import pandas as pd !pip install apyori from apyori import apriori

→ Collecting apyori

Downloading apyori-1.1.2.tar.gz (8.6 kB)

Preparing metadata (setup.py)  $\dots$  done

Building wheels for collected packages: apyori Building wheel for apyori (setup.py) ... done

 $\label{lem:condition} \textbf{Created wheel for apyori: filename=apyori-1.1.2-py3-none-any.whl size=5954 sha256=12a49e20ac7b4129dba0fe39d4886e1ebc17801f30e9b147beef} \\$ 

Stored in directory: /root/.cache/pip/wheels/c4/1a/79/20f55c470a50bb3702a8cb7c94d8ada15573538c7f4baebe2d

Successfully built apyori

Installing collected packages: apyori Successfully installed apyori-1.1.2

data=pd.read\_csv("Market\_Basket\_Optimisation.csv",header=None)

data.head(10)

<del>_</del>		0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	1
	0	shrimp	almonds	avocado	vegetables mix	green grapes	whole weat flour	yams	cottage cheese	energy drink	tomato juice	low fat yogurt	green tea	honey	salad	mineral water	salmon	antioxyda juic
	1 b	ourgers	meatballs	eggs	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	<b>2</b> c	hutney	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	3	turkey	avocado	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	4 '	mineral water	milk	energy bar	whole wheat rice	green tea	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	5	low fat yogurt	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	6	whole wheat pasta	french fries	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	7	soup	light cream	shallot	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
		frozen etables	spaghetti	green tea	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Na
	4																	<b>&gt;</b>

[data.fillna(0,inplace=True) for \_ in [data.head()]]

→ [None]

data

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	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
0	shrimp	almonds	avocado	vegetables mix	green grapes	whole weat flour	yams	cottage cheese	energy drink	tomato juice	low fat yogurt	green tea	honey	salad	mineral water	salmon
1	burgers	meatballs	eggs	0	0	0	0	0	0	0	0	0	0	0	0	0
2	chutney	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
3	turkey	avocado	0	0	0	0	0	0	0	0	0	0	0	0	0	0
4	mineral water	milk	energy bar	whole wheat rice	green tea	0	0	0	0	0	0	0	0	0	0	0
7496	butter	light mayo	fresh bread	0	0	0	0	0	0	0	0	0	0	0	0	0
7497	burgers	frozen vegetables	eggs	french fries	magazines	green tea	0	0	0	0	0	0	0	0	0	0
7498	chicken	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
7499	escalope	green tea	0	0	0	0	0	0	0	0	0	0	0	0	0	0
7500	eggs	frozen smoothie	yogurt cake	low fat yogurt	0	0	0	0	0	0	0	0	0	0	0	0
7501 r	ows × 20 co	lumns														
'who. 'yam: 'cot' 'ene 'tom 'low 'gre- 'hon 'sal: 'min 'sal: 'ant 'fro: 'spi 'oli	tage chees rgy drink' ato juice' fat yogur en tea', ey', ad', eral water mon', ioxydant j zen smooth nach', ve oil']	our',  e',  t',  uice',  nie',	3.003,min	confidence	e=0.2,min_1:	ift=3.m	in len	gth=2.ma	x lengt!	h=2)						
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result=pd.DataFrame(result)
result.head()



from mlxtend.preprocessing import TransactionEncoder
from mlxtend.frequent\_patterns import apriori,association\_rules
te=TransactionEncoder()
te\_ary=te.fit(tranct).transform(tranct)
df=pd.DataFrame(te\_ary,columns=te.columns\_)



	asparagu	s 6	almonds	antioxydant juice	asparagus	avocado	babies food	bacon	barbecue sauce	black tea	•••	turkey	vegetables mix	water spray	white wine	w f
0	) Fals	e False	True	True	False	True	False	False	False	False		False	True	False	False	
1	l Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
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3	B Fals	e True	e False	False	False	True	False	False	False	False		True	False	False	False	F
4	Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
74	96 Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
74	<b>97</b> Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
74	98 Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
74	<b>99</b> Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
75	00 Fals	e True	e False	False	False	False	False	False	False	False		False	False	False	False	F
	1 rows × 121 d	olumns														
4																<b>&gt;</b>

freqent\_itemsets=apriori(df,min\_support=0.003,use\_colnames=True)
rules=association\_rules(freqent\_itemsets,metric="confidence",min\_threshold=0.75)
antecedents=rules['antecedents'].apply(lambda x:list(x))
consequents=rules['consequents'].apply(lambda x:list(x))
df\_rule=pd.DataFrame({'antecedents':antecedents,'consequents':consequents,'support':rules['support']})
df\_rule

₹	antecedents	consequents	support
0	[almonds]	[0]	0.020264
1	[antioxydant juice]	[0]	0.008799
2	[asparagus]	[0]	0.004666
3	[avocado]	[0]	0.033196
4	[babies food]	[0]	0.004533
1440	[mineral water, spaghetti, pancakes, ground beef]	[0]	0.003066
1441	[mineral water, spaghetti, tomatoes, ground beef]	[0]	0.003066
1442	[mineral water, olive oil, spaghetti, milk]	[0]	0.003333
1443	[mineral water, spaghetti, milk, shrimp]	[0]	0.003066
1444	[mineral water, spaghetti, milk, tomatoes]	[0]	0.003333

1445 rows × 3 columns

rules[rules['antecedents'] == {'cake'}]['consequents']

consequents

15 (0)