

Theme 3 - Supervised Learning

Practical Assignment 2 - Checkpoint

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01 Dataset

<u>European Soccer Database</u> by Hugo Mathien

A soccer database that contains:

- 1. +25,000 matches
- 2. +10,000 players
- 3. 11 European Countries including their lead championship
- 4. Seasons 2008 to 2016
- 5. Players and Teams' attributes

- 6. Team line up with squad formation (X, Y coordinates)
- 7. Betting odds from up to 10 providers
- Detailed match events (goal types, possession, corner, cross, fouls, cards etc...) for +10,000 matches

02 | Specification of work

- Exploratory analysis of the dataset
- Examination of the data
 - o pre-processing
 - transformation
- Parameterization of supervised learning algorithms

- Algorithms comparison
- Analysis of the confusion matrix
- Demonstration of comparison through appropriate graphs
- Obtain best precision, recall, accuracy,
 F-measure

03 | Research

kaggle Match Outcome Prediction in Football (<u>link</u>)

kaggle EUROPEAN FOOTBALL DATA ANALYSIS (link)

ResearchGate

Exploring and modelling team performances of the Kaggle European Soccer database (<u>link</u>)

04 Tools and Algorithms

- Python
- Google Colab
- Scikit Learn
- Numpy
- MatplotLib
- Pandas
- Pipelines

- Naive Bayes Classifier
- k Nearest Neighbors Classifier (kNN)
- XGBoosts classifier (?)
- Gaussian Model
- kNN with GridsearchCV
- MPL Classifier

05 Organizing

Python notebook will be organized in 4 Sections:

Section A: Our Team

Section B: Introduction to the dataset

- Dataset description
- Data retrieval

- Remove empty values
- Label frequencies

- Over / under sample
- Split to test / train set

Section C: Baseline classification

- kNN Classifier
- Dummy Classifier

- Naive Bayes Classifier
- XGBoosts classifier (?)
- MPL Classifier
- Gaussian Model

Section D: Optimizing classifiers

- Pre-processing
- Balance dataset

- Standardization
- Variance Threshold

- Scaling
- GridSearchCV

06 | Work already implemented

NULL values in the dataset:

- removed samples (lines) from dataset
- other option: fill values with mean / most frequent values of feature

```
imr = Imputer (missing_values= 'NaN', strategy= 'mean', axis=0)
imr = imr.fit (df.values)
imputed data = imr.transform (df.values)
```

LINUX: cat data.csv | grep "?" | wc -| cat data.csv | grep -v "?" > nomissing.data.csv

Map important non-numeric values to numeric

```
Mapping = { 'of':1, 'def':2, 'gk':3}
df['size'] = df['size'].map(mapping)
```

Keep only samples and features with numeric values

06 Work already implemented

Reduce dimensions

```
selector = VarianceThreshold (threshold=0.5)
train_reduced = selector.fit_transform (C_trainData)
```

Normalize values

```
min_max_x = (x - np.min(x)) / (np.max(x) - np.min(x))
```

Possible usage:

```
imbalanced-learn to over/under-sample dataset
principal components analysis - PCA
```

```
req cols = ['overall rating', 'crossing', 'finishing', 'heading accuracy',
          'short passing', 'volleys', 'dribbling', 'curve',
          'free kick accuracy', 'long passing', 'ball control', 'acceleration',
          'sprint speed', 'agility', 'reactions', 'balance', 'shot power', 'jumping',
          'stamina', 'strength', 'long shots', 'aggression', 'interceptions',
          'positioning', 'vision', 'penalties', 'marking', 'standing tackle',
          'sliding tackle', 'gk diving', 'gk handling', 'gk kicking',
          'gk positioning', 'gk reflexes']
data = player data[req cols]
data = player data.drop(labels = ['id', 'player fifa api id', 'player api id',
'date',
                                  'potential', 'preferred foot',
                                  'attacking work rate',
                                  'defensive work rate', axis = 1)
data.fillna(0, inplace=True)
#data.isnull().values.any()
data.corr()a
```

```
from sklearn.cross validation import train test split
feature cols = ['crossing', 'finishing', 'heading accuracy', 'short passing',
       'dribbling', 'curve', 'free kick accuracy',
       'long passing', 'ball control', 'acceleration', 'sprint speed',
       'agility', 'reactions', 'balance', 'shot power', 'jumping', 'stamina',
       'strength', 'long shots', 'aggression', 'interceptions', 'positioning',
       'vision', 'penalties', 'marking', 'standing tackle', 'sliding tackle',
       'gk diving', 'gk handling', 'gk kicking', 'gk positioning',
'qk reflexes']
x = data[feature cols]
y = data.overall rating
x train, x test,
y train, y test = train test split(x, y, test size = 0.25, random state = 42)
```

Thank you!