

In [2]: `!pip install mlxtend`

```
Requirement already satisfied: mlxtend in c:\users\omnic\anaconda3\lib\site-packages (0.23.0)
Requirement already satisfied: scipy>=1.2.1 in c:\users\omnic\anaconda3\lib\site-packages (from mlxtend) (1.10.1)
Requirement already satisfied: numpy>=1.16.2 in c:\users\omnic\anaconda3\lib\site-packages (from mlxtend) (1.24.3)
Requirement already satisfied: pandas>=0.24.2 in c:\users\omnic\anaconda3\lib\site-packages (from mlxtend) (1.5.3)
Requirement already satisfied: scikit-learn>=1.0.2 in c:\users\omnic\anaconda3\lib\site-packages (from mlxtend) (1.2.2)
Requirement already satisfied: matplotlib>=3.0.0 in c:\users\omnic\anaconda3\lib\site-packages (from mlxtend) (3.7.1)
Requirement already satisfied: joblib>=0.13.2 in c:\users\omnic\anaconda3\lib\site-packages (from mlxtend) (1.2.0)
Requirement already satisfied: contourpy>=1.0.1 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (1.0.5)
Requirement already satisfied: cycler>=0.10 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (0.11.0)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (4.25.0)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (1.4.4)
Requirement already satisfied: packaging>=20.0 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (23.0)
Requirement already satisfied: pillow>=6.2.0 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (9.4.0)
Requirement already satisfied: pyparsing>=2.3.1 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (3.0.9)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\omnic\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in c:\users\omnic\anaconda3\lib\site-packages (from pandas>=0.24.2->mlxtend) (2022.7)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\omnic\anaconda3\lib\site-packages (from scikit-learn>=1.0.2->mlxtend) (2.2.0)
Requirement already satisfied: six>=1.5 in c:\users\omnic\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib>=3.0.0->mlxtend) (1.16.0)
```

In [3]: `import pandas as pd
import csv
from mlxtend.preprocessing import TransactionEncoder
from mlxtend.frequent_patterns import apriori, association_rules`

In [5]: `dataset = []
with open('Market_Basket_Optimisation.csv') as file:
 reader = csv.reader(file, delimiter=',')
 for row in reader:
 dataset += [row]`

In [6]: `len(dataset)`

Out[6]: 7501

In [8]: `te = TransactionEncoder()
x = te.fit_transform(dataset)
x`

```
Out[8]: array([[False,  True,  True, ...,  True, False, False],
        [False, False, False, ..., False, False, False],
        [False, False, False, ..., False, False, False],
        ...,
        [False, False, False, ..., False, False, False],
        [False, False, False, ..., False, False, False],
        [False, False, False, ..., False,  True, False]])
```

```
In [10]: df = pd.DataFrame(x,columns=te.columns_)
```

```
In [15]: freq_itemset = apriori(df, min_support=0.01, use_colnames=True)
```

```
In [16]: freq_itemset
```

```
Out[16]:
```

	support	itemsets
0	0.020397	(almonds)
1	0.033329	(avocado)
2	0.010799	(barbecue sauce)
3	0.014265	(black tea)
4	0.011465	(body spray)
...
252	0.011065	(ground beef, mineral water, milk)
253	0.017064	(ground beef, spaghetti, mineral water)
254	0.015731	(spaghetti, mineral water, milk)
255	0.010265	(spaghetti, olive oil, mineral water)
256	0.011465	(spaghetti, mineral water, pancakes)

257 rows × 2 columns

```
In [17]: rules = association_rules(freq_itemset, metric= 'confidence', min_threshold=0.25)
```

```
In [18]: rules
```

Out[18]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	con
0	(avocado)	(mineral water)	0.033329	0.238368	0.011598	0.348000	1.459926	0.003654	1.
1	(burgers)	(eggs)	0.087188	0.179709	0.028796	0.330275	1.837830	0.013128	1.
2	(burgers)	(french fries)	0.087188	0.170911	0.021997	0.252294	1.476173	0.007096	1.
3	(burgers)	(mineral water)	0.087188	0.238368	0.024397	0.279817	1.173883	0.003614	1.
4	(cake)	(mineral water)	0.081056	0.238368	0.027463	0.338816	1.421397	0.008142	1.
...
90	(mineral water, milk)	(spaghetti)	0.047994	0.174110	0.015731	0.327778	1.882589	0.007375	1.
91	(spaghetti, olive oil)	(mineral water)	0.022930	0.238368	0.010265	0.447674	1.878079	0.004799	1.
92	(olive oil, mineral water)	(spaghetti)	0.027596	0.174110	0.010265	0.371981	2.136468	0.005460	1.
93	(spaghetti, pancakes)	(mineral water)	0.025197	0.238368	0.011465	0.455026	1.908923	0.005459	1.
94	(mineral water, pancakes)	(spaghetti)	0.033729	0.174110	0.011465	0.339921	1.952333	0.005593	1.

95 rows × 10 columns

In [19]: `rules = rules [['antecedents', 'consequents', 'support', 'confidence']]`In [20]: `rules.head()`

Out[20]:

	antecedents	consequents	support	confidence
0	(avocado)	(mineral water)	0.011598	0.348000
1	(burgers)	(eggs)	0.028796	0.330275
2	(burgers)	(french fries)	0.021997	0.252294
3	(burgers)	(mineral water)	0.024397	0.279817
4	(cake)	(mineral water)	0.027463	0.338816

In [22]: `rules[rules['antecedents']=={'cake'}]['consequents']`

Out[22]: 4 (mineral water)
Name: consequents, dtype: object

In []: