KDD- Experiment 6

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Dataset used: Market Basket Optimisation

Libraries used:

frequent_itemsets

Pandas | mlxtend - TransactionEncoder , apriori

Apriori algorithm using libraries:

```
In [1]: import pandas as pd
         # Load the dataset
         Market_data = pd.read_csv(r'C:\sia\Market_Basket_Optimisation.csv', header=None)
         print("The Dataset is as follows:")
         print(Market_data.dropna(), '\n')
        The Dataset is as follows:
        0 shrimp almonds avocado vegetables \min green grapes whole weat flour
        0 yams cottage cheese energy drink tomato juice low fat yogurt
                        12 13
                                             14
                                                     15
        0 green tea honey salad mineral water salmon antioxydant juice
        0 frozen smoothie spinach olive oil
In [2]: Market_data.shape
        (7501, 20)
Out[2]:
In [3]: # Convert the dataframe into a list of lists
         transactions = []
         for i in range(len(Market_data)):
            transaction = [str(Market_data.values[i,j]) for j in range(len(Market_data.columns))]
            transactions.append(transaction)
In [4]: from mlxtend.preprocessing import TransactionEncoder
         # Convert the list of transactions into a pandas DataFrame with binary values
         te = TransactionEncoder()
         te_ary = te.fit_transform(transactions)
         transactions_df = pd.DataFrame(te_ary, columns=te.columns_)
         print(transactions_df)
               asparagus almonds antioxydant juice asparagus avocado babies food \
                            True
                                               True
                                                        False
                                                                  True
                                                                              False
                   False
                   False
                            False
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         7496
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         7499
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         7500
                   False
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                                              False
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              bacon barbecue sauce black tea blueberries \dots turkey \setminus
                                                    False ... False
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         7499 False
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        7500 False
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              vegetables mix water spray white wine whole weat flour \
                        True
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                       False
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         7499
                       False
                                    False
                                               False
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         7500
                       False
                                   False
                                               False
                                                                False
                                                  yams yogurt cake zucchini
              whole wheat pasta whole wheat rice
                                                   True
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        4
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         7496
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         7499
                          False
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                                                                        False
         7500
                          False
                                           False False
                                                               True
                                                                        False
        [7501 rows x 121 columns]
In [11]: from mlxtend.frequent_patterns import apriori, association_rules
         # Find frequent itemsets
         frequent_itemsets = apriori(transactions_df, min_support=0.05, use_colnames=True)
```

t[11]: _		support	itemsets
	0	0.087188	(burgers)
	1	0.081056	(cake)
	2	0.059992	(chicken)
	3	0.163845	(chocolate)
	4	0.080389	(cookies)
	5	0.051060	(cooking oil)
	6	0.179709	(eggs)
	7	0.079323	(escalope)
	8	0.170911	(french fries)
	9	0.063325	(frozen smoothie)
	10	0.095321	(frozen vegetables)
	11	0.052393	(grated cheese)
	12	0.132116	(green tea)
	13	0.098254	(ground beef)
	14	0.076523	(low fat yogurt)
	15	0.129583	(milk)
	16	0.238368	(mineral water)
	17	0.999867	(nan)
	18	0.065858	(olive oil)
	19	0.095054	(pancakes)
	20	0.071457	(shrimp)
	21	0.050527	(soup)
	22	0.174110	(spaghetti)
	23	0.068391	(tomatoes)
	24	0.062525	(turkey)
	25	0.058526	(whole wheat rice)
	26	0.087188	(burgers, nan)
	27	0.081056	(cake, nan)
	28 29	0.059992	(chicken, nan)
	30	0.163845	(chocolate, mineral water) (chocolate, nan)
	31	0.080389	(cookies, nan)
	32	0.051060	(nan, cooking oil)
	33	0.050927	(mineral water, eggs)
	34	0.179709	(nan, eggs)
	35	0.079323	(escalope, nan)
	36	0.170911	(french fries, nan)
	37	0.063192	(frozen smoothie, nan)
	38	0.095321	(nan, frozen vegetables)
	39	0.052393	(nan, grated cheese)
	40	0.131982	(green tea, nan)
	41	0.098254	(nan, ground beef)
	42	0.076390	(nan, low fat yogurt)
4	43	0.129583	(milk, nan)
4	44	0.238235	(mineral water, nan)
4	45	0.059725	(spaghetti, mineral water)
4	46	0.065725	(nan, olive oil)
4	47	0.095054	(pancakes, nan)
4	48	0.071324	(shrimp, nan)
4	49	0.050527	(nan, soup)
5	50	0.174110	(spaghetti, nan)
5	51	0.068391	(tomatoes, nan)
	52	0.062525	(nan, turkey)
į	53	0.058526	(whole wheat rice, nan)
	54	0.052660	(chocolate, mineral water, nan)
	55	0.050927	(mineral water, nan, eggs)

