# In [2]: import pandas as pd !pip install mlxtend from mlxtend.preprocessing import TransactionEncoder

Requirement already satisfied: mlxtend in c:\users\manoj\anaconda3\lib\site-pac kages (0.18.0)

Requirement already satisfied: matplotlib>=3.0.0 in c:\users\manoj\anaconda3\lib\site-packages (from mlxtend) (3.2.2)

Requirement already satisfied: numpy>=1.16.2 in c:\users\manoj\anaconda3\lib\si te-packages (from mlxtend) (1.19.5)

Requirement already satisfied: joblib>=0.13.2 in c:\users\manoj\anaconda3\lib\s ite-packages (from mlxtend) (0.16.0)

Requirement already satisfied: scipy>=1.2.1 in c:\users\manoj\anaconda3\lib\sit e-packages (from mlxtend) (1.5.0)

Requirement already satisfied: scikit-learn>=0.20.3 in c:\users\manoj\anaconda3 \lib\site-packages (from mlxtend) (0.23.1)

Requirement already satisfied: pandas>=0.24.2 in c:\users\manoj\anaconda3\lib\s ite-packages (from mlxtend) (1.0.5)

Requirement already satisfied: setuptools in c:\users\manoj\anaconda3\lib\site-packages (from mlxtend) (49.2.0.post20200714)

Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\manoj\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (1.2.0)

Requirement already satisfied: pyparsing!=2.0.4,!=2.1.2,!=2.1.6,>=2.0.1 in c:\u sers\manoj\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (2.4.7)

Requirement already satisfied: cycler>=0.10 in c:\users\manoj\anaconda3\lib\sit e-packages (from matplotlib>=3.0.0->mlxtend) (0.10.0)

Requirement already satisfied: python-dateutil>=2.1 in c:\users\manoj\anaconda3 \lib\site-packages (from matplotlib>=3.0.0->mlxtend) (2.8.1)

Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\manoj\anaconda3 \lib\site-packages (from scikit-learn>=0.20.3->mlxtend) (2.1.0)

Requirement already satisfied: pytz>=2017.2 in c:\users\manoj\anaconda3\lib\sit e-packages (from pandas>=0.24.2->mlxtend) (2020.1)

Requirement already satisfied: six in c:\users\manoj\anaconda3\lib\site-package s (from cycler>=0.10->matplotlib>=3.0.0->mlxtend) (1.15.0)

```
In [3]: te = TransactionEncoder()
te_try = te.fit(dataset).transform(dataset)
```

```
In [4]: df=pd.DataFrame(te_try,columns=te.columns_)
```

```
In [5]: df
```

#### Out[5]:

	Coffee	Milk	bhel	biscuits	bread	butter	cakes	cheese	chips	choclates	chocos	coffee
0	True	True	False	False	True	False	False	False	False	False	True	False
1	False	False	False	True	True	True	False	False	True	False	False	True
2	False	False	False	False	True	False	True	False	True	False	False	True
3	False	False	False	False	True	False	True	False	False	True	False	False
4	False	False	True	False	True	False	True	True	False	False	False	False

```
In [6]: from mlxtend.frequent_patterns import apriori
from mlxtend.frequent_patterns import association_rules
apriori(df,min_support = 0.6)
```

## Out[6]:

	support	itemsets
0	1.0	(4)
1	0.6	(6)
2	0.6	(15)
3	0.6	(4, 6)
4	0.6	(4, 15)

In [7]: from mlxtend.frequent\_patterns import apriori
from mlxtend.frequent\_patterns import association\_rules
apriori(df,min\_support = 0.8)

## Out[7]:

```
o 1.0 (4)
```

```
In [8]: frequent_itemsets = apriori(df,min_support = 0.5,use_colnames=True)
```

```
In [9]: equent_itemsets['length'] = frequent_itemsets['itemsets'].apply(lambda x:len(x))
```

In [10]: emsets[(frequent\_itemsets['length'] == 2)& (frequent\_itemsets['support']>=0.8) ]

#### Out[10]:

support itemsets length

```
In [11]: frequent itemsets[(frequent itemsets['length'] == 3)]
Out[11]:
            support itemsets length
         frequent itemsets[(frequent itemsets['support'] >=1) ]
In [12]:
Out[12]:
             support itemsets length
           0
                 1.0
                       (bread)
                                  1
          from mlxtend.frequent patterns import apriori
          from mlxtend.frequent_patterns import association_rules
          frequent_factors = apriori(df, use_colnames= True,min_support= 0.1,max_len =4)
          rules = association rules(frequent factors, metric='lift', min threshold=1)
          print(rules.head(10).sort_values(by = 'confidence', ascending= False))
            antecedents consequents
                                       antecedent support
                                                            consequent support
                                                                                  support
                                                                                            \
                  (Milk)
                            (Coffee)
                                                                                       0.2
          0
                                                       0.2
                                                                             0.2
                                                                                       0.2
          1
               (Coffee)
                               (Milk)
                                                       0.2
                                                                             0.2
          3
               (Coffee)
                                                       0.2
                                                                             1.0
                                                                                       0.2
                              (bread)
                             (Coffee)
          4
               (chocos)
                                                       0.2
                                                                             0.2
                                                                                       0.2
          5
               (Coffee)
                             (chocos)
                                                       0.2
                                                                             0.2
                                                                                       0.2
          6
                  (eggs)
                             (Coffee)
                                                       0.2
                                                                             0.2
                                                                                       0.2
          7
                                                                                       0.2
               (Coffee)
                               (eggs)
                                                       0.2
                                                                             0.2
          8
                (sugar)
                             (Coffee)
                                                       0.2
                                                                             0.2
                                                                                       0.2
          9
               (Coffee)
                                                       0.2
                                                                             0.2
                                                                                       0.2
                              (sugar)
          2
                 (bread)
                             (Coffee)
                                                       1.0
                                                                             0.2
                                                                                       0.2
             confidence
                          lift
                                leverage
                                           conviction
          0
                           5.0
                                     0.16
                                                   inf
                     1.0
                           5.0
          1
                     1.0
                                     0.16
                                                   inf
          3
                     1.0
                           1.0
                                     0.00
                                                   inf
          4
                     1.0
                           5.0
                                     0.16
                                                   inf
          5
                     1.0
                           5.0
                                     0.16
                                                   inf
          6
                     1.0
                           5.0
                                     0.16
                                                   inf
          7
                           5.0
                     1.0
                                     0.16
                                                   inf
          8
                     1.0
                           5.0
                                     0.16
                                                   inf
          9
                     1.0
                           5.0
                                     0.16
                                                   inf
          2
                     0.2
                           1.0
                                     0.00
                                                   1.0
```

```
In [14]: '{\'bread\'}'

ple = rules[(rules['consequents'].astype(str).str.contains(target, na= False))]

ple
```

# Out[14]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage
3	(Coffee)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
10	(Milk)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
18	(bhel)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
28	(biscuits)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
38	(butter)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
1143	(juices, cakes, milk)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
1155	(juices, ice cream, cheese)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
1168	(coffee, milk, chips)	(bread)	0.4	1.0	0.4	1.0	1.0	0.0
1182	(choclates, juices, milk)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0
1197	(chocos, eggs, sugar)	(bread)	0.2	1.0	0.2	1.0	1.0	0.0

90 rows × 9 columns

In [ ]:	
In [ ]:	

Type  $\it Markdown$  and LaTeX:  $\it \alpha^2$