







Move it or lose it: Exploring the relation of defensive disruptiveness and team success.

Kempe, Matthias & Goes, Floris,

University of Groningen, University Medical Center Groningen, Center for Human Movement Sciences, The Netherlands



Contact: m.kempe@umcg.nl





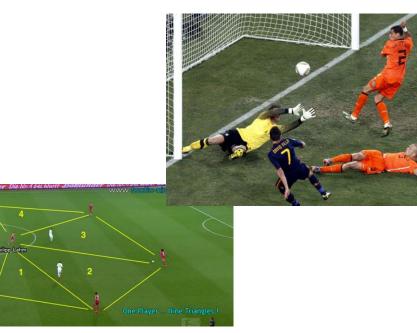




Starting Point: How to measure tactical performance?

State of the art = Notational Analysis





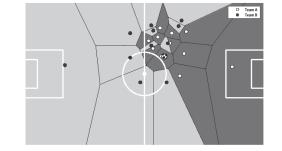






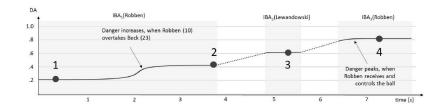


New Wave: Approaches using tracking data



Dangerousity (Link et al, 2016)

Space control (Rein et al, 2017)













The Groningen Approach- Soccer as a dynamic system?!

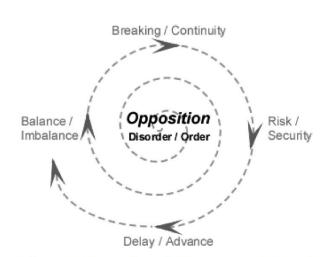


Figure 1. Dialectic aspects of game play brought about by the opposition (adapted from Gréhaigne & Godbout [2012]). Reproduced with permission from Dr Marie-Paule Poggi.



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Length, width and centroid distance as measures of teams tactical performance in youth football Hugo Folgado ^{ad}, Koen A. P. M. Lemmink ^{bc}, Wouter Frencken ^{bc} & Jaime Sampaio ^d

Variability of inter-team distances associated with match events in elite-standard soccer

WOUTER FRENCKEN¹, HARJO DE POEL¹, CHRIS VISSCHER¹, & KOEN LEMMINK^{1,2}

¹UMCG, University of Groningen, Center for Human Movement Sciences, Groningen, Netherlands, and ²Hanze University of Applied Sciences, School of Sport Studies, Groningen, Netherlands

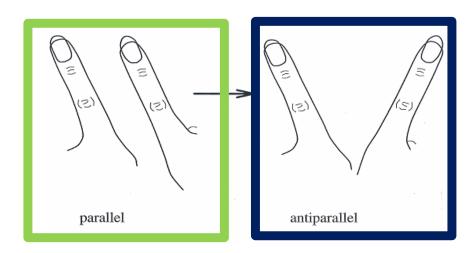


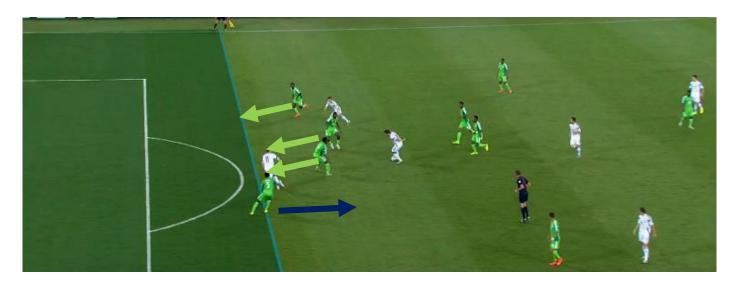






Creating Space = Dynamic System ?











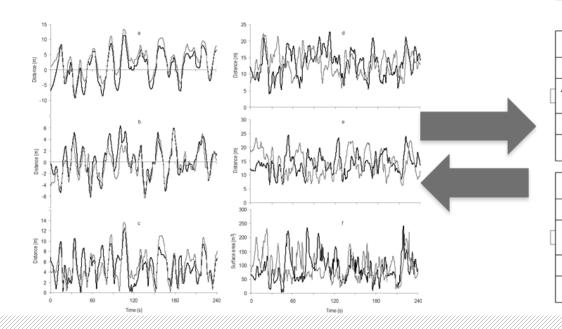


Variability of inter-team distances associated with match events in elite-standard soccer

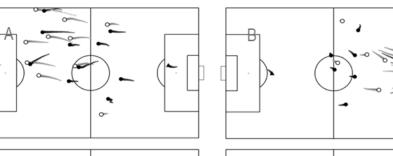
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Variability



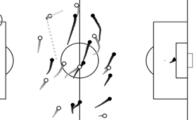
(Critical) events







2











New approach: Create moments of bad synchronization (disruption)

ORIGINAL ARTICLE

Not Every Pass Can Be an Assist: A Data-Driven Model to Measure Pass Effectiveness in Professional Soccer Matches

Floris R. Goes," Matthias Kempe, Laurentius A Meerhoff, and Koen A.P.M. Lemmink $^{\rm 1}$

1. Idea : Player Movement \uparrow = Disruption \uparrow

2. Idea : Change in Structure \uparrow = Disruption \uparrow

3. Idea : Change in Subunits \uparrow = Disruption \uparrow



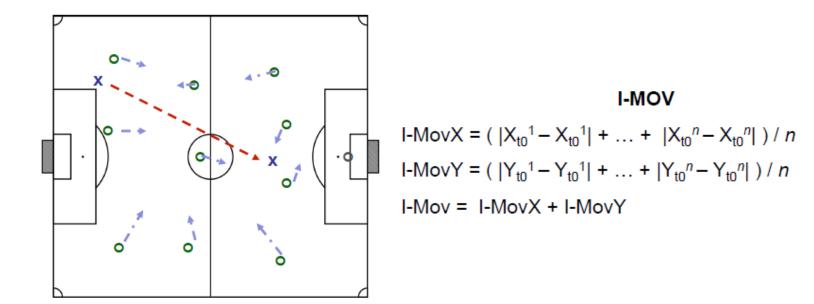






1. Idea

(Indicted) Movement \rightarrow Disruption = Passing Performance











Validating I-Mov



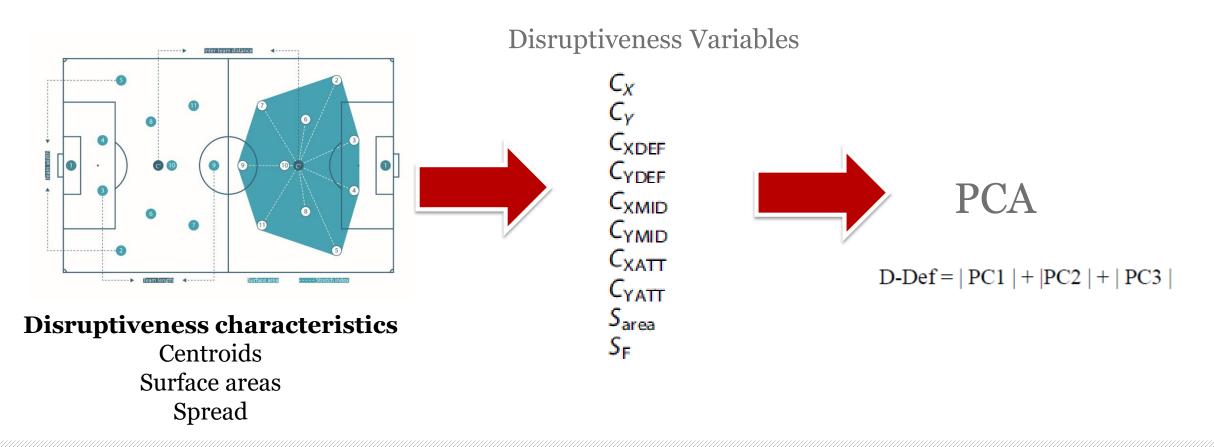








Complex Approach: Idea 2 & 3



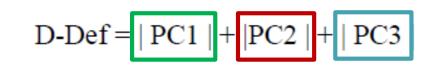


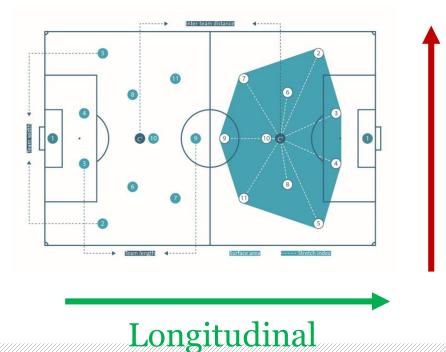






Validating D-Def



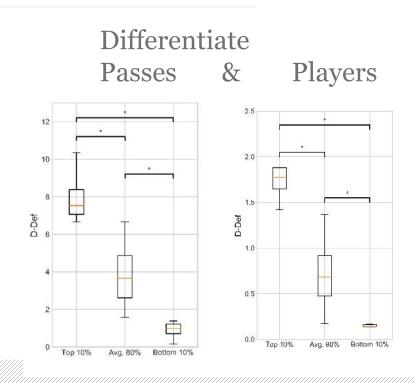


Lateral

ORIGINAL ARTICLE

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Floris R. Goes^{1,*} Matthias Kempe¹, Laurentius A Meerhoff², and Koen A.P.M. Lemmink¹











Issues with D-Def

- > Use of arbitrary 3 second window
- > Use of starting formations for subunit calculation











