# العمل في التوقع العلمي في نتائج كرة القدم عمل وتنفيذ دكتور الدرديري فضل إبراهيم

#### **FOOT-BALL GAMES (PREDICTIONS)**

import pandas as pd

import matplotlib.pyplot as plt

import numpy as np

import seaborn as sns

from scipy.stats import poisson,skellam

import matplotlib.pyplot as py

from scipy.optimize import minimize

epl\_1718 = pd.read\_csv("http://www.football-("data.co.uk/mmz4281/1718/E0.csv

[['epl\_1718 = epl\_1718[['HomeTeam','AwayTeam','FTHG','FTAG

epl\_1718 = epl\_1718.rename(columns={'FTHG': 'HomeGoals', 'FTAG': ({"AwayGoals

()epl 1718.head

AwayGoals HomeGoals AwayTeam HomeTeam

- 3 4 Leicester Arsenal 0
- 2 0 Man City Brighton 1
- 3 2 Burnley Chelsea 2
- 3 0 Huddersfield Crystal Palace 3
- 0 1 Stoke Everton 4

```
[epl_1718 = epl_1718[:-10]
()epl 1718.mean
HomeGoals 1.518919
AwayGoals 1.148649
dtype: float64
,(plt.hist(epl_1718[['HomeGoals', 'AwayGoals']].values, range(9
alpha=0.7, label=['Home', 'Away'],normed=True,
(["color=["#FFA07A", "#20B2AA
array([ 0.23783784, 0.33243243, 0.24324324, 0.08918919, ])
,0.05945946
,([ 0.0027027 , .0 ,0.03513514
array([ 0.35675676, 0.33513514, 0.17027027, 0.08918919,
.0.04054054
,[([ .0 , 0.0027027 ,0.00540541
([array([0, 1, 2, 3, 4, 5, 6, 7, 8,
(<a list of 2 Lists of Patches objects>
poisson_pred = np.column_stack([[poisson.pmf(i, epl_1718.mean()[j])
([(for i in range(8)] for j in range(2)
[pois1, = plt.plot([i-0.5 for i in range(1,9)], poisson pred[:,0])
('linestyle='-', marker='o',label="Home", color = '#CD5C5C
,[pois2, = plt.plot([i-0.5 for i in range(1,9)], poisson_pred[:,1
```

```
('linestyle='-', marker='o',label="Away", color = '#006400
(leg=plt.legend(loc='upper right', fontsize=13, ncol=2
leg.set title("Poisson
                            Actual ", prop = {'size':'14',
({"weight":'bold
([(plt.xticks([i-0.5 for i in range(1,9)],[i for i in range(9
,<matplotlib.axis.XTick at 0x20ba0bbedd8>])
,<matplotlib.axis.XTick at 0x20ba0a94358>
,<matplotlib.axis.XTick at 0x20ba0c07c88>
,<matplotlib.axis.XTick at 0x20ba0c94978>
,<matplotlib.axis.XTick at 0x20ba0c9d080>
,<matplotlib.axis.XTick at 0x20ba0c9d6d8>
,<matplotlib.axis.XTick at 0x20ba0c9dcf8>
,[<matplotlib.axis.XTick at 0x20ba0ca3320>
(<a list of 8 Text xticklabel objects>
([skellam.pmf(0.0, epl_1718.mean()[0], epl_1718.mean()[1
0.25480879523707972
(plt.xlabel("Goals per Match", size=13
(plt.ylabel("Proportion of Matches", size=13
plt.title("Number of Goals per Match (EPL 2016/17
('Season)", size=14, fontweight='bold
([plt.ylim([-0.004, 0.4
```

```
()plt.tight_layout
()plt.show
([skellam.pmf(1, epl 1718.mean()[0], epl 1718.mean()[1
0.2284260568855998
skellam_pred = [skellam.pmf(i, epl_1718.mean()[0],
[(epl 1718.mean()[1]) for i in range(-6,8
plt.hist(epl 1718[['HomeGoals']].values -
,(epl_1718[['AwayGoals']].values, range(-6,8
(alpha=0.7, label='Actual',normed=True
,array([ 0.0027027 , 0. , 0.02162162, 0.05405405, 0.05675676)
,0.10540541 ,0.21891892 ,0.26756757 ,0.14594595
,0.07027027
,([ .0 ,0.02702703 ,0.02972973
,([array([-6, -5, -4, -3, -2, -1, 0, 1, 2, 3, 4, 5, 6, 7
(<a list of 13 Patch objects>
(fig,(ax1,ax2) = plt.subplots(2, 1)
,plt.plot([i+0.5 for i in range(-6,8)], skellam pred
('linestyle='-', marker='o',label="Skellam", color = '#CD5C5C
(plt.legend(loc='upper right', fontsize=13
([(plt.xticks([i+0.5 for i in range(-6,8)],[i for i in range(-6,8)]
```

```
(plt.xlabel("Home Goals - Away Goals",size=13
(plt.ylabel("Proportion of Matches", size=13
plt.title("Difference in Goals Scored (Home Team vs Away
('Team)",size=14,fontweight='bold
([plt.ylim([-0.004, 0.26
()plt.tight_layout
()plt.show
(fig,(ax1,ax2) = plt.subplots(2, 1)
chel home =
epl_1718[epl_1718['HomeTeam']=='Chelsea'][['HomeGoals']].apply(pd.v
(alue counts,normalize=True
chel home pois =
[poisson.pmf(i,np.sum(np.multiply(chel_home.values.T,chel_home.inde
[(x.T),axis=1)[0] for i in range(8
sun home =
epl 1718[epl 1718['HomeTeam']=='Sunderland'][['HomeGoals']].apply(p
(d.value_counts,normalize=True
sun_home_pois =
[poisson.pmf(i,np.sum(np.multiply(sun home.values.T,sun home.index.
[(T),axis=1)[0]) for i in range(8
chel away =
epl 1718[epl 1718['AwayTeam']=='Chelsea'][['AwayGoals']].apply(pd.va
(lue counts,normalize=True
```

```
chel away pois =
[poisson.pmf(i,np.sum(np.multiply(chel away.values.T,chel away.index
[(.T),axis=1)[0] for i in range(8
sun away =
epl_1718[epl_1718['AwayTeam']=='Sunderland'][['AwayGoals']].apply(p
(d.value counts,normalize=True
sun away pois =
[poisson.pmf(i,np.sum(np.multiply(sun_away.values.T,sun_away.index.
[(T),axis=1)[0]) for i in range(8
import statsmodels.api as sm
import statsmodels.formula.api as smf
goal model data =
pd.concat([epl 1718[['HomeTeam','AwayTeam','HomeGoals']].assign(h
)ome=1).rename
columns={'HomeTeam':'team',
,({"AwayTeam':'opponent','HomeGoals':'goals
epl_1718[['AwayTeam','HomeTeam','AwayGoals']].assign(home=0).ren
)ame
columns={'AwayTeam':'team',
([({"HomeTeam':'opponent','AwayGoals':'goals
poisson model = smf.glm(formula="goals ~ home + team + opponent",
,data=goal model data
()family=sm.families.Poisson()).fit
```

#### (()print(poisson\_model.summary

#### Generalized Linear Model Regression Results

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\_\_\_\_\_

Dep. Variable: goals No. Observations: 740

Model: GLM Df Residuals: 700

Model Family: Poisson Df Model: 39

Link Function: log Scale: 1.0

Method: IRLS Log-Likelihood: -1020.2

Date: Wed, 26 Dec 2018 Deviance: 768.51

.Time: 01:16:59 Pearson chi2: 653

No. Iterations: 5

\_\_\_\_\_

[coef std err z P>|z| [0.025 0.975

Intercept 0.5855 0.189 3.091 0.002 0.214

0.957

team[T.Bournemouth] -0.5306 0.193 -2.754 0.006 -

0.908 -0.153

team[T.Brighton] -0.7740 0.208 -3.721 0.000 -1.182

-0.366

team[T.Burnley] -0.7418 0.206 -3.602 0.000 -1.145

-0.338

team[T.Chelsea] -0.1860 0.173 -1.074 0.283 -

0.525 0.153

team[T.Crystal Palace] -0.5203 0.193 -2.701 0.007 -

0.898 -0.143

team[T.Everton] -0.144	-0.5212	0.193	-2.706	0.007	-0.899
team[T.Huddersfield] 1.383 -0.512	-0.9476	0.222	-4.261	0.000	-
team[T.Leicester] 0.011	-0.3458	0.182	-1.901	0.057	-0.702
team[T.Liverpool] 0.400	0.0822	0.162	0.506	0.613	-0.236
team[T.Man City] 0.637	0.3378	0.153	2.211	0.027	0.038
team[T.Man United] 0.431 0.234	-0.0984	0.170	-0.580	0.562	-
team[T.Newcastle] 1.117 -0.318	-0.7174	0.204	-3.517	0.000	-
team[T.Southampton] 1.081 -0.288	-0.6843	3 0.202	2 -3.384	1 0.001	-
team[T.Stoke] -0.369	-0.7811	0.210	-3.716	0.000	-1.193
0.000					
team[T.Swansea] 1.422 -0.538	-0.9799	0.226	-4.343	0.000	-
team[T.Swansea]	-0.9799 -0.0690	0.226 0.168		0.000	-
team[T.Swansea] 1.422 -0.538 team[T.Tottenham]		0.168			- - -0.881
team[T.Swansea] 1.422 -0.538 team[T.Tottenham] 0.399 0.261 team[T.Watford]	-0.0690	0.168 0.191	-0.410 -2.648	0.682	
team[T.Swansea] 1.422 -0.538  team[T.Tottenham] 0.399 0.261  team[T.Watford] -0.132  team[T.West Brom]	-0.0690 -0.5066	0.168 0.191 0.215	-0.410 -2.648	0.682 0.008 0.000	-
team[T.Swansea] 1.422 -0.538  team[T.Tottenham] 0.399 0.261  team[T.Watford] -0.132  team[T.West Brom] 1.277 -0.435  team[T.West Ham]	-0.0690 -0.5066 -0.8563 -0.4599	0.168 0.191 0.215 0.190	-0.410 -2.648 -3.988 -2.420	0.682 0.008 0.000	-

