```
# ! pip install pandas
# ! pip install matplotlib
! pip install mlxtend
     Show hidden output
! pip install pyvis
→ Collecting pyvis
       Downloading pyvis-0.3.2-py3-none-any.whl.metadata (1.7 kB)
     Requirement already satisfied: ipython>=5.3.0 in /usr/local/lib/python3.11/dist-packages (from pyvis) (7.34.0)
     Requirement already satisfied: jinja2>=2.9.6 in /usr/local/lib/python3.11/dist-packages (from pyvis) (3.1.5)
     Requirement already satisfied: jsonpickle>=1.4.1 in /usr/local/lib/python3.11/dist-packages (from pyvis) (4.0.2)
     Requirement already satisfied: networkx>=1.11 in /usr/local/lib/python3.11/dist-packages (from pyvis) (3.4.2)
     Requirement already satisfied: setuptools>=18.5 in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (75.1.0)
     Collecting jedi>=0.16 (from ipython>=5.3.0->pyvis)
       Downloading jedi-0.19.2-py2.py3-none-any.whl.metadata (22 kB)
     Requirement already satisfied: decorator in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (4.4.2)
     Requirement already satisfied: pickleshare in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (0.7.5)
     Requirement already satisfied: traitlets>=4.2 in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (5.7.1)
     Requirement already satisfied: prompt-toolkit!=3.0.0,!=3.0.1,<3.1.0,>=2.0.0 in /usr/local/lib/python3.11/dist-packages (from ipython>=5.
     Requirement already satisfied: pygments in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (2.18.0)
     Requirement already satisfied: backcall in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (0.2.0)
     Requirement already satisfied: matplotlib-inline in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (0.1.7)
     Requirement already satisfied: pexpect>4.3 in /usr/local/lib/python3.11/dist-packages (from ipython>=5.3.0->pyvis) (4.9.0)
     Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.11/dist-packages (from jinja2>=2.9.6->pyvis) (3.0.2)
     Requirement already satisfied: parso<0.9.0,>=0.8.4 in /usr/local/lib/python3.11/dist-packages (from jedi>=0.16->ipython>=5.3.0->pyvis) (
     Requirement already satisfied: ptyprocess>=0.5 in /usr/local/lib/python3.11/dist-packages (from pexpect>4.3->ipython>=5.3.0->pyvis) (0.7
     Requirement already satisfied: wcwidth in /usr/local/lib/python3.11/dist-packages (from prompt-toolkit!=3.0.0,!=3.0.1,<3.1.0,>=2.0.0->ir
     Downloading pyvis-0.3.2-py3-none-any.whl (756 kB)
                                                 756.0/756.0 kB 18.8 MB/s eta 0:00:00
     Downloading jedi-0.19.2-py2.py3-none-any.whl (1.6 MB)
                                                 1.6/1.6 MB 34.8 MB/s eta 0:00:00
     Installing collected packages: jedi, pyvis
     Successfully installed jedi-0.19.2 pyvis-0.3.2
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from mlxtend.frequent_patterns import apriori, association_rules
from mlxtend.preprocessing import TransactionEncoder
from pyvis.network import Network
import datetime as dt
import numpy as npro
df = pd.read csv("https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBM-GPXX0VY2EN/bread%20basket.csv")
df
```

		Transaction	Item	date_time	period_day	weekday_weekend
	0	1	Bread	30-10-2016 09:58	morning	weekend
	1	2	Scandinavian	30-10-2016 10:05	morning	weekend
	2	2	Scandinavian	30-10-2016 10:05	morning	weekend
	3	3	Hot chocolate	30-10-2016 10:07	morning	weekend
	4	3	Jam	30-10-2016 10:07	morning	weekend
	20502	9682	Coffee	09-04-2017 14:32	afternoon	weekend
	20503	9682	Tea	09-04-2017 14:32	afternoon	weekend
	20504	9683	Coffee	09-04-2017 14:57	afternoon	weekend
	20505	9683	Pastry	09-04-2017 14:57	afternoon	weekend
	20506	9684	Smoothies	09-04-2017 15:04	afternoon	weekend
2	20507 rc	ows × 5 columns				

df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20507 entries, 0 to 20506

```
Data columns (total 5 columns):
                   Non-Null Count Dtype
     #
        Column
         Transaction
     0
                         20507 non-null int64
                        20507 non-null object
     1
         Item
                      20507 non-null object
       date_time
        period day
                         20507 non-null object
     4 weekday_weekend 20507 non-null object
    dtypes: int64(1), object(4)
    memory usage: 801.2+ KB
df['date_time']=pd.to_datetime(df['date_time'])
df.info()
<<class 'pandas.core.frame.DataFrame'>
    RangeIndex: 20507 entries, 0 to 20506
    Data columns (total 5 columns):
     # Column
                         Non-Null Count Dtype
                          -----
                         20507 non-null int64
     0 Transaction
     1
         Item
                         20507 non-null object
     2 date_time
                         20507 non-null datetime64[ns]
     3 period_day
                         20507 non-null object
     4 weekday_weekend 20507 non-null object
    dtypes: datetime64[ns](1), int64(1), object(3)
    memory usage: 801.2+ KB
    <ipython-input-8-7ae6461dde0f>:1: UserWarning: Parsing dates in %d-%m-%Y %H:%M format when dayfirst=False (the default) was specified. F
      df['date_time']=pd.to_datetime(df['date_time'])
df['time']=df['date_time'].dt.time
df['hour']=df['date_time'].dt.hour
df['month'] = df['date_time'].dt.month
df['month name'] = df['month'].replace([1,2,3,4,5,6,7,8,9,10,11,12],['January','February','March','April','May','June','July','August','Sept
df['day'] = df['date_time'].dt.day
df['weekday'] = df['date_time'].dt.weekday
 df['weekday 'name'] = df['weekday'].replace([0,1,2,3,4,5,6], ['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Sunday']) 
df
```

∑₹

-		Transaction	Item	date_time	period_day	weekday_weekend	time	hour	month	month name	day	weekday	weekday name
	0	1	Bread	2016-10-30 09:58:00	morning	weekend	09:58:00	9	10	October	30	6	Sunday
	1	2	Scandinavian	2016-10-30 10:05:00	morning	weekend	10:05:00	10	10	October	30	6	Sunday
	2	2	Scandinavian	2016-10-30 10:05:00	morning	weekend	10:05:00	10	10	October	30	6	Sunday
	3	3	Hot chocolate	2016-10-30 10:07:00	morning	weekend	10:07:00	10	10	October	30	6	Sunday
	4	3	Jam	2016-10-30 10:07:00	morning	weekend	10:07:00	10	10	October	30	6	Sunday
	20502	9682	Coffee	2017-04-09 14:32:00	afternoon	weekend	14:32:00	14	4	April	9	6	Sunday
:	20503	9682	Tea	2017-04-09 14:32:00	afternoon	weekend	14:32:00	14	4	April	9	6	Sunday

df.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20507 entries, 0 to 20506
Data columns (total 12 columns):
# Column Non-Null Count Dtype
```

Column Non-Null Count Dtype
-----0 Transaction 20507 non-null int64
1 Item 20507 non-null object
2 date_time 20507 non-null datetime64[ns]

3 period_day 20507 non-null object 4 weekday_weekend 20507 non-null object

```
5
                      20507 non-null object
    time
                     20507 non-null int32
6
    hour
    month
                     20507 non-null int32
                     20507 non-null object
20507 non-null int32
8
    month name
9 day
10 weekday
                      20507 non-null int32
11 weekday name
                     20507 non-null object
dtypes: datetime64[ns](1), int32(4), int64(1), object(6)
memory usage: 1.6+ MB
```

```
popular = df['Item'].value_counts()
(df['Item'].value_counts(normalize=True)*100).head(20)
```



proportion

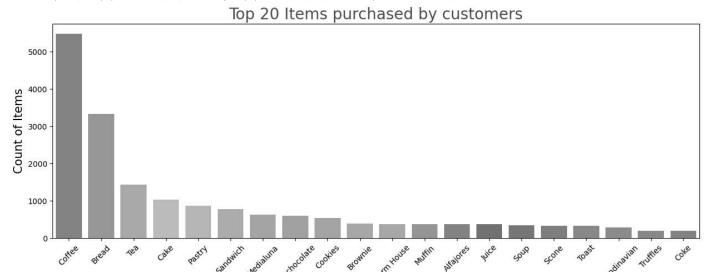
Item	
Coffee	26.678695
Bread	16.213976
Tea	6.997611
Cake	4.998293
Pastry	4.174184
Sandwich	3.759692
Medialuna	3.003852
Hot chocolate	2.877066
Cookies	2.633247
Brownie	1.848149
Farm House	1.823767
Muffin	1.804262
Alfajores	1.799386
Juice	1.799386
Soup	1.667723
Scone	1.594577
Toast	1.550690
Scandinavian	1.350758
Truffles	0.941142
Coke	0.902131

dtype: float64

```
plt.figure(figsize=(15,5))
sns.barplot(x = popular.head(20).index, y = popular.head(20).values, palette = 'hls')
plt.xlabel('Items', size = 15)
plt.xticks(rotation=45)
plt.ylabel('Count of Items', size = 15)
plt.title('Top 20 Items purchased by customers', color = 'red', size = 20)
plt.show()
```

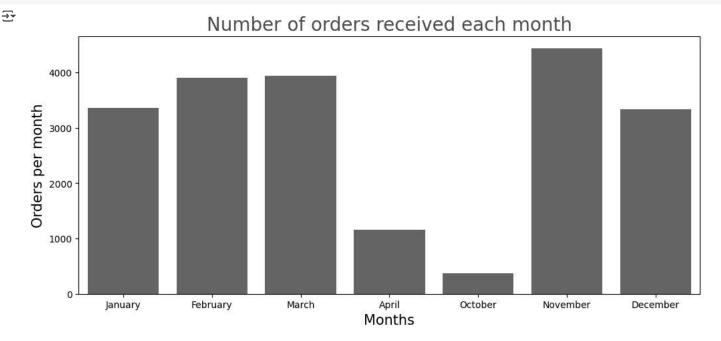
<ipython-input-15-62353c6d8660>:2: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `legenc sns.barplot(x = popular.head(20).index, y = popular.head(20).values, palette = 'hls')



Items

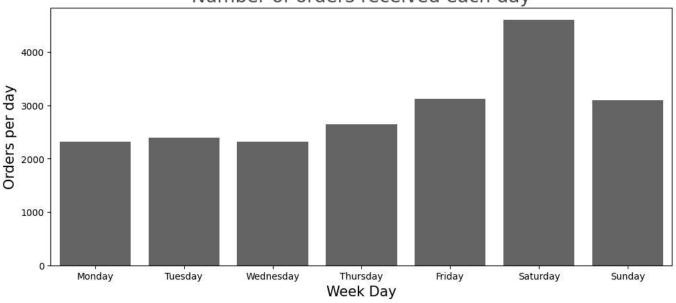
```
monthTran = df.groupby(['month','month name'])['Transaction'].count().reset_index()
plt.figure(figsize=(12,5))
sns.barplot(data = monthTran[['month name', 'Transaction']], x = "month name", y = "Transaction")
plt.xlabel('Months', size = 15)
plt.ylabel('Orders per month', size = 15)
plt.title('Number of orders received each month', color = 'red', size = 20)
plt.show()
```



```
weekTran = df.groupby(['weekday','weekday name'])['Transaction'].count().reset_index()
plt.figure(figsize=(12,5))
sns.barplot(data = weekTran[['weekday name', 'Transaction']], x = "weekday name", y = "Transaction")
plt.xlabel('Week Day', size = 15)
plt.ylabel('Orders per day', size = 15)
plt.title('Number of orders received each day', color = 'red', size = 20)
plt.show()
```



Number of orders received each day

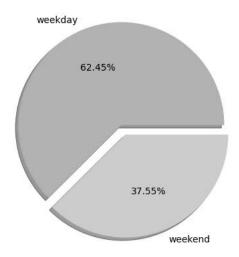


```
size = df['weekday_weekend'].value_counts()
labels = size.index.values
colors = ["cyan", "lightblue"]
explode = [0, 0.1]

plt.figure(figsize=(12,5))
plt.pie(size, labels = labels, colors = colors, explode = explode, shadow = True, autopct = "%.2f%%")
plt.title('Transaction by week period')
plt.show()
```

$\overline{\mathbf{x}}$

Transaction by week period

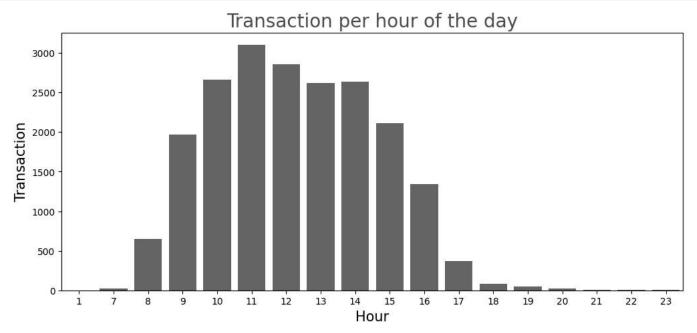


```
coutbyhour=df.groupby('hour')['Transaction'].count().reset_index()
coutbyhour.sort_values('hour',inplace=True)

plt.figure(figsize=(12,5))
sns.barplot(data=coutbyhour, x='hour', y='Transaction')
plt.xlabel('Hour', size = 15)
plt.ylabel('Transaction', size = 15)
```

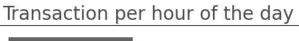
plt.title('Transaction per hour of the day', color = 'red', size = 20) plt.show()

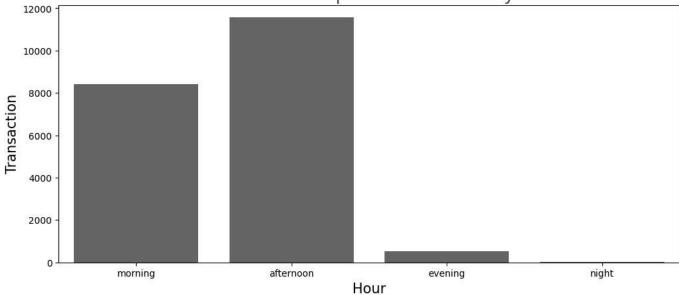




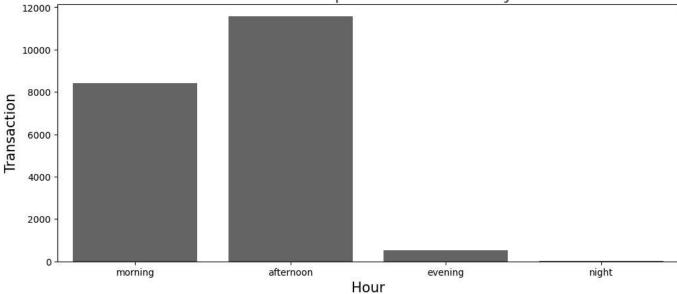
```
coutbyweekday=df.groupby('period_day')['Transaction'].count().reset_index()
coutbyweekday.loc[:,"dayorder"] = [1, 2, 0, 3]
coutbyweekday.sort_values("dayorder",inplace=True)
plt.figure(figsize=(12,5))
sns.barplot(data=coutbyweekday, x='period_day', y='Transaction')
plt.xlabel('Hour', size = 15)
plt.ylabel('Transaction', size = 15)
plt.title('Transaction per hour of the day', color = 'red', size = 20)
plt.show()
plt.figure(figsize=(12,5))
sns.barplot(data=coutbyweekday, x='period_day', y='Transaction')
plt.xlabel('Hour', size = 15)
plt.ylabel('Transaction', size = 15)
plt.title('Transaction per hour of the day', color = 'red', size = 20)
plt.show()
```











transactions = df.groupby(['Transaction', 'Item'])['Item'].count().reset_index(name ='Count') transactions

	Transaction	Item	Count
0	1	Bread	1
1	2	Scandinavian	2
2	3	Cookies	1
3	3	Hot chocolate	1
4	3	Jam	1
18882	9682	Tacos/Fajita	1
18883	9682	Tea	1
18884	9683	Coffee	1
18885	9683	Pastry	1
18886	9684	Smoothies	1

18887 rows × 3 columns

```
# Assuming 'transactions' DataFrame from your previous code holds transaction data
# Create a pivot table to represent the basket
basket = transactions.pivot_table(index='Transaction',
                                   columns='Item',
                                   values='Count',
                                   aggfunc='sum',
                                   fill_value=0)
def encode_units(x):
    # Indent the code within the function
    if(x==0):
        return False
    if(x>0):
        return True
# Now you can use the 'basket' variable, which is defined outside the function
# For example, you can apply the 'encode_units' function to the 'basket' DataFrame
basket_encoded = basket.applymap(encode_units)
🚁 <ipython-input-25-77fe47ff243f>:19: FutureWarning: DataFrame.applymap has been deprecated. Use DataFrame.map instead.
       basket_encoded = basket.applymap(encode_units)
transactions=[]
for item in df['Transaction'].unique():
    lst=list(set(df[df['Transaction']==item]['Item']))
    transactions.append(lst)
transactions[0:10]
→ [['Bread'],
       ['Scandinavian'],
      ['Cookies', 'Hot chocolate', 'Jam'],
      ['Muffin'],
['Bread', 'Coffee', 'Pastry'],
      ['Medialuna', 'Muffin', 'Pastry'],
['Coffee', 'Medialuna', 'Tea', 'Pastry'],
['Bread', 'Pastry'],
      ['Bread', 'Muffin'],
      ['Medialuna', 'Scandinavian']]
# Cell ipython-input-51-c9056906959f
# This line assigns a method object to transactions
# transactions = df.groupby(['Transaction', 'Item'])['Item'].count().reset_index
# Instead, keep the list of transactions from cell ipython-input-46-56c65ffc2649
# transactions should be the list of lists representing the items in each transaction
# Example: transactions = [['Bread']]
# Use the list of transactions created earlier in the code in cell ipython-input-46-56c65ffc2649
transactions = []
for item in df['Transaction'].unique():
    lst = list(set(df[df['Transaction'] == item]['Item']))
```

```
transactions.append(lst)
# Now, 'transactions' is the list of lists needed for TransactionEncoder
# Continue with the encoding process
te = TransactionEncoder()
encodedData = te.fit(transactions).transform(transactions)
basket_sets_2 = pd.DataFrame(encodedData, columns=te.columns_)
basket_sets_2
```



	Adjustment	Afternoon with the baker	Alfajores	Argentina Night	Art Tray	Bacon	Baguette	Bakewell	Bare Popcorn	Basket	 The BART	The Nomad	Tiffin	Toast	Trı
0	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
1	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
2	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
3	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
4	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
9460	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
9461	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
9462	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
9463	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	
9464	False	False	False	False	False	False	False	False	False	False	 False	False	False	False	

9465 rows × 94 columns

Use the original transactions DataFrame from cell ipython-input-40-f57728cee10d transactions = df.groupby(['Transaction', 'Item'])['Item'].count().reset_index

Use the list of transactions created earlier in the code in cell ipython-input-46-56c65ffc2649 transactions = [] for item in df['Transaction'].unique():

lst = list(set(df[df['Transaction'] == item]['Item']))

transactions.append(lst)

Now, 'transactions' is the list of lists needed for TransactionEncoder

Continue with the encoding process

te = TransactionEncoder()

encodedData = te.fit(transactions).transform(transactions)

basket_sets = pd.DataFrame(encodedData, columns=te.columns_) # Assign to basket_sets instead of basket_sets_2 basket_sets



	Adjustment	Afternoon with the baker	Alfajores	Argentina Night	Art Tray	Bacon	Baguette	Bakewell	Bare Popcorn	Basket	•••	The BART	The Nomad	Tiffin	Toast	Tr
0	False	False	False	False	False	False	False	False	False	False		False	False	False	False	
1	False	False	False	False	False	False	False	False	False	False		False	False	False	False	
2	False	False	False	False	False	False	False	False	False	False		False	False	False	False	
3	False	False	False	False	False	False	False	False	False	False		False	False	False	False	
4	False	False	False	False	False	False	False	False	False	False		False	False	False	False	
946	0 False	False	False	False	False	False	False	False	False	False		False	False	False	False	
946	1 False	False	False	False	False	False	False	False	False	False		False	False	False	False	
946	2 False	False	False	False	False	False	False	False	False	False		False	False	False	False	
946	3 False	False	False	False	False	False	False	False	False	False		False	False	False	False	
946	4 False	False	False	False	False	False	False	False	False	False		False	False	False	False	
0405	rows v 04 solum															

9465 rows × 94 columns

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from mlxtend.frequent_patterns import apriori, association_rules
from mlxtend.preprocessing import TransactionEncoder
from pyvis.network import Network
import datetime as dt
import numpy as npro
\# ... (rest of your code) ...
# Use the list of transactions created earlier in the code in cell ipython-input-46-56c65ffc2649
transactions = []
for item in df['Transaction'].unique():
    lst = list(set(df[df['Transaction'] == item]['Item']))
    transactions.append(lst)
\mbox{\# Now, 'transactions'} is the list of lists needed for \mbox{TransactionEncoder}
# Continue with the encoding process
te = TransactionEncoder()
encodedData = te.fit(transactions).transform(transactions)
basket\_sets = pd.DataFrame(encodedData, columns=te.columns\_) \ \# \ Assign \ to \ basket\_sets \ instead \ of \ basket\_sets\_2
# Calculate frequent itemsets using apriori BEFORE association_rules
frequentItems = apriori(basket_sets, min_support=0.01, use_colnames=True) # Adjust min_support as needed
rules = association_rules(frequentItems, metric="confidence", min_threshold=0.2)
rules.sort_values('confidence', ascending = False, inplace=True)
rules
```



	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	representativity	leverage	conviction	zhangs_metr:
24	(Toast)	(Coffee)	0.033597	0.478394	0.023666	0.704403	1.472431	1.0	0.007593	1.764582	0.3320
22	(Spanish Brunch)	(Coffee)	0.018172	0.478394	0.010882	0.598837	1.251766	1.0	0.002189	1.300235	0.2048
16	(Medialuna)	(Coffee)	0.061807	0.478394	0.035182	0.569231	1.189878	1.0	0.005614	1.210871	0.1700
18	(Pastry)	(Coffee)	0.086107	0.478394	0.047544	0.552147	1.154168	1.0	0.006351	1.164682	0.1461
1	(Alfajores)	(Coffee)	0.036344	0.478394	0.019651	0.540698	1.130235	1.0	0.002264	1.135648	0.1195
15	(Juice)	(Coffee)	0.038563	0.478394	0.020602	0.534247	1.116750	1.0	0.002154	1.119919	0.1087
19	(Sandwich)	(Coffee)	0.071844	0.478394	0.038246	0.532353	1.112792	1.0	0.003877	1.115384	0.1092
11	(Cake)	(Coffee)	0.103856	0.478394	0.054728	0.526958	1.101515	1.0	0.005044	1.102664	0.1028
20	(Scone)	(Coffee)	0.034548	0.478394	0.018067	0.522936	1.093107	1.0	0.001539	1.093366	0.0882
13	(Cookies)	(Coffee)	0.054411	0.478394	0.028209	0.518447	1.083723	1.0	0.002179	1.083174	0.0817
14	(Hot chocolate)	(Coffee)	0.058320	0.478394	0.029583	0.507246	1.060311	1.0	0.001683	1.058553	0.0604
10	(Brownie)	(Coffee)	0.040042	0.478394	0.019651	0.490765	1.025860	1.0	0.000495	1.024293	0.0262
17	(Muffin)	(Coffee)	0.038457	0.478394	0.018806	0.489011	1.022193	1.0	0.000408	1.020777	0.0225
21	(Soup)	(Coffee)	0.034443	0.478394	0.015848	0.460123	0.961807	1.0	-0.000629	0.966156	-0.0395
26	(Bread, Cake)	(Coffee)	0.023349	0.478394	0.010037	0.429864	0.898557	1.0	-0.001133	0.914880	-0.1036
29	(Cake, Tea)	(Coffee)	0.023772	0.478394	0.010037	0.422222	0.882582	1.0	-0.001335	0.902779	-0.1199
27	(Bread, Pastry)	(Coffee)	0.029160	0.478394	0.011199	0.384058	0.802807	1.0	-0.002751	0.846843	-0.2019:
23	(Tea)	(Coffee)	0.142631	0.478394	0.049868	0.349630	0.730840	1.0	-0.018366	0.802014	-0.3004
8	(Pastry)	(Bread)	0.086107	0.327205	0.029160	0.338650	1.034977	1.0	0.000985	1.017305	0.0369
0	(Alfajores)	(Bread)	0.036344	0.327205	0.010354	0.284884	0.870657	1.0	-0.001538	0.940818	-0.1335
4	(Bread)	(Coffee)	0.327205	0.478394	0.090016	0.275105	0.575059	1.0	-0.066517	0.719561	-0.5234
7	(Medialuna)	(Bread)	0.061807	0.327205	0.016904	0.273504	0.835879	1.0	-0.003319	0.926082	-0.1730
2	(Brownie)	(Bread)	0.040042	0.327205	0.010777	0.269129	0.822508	1.0	-0.002326	0.920538	-0.1835
5	(Cookies)	(Bread)	0.054411	0.327205	0.014474	0.266019	0.813004	1.0	-0.003329	0.916638	-0.1956
9	(Sandwich)	(Bread)	0.071844	0.327205	0.017010	0.236765	0.723596	1.0	-0.006498	0.881503	-0.2915
28	(Coffee, Pastry)	(Bread)	0.047544	0.327205	0.011199	0.235556	0.719901	1.0	-0.004357	0.880109	-0.2900;
6	(Hot chocolate)	(Bread)	0.058320	0.327205	0.013418	0.230072	0.703144	1.0	-0.005665	0.873841	-0.3095
12	(Cake)	(Tea)	0.103856	0.142631	0.023772	0.228891	1.604781	1.0	0.008959	1.111865	0.4205
3	(Cake)	(Bread)	0.103856	0.327205	0.023349	0.224822	0.687097	1.0	-0.010633	0.867923	-0.3369
30	(Tea, Coffee)	(Cake)	0.049868	0.103856	0.010037	0.201271	1.937977	1.0	0.004858	1.121962	0.5094
25	(Sandwich)	(Tea)	0.071844	0.142631	0.014369	0.200000	1.402222	1.0	0.004122	1.071712	0.3090
←											•

[#] Cell ipython-input-65-017c11814289

rules[rules['antecedents'] == frozenset({'Cake'})]

[#] Ensure that the cell where 'rules' is calculated (ipython-input-66-017c11814289) has been run before this cell.

import pandas as pd

import matplotlib.pyplot as plt

import seaborn as sns

 $^{{\}tt from \ mlxtend.frequent_patterns \ import \ apriori}$



	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	representativity	leverage	conviction	zhangs_metr
11	(Cake)	(Coffee)	0.103856	0.478394	0.054728	0.526958	1.101515	1.0	0.005044	1.102664	0.1028
12	(Cake)	(Tea)	0.103856	0.142631	0.023772	0.228891	1.604781	1.0	0.008959	1.111865	0.4205
3	(Cake)	(Bread)	0.103856	0.327205	0.023349	0.224822	0.687097	1.0	-0.010633	0.867923	-0.3369
4 =											

frequentItems["antecedent_len"] = frequentItems["itemsets"].apply(lambda x: len(x))
frequentItems[frequentItems["antecedent_len"]>1].sort_values(by=["antecedent_len","support"], ascending=False)

		support	itemsets	antecedent_len
	59	0.011199	(Bread, Coffee, Pastry)	3
	58	0.010037	(Bread, Cake, Coffee)	3
	60	0.010037	(Cake, Tea, Coffee)	3
	34	0.090016	(Bread, Coffee)	2
	42	0.054728	(Cake, Coffee)	2
	55	0.049868	(Tea, Coffee)	2
	50	0.047544	(Coffee, Pastry)	2
	51	0.038246	(Sandwich, Coffee)	2
	48	0.035182	(Medialuna, Coffee)	2
	46	0.029583	(Hot chocolate, Coffee)	2
	38	0.029160	(Bread, Pastry)	2
	45	0.028209	(Cookies, Coffee)	2
	40	0.028104	(Bread, Tea)	2
	44	0.023772	(Cake, Tea)	2
	56	0.023666	(Toast, Coffee)	2
	33	0.023349	(Bread, Cake)	2
	47	0.020602	(Juice, Coffee)	2
	31	0.019651	(Alfajores, Coffee)	2
	41	0.019651	(Brownie, Coffee)	2
	49	0.018806	(Muffin, Coffee)	2
	52	0.018067	(Scone, Coffee)	2
	39	0.017010	(Sandwich, Bread)	2
	37	0.016904	(Bread, Medialuna)	2
	53	0.015848	(Soup, Coffee)	2
	35	0.014474	(Bread, Cookies)	2
	57	0.014369	(Sandwich, Tea)	2
	36	0.013418	(Bread, Hot chocolate)	2
	43	0.011410	(Cake, Hot chocolate)	2
	54	0.010882	(Spanish Brunch, Coffee)	2
	32	0.010777	(Bread, Brownie)	2
	30	0.010354	(Alfajores, Bread)	2

```
Basket_Network = Network(height="1000px", width="1000px", directed=True, notebook=True)

Warning: When cdn_resources is 'local' jupyter notebook has issues displaying graphics on chrome/safari. Use cdn_resources='in_line' or Basket_Network.force_atlas_2based()
```

Basket_Network.barnes_hut()

```
# Basket_Network.hrepulsion()
Basket_Network.repulsion()
```

```
Basket_Network_Data_zip=zip(rules["antecedents"],
                            rules["consequents"],
                            rules["antecedent support"],
                            rules["consequent support"],
                            rules["confidence"])
for i in Basket Network Data zip:
    From Item = str(i[0]).replace("frozenset({\displayset}, "","").replace("'))","").replace("', '",",")
   ToItem=str(i[1]).replace("frozenset({'","").replace("'})","").replace("',
   FromWeight=i[2]
    ToWeight=i[3]
   EdgeWeight=i[4]
   Basket_Network.add_node(n_id=FromItem, shape="dot", value=FromWeight,
                            title=FromItem + "<br>Support: " + str(FromWeight))
    Basket_Network.add_node(n_id=ToItem, shape="dot", value=ToWeight,
                           title=ToItem + "<br>>Support: " + str(ToWeight))
    Basket_Network.add_edge(source=FromItem, to=ToItem, value=EdgeWeight, arrowStrikethrough=False,
                            title=FromItem + " --> " + ToItem + "<br>Confidence:" + str(EdgeWeight))
```

```
Basket_Network.set_edge_smooth(smooth_type="continuous")
Basket_Network.toggle_hide_edges_on_drag(True)
```

```
Basket_Network.save_graph("Basket_Network1.html")
Basket_Network.show("Basket_Network1.html")
```

Show hidden output

```
Basket_Network2 = Network(height="1000px", width="1000px", directed=True, notebook=True)
Basket_Network2.repulsion()
Basket_Network_Data2_zip = [] # Replace ##YOUR CODE GOES HERE## with an empty list or appropriate initialization
for i in Basket_Network_Data2_zip:
       From Item=str(i[0]).replace("frozenset({'",""}).replace("')","").replace("', '",",")
       ToItem=str(i[1]).replace("frozenset({'","").replace("'})","").replace("', '",",")
       FromWeight=i[2]
       ToWeight=i[3]
       EdgeWeight=i[4]
       Basket_Network2.add_node(n_id=FromItem, shape="dot", value=FromWeight,
                                                        title=FromItem + "<br>>Support: " + str(FromWeight))
       Basket_Network2.add_node(n_id=ToItem, shape="dot", value=ToWeight,
                                                      title=ToItem + "<br>>Support: " + str(ToWeight))
        Basket_Network2.add_edge(source=FromItem, to=ToItem, value=EdgeWeight, arrowStrikethrough=False,
                                                        title=FromItem + " --> " + ToItem + "<br>Confidence:" + str(EdgeWeight))
Basket_Network2.set_edge_smooth(smooth_type="continuous")
Basket_Network2.toggle_hide_edges_on_drag(True)
Basket_Network2.save_graph("Basket_Network2.html")
Basket_Network2.show("Basket_Network2.html")
for i in Basket_Network_Data2_zip:
       \label{lem:fromItem} From Item=str(i[0]).replace("frozenset({'',"}").replace("'),"").replace("',"",",")
        ToItem=str(i[1]).replace("frozenset({'","").replace("'})","").replace("', '",",")
        FromWeight=i[2]
       ToWeight=i[3]
       EdgeWeight=i[4]
       Basket_Network2.add_node(n_id=FromItem, shape="dot", value=FromWeight,
                                                         title=FromItem + "<br>Support: " + str(FromWeight))
       Basket_Network2.add_node(n_id=ToItem, shape="dot", value=ToWeight,
                                                      title=ToItem + "<br>Support: " + str(ToWeight))
       Basket\_Network 2. add\_edge (source=From Item, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ value=Edge Weight, \ arrow Strike through=False, \ to=ToItem, \ to=ToI
                                                        title=FromItem + " --> " + ToItem + "<br>Confidence:" + str(EdgeWeight))
Basket_Network2.set_edge_smooth(smooth_type="continuous")
Basket_Network2.toggle_hide_edges_on_drag(True)
Basket_Network2.save_graph("Basket_Network2.html")
Basket Network2.show("Basket Network2.html")
```

₹	Warning: When cdn_resources is 'local' jupyter notebook has issues displaying graphics on chrome/safari. Use cdn_resources='in_li Basket_Network2.html Basket_Network2.html	ine'

or