New Study – fixing Issues + explore relation to team success

- 1. Issue: 3s window \rightarrow normalize in change (m) per second
- 2. Issue: Arbitrary formations for subunit calculation
 - > Spilt possession in attacking or defending
 - → K-Means for 1.half of a game $(n_{clusters} = 3)$ (Bialkowski et al, 2014)
 - > Assign every player to one cluster for every timeframe of the game

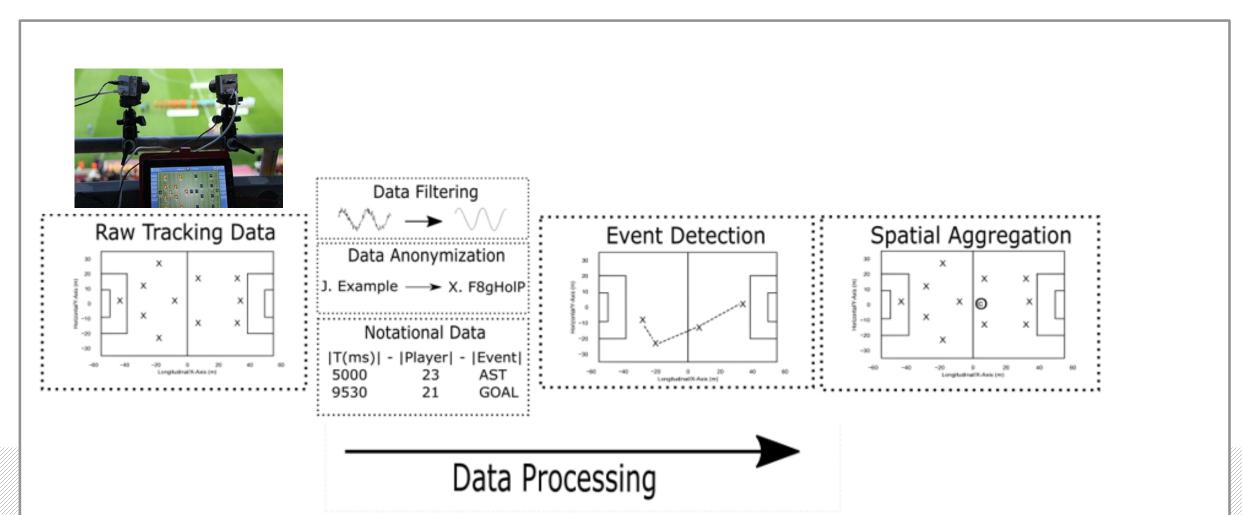








Study design











Results

Table 1 - Descriptive statistics winning and losing teams (*: p = .05 *: p < .05, *: p < .01)

	Wins (N = 89)	Losses (N = 89)	Mean Diff.	Effect Size (Cohen's d)
Individual Movement (I-Mov)				
I-Mov-X (Mean)	$0.866m \pm 0.673m$	$0.515\text{m}\pm0.675\text{m}$	+68.1%	0.52**
I-Mov-Y (Mean)	$0.772m\pm0.600m$	$0.451m\pm0.591m$	+71.2%	0.54**
I-Mov (Mean)	$1.638m\pm1.268m$	$0.966m\pm1.265m$	+69.6%	0.53***
Defensive Disruptiveness (D-Def)				
PC1 (Mean)	0.018 ± 0.015	0.013 ± 0.022	+34.1%	0.24*
PC2 (Mean)	0.010 ± 0.013	0.014 ± 0.033	-23.6%	-0.13
PC3 (Mean)	-0.026 ± 0.022	-0.021 ± 0.022	-25.5%	-0.25*
D-Def (Mean)	0.474 ± 0.048	0.484 ± 0.072	-2.0%	-0.16









Results

> 5-fold cross-validated logistic Regression

Outcome = -0.146 + 0.689 I-Mov _{Mean} + 0.172 PC1_{Mean} - 0.592 PC3_{Mean}

> predicts 69.4 % match outcomes correctly









Discussion

- > Solve the previous issues in our model
- Reconsider the lateral component of D-Def
- > Spatial-temporal variables can predict match outcome
 - So far good predictors for match outcome were goals & shots on goal's (Lago-Penas et al., 2010)
 - Passing parameters were poor game by game predictors (Collet, 2013)









Conclusion

- Key Performance Indicators build on spatial- temporal variables are good predictors for match outcome
- Spatial- temporal variables have the opportunity to model team & individual interactions
- **Stop using:**
 - purely goal related KPI's
 - Event data









Thanks to:









Floris Goes Dr. Rens Meerhoff



Koen Lemmink







Contact: m.kempe@umcg.nl











For comparision of D-Def vs other performance indicators

Wednesday 12:10 Session Inplay Prediction

"Predicting Match outcome in professional Dutch football using tactical performance metrics computed from position tracking data"



Floris Goes