_team[0][0]
          B_team= sorted(Group_standings[standings[i+1]][3].items(), key=lambda x:
       \rightarrow x[1], reverse=True)
          team 2=B team[1][0]
          #team_1_num=label_encoder.transform([team_1])[0]
          #team_2_num=label_encoder.transform([team_2])[0]
          team lst=[team 1,team 2]
          #Input_vector=np.array([[year,host_num,team_1_num,team_2_num]])
          input=create_input(team_1, team_2)
          params = create_params()
          #res=model.predict(Input vector)
          ans = predict_score(input, **params)
          #win,_=select_winning_team(res)
          qual = False
          win, prob_lst=select_winning_team(ans, qual)
          try:
                  #print("applied_score : ", prob_lst,", predicted_score: ", ans)
                  print(f"{team_1} vs {team_2} \n {team_lst[win]} wins \n")
                  # print(str(prob_lst[0]) + " : " + str(prob_lst[1]))
                  # print()
                  #print(f"
                              \{team\ lst[win]\}\ into\ the\ Quater-Finals\ \n")
                  qualified_teams_1.append(team_lst[win])
          except IndexError:
                  print(f"{team_1} vs {team_2} \n [ERROR]Match Draw ")
                  winning_team=random.choice(team_lst)
```

```
print(f"
                      {winning_team} wins at Penaly Shoot-Out ")
                        {winning_team} into the Quater-Finals \n'')
            print(f"
            qualified_teams_1.append(winning_team)
    i=i+2
##Round of 16 Section 2
qualified_teams_2=[]
standings=list(Group standings.keys())
i=0
while i < (len(standings)):</pre>
    A_team= sorted(Group_standings[standings[i]][3].items(), key=lambda x:_u
\rightarrow x[1], reverse=True)
    team_1=A_team[1][0]
    B_team= sorted(Group_standings[standings[i+1]][3].items(), key=lambda x:__
\rightarrowx[1], reverse=True)
    team_2=B_team[0][0]
    #team_1_num=label_encoder.transform([team_1])[0]
    #team_2_num=label_encoder.transform([team_2])[0]
    team_lst=[team_1,team_2]
    #Input_vector=np.array([[year,host_num,team_1_num,team_2_num]])
    input=create_input(team_1, team_2)
    params = create_params()
    #res=model.predict(Input_vector)
    ans = predict score(input, **params)
    #win, =select winning team(res)
    qual = False
    win, prob_lst=select_winning_team(ans, qual)
    try:
            #print("applied_score : ", prob_lst,", predicted_score: ", ans)
            print(f"{team 1} vs {team 2} \n {team lst[win]} wins \n")
            \# print(str(prob\_lst[0]) + " : " + str(prob\_lst[1]))
            # print()
            #print(f"
                         \{team\_lst[win]\}\ into\ the\ Quater-Finals\ \n")
            qualified_teams_2.append(team_lst[win])
    except IndexError:
            print(f"{team_1} vs {team_2} \n [ERROR]Match Draw ")
            winning_team=random.choice(team_lst)
            print(f"
                      {winning_team} wins at Penaly Shoot-Out ")
                          {winning_team} into the Quater-Finals \n'')
            #print(f"
            qualified_teams_2.append(winning_team)
```

```
i=i+2
Round of 16
Netherlands vs United States
  Netherlands wins
Argentina vs Australia
  Argentina wins
Japan vs Croatia
  Japan wins
Brazil vs South Korea
  Brazil wins
Senegal vs England
  Senegal wins
Poland vs France
  Poland wins
Spain vs Morocco
  Spain wins
```

Switzerland vs Portugal Switzerland wins

```
[47]: #Quarter Finals

Semifinal_teams=[]
i=0
print(f"Quater Final Matches\n")
while i < (len(qualified_teams_1))-1:
    team_1= qualified_teams_1[i]
    team_2= qualified_teams_1[i+1]

#team_1_num=label_encoder.transform([team_1])[0]
#team_2_num=label_encoder.transform([team_2])[0]
team_lst=[team_1,team_2]

#Input_vector=np.array([[year,host_num,team_1_num,team_2_num]])
input=create_input(team_1, team_2)</pre>
```

