

```
In [1]: get_ipython().system('pip install mlxtend')
```

```
Requirement already satisfied: mlxtend in c:\users\rohit\anaconda3\lib\site-packages (0.23.1)
Requirement already satisfied: joblib>=0.13.2 in c:\users\rohit\anaconda3\lib\site-packages (from mlxtend) (1.1.0)
Requirement already satisfied: matplotlib>=3.0.0 in c:\users\rohit\anaconda3\lib\site-packages (from mlxtend) (3.5.2)
Requirement already satisfied: scipy>=1.2.1 in c:\users\rohit\anaconda3\lib\site-packages (from mlxtend) (1.9.1)
Requirement already satisfied: pandas>=0.24.2 in c:\users\rohit\anaconda3\lib\site-packages (from mlxtend) (1.4.4)
Requirement already satisfied: scikit-learn>=1.0.2 in c:\users\rohit\anaconda3\lib\site-packages (from mlxtend) (1.0.2)
Requirement already satisfied: numpy>=1.16.2 in c:\users\rohit\anaconda3\lib\site-packages (from mlxtend) (1.21.5)
Requirement already satisfied: cycycler>=0.10 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (0.11.0)
Requirement already satisfied: packaging>=20.0 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (21.3)
Requirement already satisfied: pyparsing>=2.2.1 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (3.0.9)
Requirement already satisfied: fonttools>=4.22.0 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (4.25.0)
Requirement already satisfied: python-dateutil>=2.7 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (2.8.2)
Requirement already satisfied: pillow>=6.2.0 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (9.2.0)
Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\rohit\anaconda3\lib\site-packages (from matplotlib>=3.0.0->mlxtend) (1.4.2)
Requirement already satisfied: pytz>=2020.1 in c:\users\rohit\anaconda3\lib\site-packages (from pandas>=0.24.2->mlxtend) (2022.1)
Requirement already satisfied: threadpoolctl>=2.0.0 in c:\users\rohit\anaconda3\lib\site-packages (from scikit-learn>=1.0.2->mlxtend) (2.2.0)
Requirement already satisfied: six>=1.5 in c:\users\rohit\anaconda3\lib\site-packages (from python-dateutil>=2.7->matplotlib>=3.0.0->mlxtend) (1.16.0)
```

```
In [2]: import pandas as pd
        from mlxtend.frequent_patterns import apriori, association_rules
        import matplotlib.pyplot as plt
```

```
In [3]: movie=pd.read_csv('my_movies.csv')
        movie.head()
```

```
Out[3]:
```

	V1	V2	V3	V4	V5	Sixth Sense	Gladiator	LOTR1	Harry Potter1	Patriot	LOTR2	Harry Potter2	LOTR1
0	Sixth Sense	LOTR1	Harry Potter1	Green Mile	LOTR2	1	0	1	1	0	1	0	
1	Gladiator	Patriot	Braveheart	NaN	NaN	0	1	0	0	1	0	0	
2	LOTR1	LOTR2	NaN	NaN	NaN	0	0	1	0	0	1	0	
3	Gladiator	Patriot	Sixth Sense	NaN	NaN	1	1	0	0	1	0	0	
4	Gladiator	Patriot	Sixth Sense	NaN	NaN	1	1	0	0	1	0	0	

```
In [4]: movie=movie.drop(['V1', 'V2', 'V3', 'V4', 'V5'], axis=1)
        movie
```

Out[4]:

	Sixth Sense	Gladiator	LOTR1	Harry Potter1	Patriot	LOTR2	Harry Potter2	LOTR	Braveheart	Green Mile
0	1	0	1	1	0	1	0	0	0	1
1	0	1	0	0	1	0	0	0	1	0
2	0	0	1	0	0	1	0	0	0	0
3	1	1	0	0	1	0	0	0	0	0
4	1	1	0	0	1	0	0	0	0	0
5	1	1	0	0	1	0	0	0	0	0
6	0	0	0	1	0	0	1	0	0	0
7	0	1	0	0	1	0	0	0	0	0
8	1	1	0	0	1	0	0	0	0	0
9	1	1	0	0	0	0	0	1	0	1

In [5]: `movie.shape`

Out[5]: (10, 10)

In [6]: `frequent_itemsets = apriori(movie, min_support=0.2, use_colnames=True)`
`frequent_itemsets`

C:\Users\ROHIT\anaconda3\lib\site-packages\mlxtend\frequent_patterns\fpcommon.py:109: DeprecationWarning: DataFrames with non-bool types result in worse computational performance and their support might be discontinued in the future. Please use a DataFrame with bool type
 warnings.warn(

