

Who's Gonna Win The English Premier League?

Pythagorean Expectation

Retrieve The Data (The 20-21 Season)

```
library(worldfootballR)
```

```
EPL_Season <- get_season_team_stats(country="ENG", gender="M", season_end_year="2021", tier="1st", stat="points")
```

```
EPL_Season
```

```
##      Competition_Name Gender Country Season_End_Year      Squad
## 1      Premier League      M      ENG             2021      Arsenal
## 2      Premier League      M      ENG             2021      Aston Villa
## 3      Premier League      M      ENG             2021      Brighton
## 4      Premier League      M      ENG             2021      Burnley
## 5      Premier League      M      ENG             2021      Chelsea
## 6      Premier League      M      ENG             2021      Crystal Palace
## 7      Premier League      M      ENG             2021      Everton
## 8      Premier League      M      ENG             2021      Fulham
## 9      Premier League      M      ENG             2021      Leeds United
## 10     Premier League      M      ENG             2021      Leicester City
## 11     Premier League      M      ENG             2021      Liverpool
## 12     Premier League      M      ENG             2021      Manchester City
## 13     Premier League      M      ENG             2021      Manchester Utd
## 14     Premier League      M      ENG             2021      Newcastle Utd
## 15     Premier League      M      ENG             2021      Sheffield Utd
## 16     Premier League      M      ENG             2021      Southampton
## 17     Premier League      M      ENG             2021      Tottenham
## 18     Premier League      M      ENG             2021      West Brom
## 19     Premier League      M      ENG             2021      West Ham
## 20     Premier League      M      ENG             2021      Wolves
##      Team_or_Opponent Rk MP_Home W_Home D_Home L_Home GF_Home GA_Home GD_Home
## 1      team 8      19      8      4      7      24      21      3
## 2      team 11     19      7      4      8      29      27      2
## 3      team 16     19      4      9      6      22      22      0
## 4      team 17     19      4      6      9      14      27     -13
## 5      team 4      19      9      6      4      31      18     13
## 6      team 14     19      6      5      8      20      32     -12
## 7      team 10     19      6      4      9      24      28     -4
## 8      team 18     19      2      4     13      9      28     -19
## 9      team 9      19      8      5      6      28      21      7
## 10     team 5      19      9      1      9      34      30      4
## 11     team 3      19     10      3      6      29      20      9
```

## 12	team 1	19	13	2	4	43	17	26	
## 13	team 2	19	9	4	6	38	28	10	
## 14	team 12	19	6	5	8	26	33	-7	
## 15	team 20	19	5	1	13	12	27	-15	
## 16	team 15	19	8	3	8	28	25	3	
## 17	team 7	19	10	3	6	35	20	15	
## 18	team 19	19	3	6	10	15	39	-24	
## 19	team 6	19	10	4	5	32	22	10	
## 20	team 13	19	7	4	8	21	25	-4	
##	Pts_Home	Pts_per_G_Home	xG_Home	xGA_Home	xGD_Home	xGD_per_90_Home	MP_Away		
## 1	28	1.47	24.6	22.3	2.3	0.12	19		
## 2	25	1.32	27.7	27.8	0.0	0.00	19		
## 3	21	1.11	31.5	14.6	16.9	0.89	19		
## 4	18	0.95	20.6	27.6	-7.0	-0.37	19		
## 5	33	1.74	35.6	17.2	18.4	0.97	19		
## 6	23	1.21	13.8	26.2	-12.4	-0.65	19		
## 7	22	1.16	25.3	27.2	-1.9	-0.10	19		
## 8	10	0.53	18.8	24.9	-6.1	-0.32	19		
## 9	29	1.53	30.6	26.0	4.6	0.24	19		
## 10	28	1.47	28.9	26.2	2.7	0.14	19		
## 11	33	1.74	35.7	21.7	14.0	0.74	19		
## 12	41	2.16	39.7	15.5	24.2	1.27	19		
## 13	31	1.63	31.7	21.4	10.3	0.54	19		
## 14	23	1.21	24.1	23.1	1.0	0.05	19		
## 15	16	0.84	16.9	30.5	-13.6	-0.72	19		
## 16	27	1.42	20.8	22.4	-1.6	-0.08	19		
## 17	33	1.74	29.2	22.3	6.9	0.36	19		
## 18	15	0.79	17.2	36.1	-18.9	-0.99	19		
## 19	34	1.79	24.2	25.7	-1.5	-0.08	19		
## 20	25	1.32	23.3	20.3	3.0	0.16	19		
##	W_Away	D_Away	L_Away	GF_Away	GA_Away	GD_Away	Pts_Away	Pts_per_G_Away	xG_Away
## 1	10	3	6	31	18	13	33	1.74	28.9
## 2	9	3	7	26	19	7	30	1.58	25.2
## 3	5	5	9	18	24	-6	20	1.05	20.1
## 4	6	3	10	19	28	-9	21	1.11	19.3
## 5	10	4	5	27	18	9	34	1.79	28.4
## 6	6	3	10	21	34	-13	21	1.11	18.6
## 7	11	4	4	23	20	3	37	1.95	21.8
## 8	3	9	7	18	25	-7	18	0.95	22.5
## 9	10	0	9	34	33	1	30	1.58	26.9
## 10	11	5	3	34	20	14	38	2.00	27.1
## 11	10	6	3	39	22	17	36	1.89	36.9
## 12	14	3	2	40	15	25	45	2.37	33.6
## 13	12	7	0	35	16	19	43	2.26	28.6
## 14	6	4	9	20	29	-9	22	1.16	16.9
## 15	2	1	16	8	36	-28	7	0.37	14.5
## 16	4	4	11	19	43	-24	16	0.84	21.6
## 17	8	5	6	33	25	8	29	1.53	25.3
## 18	2	5	12	20	37	-17	11	0.58	16.6
## 19	9	4	6	30	25	5	31	1.63	29.7
## 20	5	5	9	15	27	-12	20	1.05	16.5
##	xGA_Away	xGD_Away	xGD_per_90_Away						
## 1	22.1	6.9	0.36						
## 2	25.1	0.1	0.01						

```
## 3      23.0      -2.9      -0.15
## 4      30.0     -10.7     -0.57
## 5      15.6      12.8       0.67
## 6      31.3     -12.7     -0.67
## 7      24.8      -3.0     -0.16
## 8      28.1      -5.6     -0.29
## 9      36.9     -10.0     -0.53
## 10     21.4       5.7      0.30
## 11     23.6      13.3      0.70
## 12     15.8      17.8      0.93
## 13     20.9       7.7      0.41
## 14     30.9     -14.0     -0.74
## 15     31.9     -17.4     -0.92
## 16     31.8     -10.2     -0.54
## 17     27.3      -1.9     -0.10
## 18     31.7     -15.1     -0.79
## 19     22.6       7.1      0.37
## 20     25.5      -9.0     -0.48
```