```
params = create_params()
    #res=model.predict(Input_vector)
    ans = predict_score(input, **params)
    #win,_=select_winning_team(res)
    qual = False
    win, prob_lst=select_winning_team(ans, qual)
    try:
            #print("applied_score : ", prob_lst,", predicted_score: ", ans)
            print(f"{team_1} vs {team_2} \n {team_lst[win]} wins \n")
                        \{team\ lst[win]\}\ into\ the\ Semi-Finals\ \n")
            Semifinal_teams.append(team_lst[win])
    except IndexError:
            print(f"{team_1} vs {team_2} \n [ERROR]Match Draw")
            winning_team=random.choice(team_lst)
            print(f" {winning_team} wins at Penaly Shoot-Out
            print(f"
                       {winning_team} into the Semi-Finals \n'')
            Semifinal_teams.append(winning_team)
    i=i+2
i=0
while i < (len(qualified teams 2))-1:
    team 1= qualified teams 2[i]
    team_2= qualified_teams_2[i+1]
    #team_1_num=label_encoder.transform([team_1])[0]
    #team_2_num=label_encoder.transform([team_2])[0]
    team_lst=[team_1,team_2]
    #Input_vector=np.array([[year,host_num,team_1_num,team_2_num]])
    input=create_input(team_1, team_2)
    params = create_params()
    #res=model.predict(Input_vector)
    ans = predict_score(input, **params)
    #win, =select winning team(res)
    qual = False
    win, prob lst=select winning team(ans, qual)
    try:
            #print("applied_score : ", prob_lst,", predicted_score: ", ans)
            print(f"{team_1} vs {team_2} \n {team_lst[win]} wins \n")
            \#print(f'' \{team\_lst[win]\}\ into\ the\ Semi-Finals\ \n'')
            Semifinal_teams.append(team_lst[win])
    except IndexError:
```

```
print(f"{team_1} vs {team_2} \n [ERROR]Match Draw ")
    """
    winning_team=random.choice(team_lst)
    print(f" {winning_team} wins at Penaly Shoot-Out ")
    print(f" {winning_team} into the Semi-Finals \n")
    Semifinal_teams.append(winning_team)
    """

i=i+2
```

Quater Final Matches

Netherlands vs Argentina
Netherlands wins

Japan vs Brazil
Japan wins

Senegal vs Poland
Senegal wins

Spain vs Switzerland
Spain wins

```
[48]: final_teams=[]
      third_place_match_teams=[]
      print(f"Semi Final Matches\n")
      while i < (len(Semifinal_teams))-1:</pre>
          team 1= Semifinal teams[i]
          team_2= Semifinal_teams[i+1]
          #team_1_num=label_encoder.transform([team_1])[0]
          #team_2_num=label_encoder.transform([team_2])[0]
          team_lst=[team_1,team_2]
          #Input vector=np.array([[year,host_num,team_1_num,team_2_num]])
          input=create_input(team_1, team_2)
          params = create_params()
          #res=model.predict(Input_vector)
          ans = predict_score(input, **params)
          #win,_=select_winning_team(res)
          win, prob_lst=select_winning_team(ans, qual)
```

```
try:
        #print("applied_score : ", prob_lst,", predicted_score: ", ans)
        print(f"{team_1} vs {team_2} \n {team_lst[win]} wins \n")
                     {team_lst[win]} into the FiIFA-Finals
        #print(f"
        final_teams.append(team_lst[win])
        third_place_match_teams.append(team_lst[(win+1)%2])
except IndexError:
        print(f"{team_1} vs {team_2} \n [ERROR] Match Draw ")
        winning_team=random.choice(team_lst)
        print(f"
                  {winning_team} wins at Penaly Shoot-Out
        print(f"
                    {winning_team} into the FIFA-Finals
                                                         \backslash n'')
        final_teams.append(winning_team)
        team_lst.remove(winning_team)
        third_place_match_teams.append(team_lst[0])
i=i+2
```

Semi Final Matches

Netherlands vs Japan Netherlands wins

Senegal vs Spain Senegal wins

```
[50]: #Finals and Third Place match

print(f"FIFA FINAl\n")
  team_1= final_teams[1]
  team_2= final_teams[0]

#team_1_num=label_encoder.transform([team_1])[0]
  #team_2_num=label_encoder.transform([team_2])[0]
  team_lst=[team_1,team_2]

#Input_vector=np.array([[year,host_num,team_1_num,team_2_num]])
  input=create_input(team_1, team_2)
  params = create_params()
  #res=model.predict(Input_vector)
  ans = predict_score(input, **params)
```

