

# apriori-on-market-basket

December 5, 2023

```
[1]: ! pip install mlxtend
```

```
Requirement already satisfied: mlxtend in c:\users\nikee\anaconda3\lib\site-  
packages (0.23.0)  
Requirement already satisfied: scipy>=1.2.1 in  
c:\users\nikee\anaconda3\lib\site-packages (from mlxtend) (1.10.1)  
Requirement already satisfied: numpy>=1.16.2 in  
c:\users\nikee\anaconda3\lib\site-packages (from mlxtend) (1.24.3)  
Requirement already satisfied: pandas>=0.24.2 in  
c:\users\nikee\anaconda3\lib\site-packages (from mlxtend) (1.5.3)  
Requirement already satisfied: scikit-learn>=1.0.2 in  
c:\users\nikee\anaconda3\lib\site-packages (from mlxtend) (1.3.0)  
Requirement already satisfied: matplotlib>=3.0.0 in  
c:\users\nikee\anaconda3\lib\site-packages (from mlxtend) (3.7.1)  
Requirement already satisfied: joblib>=0.13.2 in  
c:\users\nikee\anaconda3\lib\site-packages (from mlxtend) (1.2.0)  
Requirement already satisfied: contourpy>=1.0.1 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(1.0.5)  
Requirement already satisfied: cycycler>=0.10 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(0.11.0)  
Requirement already satisfied: fonttools>=4.22.0 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(4.25.0)  
Requirement already satisfied: kiwisolver>=1.0.1 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(1.4.4)  
Requirement already satisfied: packaging>=20.0 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(23.0)  
Requirement already satisfied: pillow>=6.2.0 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(9.4.0)  
Requirement already satisfied: pyparsing>=2.3.1 in  
c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(3.0.9)  
Requirement already satisfied: python-dateutil>=2.7 in
```

c:\users\nikee\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend)  
(2.8.2)  
Requirement already satisfied: pytz>=2020.1 in  
c:\users\nikee\anaconda3\lib\site-packages (from pandas>=0.24.2->mlxtend)  
(2022.7)  
Requirement already satisfied: threadpoolctl>=2.0.0 in  
c:\users\nikee\anaconda3\lib\site-packages (from scikit-learn>=1.0.2->mlxtend)  
(2.2.0)  
Requirement already satisfied: six>=1.5 in c:\users\nikee\anaconda3\lib\site-  
packages (from python-dateutil>=2.7->matplotlib>=3.0.0->mlxtend) (1.16.0)

```
[2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import re
from mlxtend.frequent_patterns import apriori
from mlxtend.frequent_patterns import association_rules
from mlxtend.preprocessing import TransactionEncoder
from mpl_toolkits.mplot3d import Axes3D
import networkx as nx
```

```
[3]: basket = pd.read_csv("Groceries_dataset.csv")
display(basket.head())
```

	Member_number	Date	itemDescription
0	1808	21-07-2015	tropical fruit
1	2552	05-01-2015	whole milk
2	2300	19-09-2015	pip fruit
3	1187	12-12-2015	other vegetables
4	3037	01-02-2015	whole milk

```
[4]: basket.itemDescription = basket.itemDescription.transform(lambda x: [x])
basket = basket.groupby(['Member_number', 'Date']).sum()['itemDescription'].
↳reset_index(drop=True)
```

```
[5]: encoder = TransactionEncoder()
transactions = pd.DataFrame(encoder.fit(basket).transform(basket),
↳columns=encoder.columns_)
display(transactions.head())
```