opponent[T.Burnley] 0.784 0.064	-0.3603	0.216	-1.666	0.096	-
opponent[T.Chelsea] 0.817 0.045	-0.3856	0.220	-1.753	0.080	-
opponent[T.Crystal Palace 0.342 0.422	0.0399	0.195	0.205	0.838	-
opponent[T.Everton] 0.328 0.436	0.0539	0.195	0.276	0.782	-
opponent[T.Huddersfield] 0.299 0.457	0.0789	0.193	0.409	0.683	-
opponent[T.Leicester] 0.305 0.459	0.0770	0.195	0.395	0.693	-
opponent[T.Liverpool] 0.710 0.132	-0.2890	0.215	-1.346	0.178	-
opponent[T.Man City] 1.064 -0.129	-0.5969	0.239	-2.503	0.012	-
opponent[T.Man United] 1.066 -0.142	-0.6041	0.236	-2.564	0.010	-
opponent[T.Newcastle] 0.509 0.285	-0.1119	0.203	-0.552	0.581	-
opponent[T.Southampton] 0.316 0.448	0.0663	0.195	0.341	0.733	-
opponent[T.Stoke] 0.136 0.594	0.2289	0.186	1.229	0.219	-
opponent[T.Swansea] 0.378 0.389	0.0055	0.196	0.028	0.977	-
opponent[T.Tottenham] 0.908 -0.022	-0.4647	0.226	-2.056	0.040	-
opponent[T.Watford] 0.167 0.573	0.2030	0.189	1.075	0.282	-
opponent[T.West Brom] 0.363 0.404	0.0204	0.196	0.104	0.917	-

```
opponent[T.West Ham] 0.2460 0.186 1.320 0.187
        0.611
0.119
home
                   0.2792
                            0.064
                                   4.339
                                            0.000
                                                    0.153
0.405
______
poisson model.predict(pd.DataFrame(data={'team': 'Arsenal',
,"opponent': 'Southampton
(([home':1],index=[1'
2.537017 1
dtype: float64
poisson_model.predict(pd.DataFrame(data={'team': 'Southampton',
,"opponent': 'Arsenal
(([home':0],index=[1'
0.905874
dtype: float64
def simulate match(foot model, homeTeam, awayTeam,
:(max goals=10
home goals avg = foot model.predict(pd.DataFrame(data={'team':
,homeTeam
,{opponent': awayTeam,'home':1'
[index=[1])).values[0
away_goals_avg = foot_model.predict(pd.DataFrame(data={'team':
,awayTeam
,{opponent': homeTeam,'home':0'
[index=[1])).values[0
```

```
team pred = [[poisson.pmf(i, team avg) for i in range(0,
[[max goals+1)] for team avg in [home goals avg, away goals avg
((([return(np.outer(np.array(team_pred[0]), np.array(team_pred[1
ars sou = simulate match(poisson model, 'Arsenal', 'Southampton',
(max_goals=10
([print(ars sou[0:5, 0:5
[0.00089708 0.00396117 0.01311828 0.02896271 0.03197214]]
[0.0022759 0.01004955 0.03328129 0.07347889 0.08111385]
[ 0.002887 0.01274794 0.0422176 0.0932086 0.10289361 ]
[0.00244146 0.01078058 0.03570226 0.07882393 0.08701428]
[[0.00154851 0.00683763 0.02264431 0.04999441 0.05518917]
from matplotlib.colors import ListedColormap
def matrix_gif(matrix, colour_matrix, colour_map, subtitle="",
:(heatmap=False, alpha=0.8
((fig, ax1 = plt.subplots(1, figsize=(5,5))
:if heatmap
(ax1.matshow(matrix, alpha=alpha
:else
(ax1.matshow(colour_matrix, cmap=colour_map, alpha=alpha
(ax1.tick params(axis=u'both', which=u'both',length=0
("=ax1.grid(which='major', axis='both', linestyle
(ax1.set xlabel('Away Team Goals', fontsize=12
(ax1.set_ylabel('Home Team Goals', fontsize=12
('ax1.xaxis.set label position('top
nrows, ncols = matrix.shape
```

```
:(for i in range(nrows
:(for j in range(ncols
[c = matrix[i][j
(ax1.text(j, i, str(round(c,4)), va='center', ha='center', size=13
,'plt.figtext(0.5, 0.05, subtitle, horizontalalignment='center
('fontsize=14, multialignment='left', fontweight='bold
return fig
(['cmap = ListedColormap(['w', '#04f5ff', '#00ff85', '#e90052
matrix = simulate match(poisson model, 'Arsenal', 'Southampton',
(max goals=5
(matn = len(matrix))
,matrix_gif(matrix, matrix, ListedColormap(['w']), heatmap=True
alpha=0.6, subtitle="Match Score Probability
("Matrix").savefig("match_matrix_0.png
()plt.close
for t,(mat,colour,subtitle) in enumerate(zip([np.zeros((matn, matn)),
,(np.tril(np.ones((matn,matn)),-1
(np.triu(np.ones((matn,matn))*2,1), np.diag([3]*matn
np.array([0 if i+j<3 else 1 for i in
,[(range(matn) for j in range(matn)]).reshape(matn,matn
,['w', '#04f5ff', '#00ff85', '#e90052','#EAF205']
Match Score Probability Matrix', 'Home Win', ']
:((["Away Win', 'Draw', 'Over 2.5 goals
matrix gif(matrix, mat, ListedColormap(['w'] + [colour]),
,heatmap=False
alpha=0.6,
((subtitle=subtitle).savefig("match_matrix_{}).png".format(t+1
```

```
:(def poiss_actual_diff(football_url, max_goals
(epl_1718 = pd.read_csv(football_url
[['epl_1718 = epl_1718[['HomeTeam','AwayTeam','FTHG','FTAG
epl_1718 = epl_1718.rename(columns={'FTHG': 'HomeGoals',
    (("FTAG': 'AwayGoals

[(team_pred = [[poisson.pmf(i, team_avg) for i in range(0, max_goals \)

for team_avg in [epl_1718['HomeGoals'].mean(),
    [[()epl_1718['AwayGoals'].mean
\ - (([return np.outer(np.array(team_pred[0]), np.array(team_pred[1 np.array([sum((epl_1718['HomeGoals']==i) &
    (((epl_1718['AwayGoals']==i) for i in range(max_goals) for j in
    (range(max_goals)]).reshape((6,6))/len(epl_1718
```

```
[] = year_arrays
:(for year in range(2005,2018
year arrays.append(poiss actual diff("http://www.football-
)data.co.uk/mmz4281/{}{}/E0.csv".format
((str(year)[-2:], str(year+1)[-2:]), 6
(cmap = sns.diverging palette(10, 133, as cmap=True
((fig, ax = plt.subplots(figsize=(5,5)
:("with sns.axes style("white
ax = sns.heatmap(np.mean(year arrays, axis=0), annot=True,
,fmt='.4f', cmap=cmap, vmin=-0.013, vmax=.013, center=0.00
square=True, linewidths=.5, annot kws={"size": 11},
({cbar kws={"shrink": .8
(ax.tick_params(axis=u'both', which=u'both',length=0
("=ax.grid(which='major', axis='both', linestyle
(ax.set xlabel('Away Team Goals', fontsize=13
(ax.set ylabel('Home Team Goals', fontsize=13
('ax.xaxis.set label position('top
('ax.xaxis.set_ticks_position('top
plt.figtext(0.45, 0.1, 'Actual Proportion - Model Probability',
,'horizontalalignment='center
('fontsize=14, multialignment='left', fontweight='bold
()plt.tight_layout
()plt.show
```

```
:(def rho_correction(x, y, lambda_x, mu_y, rho
:if x==0 and y==0
(return 1- (lambda x * mu y * rho
:elif x==0 and y==1
(return 1 + (lambda x * rho
:elif x==1 and y==0
(return 1 + (mu y * rho
:elif x==1 and y==1
return 1 - rho
:else
return 1.0
:(def dc log like(x, y, alpha x, beta x, alpha y, beta y, rho, gamma
lambda x, mu y = np.exp(alpha x + beta y + gamma),
(np.exp(alpha y + beta x)
+ ((return (np.log(rho_correction(x, y, lambda x, mu y, rho
np.log(poisson.pmf(x, lambda x)) + np.log(poisson.pmf(y, lambda x)) + np.log(poisson
(((mu_y
def solve parameters(dataset, debug = False, init vals=None,
,{options={'disp': True, 'maxiter':100
constraints = [\{'type':'eq', 'fun': lambda x: sum(x[:20])-20\}],
:(**kwargs
(()teams = np.sort(dataset['HomeTeam'].unique
check for no weirdness in dataset #
(()away_teams = np.sort(dataset['AwayTeam'].unique
```

```
:(if not np.array equal(teams, away teams
("raise ValueError("Something's not right
(n teams = len(teams))
:if init vals is None
random initialisation of model parameters #
init vals = np.concatenate((np.random.uniform(0,1,(n teams)), #
attack strength
np.random.uniform(0,-1,(n teams)), # defence
strength
np.array([0, 1.0]) # rho (score correction),
(gamma (home advantage
((
:(def dc_log_like(x, y, alpha_x, beta_x, alpha_y, beta_y, rho, gamma
lambda x, mu y = np.exp(alpha x + beta y + gamma),
(np.exp(alpha y + beta x)
+ ((return (np.log(rho correction(x, y, lambda x, mu y, rho
np.log(poisson.pmf(x, lambda x)) + np.log(poisson.pmf(y, lambda x)) + np.log(poisson
(((mu y
:(def estimate paramters(params
(([score coefs = dict(zip(teams, params[:n teams
(([(defend_coefs = dict(zip(teams, params[n_teams:(2*n_teams
[:rho, gamma = params[-2]
log like = [dc log like(row.HomeGoals, row.AwayGoals,
,[score coefs[row.HomeTeam], defend coefs[row.HomeTeam]
score coefs[row.AwayTeam],
defend coefs[row.AwayTeam], rho, gamma) for row in
[()dataset.itertuples
(return -sum(log like
```

```
opt output = minimize(estimate paramters, init vals,
(options=options, constraints = constraints, **kwargs
:if debug
sort of hacky way to investigate the output of the optimisation #
process
return opt output
:else
+ [return dict(zip(["attack_"+team for team in teams
+ [defence_"+team for team in teams"]
,['rho', 'home adv']
((opt_output.x
(params = solve_parameters(epl_1718
C:\Users\DIRDIRI\AppData\Local\Continuum\anaconda3\lib\site-
packages\ipykernel launcher.py:18: RuntimeWarning: divide by zero
encountered in log
C:\Users\DIRDIRI\AppData\Local\Continuum\anaconda3\lib\site-
packages\ipykernel launcher.py:18: RuntimeWarning: invalid value
encountered in log
(Optimization terminated successfully. (Exit mode 0
Current function value: 1018.2426976476618
Iterations: 56
Function evaluations: 2550
Gradient evaluations: 56
params
,attack Arsenal': 1.4648142298480571'}
attack Bournemouth': 0.93113641857197105'
```

```
,attack Brighton': 0.70694340992857407'
```

