Assignment on Association Rule Learning

Download Market Basket Optimization dataset from below link. Data Set: https://www.kaggle.com/hemanthkumar05/market•basket•optimization.

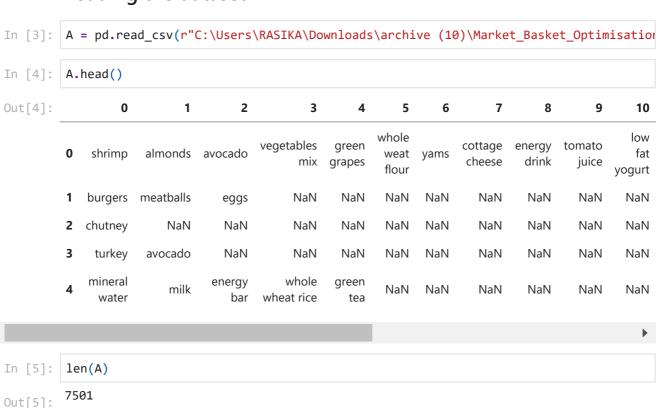
This dataset comprises the list of transactions of a retail company over the period of one week. It contains a total of 7501 transaction records where each record consists of the list of items sold in one transaction. Using this record of transactions and items in each transaction, find the association rules between items. There is no header in the dataset and the first row contains the first transaction, so mentioned header = None here while loading dataset. Follow following steps:

- 1. Data Preprocessing
- 2. Generate the list of transactions from the dataset
- 3. Train Apriori algorithm on the dataset
- 4. Visualize the list of rules
- 5. Generated rules depend on the values of hyper parameters. By increasing the minimum confidence value and find the rules accordingly

Importing Libraries

In [2]: import pandas as pd

Loading the dataset



Data Preprocessing

```
A.info()
In [6]:
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 7501 entries, 0 to 7500
         Data columns (total 20 columns):
               Column Non-Null Count Dtype
          0
               0
                        7501 non-null
                                          object
          1
               1
                        5747 non-null
                                          object
          2
                        4389 non-null
                                          object
           3
               3
                        3345 non-null
                                          object
                        2529 non-null
                                          object
               5
                        1864 non-null
          5
                                          object
          6
               6
                        1369 non-null
                                          object
          7
               7
                        981 non-null
                                          object
          8
               8
                        654 non-null
                                          object
               9
          9
                        395 non-null
                                          object
          10
               10
                        256 non-null
                                          object
               11
                        154 non-null
                                          object
          11
          12
               12
                        87 non-null
                                          object
          13
               13
                        47 non-null
                                          object
          14
               14
                        25 non-null
                                          object
          15
               15
                        8 non-null
                                          object
          16
               16
                        4 non-null
                                          object
          17
               17
                        4 non-null
                                          object
           18
               18
                        3 non-null
                                          object
               19
                        1 non-null
                                          object
         dtypes: object(20)
         memory usage: 1.1+ MB
In [7]:
         A.describe()
Out[7]:
                               1
                                        2
                                                3
                                                       4
                                                              5
                                                                     6
                                                                           7
                                                                                  8
                                                                                         9
                                                                                                10
                    7501
                                                    2529
                                                           1864
           count
                            5747
                                    4389
                                             3345
                                                                  1369
                                                                          981
                                                                                654
                                                                                       395
                                                                                               256
                                                            106
                                                                           98
         unique
                     115
                             117
                                      115
                                              114
                                                     110
                                                                   102
                                                                                 88
                                                                                                66
                                                                                               low
                                                                                     green
                                                                                                    gree
                  mineral
                          mineral
                                  mineral
                                          mineral
                                                   green
                                                          french
                                                                 green
                                                                        green
                                                                               green
                                                                                               fat
             top
                   water
                            water
                                    water
                                            water
                                                     tea
                                                           fries
                                                                   tea
                                                                          tea
                                                                                 tea
                                                                                        tea
                                                                                            yogurt
```

Generating Transaction List

484

375

freq

577

```
In [8]:
        transactions = []
        for i in range(0, 7501):
            transactions.append([str(A.values[i,j]) for j in range(0, 20)])
```

201

153

107

96

67

57

31

Creating apriori model

```
In [9]:
        from apyori import apriori
        tran_rules = apriori(transactions, min_support = 0.003, min_confidence = 0.6, min_i
```

Visualising the results

```
results = list(tran_rules)
In [10]:
          print(results[:10])
```

1

15

Ē

te

22

[RelationRecord(items=frozenset({'spaghetti', 'ground beef', 'cereals'}), support= 0.0030662578322890282, ordered statistics=[OrderedStatistic(items base=frozenset ({'ground beef', 'cereals'}), items_add=frozenset({'spaghetti'}), confidence=0.676 4705882352942, lift=3.8853031258445188)]), RelationRecord(items=frozenset({'spaghe tti', 'tomatoes', 'olive oil'}), support=0.004399413411545127, ordered_statistics= [OrderedStatistic(items_base=frozenset({'tomatoes', 'olive oil'}), items_add=froze nset({'spaghetti'}), confidence=0.6111111111111112, lift=3.5099115194827295)]), Re lationRecord(items=frozenset({'spaghetti', 'ground beef', 'nan', 'cereals'}), supp ort=0.0030662578322890282, ordered_statistics=[OrderedStatistic(items_base=frozens et({'ground beef', 'cereals'}), items_add=frozenset({'spaghetti', 'nan'}), confide nce=0.6764705882352942, lift=3.8853031258445188), OrderedStatistic(items_base=froz enset({'ground beef', 'nan', 'cereals'}), items_add=frozenset({'spaghetti'}), conf idence=0.6764705882352942, lift=3.8853031258445188)]), RelationRecord(items=frozen set({'milk', 'soup', 'frozen vegetables', 'mineral water'}), support=0.00306625783 22890282, ordered_statistics=[OrderedStatistic(items_base=frozenset({'milk', 'sou p', 'frozen vegetables'}), items_add=frozenset({'mineral water'}), confidence=0.76 up', 'frozen vegetables', 'mineral water'}), items_add=frozenset({'milk'}), confid ence=0.6052631578947368, lift=4.670863114576565)]), RelationRecord(items=frozenset ({'spaghetti', 'nan', 'tomatoes', 'olive oil'}), support=0.004399413411545127, ord ered_statistics=[OrderedStatistic(items_base=frozenset({'tomatoes', 'olive oil'}), items_add=frozenset({'spaghetti', 'nan'}), confidence=0.6111111111111112, lift=3.5 099115194827295), OrderedStatistic(items_base=frozenset({'nan', 'tomatoes', 'olive oil'}), items_add=frozenset({'spaghetti'}), confidence=0.6111111111111112, lift=3. 5099115194827295)]), RelationRecord(items=frozenset({'soup', 'frozen vegetables', 'milk', 'nan', 'mineral water'}), support=0.0030662578322890282, ordered_statistic s=[OrderedStatistic(items_base=frozenset({'milk', 'soup', 'frozen vegetables'}), i 3.218112292482745), OrderedStatistic(items_base=frozenset({'soup', 'frozen vegetab les', 'mineral water'}), items_add=frozenset({'milk', 'nan'}), confidence=0.605263 1578947368, lift=4.670863114576565), OrderedStatistic(items_base=frozenset({'mil k', 'soup', 'frozen vegetables', 'nan'}), items_add=frozenset({'mineral water'}), =frozenset({'soup', 'frozen vegetables', 'nan', 'mineral water'}), items_add=froze nset({'milk'}), confidence=0.6052631578947368, lift=4.670863114576565)])]

In [12]: resultsinDataFrame.nlargest(n = 10, columns = 'Confidence')

Out[12]:		Left Hand Side	Right Hand Side	Support	Confidence	Lift
	3	milk	mineral water	0.003066	0.766667	3.216312
	5	milk	nan	0.003066	0.766667	3.218112
	0	ground beef	spaghetti	0.003066	0.676471	3.885303
	2	ground beef	spaghetti	0.003066	0.676471	3.885303
	1	tomatoes	spaghetti	0.004399	0.611111	3.509912
	4	tomatoes	spaghetti	0.004399	0.611111	3.509912