Out[6]:

	support	itemsets
0	0.6	(Sixth Sense)
1	0.7	(Gladiator)
2	0.2	(LOTR1)
3	0.2	(Harry Potter1)
4	0.6	(Patriot)
5	0.2	(LOTR2)
6	0.2	(Green Mile)
7	0.5	(Sixth Sense, Gladiator)
8	0.4	(Sixth Sense, Patriot)
9	0.2	(Green Mile, Sixth Sense)
10	0.6	(Gladiator, Patriot)
11	0.2	(LOTR2, LOTR1)
12	0.4	(Sixth Sense, Gladiator, Patriot)

In [7]: `rules = association_rules(frequent_itemsets, metric="lift", min_threshold=0.6)`
`rules`
`rules.sort_values('lift', ascending = False)`

Out [7]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zh
8	(LOTR2)	(LOTR1)	0.2	0.2	0.2	1.000000	5.000000	0.16	inf	
9	(LOTR1)	(LOTR2)	0.2	0.2	0.2	1.000000	5.000000	0.16	inf	
4	(Green Mile)	(Sixth Sense)	0.2	0.6	0.2	1.000000	1.666667	0.08	inf	
5	(Sixth Sense)	(Green Mile)	0.6	0.2	0.2	0.333333	1.666667	0.08	1.2	
6	(Gladiator)	(Patriot)	0.7	0.6	0.6	0.857143	1.428571	0.18	2.8	
7	(Patriot)	(Gladiator)	0.6	0.7	0.6	1.000000	1.428571	0.18	inf	
11	(Sixth Sense, Patriot)	(Gladiator)	0.4	0.7	0.4	1.000000	1.428571	0.12	inf	
14	(Gladiator)	(Sixth Sense, Patriot)	0.7	0.4	0.4	0.571429	1.428571	0.12	1.4	
10	(Sixth Sense, Gladiator)	(Patriot)	0.5	0.6	0.4	0.800000	1.333333	0.10	2.0	
15	(Patriot)	(Sixth Sense, Gladiator)	0.6	0.5	0.4	0.666667	1.333333	0.10	1.5	
0	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08	1.8	
1	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08	1.4	
2	(Sixth Sense)	(Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
3	(Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
12	(Gladiator, Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
13	(Sixth Sense)	(Gladiator, Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	

In [8]:

```
rules[rules.lift>1]
```

Out [8]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zh
0	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08	1.8	
1	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08	1.4	
2	(Sixth Sense)	(Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
3	(Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
4	(Green Mile)	(Sixth Sense)	0.2	0.6	0.2	1.000000	1.666667	0.08	inf	
5	(Sixth Sense)	(Green Mile)	0.6	0.2	0.2	0.333333	1.666667	0.08	1.2	
6	(Gladiator)	(Patriot)	0.7	0.6	0.6	0.857143	1.428571	0.18	2.8	
7	(Patriot)	(Gladiator)	0.6	0.7	0.6	1.000000	1.428571	0.18	inf	
8	(LOTR2)	(LOTR1)	0.2	0.2	0.2	1.000000	5.000000	0.16	inf	
9	(LOTR1)	(LOTR2)	0.2	0.2	0.2	1.000000	5.000000	0.16	inf	
10	(Sixth Sense, Gladiator)	(Patriot)	0.5	0.6	0.4	0.800000	1.333333	0.10	2.0	
11	(Sixth Sense, Patriot)	(Gladiator)	0.4	0.7	0.4	1.000000	1.428571	0.12	inf	
12	(Gladiator, Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
13	(Sixth Sense)	(Gladiator, Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
14	(Gladiator)	(Sixth Sense, Patriot)	0.7	0.4	0.4	0.571429	1.428571	0.12	1.4	
15	(Patriot)	(Sixth Sense, Gladiator)	0.6	0.5	0.4	0.666667	1.333333	0.10	1.5	

In [9]:

```
rules.sort_values('lift', ascending = False)
```

Out[9]:

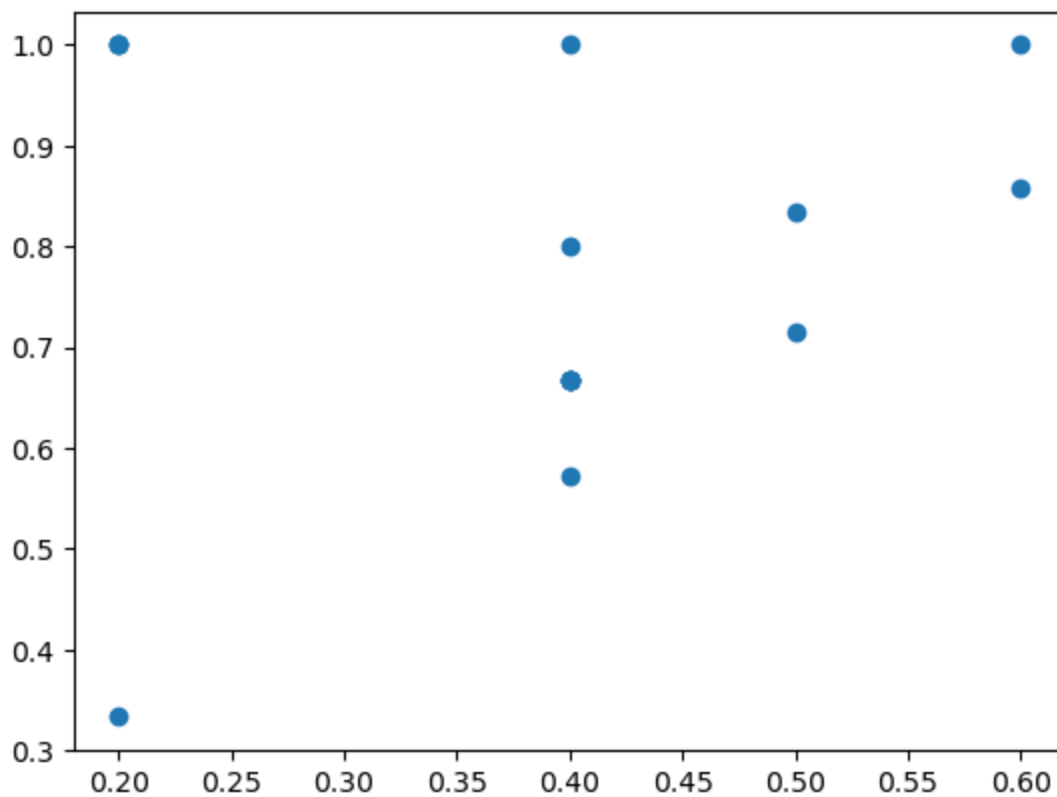
	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zh
8	(LOTR2)	(LOTR1)	0.2	0.2	0.2	1.000000	5.000000	0.16	inf	
9	(LOTR1)	(LOTR2)	0.2	0.2	0.2	1.000000	5.000000	0.16	inf	
4	(Green Mile)	(Sixth Sense)	0.2	0.6	0.2	1.000000	1.666667	0.08	inf	
5	(Sixth Sense)	(Green Mile)	0.6	0.2	0.2	0.333333	1.666667	0.08	1.2	
6	(Gladiator)	(Patriot)	0.7	0.6	0.6	0.857143	1.428571	0.18	2.8	
7	(Patriot)	(Gladiator)	0.6	0.7	0.6	1.000000	1.428571	0.18	inf	
11	(Sixth Sense, Patriot)	(Gladiator)	0.4	0.7	0.4	1.000000	1.428571	0.12	inf	
14	(Gladiator)	(Sixth Sense, Patriot)	0.7	0.4	0.4	0.571429	1.428571	0.12	1.4	
10	(Sixth Sense, Gladiator)	(Patriot)	0.5	0.6	0.4	0.800000	1.333333	0.10	2.0	
15	(Patriot)	(Sixth Sense, Gladiator)	0.6	0.5	0.4	0.666667	1.333333	0.10	1.5	
0	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08	1.8	
1	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08	1.4	
2	(Sixth Sense)	(Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
3	(Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
12	(Gladiator, Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	
13	(Sixth Sense)	(Gladiator, Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04	1.2	

In [10]:

plt.scatter('support', 'confidence', data=rules)

Out[10]:

<matplotlib.collections.PathCollection at 0x25e8e9d4a30>



```
In [11]: frequent_itemsets = apriori(movie, min_support=0.5, use_colnames=True)
frequent_itemsets
```

C:\Users\ROHIT\anaconda3\lib\site-packages\mlxtend\frequent_patterns\fpcommon.py:109: DeprecationWarning: DataFrames with non-bool types result in worse computational performance and their support might be discontinued in the future. Please use a DataFrame with bool type

```
warnings.warn(
```