	Instant food products	UHT-milk	abrasive cleaner	artif. sweetener	\
0	False	False	False	False	
1	False	False	False	False	
2	False	False	False	False	
3	False	False	False	False	
4	False	False	False	False	

	baby cosmetics	bags	baking powder	bathroom cleaner	beef	berries	\
0	False	False	False	False	False	False	
1	False	False	False	False	False	False	
2	False	False	False	False	False	False	
3	False	False	False	False	False	False	
4	False	False	False	False	False	False	

	turkey	vinegar	waffles	whipped/sour cream	whisky	white bread	\
0	False	False	False	False	False	False	
1	False	False	False	False	False	False	
2	False	False	False	False	False	False	
3	False	False	False	False	False	False	
4	False	False	False	False	False	False	

	white wine	whole milk	yogurt	zwieback
0	False	True	True	False
1	False	True	False	False
2	False	False	False	False
3	False	False	False	False
4	False	False	False	False

[5 rows x 167 columns]

```
[6]: frequent_itemsets = apriori(transactions, min_support= 6/len(basket),
    ↪ use_colnames=True, max_len = 2)
rules = association_rules(frequent_itemsets, metric="lift", min_threshold = 1.
    ↪ 5)
display(rules.head())
print("Rules identified: ", len(rules))
```

	antecedents	consequents	antecedent support	\
0	(UHT-milk)	(butter milk)	0.021386	
1	(butter milk)	(UHT-milk)	0.017577	
2	(UHT-milk)	(cream cheese )	0.021386	
3	(cream cheese )	(UHT-milk)	0.023658	
4	(artif. sweetener)	(soda)	0.001938	

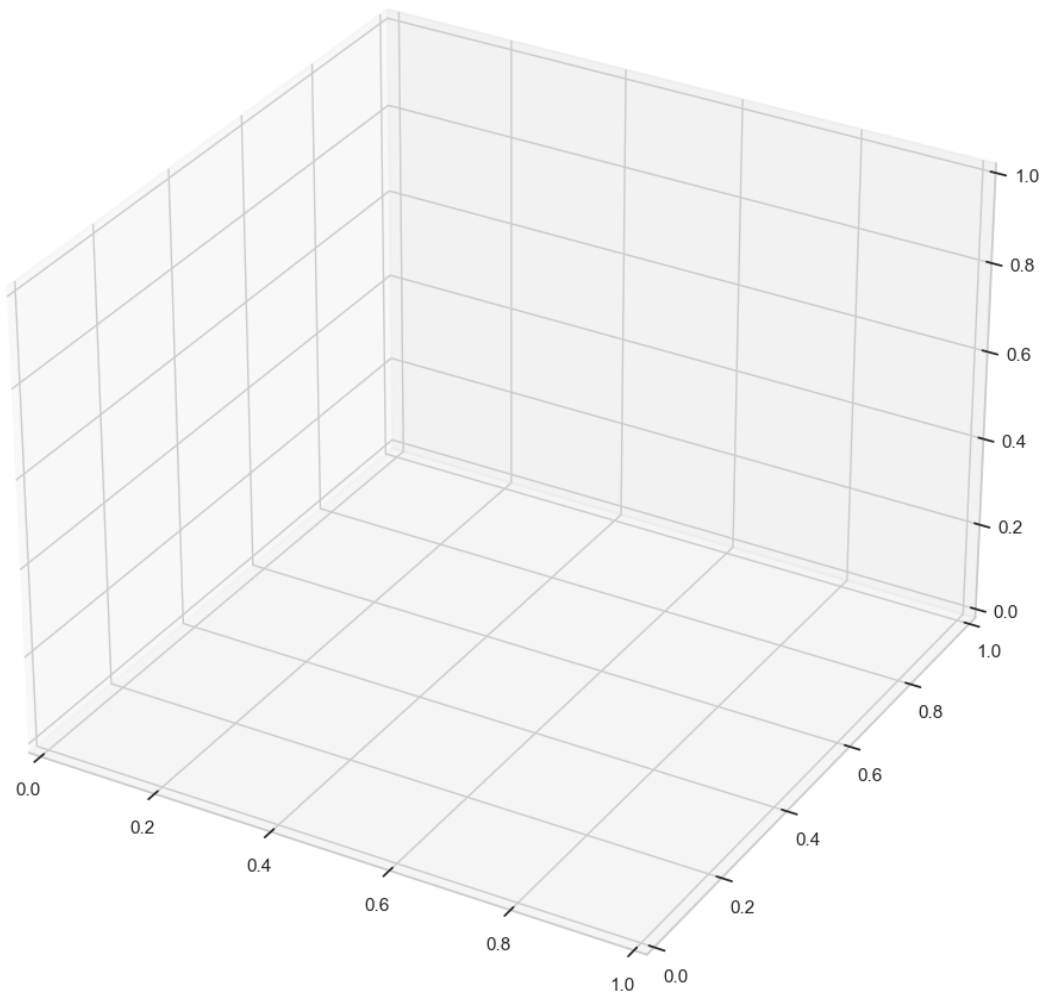
	consequent support	support	confidence	lift	leverage	conviction	\
0	0.017577	0.000601	0.028125	1.600131	0.000226	1.010854	
1	0.021386	0.000601	0.034221	1.600131	0.000226	1.013289	
2	0.023658	0.000869	0.040625	1.717152	0.000363	1.017685	
3	0.021386	0.000869	0.036723	1.717152	0.000363	1.015922	
4	0.097106	0.000468	0.241379	2.485725	0.000280	1.190178	

