

INFO6205 Ranking System Project Report

Group: NotAtAll

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Introduction

- **Background**

Due to the spread of COVID19, many industries are impacted and suffering great losses, including football. England Premier League (EPL) has announced that all the matches will be suspended. Therefore, it will be interesting to forecast the final result of EPL in this season. Although there will be a little suspense of champion because Liverpool has been leading the scoreboard too much, the suspense of the Champions League place and relegation. Therefore, we will manage to forecast the rest EPL matches and find some relationship between two teams and their match result.



- **Ranking System**

There are many superior performance ranking systems which utilized in sports. For example, ELO ranking system is one of the most commonly used in sports field such as chess, FIFA national team points and celestial system of League of Legends (LOL). As a matter of fact, ELO Ranking System is our first idea to solve the problem. However, after doing research on ELO and EPL, we found that the existing ranking system of EPL (point +3 if won, +1 if draw and +0 if lost) does not fit the ELO ranking system, that means we have to build up another ranking system to calculate the points of each teams. Furthermore, as far as I am concerned, the essence of ELO ranking system is two points. Firstly, using logistic distribution to estimate the expectation win rate point. Secondly, using a mutable K value to make the point

change more reasonable, for example, a better team will not win a lot from a very weak team. It might promote us building up a new ranking system to connect with the existing scoreboard of EPL, which will spend too much extra time. Therefore, we will utilize the match data of EPL from Season 00-01 to Season 19-20 to record the relationship between two sides (Home team and Away Team) and the probability of every results (Home Win, Draw and Away win) to estimate the probability of a match result.

Project Summarize

- **Target**

Our target can be summarized in such points. First of all, we will find out a probability of a result of an upcoming EPL match. For example, Everton will play with Liverpool at home in round 30, we will get the probability of the winning rate of home team (Everton), the draw rate and the winning rate of Away team (Liverpool). Furthermore, we will get the expectational point of both teams. We will 'Kick off' every match and get the expectational point of home and away teams, which will be added in their total points in scoreboard. Therefore, we will get a final expectational scoreboard. Finally, we will simulate every next match randomly, using the probability of each match calculated before. For example, if the match Everton vs Liverpool is calculated that the winning rate of Everton is 22%, draw rate is 16% and winning rate of Liverpool is 62%, we will use `java.util.Random` to simulate the match with such probability and get the result randomly.

- **Basic Frame**

The package `League` is used to simulate the EPL system, which Java classes `Team`, `Match` can create every team and match objects. Package `Rank` is used to build up some basic algorithms and package `Main` is used to simulate the season and calculate the final point expectation.

To calculate the ability gap between, we cannot just use the current point gap because it may occur that two teams have finished different count of rounds. For example, Team A has finished 10 matches and its point is 25. Team B has finished 15 matches while its point is also 25. We cannot say that team A and B are at the same level because they have different match count although they have the same point. Therefore, we have to calculate point per match to judge the ability gap between two teams. As a matter of fact, we will use floor of $5 * \text{point} / \text{finished}$

match to estimate the team point, because we want a beautiful probability distribution of match results of two teams.

