# Association Rule Mining via Apriori Algorithm in Python

# Installing Apyori package

```
!pip install apyori

Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/pub
Collecting apyori
    Downloading apyori-1.1.2.tar.gz (8.6 kB)
Building wheels for collected packages: apyori
    Building wheel for apyori (setup.py) ... done
    Created wheel for apyori: filename=apyori-1.1.2-py3-none-any.whl size=5974 sha256=f19
    Stored in directory: /root/.cache/pip/wheels/cb/f6/e1/57973c631d27efd1a2f375bd6a83b2a
Successfully built apyori
Installing collected packages: apyori
Successfully installed apyori-1.1.2
```

## Import the Libraries

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from apyori import apriori
```

#### Importing the Dataset

|   | shrimp            | almonds   | avocado       | vegetables<br>mix | green<br>grapes | whole<br>weat<br>flour | yams | cottage<br>cheese | energy<br>drink | tomato<br>juice |
|---|-------------------|-----------|---------------|-------------------|-----------------|------------------------|------|-------------------|-----------------|-----------------|
| 0 | burgers           | meatballs | eggs          | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             |
| 1 | chutney           | NaN       | NaN           | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             |
| 2 | turkey            | avocado   | NaN           | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             |
| 3 | mineral<br>water  | milk      | energy<br>bar | whole wheat rice  | green<br>tea    | NaN                    | NaN  | NaN               | NaN             | NaN             |
| 4 | low fat<br>yogurt | NaN       | NaN           | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             |

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store\_data = pd.read\_csv('store\_data.csv', header=None)

store\_data.head()

|   | 0                | 1         | 2             | 3                 | 4               | 5                      | 6    | 7                 | 8               | 9               |     |
|---|------------------|-----------|---------------|-------------------|-----------------|------------------------|------|-------------------|-----------------|-----------------|-----|
| 0 | shrimp           | almonds   | avocado       | vegetables<br>mix | green<br>grapes | whole<br>weat<br>flour | yams | cottage<br>cheese | energy<br>drink | tomato<br>juice | yog |
| 1 | burgers          | meatballs | eggs          | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             | Ν   |
| 2 | chutney          | NaN       | NaN           | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             | Ν   |
| 3 | turkey           | avocado   | NaN           | NaN               | NaN             | NaN                    | NaN  | NaN               | NaN             | NaN             | Ν   |
| 4 | mineral<br>water | milk      | energy<br>bar | whole wheat rice  | green<br>tea    | NaN                    | NaN  | NaN               | NaN             | NaN             | N   |



### **Data Proprocessing**

```
records = []
for i in range(0, 7501):
    records.append([str(store_data.values[i,j]) for j in range(0, 20)])
```

# **Applying Apriori**

association\_rules = apriori(records, min\_support=0.0045, min\_confidence=0.2, min\_lift=3, min\_ association\_results = list(association\_rules)

```
print(len(association_results))

48

print(association_results[0])

RelationRecord(items=frozenset({'light cream', 'chicken'}), support=0.00453272896947073
```

#### Viewing the Results

```
for item in association_results:
   # first index of the inner list
   # Contains base item and add item
   pair = item[0]
   items = [x for x in pair]
   print("Rule: " + items[0] + " -> " + items[1])
   #second index of the inner list
   print("Support: " + str(item[1]))
   #third index of the list located at 0th
   #of the third index of the inner list
   print("Confidence: " + str(item[2][0][2]))
   print("Lift: " + str(item[2][0][3]))
   print("======="")
    Lift: 3.2819951870487856
    _____
    Rule: escalope -> nan
    Support: 0.005732568990801226
    Confidence: 0.3006993006993007
    Lift: 3.790832696715049
    Rule: escalope -> nan
    Support: 0.005865884548726837
    Confidence: 0.3728813559322034
    Lift: 4.700811850163794
    _____
    Rule: frozen vegetables -> ground beef
    Support: 0.008665511265164644
    Confidence: 0.31100478468899523
    Lift: 3.165328208890303
    _____
    Rule: olive oil -> frozen vegetables
    Support: 0.004799360085321957
    Confidence: 0.20338983050847456
    Lift: 3.088314005352364
```

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Rule: frozen vegetables -> shrimp Support: 0.007199040127982935 Confidence: 0.30508474576271183

Lift: 3.200616332819722

Rule: olive oil -> frozen vegetables

Support: 0.005732568990801226 Confidence: 0.20574162679425836

Lift: 3.1240241752707125

Lift: 3.0131489680782684

Rule: frozen vegetables -> tomatoes

Support: 0.006665777896280496 Confidence: 0.23923444976076558

Lift: 3.4980460188216425

Rule: grated cheese -> ground beef
Support: 0.005332622317024397

Confidence: 0.3225806451612903 Lift: 3.283144395325426

Rule: ground beef -> herb & pepper Support: 0.006665777896280496

Rule: nan -> ground beef Support: 0.015997866951073192 Confidence: 0.3234501347708895

Lift: 3.2919938411349285

Rule: ground beef -> herb & pepper