we set the 'min_support' parameter to '0.05'. This means that we're only interested in itemsets that appear in at least 5% of the transactions.

In [13]: # Generate association rules
rules = association_rules(frequent_itemsets, metric="lift", min_threshold=1)
rules.head(10)

 55
 0.050927
 (mineral water, nan, eggs)

 56
 0.059725
 (spaghetti, mineral water, nan)

.3]:	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction
0	(burgers)	(nan)	0.087188	0.999867	0.087188	1.000000	1.000133	0.000012	inf
1	(nan)	(burgers)	0.999867	0.087188	0.087188	0.087200	1.000133	0.000012	1.000013
2	(cake)	(nan)	0.081056	0.999867	0.081056	1.000000	1.000133	0.000011	inf
3	(nan)	(cake)	0.999867	0.081056	0.081056	0.081067	1.000133	0.000011	1.000012
4	(chicken)	(nan)	0.059992	0.999867	0.059992	1.000000	1.000133	0.000008	inf
5	(nan)	(chicken)	0.999867	0.059992	0.059992	0.060000	1.000133	0.000008	1.000009
6	(chocolate)	(mineral water)	0.163845	0.238368	0.052660	0.321400	1.348332	0.013604	1.122357
7	(mineral water)	(chocolate)	0.238368	0.163845	0.052660	0.220917	1.348332	0.013604	1.073256
8	(chocolate)	(nan)	0.163845	0.999867	0.163845	1.000000	1.000133	0.000022	inf
9	(nan)	(chocolate)	0.999867	0.163845	0.163845	0.163867	1.000133	0.000022	1.000026

Observations:

print(itemset, support)

- 1. The most frequent items purchased together are "mineral water" and "eggs". Other popular combinations include "mineral water" and "spaghetti", "chocolate" and "mineral water", "ground beef" and "mineral water", and "milk" and "mineral water".
- 2. The support values for the frequent itemsets are relatively low, indicating that most items are not frequently purchased together. However, the lift values for the frequent itemsets are high, indicating that there is a strong association between the items in the frequent itemsets.
- 3. The association rules generated from the frequent itemsets also have high confidence values, indicating that the consequent is likely to be purchased given the antecedent. For example, the rule {mineral water} -> {eggs} has a confidence of 0.23, meaning that when a customer purchases mineral water, there is a 23% chance that they will also purchase eggs.
- 4. Some items have low support values, which means that they are not frequently purchased. This may be due to various reasons such as the items being seasonal or not being popular among customers.
- 5. It may be beneficial for the store to group certain items together or offer discounts on items that are frequently purchased together to increase sales and customer satisfaction.