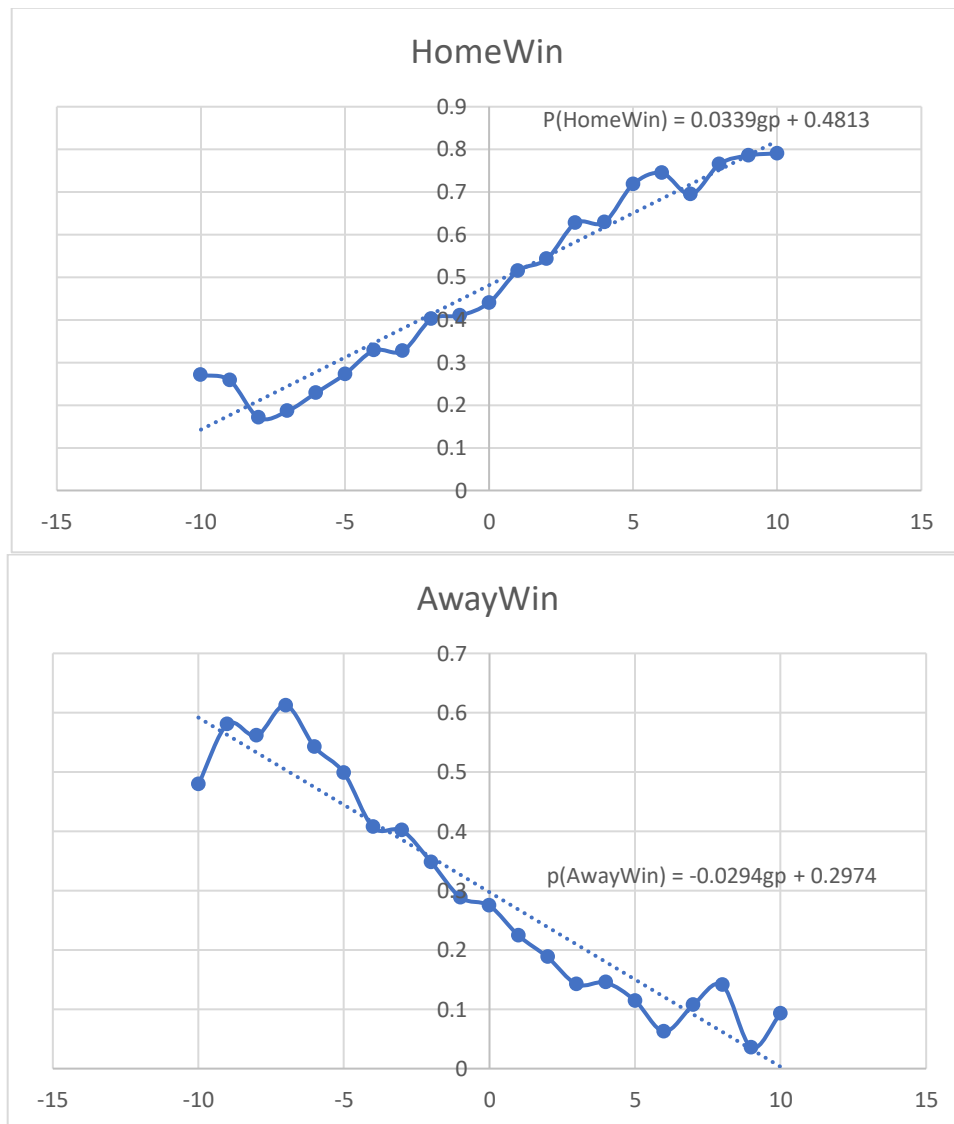
```
public int getPointPer5Match() {
    return finishedMatch == 0 ? 0 : point * 5 / finishedMatch;
}
```

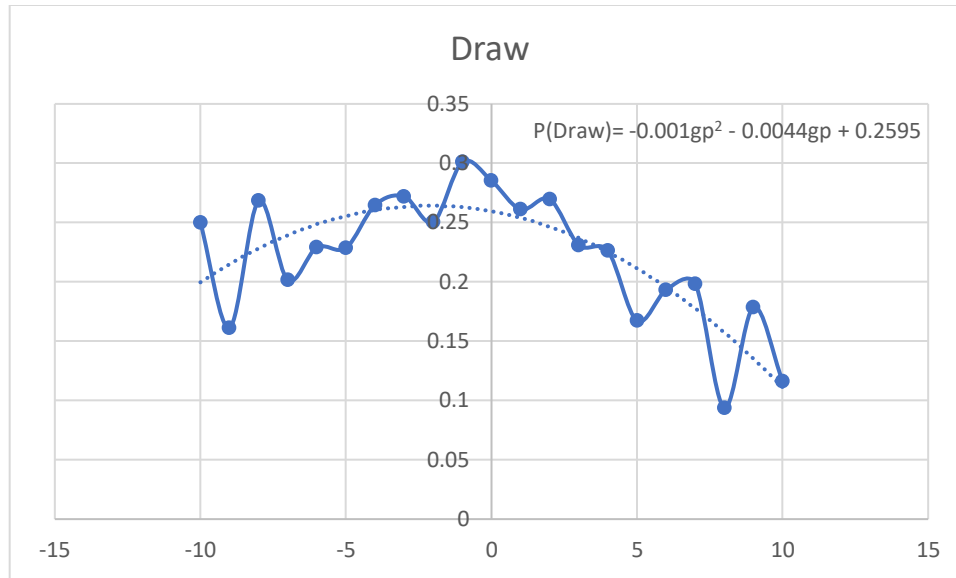
Actually, we can distribute the point gap between two against teams from -15 to 15 according to the equation $\left[\frac{5 * pointOfHome}{finishedMatchOfHome} - \frac{5 * pointOfAway}{finishedMatchOfAway} \right]$. To simplify, we will define $\left[\frac{5 * pointOfHome}{finishedMatchOfHome} - \frac{5 * pointOfAway}{finishedMatchOfAway} \right]$ as *gp*.

