

Predictions for the EUROPEAN FOOTBALL CHAMPIONSHIP 2024 based on Statistical Analytical Football Models

AUEB & Trieste Sports Analytics Research Group,

Athens University of Economics and Business and University of Trieste



eams

Scienze Economiche, Aziendali, Matematiche e Statistiche "Bruno de Finetti"

This article was edited and co-authored by Ioannis Ntzoufras, Professor of Statistics at AUEB, and Argyro Damoulaki, PhD Candidate in the same department. The article is based on the analysis of the collaborating team of Trieste (Professors Leonardo Egidi and Nicola Torelli, PhD candidate Roberto Macrì Demartino, and data science master student Giulio Fantuzzi) with the assistance of V. Palaskas (OpenBet, application development) D. Karlis (AUEB Statistics, analysis consultant). The final result is a cooperation between the research teams of the two universities on Sports Analytics.

The group stage ended with several surprises, and we are entering the knockout round. As "classic" favorites to win we would characterize Germany against Georgia, which however is a tough team and will not be an easy opponent. Also of great interest is how England, the Netherlands, Portugal and France will present themselves, who did not let us with good impression in the first round. So, we start with a brief review of our predictions for the group stage and then we present our "predictions" for the round of 16 where we believe that the favorites will start to show their potential and whether they can reach the trophy.

Reminder for friends of Statistics

The use of statistical techniques to predict football matches first appeared in the scientific literature in 1968 with the pioneering scientific publication of Reep & Benjamin. The next real innovations appear in the 80s (with the work of Michael Maher) and the 90s (with the work of Lee in 1997). However, the first important publications in the field, introducing models on which models are based and which we still use today, were the works of Dixon & Coles in 1997 and the bivariate Poisson model of Karlis and Ntzoufra in 2003 (two of the authors of this analysis). These two models formed the basis of modern models for predicting football match outcomes.

In this analysis we use the model of Karlis and Ntzoufras through the package "footbayes" in the statistical programming language R developed by Professor Leonardo Egidi from the University of Trieste with the assistance of Vasilis Palaskas (Analyst at Open Bet and active member of AUEB Sports Analytics Group). The model also includes the estimation of parameters that estimate the performance of each group that change over time. To learn the

model, all international matches of the 2020-2024 period were used. The main explanatory variable is the difference between the two teams in the Coca-Cola/FIFA ranking. The model, first proposed by Karlis & Ntzoufras in 2003, extends the usual two-variate Poisson model. Details of the statistical and machine learning model used can be found at the end of this article.

Review of matchday 3 and group round.

The results of Matchday 3 find our model moving in shallow water as the favorites (Germany, England, France, Netherlands, Belgium and Portugal) failed to win. The biggest surprise was Georgia's victory over Portugal, which had a probability of appearing just 6% based on the model. It should be noted that matchday 3 is traditionally the most difficult in terms of predictions as many teams play expediency football in order to get the draw for the qualification or even some teams that have secured the qualification do not use their key players.

Overall, the model in all group stage matches did relatively well as it caught or visualized the flow of the match at a rate of 56%.

Rival teams (A-B)		Odds			Prevalent	
		Win A team	Draw	Win B team	Result (Probability)	Final Result
Switzerland	Germany	0.283	0.261	0.455	0-1 (0.116)	1 – 1
Scotland	Hungary	0.262	0.288	0.450	0-1 (0.140)	0 – 1
Albania	Spain	0.058	0.170	0.772	0-2 (0.171)	0 – 1
Croatia	Italy	0.308	0.278	0.414	0-1 (0.120)	1 – 1
France	Poland	0.714	0.189	0.097	2-0 (0.138)	1 – 1
Netherlands	Austria	0.482	0.250	0.267	1-0 (0.108)	2 – 3
Denmark	Serbia	0.442	0.289	0.269	1-0 (0.143)	0 - 0
England	Slovenia	0.735	0.190	0.076	1-0 (0.167)	0 - 0
Slovakia	Romania	0.319	0.312	0.369	0-0 (0.161)	1 – 1
Ukraine	Belgium	0.135	0.218	0.647	0-1 (0.128)	0 - 0
Georgia	Portugal	0.060	0.130	0.810	0-3 (0.105)	2 – 0
Czech	Turkey	0.401	0.263	0.336	1-1 (0.109)	1 – 2

Table 1: Table with the odds of the outcome of the matches for Matchday 3 of the European Championship 2024.

Predictions for the Round of 16

From Table 2 with the possible results, the following teams stand out as favorites:

- 1. Spain with a 73% chance of winning over Georgia
- 2. England with a 71% chance of winning against Slovakia
- 3. Portugal with a 65% chance of winning against Slovenia
- 4. The Netherlands with a 62% chance of winning against Romania

Of these four favorites, Spain, based on their performances at the EURO, does indeed seem to be the undisputed favorite. The model seems to overestimate England who have done the job so far, but not in a particularly impressive way. Portugal also seems to be able to beat Slovenia relatively easily based on their pre-match performance against Georgia. The last race put us in some doubts. Finally, the Netherlands has disappointed so far and probably the 62% probability overestimates the current state of the race (remember that the model also uses races from previous races).

Finally, the remaining four races (half of them!) are closer but with a slight lead of one of the two teams. In these matches we consider that the teams are relatively close and may even draw due to tactics and strategy. In particular, we have

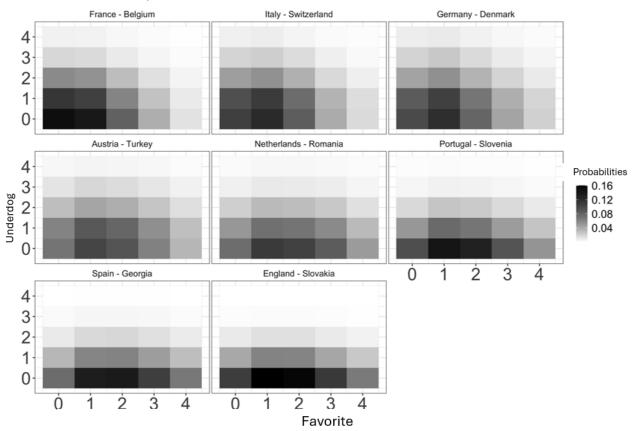
- 1. Austria (55%) beating Turkey (22%) but with an increased probability of a draw (23%)
- 2. Italy (44%) prevailing over Switzerland (29%)
- 3. Germany (45%) prevailing over Denmark (29%)
- 4. France (41%) beat Belgium (29%) but with an increased likelihood of a draw (30%)

From these matches, Switzerland has shown itself very strong as an opponent and it seems based on the matches that it will make it much more difficult for Italy than the model predicts.

Rival teams (A-B)			Prevalent		
		Win A team	Draw	Win B team	Result (Probability)
Switzerland	Italy	0.288	0.273	0.439	0-1 (0.123)
Germany	Denmark	0.448	0.263	0.289	1-0 (0.120)
England	Slovakia	0.714	0.206	0.080	1-0 (0.160)
Spain	Georgia	0.726	0.186	0.088	2-0 (0.139)
France	Belgium	0.406	0.301	0.293	0-0 (0.152)
Portugal	Slovenia	0.653	0.220	0.127	1-0 (0.145)
Romania	Netherland	0.163	0.213	0.624	0-1 (0.109)
Austria	Turkey	0.550	0.231	0.218	1-0 (0.101)

Table 2: Table with the odds of the outcome of the matches for the round of 16 of the European Championship 2024.

Figure 1 gives in more detail the odds for each score for each of the 8 matches of the round of 16.



Posterior match probabilities

Figure 1: Probability Chart of possible scores for the round of 16 of the 2024 European Championship.