,attack\_Burnley': 0.70605333209736354'

,attack Chelsea': 1.278105977670855'

,attack\_Crystal Palace': 0.94096897228372833'

,attack Everton': 0.94842474860340364'

,attack Huddersfield': 0.53412121559506731'

,attack Leicester': 1.1329459072887267'

,attack\_Liverpool': 1.5514657210419127'

,attack\_Man City': 1.8072646940822659'

,attack\_Man United': 1.3577410549133089'

,attack\_Newcastle': 0.70833924593956421'

,attack\_Southampton': 0.79261324548106593'

,attack Stoke': 0.69025628983131071'

,attack Swansea': 0.47027703829482476'

,attack Tottenham': 1.391314706341928'

,attack\_Watford': 0.95342929365886531'

,attack West Brom': 0.61172499584256623'

.attack West Ham': 1.0220595026846429'

,defence Arsenal': -0.87841529803336293'

,defence\_Bournemouth': -0.75049065773546941'

,defence\_Brighton': -0.91734945989907735'

,defence Burnley': -1.2558876859373314'

,defence Chelsea': -1.2759760496035195'

,defence Crystal Palace': -0.83340671172432856'

,defence Everton': -0.82799108800611176'

,defence Huddersfield': -0.80377021526267012'

,defence Leicester': -0.796836258660654'

,defence Liverpool': -1.1534397907896747'

,defence\_Man City': -1.4807174326196608'

,defence\_Man United': -1.4878945411081537'

,defence Newcastle': -1.012445183194113'

,defence Southampton': -0.81379964067196986'