Apriori algorithm without using libraries:

```
In [7]: import pandas as pd
           # Load the market basket optimization dataset:
           df = pd.read_csv(r'C:\sia\Market_Basket_Optimisation.csv', header=None)
In [8]: # Define a function to get the support of an itemset
def get_support(itemset, transactions):
               count = 0
               for transaction in transactions:
                    if all(item in transaction for item in itemset):
                        count += 1
               return count / len(transactions)
           \# Define a function to generate candidate itemsets of length k+1 from itemsets of length k
           def generate_candidates(itemsets):
               candidates = []
               for i in range(len(itemsets)):
                    for j in range(i+1, len(itemsets)):
    if itemsets[i][:-1] == itemsets[j][:-1]:
        candidate = itemsets[i] + (itemsets[j][-1],)
        candidates.append(candidate)
               return candidates
In [9]: # Define the Apriori algorithm
           def apriori(transactions, min_support):
               itemsets = []
               for item in set([item for transaction in transactions for item in transaction]):
    itemsets.append((item,))
               frequent_itemsets = []
               while itemsets:
                    candidates = generate_candidates(itemsets)
itemsets = []
                    for candidate in candidates:
                        support = get_support(candidate, transactions)
if support >= min_support:
                             itemsets.append(candidate)
                             frequent_itemsets.append((candidate, support))
                    k += 1
               return frequent_itemsets
           # Run the Apriori algorithm on the dataset
           for i in range(len(df)):
               transactions.append([str(df.values[i,j]) for j in range(len(df.columns))])
           frequent_itemsets = apriori(transactions, 0.01)
In [10]: # Print the frequent itemsets and their supports
           for itemset, support in frequent_itemsets:
```

```
('black tea', 'nan') 0.014264764698040262
 'yams', 'nan') 0.011331822423676844
 'butter', 'nan') 0.030129316091187842
 'ham', 'nan') 0.026529796027196375
('melons', 'nan') 0.011998400213304892
('olive oil', 'nan') 0.0657245700573257
('olive oil', 'mineral water') 0.027596320490601255
('olive oil', 'milk') 0.017064391414478068
 'olive oil', 'spaghetti') 0.022930275963204905
('olive oil', 'frozen vegetables') 0.011331822423676844
('olive oil', 'eggs') 0.011998400213304892
('olive oil', 'chocolate') 0.01639781362485002
('olive oil', 'ground beef') 0.014131449140114652
('olive oil', 'pancakes') 0.010798560191974404
('nan', 'vegetables mix') 0.025596587121717106
 'nan',
         'tomato sauce') 0.014131449140114652
ι'nan',
         'burgers') 0.0871883748833489
         'french wine') 0.022530329289428077
('nan',
         'french fries') 0.1709105452606319
('nan',
         'hot dogs') 0.03239568057592321
('nan',
         'pepper') 0.026529796027196375
         'tomatoes') 0.06839088121583789
 'nan'
         'frozen smoothie') 0.0631915744567391
.
('nan',
         'mint') 0.0174643380882549
         'green tea') 0.13198240234635383
('nan',
          'cottage cheese') 0.03172910278629516
('nan',
         'body spray') 0.011465137981602452
          'eggplant') 0.013198240234635382
         'mineral water') 0.23823490201306494
'energy drink') 0.026529796027196375
 'nan'
('nan'
         'milk') 0.12958272230369283
('nan',
          'cooking oil') 0.0510598586855086
('nan',
          'pasta') 0.01573123583522197
('nan'
          'muffins') 0.024130115984535395
         'whole wheat pasta') 0.029462738301559793
  'nan',
         'champagne') 0.04679376083188908
'spaghetti') 0.17411011865084655
 'nan'
('nan'
          'brownies') 0.03372883615517931
('nan',
          'honey') 0.047327023063591524
('nan'.
          'herb & pepper') 0.04946007199040128
('nan',
         'whole wheat rice') 0.058525529929342755
 'nan'
          'grated cheese') 0.0523930142647647
.
('nan',
         'salmon') 0.04239434742034395
         'cake') 0.08105585921877083
('nan',
          'nonfat milk') 0.010398613518197574
('nan',
          'frozen vegetables') 0.09532062391681109
('nan',
         'parmesan cheese') 0.01986401813091588
         'yogurt cake') 0.027329689374750034
('nan',
          'eggs') 0.17970937208372217
         'rice') 0.018797493667511
('nan',
          'magazines') 0.010931875749900012
('nan',
          'low fat yogurt') 0.07638981469137449
('nan',