Bibliography for reading fans

- Dixon, M.J. and Coles, S.G. (1997), Modelling Association Football Scores and Inefficiencies in the Football Betting Market. *Journal of the Royal Statistical Society: Series C (Applied Statistics)*, **46**, 265-280.
- Karlis, D. and Ntzoufras, I. (2003), Analysis of sports data by using bivariate Poisson models. *Journal of the Royal Statistical Society: Series D (The Statistician)*, **52**, 381-393.
- Lee A.J. (1997). Modeling Scores in the Premier League: Is Manchester United Really the Best? *Chance*, **10**, 15-19.
- Maher, M.J. (1982), Modelling association football scores. *Statistica Neerlandica*, **36**, 109-118.
- Reep, C., & Benjamin, B. (1968). Skill and Chance in Association Football. *Journal of the Royal Statistical Society. Series A (General)*, **131**, 581-585.

The Magic Equations of the statistical model

$$(X_i, Y_i) \sim \begin{cases} (1-p) \operatorname{BP}(x_i, y_i | \lambda_1, \lambda_2, \lambda_3) & \text{if } x \neq y \\ (1-p) \operatorname{BP}(x_i, y_i | \lambda_1, \lambda_2, \lambda_3) + p D(x, \eta) & \text{if } x = y, \end{cases}$$
(1)

$$\begin{aligned} \log(\lambda_{1i}) &= \operatorname{att}_{h_{i,t}} + \operatorname{def}_{a_{i,t}} + \frac{\gamma}{2}(\operatorname{ranking}_{h_{i}} - \operatorname{ranking}_{a_{i}}) \end{aligned} \tag{2} \\ \log(\lambda_{2i}) &= \operatorname{att}_{a_{i,t}} + \operatorname{def}_{h_{i,t}} - \frac{\gamma}{2}(\operatorname{ranking}_{h_{i}} - \operatorname{ranking}_{a_{i}}), \quad i = 1, \dots, n \ (\text{matches}), \end{aligned} \tag{3} \\ \log(\lambda_{3i}) &= \rho, \end{aligned} \tag{4} \\ \operatorname{att}_{k,t} \sim \mathcal{N}(\operatorname{att}_{k,t-1}, \sigma^{2}), \end{aligned} \tag{5} \\ \operatorname{def}_{k,t} \sim \mathcal{N}(\operatorname{def}_{k,t-1}, \sigma^{2}), \end{aligned} \tag{6} \\ \rho, \ \gamma \sim \mathcal{N}(0, 1) \cr p \sim \operatorname{Uniform}(0, 1) \cr \sum_{k=1}^{n_{t}} \operatorname{att}_{k, k} = 0, \ \ k = 1, \dots, n_{t} \ (\text{teams}), \ t = 1, \dots, T \ (\text{times}). \end{aligned} \tag{9} \end{aligned}$$

- *i* is the indicator of the struggle
- X_i and is the number of goals scored between the 1st Y_i and 2nd teams in match i
- h_i and are the 1st a_i and 2nd teams respectively (or home and away teams where applicable) for the *i* match.
- *att_{k,t}* and parameters estimating the offensive and defensive capacity/capability of group k at time *def_{k,t}* t (dynamic parameters that change over time)
- *ranking*_k Coca-Cola FIFA ranking on April 4, 2024 for team k.

A few words about the Authors

AUEB Sports Analytics Group



loannis Ntzoufras is Professor of Statistics and Chairman of the Department of Statistics of the Athens University of Economics and Business. He is a founding member of AUEB Sports Analytics Group along with Dimitris Karlis. He has recognized scientific activity in areas such as Bayesian statistical methodology, computational statistics, biostatistics, psychometrics and sports analytics.



Dimitris Karlis is Professor of Statistics and Deputy Chairman of the Department of Statistics of the Athens University of Economics and Business. He is a founding member of AUEB Sports Analytics Group along with Ioannis Ntzoufras. He has recognized scientific activity in areas such as statistical methodology, computational statistics, biostatistics, and sports analytics.