The next step of the system is read the data from past 19 seasons. We will ignore the irrelevant information such as index of the gambling and keep only home team, away team and the match result (H, D or A, which represent home win, draw and away win). Because of lack of scoreboard in this data, we have to calculate the point of each team before match began. Afterwards, we will put the point gaps between two against teams as the key and an integer array as the value in a HashMap. The array contains three numbers, which respectively correspond to home won counts, draw counts and away won counts when the point gaps between two against teams is the key integer. The result will be written in relationship.csv file, in which the corresponding Home Win / Draw / Away Win rate will be calculated.

Score Gap: -15 Result: 20 15 11	PointGap	HomeWin	Draw	AwayWin	HomeWinRate	DrawRate	AwayWinRate
Score Gap: -14 Result: 0 2 1	-15	20	15	11	0.434782609	0.326086957	0.239130435
Score Gap: -13 Result: 1 2 3	-14	0	2	1	0	0.666666667	0.333333333
Score Gap: -12 Result: 0 1 2	-13	1	2	3	0.166666667	0.333333333	0.5
Score Gap: -11 Result: 1 1 5	-12	0	1	2	0	0.333333333	0.666666667
Score Gap: -10 Result: 13 12 23	-11	1	1	5	0.142857143	0.142857143	0.714285714
Score Gap: -9 Result: 8 5 18	-10	13	12	23	0.270833333	0.25	0.479166667
Score Gap: -8 Result: 14 22 46	-9	8	5	18	0.258064516	0.161290323	0.580645161
Score Gap: -7 Result: 24 26 79	-8	14	22	46	0.170731707	0.268292683	0.56097561
Score Gap: -6 Result: 52 52 123	-7	24	26	79	0.186046512	0.201550388	0.612403101
Score Gap: -5 Result: 86 72 157	-6	52	52	123	0.22907489	0.22907489	0.54185022
Score Gap: -4 Result: 134 108 166	-5	86	72	157	0.273015873	0.228571429	0.498412698
Score Gap: -3 Result: 167 139 205	-4	134	108	166	0.328431373	0.264705882	0.406862745
Score Gap: -2 Result: 239 149 207	-3	167	139	205	0.326810176	0.272015656	0.401174168
Score Gap: -1 Result: 300 220 211	-2	239	149	207	0.401680672	0.250420168	0.34789916
Score Gap: 0 Result: 491 318 306	-1	300	220	211	0.410396717	0.300957592	0.288645691
Score Gap: 1 Result: 363 184 158	0	491	318	306	0.440358744	0.285201794	0.274439462
Score Gap: 2 Result: 318 158 110	1	363	184	158	0.514893617	0.260992908	0.224113475
Score Gap: 3 Result: 296 109 67	2	318	158	110	0.542662116	0.269624573	0.187713311
Score Gap: 4 Result: 256 92 59	3	296	109	67	0.627118644	0.230932203	0.141949153
Score Gap: 5 Result: 232 54 37	4	256	92	59	0.628992629	0.226044226	0.144963145
Score Gap: 6 Result: 131 34 11	5	232	54	37	0.718266254	0.167182663	0.114551084
Score Gap: 7 Result: 84 24 13	6	131	34	11	0.744318182	0.193181818	0.0625
Score Gap: 8 Result: 49 6 9	7	84	24	13	0.694214876	0.198347107	0.107438017
Score Gap: 9 Result: 22 5 1	8	49	6	9	0.765625	0.09375	0.140625
Score Gap: 10 Result: 34 5 4	9	22	5	1	0.785714286	0.178571429	0.035714286
Score Gap: 11 Result: 4 0 0	10	34	5	4	0.790697674	0.11627907	0.093023256
Score Gap: 12 Result: 0 0 2	11	4	0	0	1	0	0
Score Gap: 13 Result: 5 2 2	12	0	0	2	0	0	1
Score Gap: 14 Result: 14 5 4	13	5	2	2	0.555555556	0.222222222	0.222222222
Score Gap: 15 Result: 14 5 4	15	14	5	4	0.608695652	0.217391304	0.173913043