	zhangs_metric
0	0.383247
1	0.381761
2	0.426767

```
3      0.427761
4      0.598864
```

Rules identified: 190

```
[7]: sns.set(style = "whitegrid")
fig = plt.figure(figsize=(12, 12))
ax = fig.add_subplot(projection = '3d')
```



```
[8]: x = rules['support']
y = rules['confidence']
z = rules['lift']
```

```
[9]: ax.set_xlabel("Support")
ax.set_ylabel("Confidence")
ax.set_zlabel("Lift")
```

```
[9]: Text(0.09107777124005584, 0.012202107626099664, 'Lift')
```

```
[10]: ax.scatter(x, y, z)
ax.set_title("3D Distribution of Association Rules")
```

```
[10]: Text(0.5, 0.92, '3D Distribution of Association Rules')
```

```
[11]: plt.show()
```

```
[20]: def draw_network(rules, rules_to_show):
    # Directional Graph from NetworkX
    network = nx.DiGraph()

    # Loop through number of rules to show
    for i in range(rules_to_show):

        # Add a Rule Node
        network.add_nodes_from(["R"+str(i)])
        for antecedents in rules.iloc[i]['antecedents']:
            # Add antecedent node and link to rule
            network.add_nodes_from([antecedents])
            network.add_edge(antecedents, "R"+str(i), weight = 2)

        for consequents in rules.iloc[i]['consequents']:
            # Add consequent node and link to rule
            network.add_nodes_from([consequents])
            network.add_edge("R"+str(i), consequents, weight = 2)

    color_map=[]

    # For every node, if it's a rule, colour as Black, otherwise Orange
    for node in network:
        if re.compile("^R\d+$").fullmatch(node) != None:
            color_map.append('black')
        else:
            color_map.append('orange')