Argyro Damoulaki is a PhD candidate at the Department of Statistics of AUEB. Her current research focuses on performance evaluation models using Bayesian statistics. He is involved in sports analytics and has been an active member of AUEB Sports Analytics Group since 2022.



Vasilis Palaskas is a Statistical Analyst and Data Scientist at Open Bet. He previously worked as head of analysis and forecasting at Fantasy Sports Interactive (FSI). He has been an active member of the AUEB Sports Analytics research team since 2019 when he graduated from M.Sc. in Statistics of the Athens University of Economics and Business.

The research team of the University of Trieste



Dipartemento d Scienze Economiche, Aziendali, Matematiche e Statistiche "Bruno de Finetti"



Leonardo Egidi is Assistant Professor of Statistics at the University of Trieste in Italy and a member of the AUEB Sports Analytics Group of the Athens University of Economics and Business. He has a PhD in football modeling and analytics and intense research activity in Bayesian Statistical methodology.



Nicola Torelli is Full Professor of Statistics at the University of Trieste. His main research fields of statistical classification, clustering and Bayesian modeling. He is the former president of the Italian Statistical Society.



Roberto Macri Demartino is a PhD candidate at the University of Padua. His main research interests focus on Bayesian statistical methodology with applications in a variety of fields, such as sports analytics.



Giulio Fantuzzi received a degree in Statistics and is now a postgraduate student in Data Science and Artificial Intelligence at the University of Trieste. He is very active in the field of sports modeling mainly using R and Python.

Active Author Collaborations

The three authors (L. Egidi, I. Ntzoufras and D. Karlis) of the article are currently collaborating on the writing of a book on *Football Analytics* in an international scientific house while in the last workshop of the team they gave a seminar course in Football analytics.

The two universities are also preparing a postgraduate program in Sports Analytics with plans to start in October 2025.

L. Egidi and **V. Palaskas** collaborate on the development of the software "footbayes" (library of the statistical programming language R).

L. Egidi, I. Ntzoufras and V. Palaskas collaborate on the writing of a scientific article evaluating players in Volleyball.

V. Palaskas is a graduate of AUEB M.Sc. in Sports Analytics, an active member of AUEB Sports Analytics and has cosupervised a diploma thesis in the framework of AUEB's M.Sc. in Statistics and his previous collaboration with FSI (Fantasy Sports Interactive).

Η Ομάδα AUEB Sports Analytics



AUEB **Sports Analytics Group** was founded in 2015 by professors loannis Ntzoufras and Dimitris Karlis. Its members are important members of the sports analytics community such as Leonardo Egidi (University of Trieste), Ioannis Kosmidis (Warwick), Konstantinos Pelechrinis (Pittsburg), Nial Friel (UCD) and Gianluca Baio (UCL) as well as the former coach of the Greek national volleyball team. Sotiris Drikos and Sacramento Kings scouter Christos Marmarinos. The research team is responsible for the series of annual conferences named AUEB Sports Analytics Workshop (6 in total) and in 2019 organized the

international conference MathSport 2019 with 200 participating scientists from around the world. The team has a number of important scientific publications in the field of sports analytics. Finally, we would like to mention that the team was founded in 2015 due to the visit of Professor Stefan Kesenne (University of Antwerp & Leuven), a great Sports Economist who played an active role in the Bosman case. Stefan Kesenne actively supported the team until 2021 when he suddenly passed away. The existence of AUEB Sports Analytics Group is largely due to the contribution and inspiration that Mr. Kesenne gave us.

Website tou group https://aueb-analytics.wixsite.com/sports

Website του AUEB Sports Analytics Workshop https://aueb-analytics.wixsite.com/saw2024

Listen to the podcast on Basketball Analytics by Ioannis Ntzoufras, Argyro Damoulakis and Christos Marmarinos <u>here</u>.