

# **Assignment Project Report**

## **Association Rule Mining: Market Basket Analysis**

**Name:** Pavan Tiwari

**Course:** AI and ML

(Batch 4)

- **Problem Statement**

Apriori is a statistical algorithm for implementing associate rule mining, that primarily relies on three components: Life, Support and Confidence. Using this algorithm try to find the rules that describe the relation between each of the products that were brought by the customers as described in

### **Prerequisites**

- Software:
  - Python 3 (Use anaconda as your python distributor as well)
- Tools:
  - Pandas
  - Numpy
  - Matplotlib
  - Seaborn
- Dataset: : Store Data  
<https://drive.google.com/file/d/1y5DYn0dGoSbC22xowBq2d4po6h1JxcTQ/view?usp=sharing>

- **Method Used**

Association Rule Mining is used when we want to find an association between different objects in a set or find frequent patterns in a transaction database or relational databases. The applications of Association Rule Mining are found in Marketing, Basket Data Analysis (or Market Basket Analysis) in retailing, clustering and classification. It can be used to find what items do customers frequently buy together by generating a set of rules called Association Rules.

- **Implementation:**

## 1. Load all required libraries and Dataset

```
In [1]: 1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
```

```
In [2]: 1 data = pd.read_csv('store_data.csv')
2 data.head()
```

## 2. Preprocessing data

```
In [5]: 1 transactions = []
2 for i in range(0, data.shape[0]):
3     transactions.append([str(data.values[i, j]) for j in range(0, 20)])
4
5 print(transactions[0])

['burgers', 'meatballs', 'eggs', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan', 'nan']
```

## 3. Using Model

```
In [6]: 1 from apyori import apriori
2
3 rules = apriori(transactions, min_support = 0.003, min_confidence = 0.2, min_lift = 3, min_length = 2)
4 # Support: number of transactions containing set of items / total number of transactions
5 # . --> products that are bought at least 3 times a day --> 21 / 7501 = 0.0027
6 # Confidence: Should not be too high, as then this will lead to obvious rules
7
8 #Try many combinations of values to experiment with the model.
9
10 #viewing the rules
11 results = list(rules)
```

## 4. Output

```
In [7]: 1 #Transferring the list to a table
2
3 results = pd.DataFrame(results)
4 results.head(5)
```

```
Out[7]:
```

	items	support	ordered_statistics
0	(chicken, light cream)	0.004533	[((light cream), (chicken), 0.2905982905982906...
1	(mushroom cream sauce, escalope)	0.005733	[((mushroom cream sauce), (escalope), 0.300699...
2	(pasta, escalope)	0.005867	[((pasta), (escalope), 0.37288135593220345, 4....
3	(honey, fromage blanc)	0.003333	[((fromage blanc), (honey), 0.2450980392156863...
4	(herb & pepper, ground beef)	0.016000	[((herb & pepper), (ground beef), 0.3234501347...

```
In [9]: 1 results['ordered_statistics'][0]
```

```
Out[9]: [OrderedStatistic(items_base=frozenset({'light cream'}), items_add=frozenset({'chicken'}), confidence=0.2905982905982906, lift=4.843304843304844)]
```

