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import pandas as pd
from mlxtend.frequent_patterns import fpgrowth, association_rules
from mlxtend.preprocessing import TransactionEncoder

# Given transaction set
transactions = [
    [1, 3, 4],
    [2, 3, 5],
    [1, 2, 3, 5],
    [2, 5]
]

# Step 1: One-hot encode the transaction data
te = TransactionEncoder()
te_ary = te.fit(transactions).transform(transactions)
df = pd.DataFrame(te_ary, columns=te.columns_)

# Step 2: Apply FP-Growth algorithm
frequent_itemsets = fpgrowth(df, min_support=0.5, use_colnames=True)

# Step 3: Generate Association Rules
rules = association_rules(frequent_itemsets, metric="confidence", min_threshold=0.75)

# Step 4: Print results
print("★ Frequent Itemsets:")
print(frequent_itemsets)

print("\n★ Association Rules:")
print(rules[['antecedents', 'consequents', 'support', 'confidence', 'lift']])

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➡ ★ Frequent Itemsets:
  support  itemsets
0    0.75      (3)
1    0.50      (1)
2    0.75      (5)
3    0.75      (2)
4    0.50    (3, 5)
5    0.50    (1, 3)
6    0.75    (2, 5)
7    0.50    (2, 3)
8    0.50  (2, 3, 5)

★ Association Rules:
  antecedents consequents support confidence lift
0      (1)      (3)    0.50      1.0  1.333333
1      (2)      (5)    0.75      1.0  1.333333
2      (5)      (2)    0.75      1.0  1.333333
3    (2, 3)      (5)    0.50      1.0  1.333333
4    (3, 5)      (2)    0.50      1.0  1.333333

```

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import pandas as pd
from mlxtend.frequent_patterns import apriori, association_rules

# Define transactions as a list of lists
transactions = [
    [1, 3, 4],
    [2, 3, 5],
    [1, 2, 3, 5],
    [2, 5]
]

# Convert to a one-hot encoded DataFrame
df = pd.DataFrame([[item in transaction for item in range(1, 6)] for transaction in transactions],
                  columns=[1, 2, 3, 4, 5])

# Step 1: Apply Apriori algorithm to find frequent itemsets
frequent_itemsets = apriori(df, min_support=2/4, use_colnames=True) # Minimum support = 2 (out of 4 transactions)

# Step 2: Generate association rules with minimum confidence of 75%

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rules = association_rules(frequent_itemsets, metric="confidence", min_threshold=0.75)
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# Step 3: Drop unwanted columns (lift, leverage, conviction)
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rules = rules.drop(columns=['lift', 'leverage', 'conviction', 'antecedent support', 'consequent support', 'support'])
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# Step 3: Display the results
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```
print("Frequent Itemsets:")
```

```
print(frequent_itemsets)
```

```
print("\nAssociation Rules:")
```

```
print(rules)
```



Frequent Itemsets:

	support	itemsets
0	0.50	(1)
1	0.75	(2)
2	0.75	(3)
3	0.75	(5)
4	0.50	(1, 3)
5	0.50	(2, 3)
6	0.75	(2, 5)
7	0.50	(3, 5)
8	0.50	(2, 3, 5)

Association Rules:

	antecedents	consequents	confidence	representativity	zhangs_metric	\
0	(1)	(3)	1.0	1.0	0.5	
1	(2)	(5)	1.0	1.0	1.0	
2	(5)	(2)	1.0	1.0	1.0	
3	(2, 3)	(5)	1.0	1.0	0.5	
4	(3, 5)	(2)	1.0	1.0	0.5	

	jaccard	certainty	kulczynski
0	0.666667	1.0	0.833333
1	1.000000	1.0	1.000000
2	1.000000	1.0	1.000000
3	0.666667	1.0	0.833333
4	0.666667	1.0	0.833333