However, there is some “unreasonable” data. For example, we can notice that when gp is -15, which means that the away team is too much stronger than home team, there are more circumstances that home team defeated away team. However, it’s easy to understand. We can imagine that after Round 1 of a season, team A won the match in Round 1. Therefore, its score is 3, meanwhile, team B lost the match in Round 1 so its score is 0. In this case, if B would encounter A at home in Round 2, we can calculate the gp between B and A is -15. Therefore, when the absolute value of gp is big enough (in this case, larger than 10), it will stand a good chance that it happened in the first few rounds of the season, which will course extreme gp . Therefore, we will choose gp only from -10 to 10, which will be much easy to fit the data.





$$\begin{aligned}
 P(\text{HomeWin}) &= 0.0339gp + 0.4813 \\
 P(\text{AwayWin}) &= -0.0294gp + 0.2974 \\
 P(\text{Draw}) &= -0.001gp^2 - 0.0044gp + 0.2595
 \end{aligned}$$

From the data and the corresponding fitting curve, we can find out that the change tendency of home win rate and away win rate are easy to understand. With the growth of gp , the ability gap between home and away teams will increase. Of course, the home win rate will increase, too. The same reason is suitable for the away win rate. However, although the fitting curve of draw rate is also easy to understand, which is when gp between home and away team tend to 0, the more similar their strength are, the more possible the game result would be draw. However, we can observe that the degree of fitting is not very high. The fitting curve can only represent the general trend of primitive curve. As a matter of fact, this is the problem of different style of different teams. For some teams such as Arsenal, the attack is weak, but the defense is very stable. The ability for Arsenal to draw the game is much higher than some other teams which is more likely to devote their energy to attack. Therefore, we can only get a tendency of the relationship between gp and draw rate. Because it will be much more complete to analyze different teams by their style. We can fix that in this project by calculate the draw rate of a match by the equation: $P(\text{Draw}) = 1 - P(\text{HomeWin}) - P(\text{AwayWin})$. However, in such case, the draw rate will be a lineage tendency, which cannot fit our fitting curve. In order to satisfy the basic equation: $P(\text{HomeWin}) + P(\text{AwayWin}) + P(\text{Draw}) = 1$, we decide to use equation: $P(\text{Draw}) = 1 - P(\text{HomeWin}) - P(\text{AwayWin})$, because we find that the

squared coefficient of the $P(\text{Draw})$ equation is only -0.001, which means the tendency curve approach to a linear line.

$$P(\text{Draw}) = -0.0045gp + 0.2213$$

For every match which has been scheduled but haven't kickoff, we can estimate the possibility of home team win, draw or away team win by utilizing the equation above. According to the possibility, we can calculate the expectational point of each team in each match. The home team's expectational point is $3 * P(\text{HomeWin}) + 1 * P(\text{Draw})$, while the away team's expectational point is $3 * P(\text{AwayWin}) + 1 * P(\text{Draw})$.

$$E(H) = 3 * P(\text{HomeWin}) + 1 * P(\text{Draw})$$

$$E(A) = 3 * P(\text{AwayWin}) + 1 * P(\text{Draw})$$

With the expectational point equation, we can calculate the expectational point in the end of the season by calculate every match's expectational point of both teams.

Team	Exception Points
Liverpool	98.58170000000001
Manchester City	72.8679
Leicester City	66.52529999999999
Chelsea	60.31119999999999
Manchester Unite	57.12580000000001
Wolverhampton Wanderers	55.1078
Sheffield United	55.101400000000005
Tottenham Hotspur	52.81170000000001
Arsenal	51.8143
Burnley	49.51839999999999
Crystal Palace	49.054899999999996
Everton	47.385200000000005
Newcastle Unite	46.316700000000004
Southampton	44.66780000000001
Brighton and Hove Albion	38.9172
WestHam United	37.0234
Watford	36.444700000000005
Bournemouth	36.1756
Aston Villa	34.9855
Norwich City	30.064899999999998

Because the contingency will be excluded by calculating the expectational point, which means the score board will not looks more dispersedly. However, it's enough for us to estimate the general ranking in the end of the incomplete season. There is no doubt that Liverpool will win the champion, Norwich City and Aston Villa will fall on eval days, they cannot stay in EPL next season. Bournemouth is very likely to face promotion and relegation play-off.