```
#win,_=select_winning_team(res)
win, prob_lst=select_winning_team(ans, qual)
try:
    #print("applied_score : ", prob_lst,", predicted_score: ", ans)
    if (ans[0] > ans[1]):
       print("[predicted_score] ", round(ans[0]), ":", round(ans[1])-1)
   else:
        print("[predicted_score] ", round(ans[0]-1), ":", round(ans[1]))
   print(f"{team_1} vs {team_2} \n {team_lst[win]} are the Winners \n\n")
   winner=team lst[win]
   place_2=team_lst[(win+1)%2]
except IndexError:
   print(f"{team_1} vs {team_2} \n [ERROR]Match Draw ")
   winning_team=random.choice(team_lst)
   print(f" {winning_team} wins at Penaly Shoot-Out ")
   print(f"
              {winning_team} are the Winners \n\n'')
   winner=winning_team
    team lst.remove(winning team)
   place_2=team_lst[0]
print(f"Third Place match\n")
team_1= third_place_match_teams[1]
team_2= third_place_match_teams[0]
#team_1_num=label_encoder.transform([team_1])[0]
#team_2_num=label_encoder.transform([team_2])[0]
team_lst=[team_1,team_2]
#Input_vector=np.array([[year,host_num,team_1_num,team_2_num]])
input=create_input(team_1, team_2)
params = create_params()
#res=model.predict(Input vector)
ans = predict_score(input, **params)
#win, =select winning team(res)
win, prob_lst=select_winning_team(ans, qual)
try:
   print(f"{team_1} vs {team_2} \n {team_lst[win]} Wins the 3rd Place \n")
   place_3=team_lst[win]
except IndexError:
   print(f"{team_1} vs {team_2} \n [ERROR]Match Draw ")
```

```
winning_team=random.choice(team_lst)
                     {winning_team} wins at Penaly Shoot-Out
                      {winning_team} Wins the 3rd Place \n")
          print(f"
          place\_3 = winning\_team
      print(f"\n\nWinner is {winner} ")
      print(f"Runner-up is {place_2} ")
      print(f"3rd Place is {place_3} ")
     FIFA FINAL
     [predicted_score] 3 : 2
     Senegal vs Netherlands
       Senegal are the Winners
     Third Place match
     Spain vs Japan
       Spain Wins the 3rd Place
     Winner is Senegal
     Runner-up is Netherlands
     3rd Place is Spain
[43]: | !apt-get update
      !apt-get install texlive texlive-xetex texlive-latex-extra pandoc
      !pip install pypandoc
     Ign:1 https://developer.download.nvidia.com/compute/machine-
     learning/repos/ubuntu1804/x86_64 InRelease
     Hit:2 https://developer.download.nvidia.com/compute/cuda/repos/ubuntu1804/x86_64
     InRelease
     Hit:3 https://cloud.r-project.org/bin/linux/ubuntu bionic-cran40/ InRelease
     Hit:4 https://developer.download.nvidia.com/compute/machine-
     learning/repos/ubuntu1804/x86_64 Release
     Get:5 http://security.ubuntu.com/ubuntu bionic-security InRelease [88.7 kB]
     Hit:6 http://ppa.launchpad.net/c2d4u.team/c2d4u4.0+/ubuntu bionic InRelease
     Hit:7 http://archive.ubuntu.com/ubuntu bionic InRelease
     Hit:9 http://archive.ubuntu.com/ubuntu bionic-updates InRelease
     Hit:10 http://ppa.launchpad.net/cran/libgit2/ubuntu bionic InRelease
```

```
Get:11 http://archive.ubuntu.com/ubuntu bionic-backports InRelease [83.3 kB]
     Hit:12 http://ppa.launchpad.net/deadsnakes/ppa/ubuntu bionic InRelease
     Hit:13 http://ppa.launchpad.net/graphics-drivers/ppa/ubuntu bionic InRelease
     Fetched 172 kB in 2s (88.3 kB/s)
     Reading package lists... Done
     Reading package lists... Done
     Building dependency tree
     Reading state information... Done
     pandoc is already the newest version (1.19.2.4~dfsg-1build4).
     texlive is already the newest version (2017.20180305-1).
     texlive-latex-extra is already the newest version (2017.20180305-2).
     texlive-xetex is already the newest version (2017.20180305-1).
     The following package was automatically installed and is no longer required:
       libnvidia-common-460
     Use 'apt autoremove' to remove it.
     0 upgraded, 0 newly installed, 0 to remove and 19 not upgraded.
     Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-
     wheels/public/simple/
     Requirement already satisfied: pypandoc in /usr/local/lib/python3.8/dist-
     packages (1.10)
[44]: ||!|jupyter nbconvert --to PDF '/content/drive/MyDrive/2022/BigData/Code/
      →model2022 final.ipynb'
     [NbConvertApp] Converting notebook
     /content/drive/MyDrive/2022/BigData/Code/model2022_final.ipynb to PDF
     [NbConvertApp] Writing 127387 bytes to ./notebook.tex
     [NbConvertApp] Building PDF
     [NbConvertApp] Running xelatex 3 times: ['xelatex', './notebook.tex', '-quiet']
     [NbConvertApp] Running bibtex 1 time: ['bibtex', './notebook']
     [NbConvertApp] WARNING | bibtex had problems, most likely because there were no
     citations
     [NbConvertApp] PDF successfully created
     [NbConvertApp] Writing 100654 bytes to
     /content/drive/MyDrive/2022/BigData/Code/model2022_final.pdf
[44]:
```