Prepare The Data

```
library(tidyverse)
```

```
## -- Attaching packages ----- tidyverse 1.3.1 --
```

```
## v ggplot2 3.3.5    v purrr  0.3.4
## v tibble  3.1.6    v dplyr  1.0.7
## v tidyr   1.1.4    v stringr 1.4.0
## v readr   2.1.0    v forcats 0.5.1
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
```

```
EPL_Season <- EPL_Season %>%
  select(Squad, MP_Home, MP_Away, GF_Home, GF_Away, GA_Home, GA_Away, Pts_Home, Pts_Away)

EPL_Season <- EPL_Season %>%
  mutate(Total_Matches=MP_Home + MP_Away) %>%
  mutate(Total_points=Pts_Home + Pts_Away) %>%
  mutate(Goals_for=GF_Home + GF_Away) %>%
  mutate(Goals_against=GA_Home + GA_Away) %>%
  select(Squad, Total_Matches, Total_points, Goals_for, Goals_against)
```

```
EPL_Season
```

```
##           Squad Total_Matches Total_points Goals_for Goals_against
## 1      Arsenal           38           61       55           39
## 2  Aston Villa           38           55       55           46
## 3    Brighton           38           41       40           46
```

## 4	Burnley	38	39	33	55
## 5	Chelsea	38	67	58	36
## 6	Crystal Palace	38	44	41	66
## 7	Everton	38	59	47	48
## 8	Fulham	38	28	27	53
## 9	Leeds United	38	59	62	54
## 10	Leicester City	38	66	68	50
## 11	Liverpool	38	69	68	42
## 12	Manchester City	38	86	83	32
## 13	Manchester Utd	38	74	73	44
## 14	Newcastle Utd	38	45	46	62
## 15	Sheffield Utd	38	23	20	63
## 16	Southampton	38	43	47	68
## 17	Tottenham	38	62	68	45
## 18	West Brom	38	26	35	76
## 19	West Ham	38	65	62	47
## 20	Wolves	38	45	36	52

Calculate The Pythagorean Expectation

From Soccer Stats

```
EPL_Season$SocStatsPyt <- (1.7 * (EPL_Season$Goals_for - EPL_Season$Goals_against) / (EPL_Season$Goals_for + EPL_Season$Goals_against))
```

From Penalt.y

```
a <- 1.22777 # exponent of goals for
b <- 1.072388 # exponent for goals for in sum of GF and GA
c <- 1.127248 # exponent for GA
n <- 15 # number of matches played

EPL_Season$PenaltyPytExp <- (EPL_Season$Goals_for ** a / (EPL_Season$Goals_for ** b + EPL_Season$Goals_against ** c))
```

From StatsBomb

```
EPL_Season$StatsBombPyt <- 0.677 * (EPL_Season$Goals_for - EPL_Season$Goals_against) + 52.29
```

```
EPL_Season <- EPL_Season %>%
  select(Squad, Total_points, SocStatsPyt, PenaltyPytExp, StatsBombPyt) %>%
  arrange(desc(Total_points))
```

EPL_Season

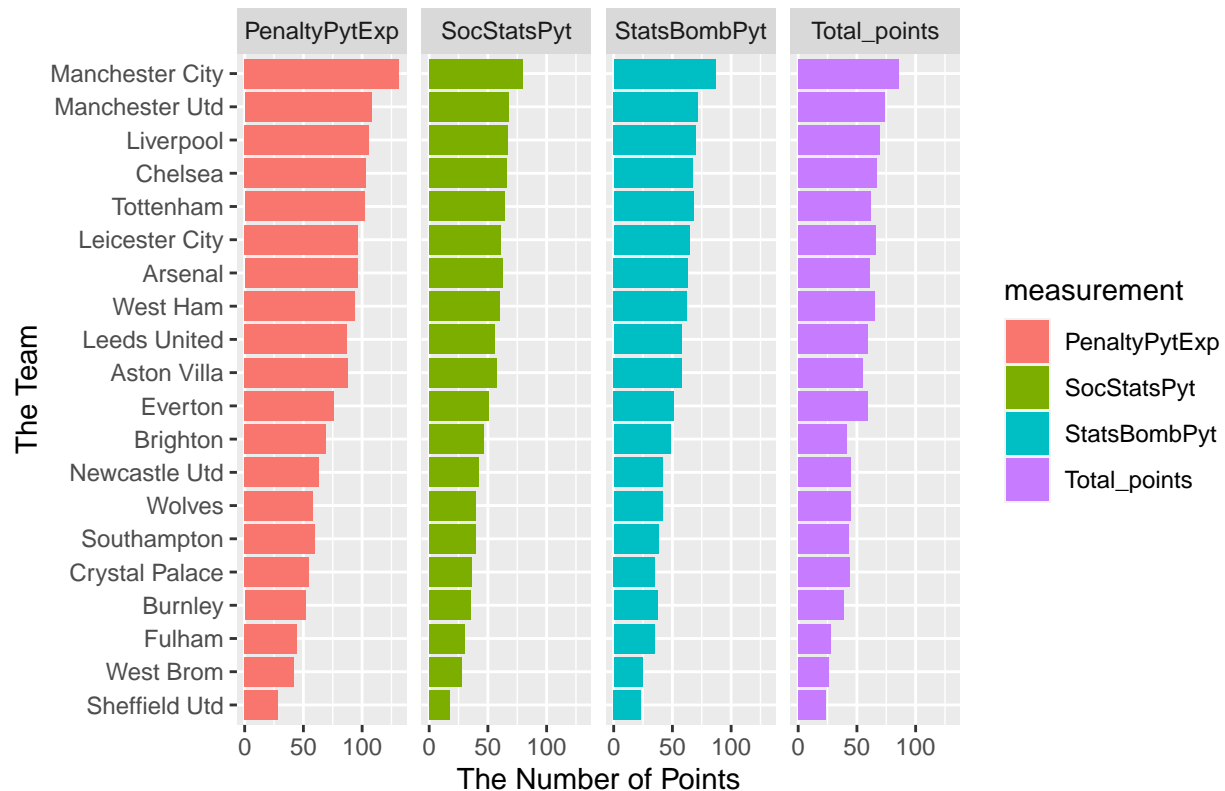
##	Squad	Total_points	SocStatsPyt	PenaltyPytExp	StatsBombPyt
## 1	Manchester City	86	79.94870	131.52220	86.817
## 2	Manchester Utd	74	67.31197	107.88310	71.923
## 3	Liverpool	69	66.56909	105.64617	69.892

## 4	Chelsea	67	66.41915	103.20586	67.184
## 5	Leicester City	66	61.15424	96.76285	64.476
## 6	West Ham	65	60.18991	94.06538	62.445
## 7	Tottenham	62	64.44867	102.15384	67.861
## 8	Arsenal	61	62.29574	95.93903	63.122
## 9	Everton	59	50.62000	76.29461	51.613
## 10	Leeds United	59	55.75517	87.01099	57.706
## 11	Aston Villa	55	57.05644	87.71894	58.383
## 12	Newcastle Utd	45	41.72963	63.14819	41.458
## 13	Wolves	45	39.55455	58.32354	41.458
## 14	Crystal Palace	44	36.20654	54.62733	35.365
## 15	Southampton	43	39.50348	60.14318	38.073
## 16	Brighton	41	46.79302	69.25804	48.228
## 17	Burnley	39	35.15000	51.84565	37.396
## 18	Fulham	28	30.30500	44.49791	34.688
## 19	West Brom	26	27.43874	42.18491	24.533
## 20	Sheffield Utd	23	17.83253	28.56955	23.179