    # Position nodes using spring layout
    pos = nx.spring_layout(network, k=16, scale=1)
    # Draw the network graph
    nx.draw(network, pos, node_color = color_map, font_size=8)

    # Shift the text position upwards
```

```

for p in pos:
    pos[p][1] += 0.12

nx.draw_networkx_labels(network, pos)
plt.title("Network Graph for Association Rules")
plt.show()

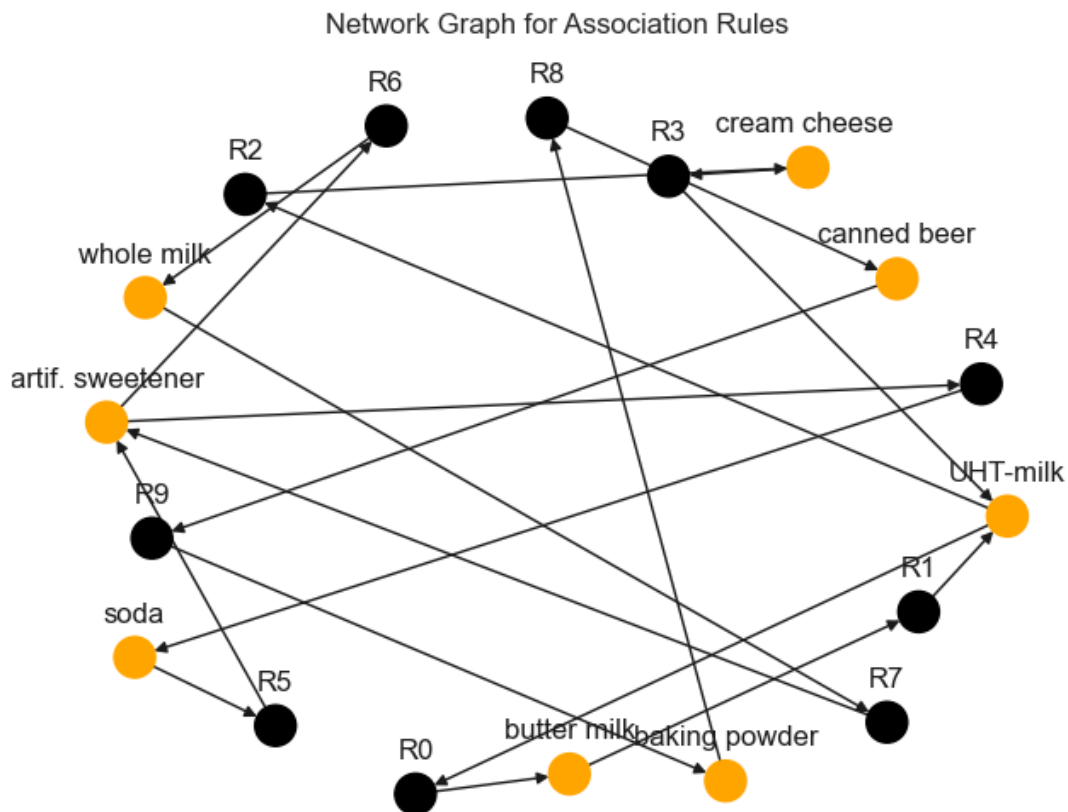
```

```

<>:24: DeprecationWarning: invalid escape sequence '\d'
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C:\Users\nikee\AppData\Local\Temp\ipykernel_44280\34899696.py:24:
DeprecationWarning: invalid escape sequence '\d'
    if re.compile("^([R]\d+$)").fullmatch(node) != None:

```

```
[21]: draw_network(rules, 10)
```



```

[22]: milk_rules = rules[rules['consequents'].astype(str).str.contains('whole milk')]
milk_rules = milk_rules.sort_values(by=['lift'], ascending = [False]).
↪reset_index(drop = True)

```

```
display(milk_rules.head())
```

	antecedents	consequents	antecedent support	consequent support	\
0	(brandy)	(whole milk)	0.002540	0.157923	
1	(softener)	(whole milk)	0.002740	0.157923	
2	(canned fruit)	(whole milk)	0.001403	0.157923	
3	(syrup)	(whole milk)	0.001403	0.157923	
4	(artif. sweetener)	(whole milk)	0.001938	0.157923	

	support	confidence	lift	leverage	conviction	zhangs_metric
0	0.000869	0.342105	2.166281	0.000468	1.279957	0.539750
1	0.000802	0.292683	1.853328	0.000369	1.190523	0.461695
2	0.000401	0.285714	1.809201	0.000179	1.178908	0.447899
3	0.000401	0.285714	1.809201	0.000179	1.178908	0.447899
4	0.000535	0.275862	1.746815	0.000229	1.162868	0.428360

```
[ ]:
```