

Visual Soccer Analytics:

Understanding the Characteristics of Collective Team Movement Based on Feature-Driven Analysis and Abstraction

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ISPRS International Journal of Geo-Information, Special Issue Advances in Spatio-Temporal Data Analysis and Mining, 2015

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What: Data

Single Soccer Game

- Mainly geometrical data
- Data every 100 milliseconds
- Manually annotated events (fouls, goals ...)

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Overview

Data 5



<https://www.janetzko.eu/project/soccer/>

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Why

The need of a software

- Increasing demand from clubs
- Now we can
- Video analyst: 3 working days per opponent team
- Current support from system is limited
- Visualisation to not get overwhelmed by data

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Why

Improve previous work

- No (good) automatic identification of situations
 - Need expert verifications
 - Doesn't support domain knowledge
 - 1: classification method but no explanation

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Why

Tasks

- Support experts in exploring characteristics of situations
- Incorporation of meaningful features describing situation
- Visualisation with interactive re-ranking of features and search for similar situations

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How

Workflow

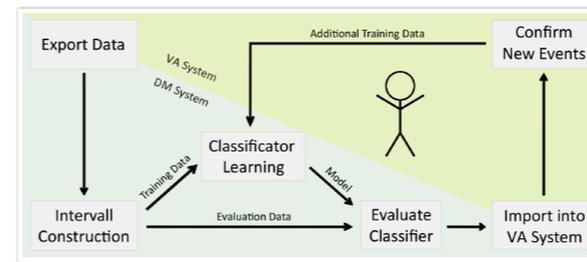


Figure 1. Previous workflow

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How

Workflow

- Intervals: General time interval
- Move: Ball possession
- Event: Foul / goal / ...

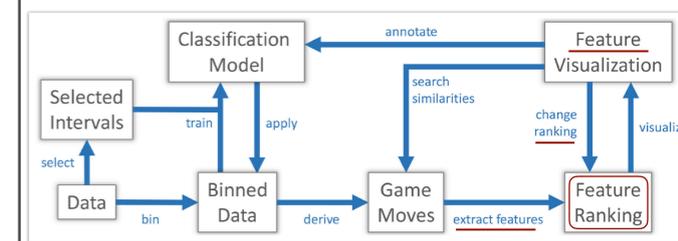
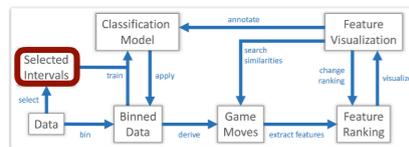


Figure 3. New workflow

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How

Workflow



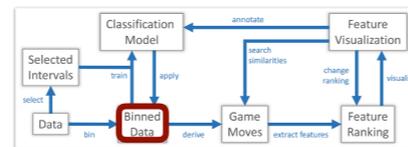
Interval selection:

- Manual or automatic
- Shows data of interest
- Main reason of use

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How

Workflow



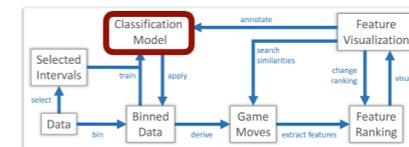
Binning:

- Smooth out noise => better classification
- Less Data
- 100 milliseconds -> 2 seconds time frame

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How

Workflow



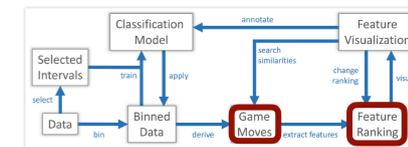
Classification model:

- Compute features of binned data
- 5 classification algorithms:
 - Logistic model trees, Logistic base, Functional trees, decision stump and Support vector machines
- Training set: 33% of intervals
- Returns classified set of 2s intervals

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How

Workflow



Game moves and Feature ranking:

- Derive Game moves from interesting 2s intervals
- Extract interpretable features of each moves
- Relevant if unusual values

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How

Workflow

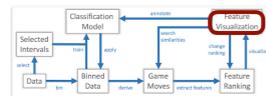


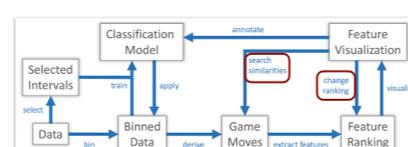
Table 1. Meaningful features

Glyph	Description	Expert Interpretation
	Covered distance of the ball	Build up play or win of the ball
	Number of passes	Team or solo action
	Number of players with touch of the ball	Team or solo action
	Straightness of the ball	Straight direct or on the scout play
	Distance of passes	Short passing game or long passes
	Speed of move	Fast counterattack or careful build up play
	Number of overcome players	Combination play or counterattack

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How

Workflow



Ranking change:

- User can re-ranking features

Similarity search:

- Search similar moves based on events and ranking features

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How

Visual design

Time:

- Navigation and Show events

Move:

- Show moves duration and main feature

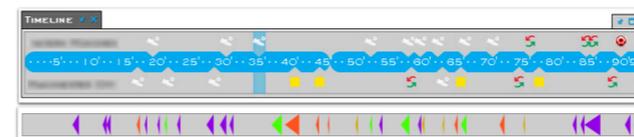


Figure 4. and 5.

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How

Visual design

Move characteristic:

- Shows ranked features
- Connector to see better
- Drag and drop re-ranking

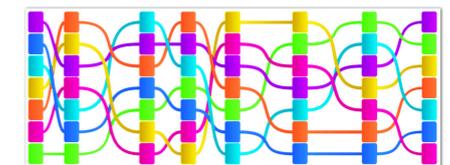


Figure 6.

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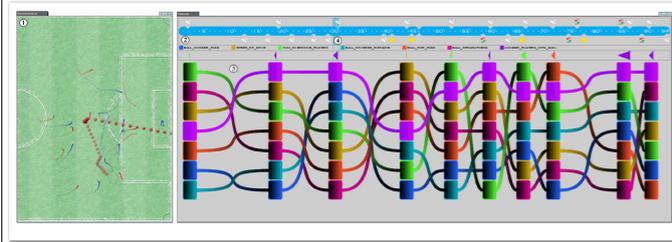


Figure 1.

Data

- 66 professional soccer matches
- Manually annotated events (foul, pass, cross...)
- Temporal resolution: 100 milliseconds

Expert evaluation

- 2 experts : involved in pre-study and expert study
- Coach working at Bayern Munich
- Official referee
- "Ground truth" by additional expert: 35 situations

Results

Table 2. Evaluations results

	Precision	Recall	F-Measure
First Classification	61.53% (8 of 13)	22.85% (8 of 35)	33%
Second Classification (First round of user feedback)	58.82% (20 of 34)	57.14% (20 of 35)	57%
Third Classification (Second round of user feedback)	55.76% (29 of 52)	82.85% (29 of 35)	66%

Results

- Experts liked reducing complexity with meaningful features
- Agreed on features
- Proposed to add information on outcome
- Really liked similarity search (and re-ranking)
- Think that video analyst would use it

+ strengths

- Answer well their task
- Method that you can tweak (reranking) but default => not overwhelming
- Very detailed
- Features seem meaningful

- weakness

- No video for double check
- Unnecessarily long
- Need to read 1st paper to understand some features
- I would use air / ground and not straightness of ball

- weakness

- Validation by 2 "experts" but no video analyst
- 66 games dataset in validation but only use 1
- Very important to have a global view of a tactic not precise movement every 2 seconds
- Only single game
- Do not critique their paper

Thank you !