Visualize The Result

```
EPL_Season %>%
  pivot_longer(2:(length(colnames(EPL_Season))), names_to='measurement', values_to='total') %>%
  arrange(desc(total)) %>%
  ggplot(aes(x=total, y=reorder(Squad, total, sum), fill=measurement)) +
  geom_bar(position='stack', stat='identity') +
  facet_grid(~measurement) +
  ggtitle("The Number of Points at the End of EPL Season 20-21") +
  xlab("The Number of Points") +
  ylab("The Team")
```

The Number of Points at the End of EPL Season 20–21



Find The Best Equation

```
EPL_Season %>%
  mutate(ptsSoc = (Total_points - SocStatsPyt)^2,
         ptsPen = (Total_points - PenaltyPytExp)^2,
         ptsStats = (Total_points - StatsBombPyt)^2) %>%
  summarise(meanSoc = mean(ptsSoc) ** 0.5, meanPen = mean(ptsPen) ** 0.5, meanStats = mean(ptsStats) ** 0.5)

##      meanSoc meanPen meanStats
## 1 4.597008 27.4727 4.164511
```

Test it on Current Season of EPL (21-22)

```
EPL_Season <- get_season_team_stats(country="ENG", gender="M", season_end_year="2022", tier="1st", stat="pts")

EPL_Season <- EPL_Season %>%
  select(Squad, MP_Home, MP_Away, GF_Home, GF_Away, GA_Home, GA_Away, Pts_Home, Pts_Away)

EPL_Season <- EPL_Season %>%
  mutate(Total_Matches=MP_Home + MP_Away) %>%
  mutate(Total_points=Pts_Home + Pts_Away) %>%
```

```
mutate(Goals_for=GF_Home + GF_Away) %>%
mutate(Goals_against=GA_Home + GA_Away) %>%
select(Squad, Total_Matches, Total_points, Goals_for, Goals_against)
```

EPL_Season

##	Squad	Total_Matches	Total_points	Goals_for	Goals_against
## 1	Arsenal	12	20	13	17
## 2	Aston Villa	12	13	16	20
## 3	Brentford	12	13	16	17
## 4	Brighton	12	17	12	14
## 5	Burnley	12	9	14	20
## 6	Chelsea	12	29	30	4
## 7	Crystal Palace	12	16	18	17
## 8	Everton	11	15	16	16
## 9	Leeds United	11	11	11	18
## 10	Leicester City	12	15	16	21
## 11	Liverpool	12	25	35	11
## 12	Manchester City	11	23	22	6
## 13	Manchester Utd	12	17	20	21
## 14	Newcastle Utd	12	6	15	27
## 15	Norwich City	12	8	7	27
## 16	Southampton	12	14	11	14
## 17	Tottenham	11	16	9	16
## 18	Watford	12	13	16	20
## 19	West Ham	12	23	23	14
## 20	Wolves	12	19	12	12

```
EPL_Season$StatsBombPyt <- 0.677 * (EPL_Season$Goals_for - EPL_Season$Goals_against) + 52.29
```

```
EPL_Season %>%
  select(Squad, Total_points, StatsBombPyt) %>%
  arrange(desc(StatsBombPyt))
```

##	Squad	Total_points	StatsBombPyt
## 1	Chelsea	29	69.892
## 2	Liverpool	25	68.538
## 3	Manchester City	23	63.122
## 4	West Ham	23	58.383
## 5	Crystal Palace	16	52.967
## 6	Everton	15	52.290
## 7	Wolves	19	52.290
## 8	Brentford	13	51.613
## 9	Manchester Utd	17	51.613
## 10	Brighton	17	50.936
## 11	Southampton	14	50.259
## 12	Arsenal	20	49.582
## 13	Aston Villa	13	49.582
## 14	Watford	13	49.582
## 15	Leicester City	15	48.905
## 16	Burnley	9	48.228
## 17	Leeds United	11	47.551

## 18	Tottenham	16	47.551
## 19	Newcastle Utd	6	44.166
## 20	Norwich City	8	38.750

```
EPL_Season %>%
  select(Squad, Total_points, StatsBombPyt) %>%
  ggplot(aes(x=StatsBombPyt, y=reorder(Squad, StatsBombPyt, sum))) +
  geom_bar(stat='identity', fill='steelblue')
```