,defence\_Stoke': -0.65840192921987362'

,defence\_Swansea': -0.88267148954438734'

,defence\_Tottenham': -1.3648071533103838'

,defence\_Watford': -0.67810524317000365'

,defence West Brom': -0.8765267745724804'

,defence\_West Ham': -0.63415373318843293'

,home adv': 0.2860381283717664'

{rho': -0.14562632628323352'

(params = solve parameters(epl 1718

C:\Users\DIRDIRI\AppData\Local\Continuum\anaconda3\lib\site-packages\ipykernel\_launcher.py:18: RuntimeWarning: divide by zero encountered in log

C:\Users\DIRDIRI\AppData\Local\Continuum\anaconda3\lib\site-packages\ipykernel\_launcher.py:18: RuntimeWarning: invalid value encountered in log

(Optimization terminated successfully. (Exit mode 0

Current function value: 1018.2426976767588

Iterations: 55

Function evaluations: 2502

Gradient evaluations: 55

#### params

,attack\_Arsenal': 1.4648142298480571'}

,attack\_Bournemouth': 0.93113641857197105'

,attack Brighton': 0.70694340992857407'

,attack Burnley': 0.70605333209736354'

,attack Chelsea': 1.278105977670855'

,attack\_Crystal Palace': 0.94096897228372833'

,attack Everton': 0.94842474860340364'

,attack\_Huddersfield': 0.53412121559506731'

,attack Leicester': 1.1329459072887267'

,attack\_Liverpool': 1.5514657210419127'

,attack\_Man City': 1.8072646940822659'

,attack Man United': 1.3577410549133089'

,attack Newcastle': 0.70833924593956421'

.attack Southampton': 0.79261324548106593'

,attack\_Stoke': 0.69025628983131071'

.attack Swansea': 0.47027703829482476'

,attack Tottenham': 1.391314706341928'

,attack\_Watford': 0.95342929365886531'

,attack West Brom': 0.61172499584256623'

,attack West Ham': 1.0220595026846429'

,defence\_Arsenal': -0.87841529803336293'

,defence\_Bournemouth': -0.75049065773546941'

,defence Brighton': -0.91734945989907735'

,defence Burnley': -1.2558876859373314'

```
,defence Chelsea': -1.2759760496035195'
,defence Crystal Palace': -0.83340671172432856'
,defence_Everton': -0.82799108800611176'
,defence Huddersfield': -0.80377021526267012'
,defence Leicester': -0.796836258660654'
,defence Liverpool': -1.1534397907896747'
,defence Man City': -1.4807174326196608'
,defence Man United': -1.4878945411081537'
,defence Newcastle': -1.012445183194113'
,defence Southampton': -0.81379964067196986'
,defence Stoke': -0.65840192921987362'
,defence Swansea': -0.88267148954438734'
,defence Tottenham': -1.3648071533103838'
,defence Watford': -0.67810524317000365'
,defence West Brom': -0.8765267745724804'
,defence West Ham': -0.63415373318843293'
,home_adv': 0.2860381283717664'
{rho': -0.14562632628323352'
:(def calc means(param dict, homeTeam, awayTeam
return [np.exp(param_dict['attack '+homeTeam] +
,(['param dict['defence '+awayTeam] + param dict['home adv
np.exp(param_dict['defence_'+homeTeam] +
[([param dict['attack '+awayTeam
def dixon coles simulate match(params dict, homeTeam, awayTeam,
:(max goals=10
```

```
(team avgs = calc means(params dict, homeTeam, awayTeam
team pred = [[poisson.pmf(i, team avg) for i in range(0,
[max_goals+1)] for team_avg in team_avgs
output matrix = np.outer(np.array(team pred[0]),
(([np.array(team_pred[1
correction_matrix = np.array([[rho_correction(home goals,
,[away_goals, team avgs[0
team avgs[1], params['rho']) for
[(away goals in range(2
([(for home goals in range(2
output matrix[:2,:2] = output matrix[:2,:2] * correction matrix
return output matrix
ars sou dc = dixon coles simulate match(params, 'Arsenal',
('Southampton', max goals=10
[Simple Poisson, Dixon-Coles] #
("print("Arsenal Win
print('; '.join("{0}: {1:.5f}".format(model, prob) for model,prob in
zip(["Basic Poisson", "Dixon-Coles"], list(map(lambda
((((([x:np.sum(np.tril(x, -1)), [ars_sou, ars_sou_dc
("print("Southampton Win
print('; '.join("{0}: {1:.5f}".format(model, prob) for model,prob in
zip(["Basic Poisson", "Dixon-Coles"], list(map(lambda
((((([x:np.sum(np.triu(x, 1)), [ars_sou, ars_sou_dc
("print("Draw
print('; '.join("{0}: {1:.5f}".format(model, prob) for model,prob in
zip(["Basic Poisson", "Dixon-Coles"], list(map(lambda
((((([x:np.sum(np.diag(x)), [ars sou, ars sou dc
```