Out[11]:

	support	itemsets
0	0.6	(Sixth Sense)
1	0.7	(Gladiator)
2	0.6	(Patriot)
3	0.5	(Sixth Sense, Gladiator)
4	0.6	(Gladiator, Patriot)

```
In [12]: rules1 = association_rules(frequent_itemsets, metric="lift", min_threshold=1)
rules1
rules1.sort_values('lift', ascending = False)
```

Out[12]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zha
2	(Gladiator)	(Patriot)	0.7	0.6	0.6	0.857143	1.428571	0.18	2.8	
3	(Patriot)	(Gladiator)	0.6	0.7	0.6	1.000000	1.428571	0.18	inf	
0	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08	1.8	
1	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08	1.4	

```
In [13]: rules1.sort_values('lift', ascending = False)
```

Out[13]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zha
2	(Gladiator)	(Patriot)	0.7	0.6	0.6	0.857143	1.428571	0.18		2.8
3	(Patriot)	(Gladiator)	0.6	0.7	0.6	1.000000	1.428571	0.18		inf
0	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08		1.8
1	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08		1.4

In [14]: `rules1[rules.lift>1]`

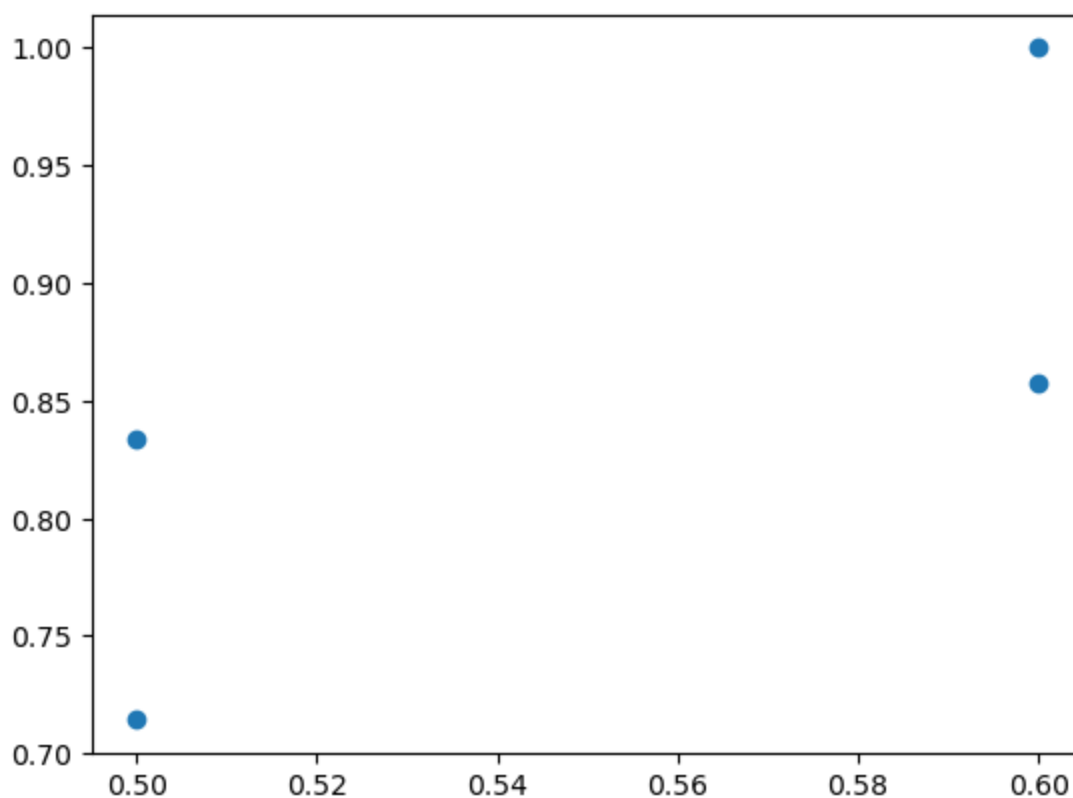
C:\Users\ROHIT\AppData\Local\Temp\ipykernel_16808\2275511782.py:1: UserWarning: Boolean Series key will be reindexed to match DataFrame index.
`rules1[rules.lift>1]`

Out[14]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage	conviction	zha
0	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08		1.8
1	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08		1.4
2	(Gladiator)	(Patriot)	0.7	0.6	0.6	0.857143	1.428571	0.18		2.8
3	(Patriot)	(Gladiator)	0.6	0.7	0.6	1.000000	1.428571	0.18		inf

In [15]: `plt.scatter('support', 'confidence', data=rules1)`

Out[15]: `<matplotlib.collections.PathCollection at 0x25e8e26c4c0>`



In []: