

Eclat

January 17, 2025

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[ ]: '''  
    Eclat -->  
  
    The ECLAT (Equivalence Class Clustering and Bottom-Up Lattice Traversal)  
    algorithm is a frequent itemset mining algorithm widely used in association  
    rule learning.  
    Unlike the Apriori algorithm, which generates candidate itemsets level  
    by level, ECLAT operates using a vertical data format and is typically  
    faster in scenarios with dense datasets.  
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[ ]: '''  
    Steps in ECLAT Algorithm -->  
  
    Transform Dataset :  
    Convert the dataset into a vertical format (item-TID).  
  
    Generate Frequent Items :  
    Identify items that satisfy the minimum support threshold.  
  
    Combine Items :  
    Use the intersection of TID lists to generate k-itemsets.  
  
    Prune :  
    Discard itemsets that do not meet the minimum support threshold.  
  
    Repeat :  
    Continue until no more frequent itemsets can be generated.  
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| Feature | Apriori | ECLAT |
|-----------------|-----------------------|-----------------------|
| Data Format | Horizontal | Vertical |
| Search Strategy | BFS | DFS |
| Efficiency | Slower for dense data | Faster for dense data |
| Pruning | Apriori property | TID list intersection |
| Memory Use | Scans multiple times | Stores TID lists |

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[ ]: # Eclat uses support so confidence and lift is removed in Apriori approach !
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[1]: # Importing Libraries -->

import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from apyori import apriori
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[3]: # Importing Dataset -->

data = pd.read_csv('Data/Market_Basket_Optimisation.csv', header=None)
data.head(5)
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[3]:
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| | 0 | 1 | 2 | 3 | 4 | \ |
|---|---------------|-----------|------------|------------------|--------------|---|
| 0 | shrimp | almonds | avocado | vegetables mix | green grapes | |
| 1 | burgers | meatballs | eggs | NaN | NaN | |
| 2 | chutney | NaN | NaN | NaN | NaN | |
| 3 | turkey | avocado | NaN | NaN | NaN | |
| 4 | mineral water | milk | energy bar | whole wheat rice | green tea | |

| | 5 | 6 | 7 | 8 | 9 | \ |
|---|------------------|------|----------------|--------------|--------------|---|
| 0 | whole weat flour | yams | cottage cheese | energy drink | tomato juice | |
| 1 | NaN | NaN | NaN | NaN | NaN | |
| 2 | NaN | NaN | NaN | NaN | NaN | |
| 3 | NaN | NaN | NaN | NaN | NaN | |
| 4 | NaN | NaN | NaN | NaN | NaN | |

| | 10 | 11 | 12 | 13 | 14 | 15 | \ |
|---|----------------|-----------|-------|-------|---------------|--------|---|
| 0 | low fat yogurt | green tea | honey | salad | mineral water | salmon | |
| 1 | NaN | NaN | NaN | NaN | NaN | NaN | |
| 2 | NaN | NaN | NaN | NaN | NaN | NaN | |
| 3 | NaN | NaN | NaN | NaN | NaN | NaN | |
| 4 | NaN | NaN | NaN | NaN | NaN | NaN | |

| | 16 | 17 | 18 | 19 |
|---|-------------------|-----------------|---------|-----------|
| 0 | antioxydant juice | frozen smoothie | spinach | olive oil |
| 1 | NaN | NaN | NaN | NaN |
| 2 | NaN | NaN | NaN | NaN |
| 3 | NaN | NaN | NaN | NaN |
| 4 | NaN | NaN | NaN | NaN |

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[4]: # Creating Transaction List -->

transactions = []

for i in range(0,7501):
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transactions.append([str(data.values[i,j]) for j in range(0,20)])
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[5]: # Apriori Rules -->
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rules = apriori(transactions=transactions, min_support=0.003, min_confidence=0.  
↳2, min_lift=3, min_length=2, max_length=2)
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[6]: # Results -->
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results = list(rules)  
results
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[6]: [RelationRecord(items=frozenset({'chicken', 'light cream'}),  
support=0.004532728969470737,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'light cream'}),  
items_add=frozenset({'chicken'}), confidence=0.29059829059829057,  
lift=4.84395061728395)]),  
RelationRecord(items=frozenset({'escalope', 'mushroom cream sauce'}),  
support=0.005732568990801226,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'mushroom cream  
sauce'}), items_add=frozenset({'escalope'}), confidence=0.3006993006993007,  
lift=3.790832696715049)]),  
RelationRecord(items=frozenset({'escalope', 'pasta'}),  
support=0.005865884548726837,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'pasta'}),  
items_add=frozenset({'escalope'}), confidence=0.3728813559322034,  
lift=4.700811850163794)]),  
RelationRecord(items=frozenset({'honey', 'fromage blanc'}),  
support=0.003332888948140248,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'fromage blanc'}),  
items_add=frozenset({'honey'}), confidence=0.2450980392156863,  
lift=5.164270764485569)]),  
RelationRecord(items=frozenset({'ground beef', 'herb & pepper'}),  
support=0.015997866951073192,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'herb & pepper'}),  
items_add=frozenset({'ground beef'}), confidence=0.3234501347708895,  
lift=3.2919938411349285)]),  
RelationRecord(items=frozenset({'ground beef', 'tomato sauce'}),  
support=0.005332622317024397,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'tomato sauce'}),  
items_add=frozenset({'ground beef'}), confidence=0.3773584905660377,  
lift=3.840659481324083)]),  
RelationRecord(items=frozenset({'light cream', 'olive oil'}),  
support=0.003199573390214638,  
ordered_statistics=[OrderedStatistic(items_base=frozenset({'light cream'}),  
items_add=frozenset({'olive oil'}), confidence=0.20512820512820515,  
lift=3.1147098515519573)])]
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RelationRecord(items=frozenset({'whole wheat pasta', 'olive oil'}),
support=0.007998933475536596,
ordered_statistics=[OrderedStatistic(items_base=frozenset({'whole wheat
pasta'}), items_add=frozenset({'olive oil'}), confidence=0.2714932126696833,
lift=4.122410097642296)]],
RelationRecord(items=frozenset({'shrimp', 'pasta'}),
support=0.005065991201173177,
ordered_statistics=[OrderedStatistic(items_base=frozenset({'pasta'}),
items_add=frozenset({'shrimp'}), confidence=0.3220338983050847,
lift=4.506672147735896)]])

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[7]: # Putting Results into DataFrame -->

def inspect(results):
    lhs = [tuple(result[2][0][0])[0] for result in results]
    rhs = [tuple(result[2][0][1])[0] for result in results]
    supports = [result[1] for result in results]
    return list(zip(lhs, rhs, supports))

resultDF = pd.DataFrame(inspect(results), columns=['Left Hand Side', 'Right_
↳Hand Side', 'Support'])
resultDF.head(10)

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[7]:
      Left Hand Side Right Hand Side  Support
0      light cream      chicken  0.004533
1  mushroom cream sauce      escalope  0.005733
2           pasta      escalope  0.005866
3    fromage blanc          honey  0.003333
4    herb & pepper    ground beef  0.015998
5    tomato sauce    ground beef  0.005333
6      light cream      olive oil  0.003200
7  whole wheat pasta      olive oil  0.007999
8           pasta      shrimp  0.005066

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[8]: # Display Results by Descending Support -->

resultDF.nlargest(n=10, columns='Support')

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[8]:
      Left Hand Side Right Hand Side  Support
4    herb & pepper    ground beef  0.015998
7  whole wheat pasta      olive oil  0.007999
2           pasta      escalope  0.005866
1  mushroom cream sauce      escalope  0.005733
5    tomato sauce    ground beef  0.005333
8           pasta      shrimp  0.005066
0      light cream      chicken  0.004533
3    fromage blanc          honey  0.003333

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6 light cream olive oil 0.003200