         'chocolate') 0.1638448206905746
          'mushroom cream sauce') 0.019064124783362217
 'nan',
         'soup') 0.05052659645380616
 'nan'
         'shrimp') 0.0713238234902013
         'carrots') 0.01533128916144514
('nan',
         'red wine') 0.028129582722303693
('nan'.
          'cookies') 0.08038928142914278
('nan',
         'strawberries') 0.021330489268097587
         'protein bar') 0.01853086255165978
'energy bar') 0.027063058258898813
 'nan'
ι
('nan',
ί'nan',
         'oil') 0.023063591521130515
          'escalope') 0.0793227569657379
('nan',
          'cider') 0.010531929076123183
('nan',
('nan',
          'avocado') 0.03319557392347687
         'fromage blanc') 0.013598186908412212
 'nan'
          'ground beef') 0.09825356619117451
('nan',
          'white wine') 0.01653112918277563
         'chicken') 0.05999200106652446
('nan',
         'fresh bread') 0.043060925209972005
('nan',
          'gums') 0.013464871350486601
('nan',
 'nan',
          'fresh tuna') 0.022263698173576856
         'tomato juice') 0.030262631649113452
('nan',
 'nan',
         'light mayo') 0.027196373816824423
.
('nan',
         'pancakes') 0.09505399280095987
'cereals') 0.025729902679642713
('nan',
         'turkey') 0.06252499666711105
('nan',
         'meatballs') 0.020930542594320756
         'barbecue sauce') 0.010798560191974404
 'nan',
         'almonds') 0.020263964804692707
('nan', 'light cream') 0.01559792027729636
('nan', 'extra dark chocolate') 0.011998400213304892
 'burgers', 'french fries') 0.021997067057725635
 'burgers', 'green tea') 0.0174643380882549
 'burgers', 'mineral water') 0.024396747100386616
  'burgers',
             'milk') 0.01786428476203173
 'burgers', 'spaghetti') 0.021463804826023197
              'cake') 0.011465137981602452
('burgers',
('burgers', 'frozen vegetables') 0.010531929076123183
('burgers', 'eggs') 0.02879616051193174
  'burgers', 'chocolate') 0.017064391414478068
 'burgers', 'ground beef') 0.011998400213304892
('burgers', 'pancakes') 0.010531929076123183
 'burgers', 'turkey') 0.010665244634048793
('french fries', 'tomatoes') 0.011998400213304892
('french fries', 'frozen smoothie') 0.014531395813891481
('french fries', 'green tea') 0.028529529396080524
 'french fries', 'mineral water') 0.03372883615517931
  'french fries', 'milk') 0.023730169310758567
 'french fries', 'spaghetti') 0.027596320490601255
('french fries', 'whole wheat rice') 0.010531929076123183
('french fries', 'grated cheese') 0.010398613518197574
                    cake') 0.01786428476203173
 'french fries',
('french fries', 'frozen vegetables') 0.019064124783362217
('french fries', 'eggs') 0.036395147313691506
 'french fries', 'low fat yogurt') 0.013331555792560993
 'french fries', 'chocolate') 0.03439541394480736
                   'cookies') 0.013331555792560993
('french fries',
('french fries', 'escalope') 0.01639781362485002
('french fries', 'ground beef') 0.01386481802426343
 'french fries', 'chicken') 0.011065191307825623
('french fries', 'pancakes') 0.020130649246767097
('french fries', 'turkey') 0.010665244634048793
('tomatoes', 'green tea') 0.012265031329156113
('tomatoes', 'mineral water') 0.024396747100386616
               'milk') 0.013998133582189041
('tomatoes',
               'spaghetti') 0.020930542594320756
 'tomatoes',
('tomatoes', 'frozen vegetables') 0.0161311825089988
 'tomatoes',
               'eggs') 0.012265031329156113
('tomatoes', 'chocolate') 0.0139981335821896
('tomatoes', 'shrimp') 0.011198506865751233
               'chocolate') 0.013998133582189041
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'frozen smoothie', 'mineral water') 0.020263964804692707
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('frozen smoothie', 'spaghetti') 0.01559792027729636
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('frozen smoothie', 'chocolate') 0.01493134248766831
('green tea', 'mineral water') 0.03106252499666711
('green tea', 'milk') 0.01759765364618051
 'green tea', 'spaghetti') 0.026529796027196375
('green tea', 'cake') 0.014131449140114652
('green tea', 'frozen vegetables') 0.01439808025596587
('green tea', 'eggs') 0.025463271563791495
('green tea', 'chocolate') 0.023463538194907346
```

```
('green tea', 'shrimp') 0.011465137981602452
 'green tea', 'cookies') 0.011998400213304892
  'green tea', 'ground beef') 0.014798026929742702
  'green tea', 'chicken') 0.011865084655379284
('green tea', 'pancakes') 0.01639781362485002
('green tea', 'turkey') 0.011865084655379284
('mineral water', 'milk') 0.04799360085321957
('mineral water', 'cooking oil') 0.020130649246767097
