





PASSING TACTICAL RELEVANCE DURING SOCCER MATCHES DETERMINED BY MACHINE LEARNING

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SPORT AND MATHEMATICS



What is the importance of sport for mathematics?



http://www.rtv.unicamp.br/?video_listing=educacao-fisica-e-matematica



SPORT AND MATHEMATICS





SPORT



















MACHINE LEARNING







SOCCER

- Intermittent team sports.
- Machine Learning techniques used in soccer matches.
- Training to Match Match to Training.



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SOCCER PASSES

- Passes are the most used technical-tactical action during the match and has been considered an important performance indicator.
- Usually, the pass is analyzed based on effectiveness (successful or not successful) limiting the quality of information.





SOCCER PASSES







SOCCER PASSES

- GRAPH VISUAL RHYTHMS (Rodrigues *et al.*, 2019): successful tactical strategies used in soccer matches (passes between players).
- Classifying passes based on most significant concept to performance would allow more relevant analysis about the match.
- Mathematics can help to answer the importance of passes for soccer without human subjectivity





SOCCER PASSES

- Passing TACTICAL RELEVANCE (TR)
- Machine learning to classify a soccer problem (passes complexity - very bad to very good)
- Tactical advantage provided by the passing in a given real match context.







 Propose an automatic pass classification model in soccer matches based on tactical relevance concept, from spatiotemporal data.





- 4 official matches of Professional Brazilian Football Championship 2016.
- Tracking by Dvideo 15 Hz (0.3 m BARROS et al. 2007).
- Smoothing by digital filter *Butterworth* (Matlab[®]2017).
- Scout by DVideo interface.











- Supervised dataset framework:
- Passing predictive variables.
- Passing response variables (labeling process).





Passing predictive variables



- Density
- Nearest opponent
- Outplayed opponent
- Distance, angle
- Opponent btw receiver and target
- Key pass
- Assistance, etc.

33 predictive variables











- Classification process
- Software Python 3.7 scikit-learn v0.20.3 library;
- Random Forest Classifier (RF);
- Split stratified K-fold (n = 3);
- Normalized and balanced accuracy.







RESULTS

Labeling process (experts)

	Very Bad	Bad	Medium	Good	Very Good	Total
Very Bad	7	0	0	0	0	7
Bad	9	46	8	1	0	64
Medium	0	5	262	56	0	323
Good	0	0	5	59	2	66
Very Good	0	0	0	0	5	5

✓ Agreement = 81,5%
✓ *Kappa* = 0,65 (substancial)





RESULTS

• Passing Tactical Relevance:

Normalized accuracy (K-fold) = 0.80 ± 0.01

Balanced accuracy (K-fold) = 0.50 ± 0.04





RESULTS

accuracy = 80.0% (passes classification - 5 classes) - similar Horton & Gudmundsson (2014) accuracy = 85.0% (passes quality classification - 3 classes).







FINAL CONSIDERATIONS

- Passing Tactical Relevance: multivariate and nonlinear problem.
- Predictor variables based on the opinion of experts, and a classification model based on a nonlinear algorithm.







FINAL CONSIDERATIONS

- Test the classifier with three classes.
- Increase number of labeled samples from the more extreme classes.
- Test other classifiers, parameters and performance evaluation of the classifiers.





TAKE HOME MESSAGE

- Practical applications: player ranking, offensive sequence of actions (time series) and training planning.
- Data to Players: how can we do this?
- Quality of Data Relevance of Analysis.
- Talent identification and development.







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