## **Arsenal Win** Basic Poisson: 0.72700: Dixon-Coles: 0.71626 Southampton Win Basic Poisson: 0.11278; Dixon-Coles: 0.10287 Draw Basic Poisson: 0.16015; Dixon-Coles: 0.18079 (cmap = sns.diverging palette(10, 133, as cmap=True ((fig. ax = plt.subplots(figsize=(5,5) :("with sns.axes style("white ax = sns.heatmap(simulate match(poisson model, 'Arsenal', \- ('Southampton', max goals=5 dixon coles simulate match(params, 'Arsenal', ,('Southampton', max\_goals=5 annot=True, fmt='.4f', cmap=cmap, vmin=-0.013, ,vmax=.013, center=0.00 square=True, linewidths=.5, annot kws={"size": 11}, ({cbar kws={"shrink": .8 (ax.tick params(axis=u'both', which=u'both',length=0 ("=ax.grid(which='major', axis='both', linestyle (ax.set xlabel('Away Team Goals', fontsize=13 (ax.set ylabel('Home Team Goals', fontsize=13 ('ax.xaxis.set label position('top ('ax.xaxis.set\_ticks\_position('top

plt.figtext(0.45, 0.07, 'BP Probs - DC Probs \nArsenal v

,'Southampton', horizontalalignment='center

```
('fontsize=14, multialignment='left', fontweight='bold
()plt.tight layout
()plt.show
((fig,(ax1,ax2) = plt.subplots(2, 1, figsize=(10,5)))
ax1.plot(range(1000), [0 if y > 600 else 1 for y in range(1000)],
("=label='Component 1', color='#38003c', marker
ax2.plot(range(1000), np.exp([y*-0.005 for y in range(1000)]),
("=label='Component 1', color='#07F2F2', marker
ax2.plot(range(1000), np.exp([y*-0.003 for y in range(1000)]),
("=label='Component 1', color='#05F26C', marker
ax2.plot(range(1000), np.exp([y*-0.001 for y in range(1000)]),
("=label='Component 1', color='#e90052', marker
([ax1.set ylim([-0.05, 1.05]
([ax2.set ylim([-0.05,1.05
([ax1.set_xlim([-0.5,1000]
([ax2.set xlim([-0.5,1000]
([])ax1.set_xticklabels
(ax2.xaxis.set tick params(labelsize=12
(ax1.yaxis.set tick params(labelsize=12
(ax2.yaxis.set tick params(labelsize=12
ax1.set title("Time Decay Weighting
('Functions", size=14, fontweight='bold
(ax2.set_xlabel("Number of Days Ago",size=13
```

```
(ax1.set ylabel("\phi(t)",size=13
(ax2.set_ylabel("\phi(t)",size=13)
ax1.text(830, 0.5, '1
                        $t \leq \mathregular{t_0}$\n0
                                                          $t >
,'${\mathregular{t_0}
,'verticalalignment='bottom', horizontalalignment='left
(color='black', fontsize=15
,'}' ,ax1.text(800, 0.5
,'verticalalignment='bottom', horizontalalignment='left
(color='black', fontsize=44
' = (ax1.text(730, 0.62, '\phi(t)))
,'verticalalignment='bottom', horizontalalignment='left
(color='black', fontsize=15
(ax2.text(730, 0.62, '\phi(t) = exp(-\xi t))
,'verticalalignment='bottom', horizontalalignment='left
(color='black', fontsize=15
,'ax2.text(250, 0.8, '\xi = 0.001
,'verticalalignment='bottom', horizontalalignment='left
(color='#e90052', fontsize=15
'ax2.text(250, 0.5, '\xi = 0.003
,'verticalalignment='bottom', horizontalalignment='left
(color='#05F26C', fontsize=15
'ax2.text(250, 0.0, '\xi = 0.005
,'verticalalignment='bottom', horizontalalignment='left
(color='#07F2F2', fontsize=15
()plt.tight_layout
()plt.show
```

```
def dc_log_like_decay(x, y, alpha_x, beta_x, alpha_y, beta_y, rho,
:(gamma, t, xi=0
lambda x, mu y = np.exp(alpha x + beta y + gamma),
(np.exp(alpha y + beta x)
return np.exp(-xi*t) * (np.log(rho correction(x, y, lambda x, mu y,
+ ((rho
np.log(poisson.pmf(x, lambda x)) +
(((np.log(poisson.pmf(y, mu_y
epl 1718 = pd.read csv("http://www.football-
("data.co.uk/mmz4281/1718/E0.csv
epl 1718['Date'] = pd.to datetime(epl 1718['Date'],
('format='%d/%m/%y
epl_1718['time_diff'] = (max(epl_1718['Date']) - epl_1718['Date']).dt.days
epl 1718 = epl 1718[['HomeTeam','AwayTeam','FTHG','FTAG', 'FTR',
[["time_diff
epl 1718 = epl 1718.rename(columns={'FTHG': 'HomeGoals', 'FTAG':
({"AwayGoals
()epl 1718.head
time_diff FTR AwayGoals HomeGoals AwayTeam HomeTeam
275
       н
              3
                    4
                        Leicester
                                     Arsenal
                                                 0
274
              2
                        Man City
                                     Brighton
                                                 1
        Α
                    0
274
              3
                    2
                         Burnley
                                     Chelsea
                                                 2
        Α
274
              3
                           Huddersfield
                                           Crystal Palace
        Α
                    0
                                                             3
274
       Н
                    1 Stoke
                               Everton
                                           4
              0
```

```
def solve parameters decay(dataset, xi=0.001, debug = False,
,{init_vals=None, options={'disp': True, 'maxiter':100
constraints = [\{'type':'eq', 'fun': lambda x: sum(x[:20])-20\}],
:(**kwargs
(()teams = np.sort(dataset['HomeTeam'].unique
check for no weirdness in dataset #
(()away teams = np.sort(dataset['AwayTeam'].unique
:(if not np.array_equal(teams, away_teams
("raise ValueError("something not right
(n teams = len(teams
:if init vals is None
random initialisation of model parameters #
init_vals = np.concatenate((np.random.uniform(0,1,(n_teams)), #
attack strength
np.random.uniform(0,-1,(n teams)), # defence
strength
np.array([0,1.0]) # rho (score correction),
(gamma (home advantage
((
def dc log like decay(x, y, alpha x, beta x, alpha y, beta y, rho,
:(gamma, t, xi=xi
lambda_x, mu_y = np.exp(alpha_x + beta_y + gamma),
(np.exp(alpha y + beta x)
return np.exp(-xi*t) * (np.log(rho_correction(x, y, lambda_x, mu y,
+ ((rho
np.log(poisson.pmf(x, lambda x)) +
(((np.log(poisson.pmf(y, mu_y
```

```
:(def estimate paramters(params
(([score coefs = dict(zip(teams, params[:n teams
(([(defend_coefs = dict(zip(teams, params[n_teams:(2*n_teams
[:rho, gamma = params[-2
log like = [dc log like decay(row.HomeGoals, row.AwayGoals,
,[score coefs[row.HomeTeam], defend coefs[row.HomeTeam]
score_coefs[row.AwayTeam],
,[defend coefs[row.AwayTeam
rho, gamma, row.time diff, xi=xi) for row in
[()dataset.itertuples
(return -sum(log like
opt_output = minimize(estimate_paramters, init_vals,
(options=options, constraints = constraints
:if debug
sort of hacky way to investigate the output of the optimisation #
process
return opt output
:else
+ [return dict(zip(["attack_"+team for team in teams
+ [defence "+team for team in teams"]
,['rho', 'home adv']
((opt_output.x
(params xi= solve parameters decay(epl 1718, xi=0.0018
C:\Users\DIRDIRI\AppData\Local\Continuum\anaconda3\lib\site-
packages\ipykernel launcher.py:19: RuntimeWarning: divide by zero
encountered in log
```