Next step we will use the `java.util.Random` to simulate the matches left. Because of uncertainty is getting involved, the scoreboard will look more differently:

Team	Win	Draw	Lose	Points
Liverpool	***** 34	***** 3	***** 1	***** 105
Manchester City	***** 23	***** 6	***** 9	***** 75
Leicester City	***** 21	***** 6	***** 11	***** 69
Manchester Unite	***** 17	***** 10	***** 11	***** 61
Wolverhampton Wanderers	***** 15	***** 14	***** 9	***** 59
Sheffield United	***** 15	***** 12	***** 11	***** 57
Tottenham Hotspur	***** 16	***** 9	***** 13	***** 57
Chelsea	***** 15	***** 11	***** 12	***** 56
Burnley	***** 14	***** 9	***** 15	***** 51
Arsenal	***** 11	***** 15	***** 12	***** 48
Newcastle Unite	***** 12	***** 10	***** 16	***** 46
Crystal Palace	***** 11	***** 12	***** 15	***** 45
Watford	***** 11	***** 11	***** 16	***** 44
Everton	***** 12	***** 7	***** 19	***** 43
Southampton	***** 12	***** 7	***** 19	***** 43
Aston Villa	***** 11	***** 8	***** 19	***** 41
Brighton and Hove Albion	***** 8	***** 16	***** 14	***** 40
Norwich City	***** 10	***** 8	***** 20	***** 38
Bournemouth	***** 9	***** 8	***** 21	***** 35
WestHam United	***** 8	***** 8	***** 22	***** 32

We can see that Liverpool can almost win the champion in every simulation, because its advantage accumulated is very huge. However, when we focus on the relegation zone, which is the last three teams, we can find that it may change because of the uncertainty. Even from the expectational point table, the gap of the expectational points of the relegation zone is only several points, which means that the situation may change a lot if one of the teams won one more match. As the prober says: The football is round, everything could happen. Therefore, the expectational scoreboard can just provide us a possibility of the last result. We cannot deny that the result must be filled with complexity.

Java Implementation

In order to statistic the relationship between *gp* and the match result, we will put every relationship in a HashMap whose key is *gp* and value is an array with counts of three different results:

```
public static void relationStat(HashMap<String, Team> t, HashMap<Integer, int[]> r, String f) {
    for(int i = 1; i < 381; i++) {
        ReadFile.readLine(i, f);
        String[] match = ReadFile.getLineResult();
        String home = match[1];
        String away = match[2];
        String result = match[3];
        int homePiontPer5Match = t.get(home).getPointPer5Match();
        int awayPiontPer5Match = t.get(away).getPointPer5Match();
        if(!r.containsKey(homePiontPer5Match - awayPiontPer5Match)) {
            r.put(homePiontPer5Match - awayPiontPer5Match, new int[3]);
            if(result.equals("H")) r.get(homePiontPer5Match - awayPiontPer5Match)[0]++;
            else if(result.equals("D")) r.get(homePiontPer5Match - awayPiontPer5Match)[1]++;
            else r.get(homePiontPer5Match - awayPiontPer5Match)[2]++;
        }
        else {
            if(result.equals("H")) r.get(homePiontPer5Match - awayPiontPer5Match)[0]++;
            else if(result.equals("D")) r.get(homePiontPer5Match - awayPiontPer5Match)[1]++;
            else r.get(homePiontPer5Match - awayPiontPer5Match)[2]++;
        }
        if(result.equals("H")) {
            t.get(home).win();
            t.get(away).lose();
        }
        else if(result.equals("D")) {
            t.get(home).draw();
            t.get(away).draw();
        }
        else {
            t.get(home).lose();
            t.get(away).win();
        }
    }
}
```

To calculate the probability of three results:

```
public static double HW(Team rHome, Team rAway) {
    return 0.0339 * (rHome.getPointPer5Match() - rAway.getPointPer5Match()) + 0.4813;
}

public static double AW(Team rHome, Team rAway) {
    return -0.0294 * (rHome.getPointPer5Match() - rAway.getPointPer5Match()) + 0.2974;
}

public static double D(Team rHome, Team rAway) {
    return -0.0045 * (rHome.getPointPer5Match() - rAway.getPointPer5Match()) + 0.2273;
}
```


To simulate one match with calculated probability of three results:

```
public static Match kickOff(Match match, HashMap<Integer, int[]> r) {
    double HWR = HW(match.getH(), match.getA());
    double HDR = D(match.getH(), match.getA());
    double HLR = AW(match.getH(), match.getA());
    double rand = random.nextDouble();
    if (rand <= HWR) {
        match.setHWin(true);
        match.getH().win();
        match.getA().lose();
    } else if (rand > HWR && rand <= HDR + HWR) {
        match.setDraw(true);
        match.getA().draw();
        match.getH().draw();
    } else {
        match.setAWin(true);
        match.getA().win();
        match.getH().lose();
    }
    return match;
}
```

