Football Match
Outcome
Prediction

using DeepSet Player Aggregation

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- Problem Understanding Phase
 - Problem Definition
 - Previous Work
 - Our Idea
- Data Preparation
- Exploratory Data Analysis

Problem Understanding Phase

Problem Definition

- Association Football is the world's most popular sport.
- How is football Played?
- Home Team Advantage
- What is the players' role?
- The Conditions of aggregating the players

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Problem Understanding Phase

Previous Work

- Elo Rating System
- Blade Chest
- Simple GNN

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Problem Understanding Phase

- Our Idea
 - Our Data
 - Our Solution: Deepset

$$V_T = \mathrm{MLP}_{\theta}(\sum_{p \in T} \mathrm{MLP}_{\phi}(p))$$

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 - Reclassifying the Categorical Features
 - Binning
- Exploratory Data Analysis

Data Preparation

Outlier Detection

Z-Score

• IQR

	postGame goals		
37467			2
50688			2
50757			2
50877			2
51107			2
67259			4
47229			4
27278			4
26430			4
18678			5
[944 r	ows x	1	columns]

	posi	tGa	ame	e goals
39708				1
51438				1
51409				1
51407				1
51406				1
27278				4
47229				4
49567				4
67259				4
18678				5
[8211	rows	Χ	1	columns]

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Data Preparation

Transformation and Standardization

```
preGame overall
Original data skewness: 0.11
After transformation if necessary data skewness: -0.053
preGame_potential
Original data skewness: 0.094
After transformation if necessary data skewness: -0.007
preGame marketValueMilEuro
Original data skewness: 9.837
After transformation if necessary data skewness: 0.104
preGame ageDays
Original data skewness: 0.28
After transformation if necessary data skewness: 0.067
```

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Data Preparation

Reclassifying the Categorical Features

	<pre>preGame_position</pre>	<pre>preGame_rc_position</pre>
0	DC	D
1	FW	F
2	МС	М
3	MR	М
4	МС	М
80075	AMC	М
80076	DMC	М
80077	FW	F
80078	FW	F
80079	AMC	М

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Data Preparation

Binning

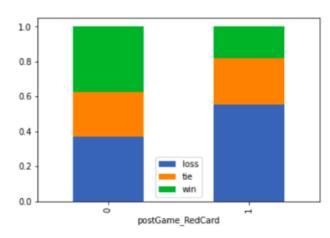
Methods: Equal width, Equal Frequency #bins = random between 3 and 9

```
umerical fp = ['postGame minPlayed',
 'preGame overall', 'preGame potential', 'preGame marketValueMilEuro',
 'preGame ageDays', 'postGame error',
  'postGame_clearance', 'postGame_index', 'postGame_shots',
  'postGame shots on target', 'postGame shots left foot',
  'postGame shots right foot', 'postGame shots head',
  'postGame shots other', 'postGame goals', 'postGame goals left foot',
 'postGame_goals_right_foot', 'postGame_goals_head',
  'postGame goals other', 'postGame xG', 'postGame cross',
  'postGame cross success', 'postGame pass', 'postGame pass success',
  'postGame pass final third', 'postGame pass final third success',
  'postGame_pass_forward', 'postGame_pass_forward_success',
  'postGame_dribble', 'postGame_dribble_success', 'postGame_tackle',
  'postGame tackle success', 'postGame interception',
  'postGame challenge', 'postGame ball recovery', 'postGame ball lost',
  'postGame key pass', 'preGame xgpm', 'preGame xppm',]
numerical gk = ['postGame minPlayed',
     'preGame overall',
      'preGame_potential', 'preGame_marketValueMilEuro',
      'preGame_ageDays', 'preGame_xgpm',
      'preGame xppm', 'postGame error', 'postGame clearance',
      'postGame index', 'postGame pickUp', 'postGame punch', 'postGame save']
```

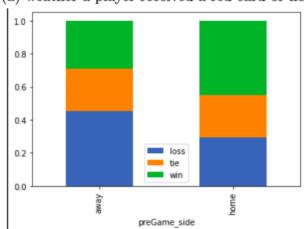
	<pre>postGame_pass</pre>	<pre>postGame_pass_binned_6</pre>
0	52	(32.667, 65.333]
1	16	(-0.196, 32.667]
2	51	(32.667, 65.333]
3	46	(32.667, 65.333]
4	98	(65.333, 98.0]
80075	14	(-0.196, 32.667]
80076	9	(-0.196, 32.667]
80077	22	(-0.196, 32.667]
80078	8	(-0.196, 32.667]
80079	10	(-0.196, 32.667]

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- Exploratory Data Analysis
 - Univariate Relations with the Target Value
 - Multivariate Relations
 - Binning Based on Predictive Value
 - Extracting New Features

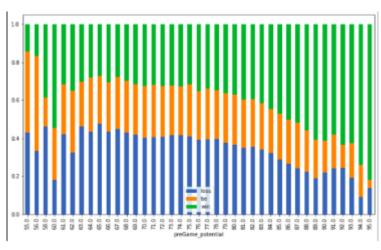
Univariate Relations with the Target Value



(a) weather a player received a red card or not

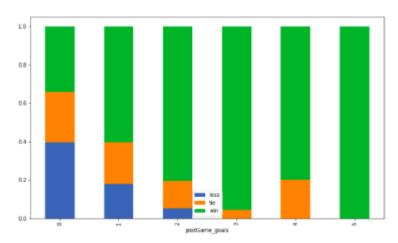


(b) weather a player is playing for the home team or not



(c) pre-game potential from the video game features



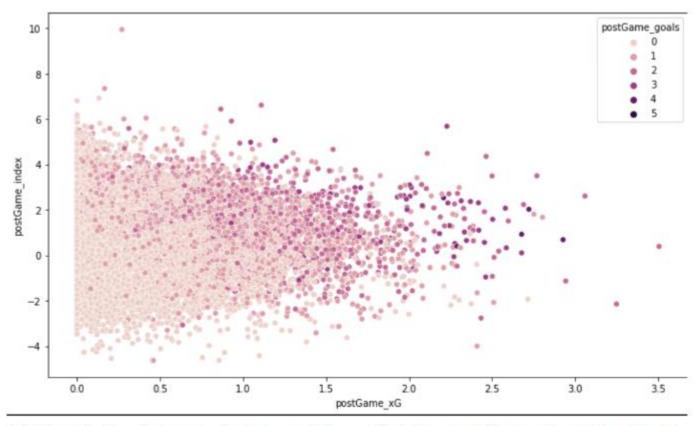


(e) the number of scored goals by a player



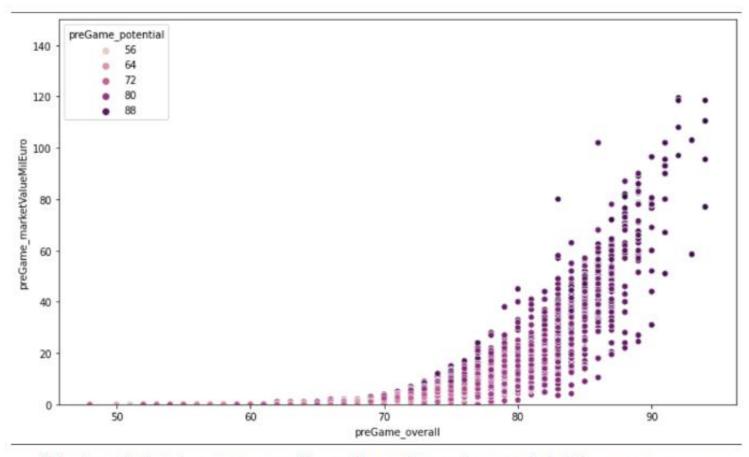
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Multivariate Relations



(a) the relation between stephenson index, xG index and the number of goals of a player

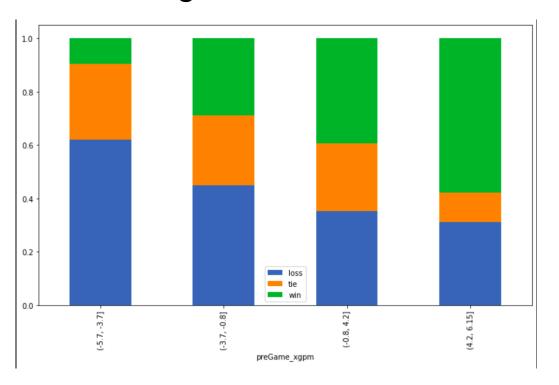
Multivariate Relations

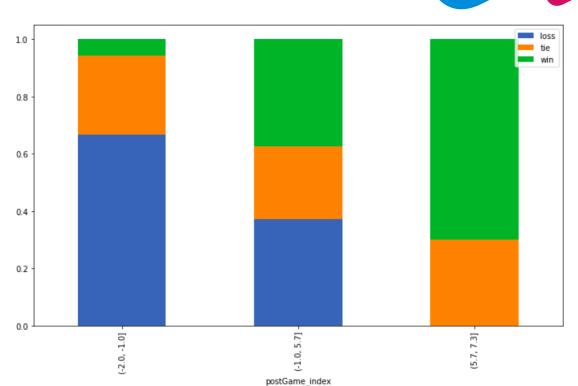


(b) the relation between overall, market value and potential video game scores

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Binning Based on Predictive Value





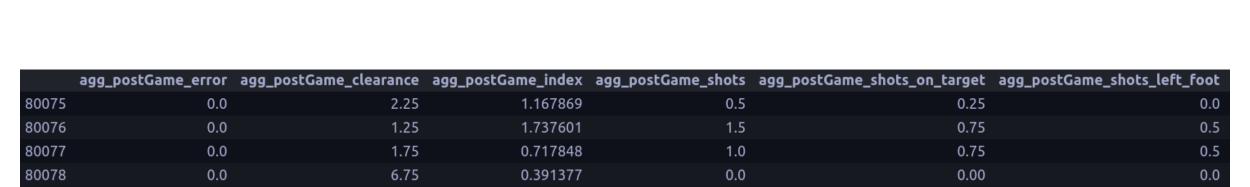
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0.00

Extracting New Features

0.0

80079



0.0

1.234975

0.00

THANK YOU!