C:\Users\DIRDIRI\AppData\Local\Continuum\anaconda3\lib\site-packages\ipykernel\_launcher.py:19: RuntimeWarning: invalid value encountered in log

(Optimization terminated successfully. (Exit mode 0

Current function value: 832.6598925477288

Iterations: 43

Function evaluations: 1956

Gradient evaluations: 43

```
params_xi
```

,attack\_Arsenal': 1.4593651460722956'}

,attack\_Bournemouth': 0.98551283613765683'

,attack\_Brighton': 0.69928781168010956'

,attack\_Burnley': 0.70431831190395433'

,attack Chelsea': 1.2374364621077387'

attack Crystal Palace': 1.0097589903242024'

,attack Everton': 0.94292095612817972'

,attack Huddersfield': 0.46239735411738703'

,attack Leicester': 1.1875029236609327'

,attack\_Liverpool': 1.5541275187521442'

,attack\_Man City': 1.7732021459388028'

attack Man United': 1.2929629529047928'

,attack Newcastle': 0.78053285033641984'

,attack\_Southampton': 0.77003553613634246'

,attack Stoke': 0.70057070734386662'

,attack Swansea': 0.46822884080600669'

```
,attack_Tottenham': 1.4286245703681737'
```

,attack Watford': 0.88735796685491264'

,attack West Brom': 0.59796943780667744'

,attack West Ham': 1.0578866806194034'

,defence Arsenal': -0.90350688807197488'

,defence Bournemouth': -0.74359641563485857'

,defence\_Brighton': -0.88576784878115034'

,defence\_Burnley': -1.1823770286264468'

,defence Chelsea': -1.1924572572147196'

,defence\_Crystal Palace': -0.87595090531915076'

,defence Everton': -0.82109862955384794'

,defence\_Huddersfield': -0.84872681894152968'

,defence Leicester': -0.73455540299397404'

,defence Liverpool': -1.2116005980564268'

,defence Man City': -1.5083062975356949'

.defence Man United': -1.5143203684750592'

,defence\_Newcastle': -1.0661409709636933'

,defence Southampton': -0.85568382082476147'

,defence Stoke': -0.6876953305361172'

,defence\_Swansea': -0.85747531921329301'

,defence\_Tottenham': -1.2568197505130652'

,defence Watford': -0.72640660682168368'

,defence West Brom': -0.87337838654160449'

,defence\_West Ham': -0.66515172274413292'

,home adv': 0.30317368503677694'

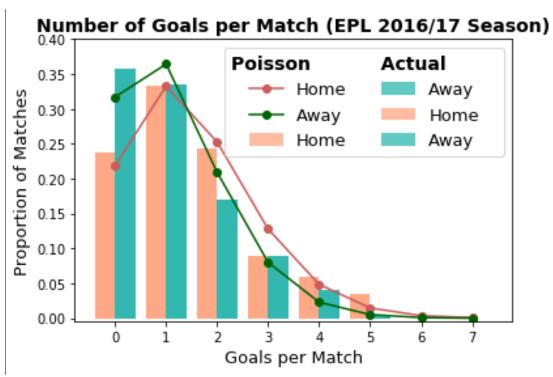
{rho': -0.13185247769791564'

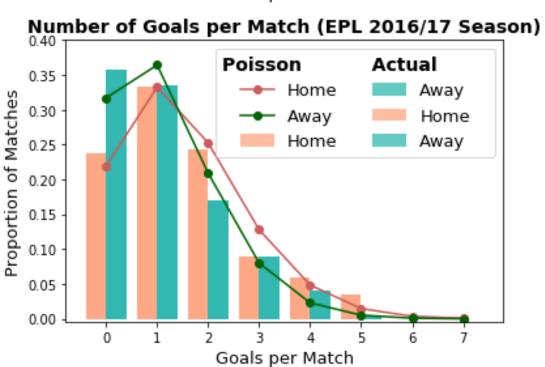
```
xi vals = [0.0, 0.0002, 0.0004, 0.0006, 0.0008, 0.001, 0.0012, 0.0014, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.0018, 0.001
,0.0016, 0.0018
[0.005,0.0045,0.004,0.0035,0.0035,0.003,0.0025,0.002
I pulled the scores from files on my computer that had been generated #
seperately
[] = xi scores#
:for xi in xi vals#
:with open ('find xi {}.txt'.format(str(xi)[2:]), 'rb') as fp
(((xi scores.append(sum(pickle.load(fp
xi scores = [-125.38424297397718, -125.3994150871104, -
,125.41582329299528, -125.43330024318175, -125.45167361727589
- ,125.49165987944551- ,125.47148572476918-
,125.5588181265923- ,125.53570389317336- ,125.51283291929082
- ,125.64545123148538- ,125.58171066742123-
,125.78763678848986- ,125.78763678848986- ,125.71506317675832
- ,125.94721517841089- ,125.8651515986525-
[126.03247674382676
((fig, ax1 = plt.subplots(1, 1, figsize=(10,4))
ax1.plot(xi_vals, xi_scores, label='Component 1', color='#F2055C',
('marker='o
([ax1.set_ylim([-126.20, -125.20
([ax1.set xlim([-0.0001,0.0051
([])ax1.set xticklabels#
```