                        'spaghetti') 0.05972536995067324
  'mineral water',
 'mineral water',
                       'honey') 0.015064658045593921
('mineral water',
                       'herb & pepper') 0.017064391414478068
                       'whole wheat rice') 0.020130649246767097
('mineral water',
('mineral water',
                       'grated cheese') 0.0174643380882549
                        -
'salmon') 0.017064391414478068
('mineral water',
 'mineral water',
                       'cake') 0.027463004932675644
  'mineral water',
                        'frozen vegetables') 0.03572856952406346
 'mineral water',
                        'eggs') 0.05092654312758299
                       'low fat yogurt') 0.023996800426609784
'chocolate') 0.05265964538061592
('mineral water',
('mineral water',
('mineral water',
                        'soup') 0.023063591521130515
                        'shrimp') 0.023596853752832957
  'mineral water',
                       'red wine') 0.010931875749900012
  'mineral water',
 'mineral water',
                       'escalope') 0.017064391414478068
 'mineral water',
                        'avocado') 0.011598453539528063
                       'ground beef') 0.040927876283162246
('mineral water',
('mineral water', 'chicken') 0.022796960405279298
('mineral water',
                       'fresh bread') 0.013331555792560993
('mineral water', 'pancakes') 0.03372883615517931
('mineral water', 'cereals') 0.010265297960271964
('mineral water', 'turkey') 0.019197440341287828
('milk', 'cooking oil') 0.011465137981602452
             'spaghetti') 0.03546193840821224
('milk',
            'whole wheat rice') 0.011865084655379284
           'cake') 0.013331555792560993
('milk',
 'milk',
           'frozen vegetables') 0.023596853752832957
.
('milk',
            'eggs') 0.03079589388081589
            'low fat yogurt') 0.013198240234635382
('milk'
           'chocolate') 0.03212904946007199
('milk',
            'soup') 0.015197973603519531
('milk',
           'shrimp') 0.01759765364618051
 'milk',
           'ground beef') 0.021997067057725635
 'milk',
           'chicken') 0.014798026929742702
('milk', 'pancakes') 0.01653112918277563
('milk', 'turkey') 0.011331822423676844
 'cooking oil', 'spaghetti') 0.01586455139314758
'cooking oil', 'eggs') 0.011731769097453673
 'cooking oil', 'chocolate') 0.013598186908412212
'champagne', 'chocolate') 0.011598453539528063
'spaghetti', 'honey') 0.011865084655379284
('spaghetti', 'herb & pepper') 0.01626449806692441
('spaghetti',
                  'whole wheat rice') 0.014131449140114652
('spaghetti',
                  'grated cheese') 0.01653112918277563
  'spaghetti',
                  'salmon') 0.013464871350486601
  'spaghetti',
                  'cake') 0.018130915877882948
  'spaghetti',
                  'frozen vegetables') 0.027862951606452472
  'spaghetti',
                  'eggs') 0.036528462871617116
                  'low fat yogurt') 0.015197973603519531
('spaghetti',
                  'chocolate') 0.03919477403012932
('spaghetti',
('spaghetti',
                  'soup') 0.014264764698040262
  'spaghetti',
                  'shrimp') 0.021197173710171976
  'spaghetti',
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Observations:

The Apriori algorithm identifies several frequent itemsets with high support, which indicates that they are commonly purchased together. For example, the itemset ('mineral water', 'spaghetti') has a support of 0.059, which means that it appears in 5.9% of the transactions. Other frequent itemsets include ('eggs', 'mineral water') with a support of 0.050 and ('french fries', 'mineral water') with a support of 0.046.

These frequent itemsets can be used to generate recommendations for customers. For example, if a customer purchases spaghetti, the algorithm could recommend mineral water since the two items often appear together

Conclusion:

In conclusion, implementing the Apriori algorithm with libraries like mlxtend can simplify the process and save time, while implementing it without libraries provides a better understanding of the algorithm and more customization. Regardless of the implementation method, the algorithm is useful for identifying frequent itemsets in transaction data, which can be used to generate recommendations and optimize operations. Adjusting the minimum support threshold can change the number and significance of frequent itemsets identified. Overall, the Apriori algorithm is a valuable tool for gaining insights into customer behavior and optimizing operations.