To calculate the expectational scoreboard:

```
public static Match exceptionKickOff(Match match, HashMap<Integer, int[]> r) {
    double HWR = HW(match.getH(), match.getA());
    double HDR = D(match.getH(), match.getA());
    double HLR = AW(match.getH(), match.getA());
    match.getH().setExpectationPoint(match.getH().getExpectationPoint() + 3*HWR + HDR);
    match.getA().setExpectationPoint(match.getA().getExpectationPoint() + 3*HLR + HDR);
    return match;
}
```

Limitation of the Algorithm

- **Computational accuracy**

When Java calculating float and double numbers, it will cause precision loss because its own malpractice. For example, when calculating the expectational scoreboard, we can find that several teams who has very similar expectational points will get different results, like these:

Team	Exception Points	Team	Exception Points
Liverpool	98.29910000000001	Liverpool	98.8688
Manchester City	72.9651	Manchester City	72.96600000000001
Leicester City	66.6225	Leicester City	66.52529999999999
Chelsea	60.31119999999999	Chelsea	60.21849999999999
Manchester Unite	56.94040000000001	Manchester Unite	57.412900000000015
Sheffield United	55.194100000000006	Wolverhampton Wanderers	55.205
Wolverhampton Wanderers	55.0151	Sheffield United	55.194100000000006
Tottenham Hotspur	53.0016	Tottenham Hotspur	52.80720000000001
Arsenal	52.101400000000005	Arsenal	52.096900000000005
Burnley	49.898199999999996	Burnley	49.52289999999999
Crystal Palace	49.054899999999996	Crystal Palace	48.67509999999999
Everton	47.482400000000005	Everton	47.385200000000005
Newcastle Unite	46.59930000000001	Newcastle Unite	46.312200000000004
Southampton	44.473400000000005	Southampton	44.385200000000005
Brighton and Hove Albion	38.82	Brighton and Hove Albion	38.6301
WestHam United	36.7363	WestHam United	36.6436
Watford	36.347500000000004	Watford	36.444700000000005
Bournemouth	35.893	Bournemouth	36.272800000000004
Aston Villa	34.98100000000001	Aston Villa	35.07820000000001
Norwich City	30.064899999999998	Norwich City	30.1621

We can find that the position of Sheffield United and Wolverhampton Wanderers are different.

To fix this problem, we can use `java.math.BigDecimal` to get more accurate calculation. However, it is proved that precision loss will still happen when we handle double value with `java.math.BigDecimal`. Therefore, we have to accept the error.

- **Not Suitable for Initial Phase**

Our estimation is based on current scoreboard after round 29, which means we can get a general level of each team. That means, we can get a more reasonable *gp* because the current scoreboard and finished matches can embody the general level of each team. However, try to image that if we want to use this algorithm from the first several rounds of one season. We will find that it is hard for us to get a reasonable result, because there will be more contingency. For example, the defending champion Liverpool lost the round 1 unfortunately while Norwich City won the first match luckily. If these two teams met in round two, we will get an

extreme gp which is 15, which means it's very hard for Liverpool to win. However, the truth tells us that Liverpool would still have a very high probability to win because the ability gap. Therefore, we can conclude that the algorithm is suitable when there are enough training data which can embody the level of each teams.

Conclusion

When facing an upcoming EPL match, we can get such equations to estimate the probability of result:

$$P(HomeWin) = 0.0339 \left[\frac{5 * pointOfHome}{finishedMatchOfHome} - \frac{5 * pointOfAway}{finishedMatchOfAway} \right] + 0.4813$$

$$P(AwayWin) = -0.0294 \left[\frac{5 * pointOfHome}{finishedMatchOfHome} - \frac{5 * pointOfAway}{finishedMatchOfAway} \right] + 0.2974$$

$$P(Draw) = -0.001gp \left[\frac{5 * pointOfHome}{finishedMatchOfHome} - \frac{5 * pointOfAway}{finishedMatchOfAway} \right]^2 - 0.0044 \left[\frac{5 * pointOfHome}{finishedMatchOfHome} - \frac{5 * pointOfAway}{finishedMatchOfAway} \right] + 0.2595$$