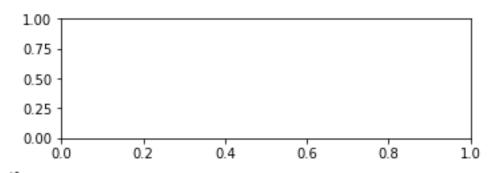
```
(ax1.set ylabel('S(\xi)', fontsize=13
(ax1.set xlabel('ξ', fontsize=13
(ax1.xaxis.set_tick_params(labelsize=12
(ax1.yaxis.set tick params(labelsize=12
ax1.set title("Predictive Profile Log-Likelihood (EPL 2017/18
('Season)", size=14, fontweight='bold
()plt.show
()epl 1318 = pd.DataFrame
:(for year in range(13,18
epl 1318 = pd.concat((epl 1318, pd.read csv("http://www.football-
((((data.co.uk/mmz4281/{}{E0.csv".format(year, year+1
epl 1318['Date'] = pd.to datetime(epl 1318['Date'],
('format='%d/%m/%y
epl_1318['time_diff'] = (max(epl_1318['Date']) - epl_1318['Date']).dt.days
epl_1318 = epl_1318[['HomeTeam','AwayTeam','FTHG','FTAG', 'FTR',
[["time diff
epl_1318 = epl_1318.rename(columns={'FTHG': 'HomeGoals', 'FTAG':
({"AwayGoals
('epl 1318 = epl 1318.dropna(how='all
()epl_1318.head
time diff FTR AwayGoals HomeGoals AwayTeam HomeTeam
                                         Arsenal
1730.0
               3.0
                     1.0 Aston Villa
           Α
                                                     0
1730.0
           н
               0.0
                     1.0 Stoke
                                 Liverpool
1730.0
               2.0
                     2.0
                             Everton
                                        Norwich
                                                     2
           D
               1.0
                     0.0
                             Fulham Sunderland
                                                     3
1730.0
           Α
```

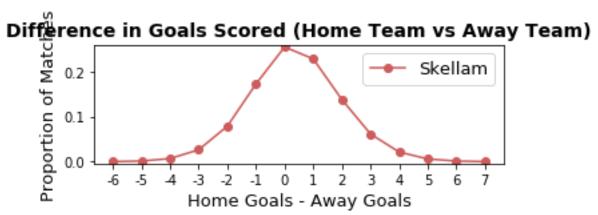
```
xi vals = [0.0, 0.0005, 0.001, 0.0015, 0.002, 0.0025, 0.00275, 0.003, 0.00275, 0.00275, 0.003, 0.00275, 0.00275, 0.00275, 0.003, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275, 0.00275
.0.00325
[0.006, 0.0055, 0.005, 0.0045, 0.00425, 0.004, 0.00375, 0.0035
I pulled the scores from files on my computer that had been generated #
seperately
[] = xi scores#
:for xi in xi vals#
:with open ('find xi 5season {}.txt'.format(str(xi)[2:]), 'rb') as fp #
(((xi scores.append(sum(pickle.load(fp
xi_scores = [-127.64548699733858, -126.88558052909376, -
,126.24253680407995, -125.75657140537645, -125.43198691100818
- ,125.1929173322124- ,125.24473381373896-
,125.15741294807299- ,125.15259048041912- ,125.16314084998176
- ,125.20427802084305- ,125.17611832471187-
,125.39161839279092- ,125.2863163741079- ,125.24143128833828
[125.64269122223465-,125.51241118364625-
((fig, ax1 = plt.subplots(1, 1, figsize=(10,4
ax1.plot(xi vals, xi scores, label='Component 1', color='#F2055C',
('marker='o
([ax1.set ylim([-0.05,1.05#
([ax1.set xlim([-0.0001, 0.0061
([])ax1.set xticklabels#
```

```
(ax1.set_ylabel('S(ξ)', fontsize=13
(ax1.set_xlabel('ξ', fontsize=13
(ax1.xaxis.set_tick_params(labelsize=12
(ax1.yaxis.set_tick_params(labelsize=12
ax1.set_title("Predictive Profile Log-Likelihood (EPL 13/14 - 17/18 ('Seasons)",size=14,fontweight='bold
```

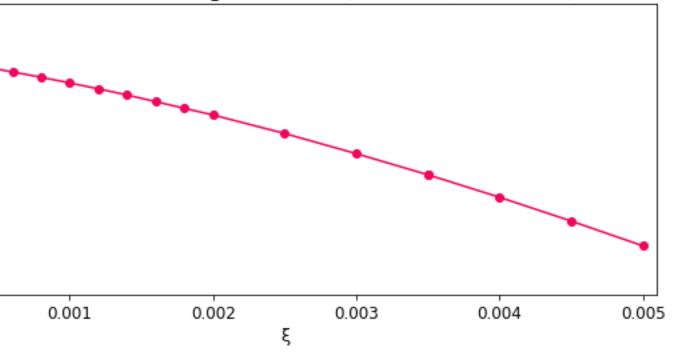








Predictive Profile Log-Likelihood (EPL 2017/18 Season)



### **Time Decay Weighting Functions**

