

## TD Association Mining

You need to install firstly the package **mlxtend** under **Python3.6**.

The package **mlxtend** doesn't work with the pre-installed **python3** in some systems, since it uses `print(f " ")` in the package. Check the version of the **python3** before the installation.

You can find more information related to the package **mlxtend** from

[http://rasbt.github.io/mlxtend/user\\_guide/frequent\\_patterns/association\\_rules/](http://rasbt.github.io/mlxtend/user_guide/frequent_patterns/association_rules/)

### Part I Examples

Write the comments to explain how the following programs work.

#### Example I.

```
import pandas as pd
from mlxtend.preprocessing import TransactionEncoder

dataset = [['f', 'a', 'c', 'd', 'g', 'i', 'm', 'p'],
           ['a', 'b', 'c', 'f', 'l', 'm', 'o'],
           ['b', 'f', 'h', 'j', 'o'],
           ['b', 'c', 'k', 's', 'p'],
           ['a', 'f', 'c', 'e', 'l', 'p', 'm', 'n']]

te = TransactionEncoder()
te_ary = te.fit(dataset).transform(dataset)
df = pd.DataFrame(te_ary, columns=te.columns_)
df

from mlxtend.frequent_patterns import apriori

#apriori(df, min_support=0.5, use_colnames=True)
frequent_itemsets = apriori(df, min_support=0.5, use_colnames=True)
frequent_itemsets['length'] = frequent_itemsets['itemsets'].apply(lambda x: len(x))
frequent_itemsets

frequent_itemsets[ (frequent_itemsets['length'] == 2) &
                   (frequent_itemsets['support'] >= 0.6) ]
frequent_itemsets[ frequent_itemsets['itemsets'] == {'f', 'm'} ]
frequent_itemsets[frequent_itemsets['itemsets'].str.contains('m', regex=False)]
```

#### Example II.

```
import pandas as pd
from mlxtend.frequent_patterns import apriori
from mlxtend.frequent_patterns import association_rules

Jogging_data=pd.read_csv('JoggingTitre.csv', sep=',')
x_dum=pd.get_dummies(Jogging_data)
x_dum

frequent_itemsets02=apriori(x_dum, min_support=0.2, use_colnames=True)
frequent_itemsets02
```

```
ARules02=association_rules(frequent_itemsets02, metric="confidence", min_threshold=0.8)
ARules02
ARules02.iloc[0:, 0:4]
ARules02.iloc[0:, [0, 1, 4, 5]]
ARules02[(ARules02['consequents'] == {'Jogging_Oui'})|(ARules02['consequents'] ==
{'Jogging_Non'})]
ARules02[(ARules02['consequents'] == {'Jogging_Oui'})].iloc[0:, [0, 1, 4, 5]]
```

## Part II Exercises

Execute the following analyses and observations with at least 3 of the following data sets :

***car.data,***  
***flag.data,***  
***agaricus-lepiota.data,***  
***tic-tac-toe.data,*** and  
***zoo.data***

It's better to analyze the same data sets used in the last TP in order to make some comparaisons

### Exercise1

- Find the frequent patterns and association rules from the data set with *minSup*=10% and *minConf*=80%. You have to give your attention to the following points.
  - a) It's not necessary to use all attributes in the data set, for example, the attributes as identifier should not be used in your analyses.
  - b) You may need to do some preprocessing steps to preparer your data.
  - c) You should use categorical data, at least ordinal data, in your analyses. You can try some kinds of discrimination to convert your numerical data to categorical data.
  - d) The first line in the data set should be the attribute name for your data when you use *mlxtend* package.
- Mine the 3 most interesting phenomena from your association rule set.

### Exercise2

- A decision tree can be transformed to a set of " If...Then..." rules. In this exercise, you are asked firstly to transform the highest accuracy decision tree to a such set of rules for each analyzed data set in last TP.
- Compare and observe the rules from the last TP and the association rules found in Exercise1.

## Final Report

In your report, you have to return

- the programs with comments of Part I;
- Python programs with comments used for the exercises in Part II;
- descriptions of the interesting phenomena found in Exercise1 of Part II;
- observations from Exercise 2 of Part II