In [1]: import pandas as pd
 import numpy as np
 import matplotlib.pyplot as plt
 import seaborn as sns
 from mlxtend.frequent_patterns import apriori,association_rules
 from mlxtend.preprocessing import TransactionEncoder

In [2]: movie=pd.read_csv('my_movies.csv')
movie

Out[2]:

	V1	V2	V 3	V4	V5	Sixth Sense	Gladiator	LOTR1	Harry Potter1	Patriot	LOTR2
0	Sixth Sense	LOTR1	Harry Potter1	Green Mile	LOTR2	1	0	1	1	0	1
1	Gladiator	Patriot	Braveheart	NaN	NaN	0	1	0	0	1	0
2	LOTR1	LOTR2	NaN	NaN	NaN	0	0	1	0	0	1
3	Gladiator	Patriot	Sixth Sense	NaN	NaN	1	1	0	0	1	0
4	Gladiator	Patriot	Sixth Sense	NaN	NaN	1	1	0	0	1	0
5	Gladiator	Patriot	Sixth Sense	NaN	NaN	1	1	0	0	1	0
6	Harry Potter1	Harry Potter2	NaN	NaN	NaN	0	0	0	1	0	0
7	Gladiator	Patriot	NaN	NaN	NaN	0	1	0	0	1	0
8	Gladiator	Patriot	Sixth Sense	NaN	NaN	1	1	0	0	1	0
9	Sixth Sense	LOTR	Gladiator	Green Mile	NaN	1	1	0	0	0	0
4											•

In [3]: movie.shape

Out[3]: (10, 15)

```
In [6]: movie.info()
         <class 'pandas.core.frame.DataFrame'>
        RangeIndex: 10 entries, 0 to 9
        Data columns (total 15 columns):
                              Non-Null Count Dtype
          #
              Column
          0
              V1
                              10 non-null
                                               object
          1
              V2
                              10 non-null
                                               object
          2
              V3
                              7 non-null
                                              object
          3
              ۷4
                              2 non-null
                                               object
          4
              V5
                              1 non-null
                                               object
          5
              Sixth Sense
                              10 non-null
                                               int64
          6
              Gladiator
                              10 non-null
                                               int64
          7
              LOTR1
                              10 non-null
                                               int64
          8
              Harry Potter1
                             10 non-null
                                               int64
          9
              Patriot
                              10 non-null
                                               int64
          10
                              10 non-null
              LOTR2
                                               int64
          11
             Harry Potter2
                             10 non-null
                                               int64
          12
              LOTR
                              10 non-null
                                               int64
          13
              Braveheart
                              10 non-null
                                               int64
          14 Green Mile
                              10 non-null
                                               int64
        dtypes: int64(10), object(5)
        memory usage: 1.3+ KB
```

Out[7]:

In [7]:

movie2

Sixth Harry Harry Green Gladiator LOTR1 Patriot LOTR2 LOTR Braveheart Potter2 Mile Sense Potter1

Apriori Algorithm

movie2=movie.iloc[:,5:]

1. Association rules with 10% Support and 70% confidence

In [8]: # with 10% support
 frequent_itemsets=apriori(movie2,min_support=0.1,use_colnames=True)
 frequent_itemsets

Out[8]:

	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	0.1	(Harry Potter2)
7	0.1	(LOTR)
8	0.1	(Braveheart)
9	0.2	(Green Mile)
10	0.5	(Gladiator, Sixth Sense)
11	0.1	(LOTR1, Sixth Sense)
12	0.1	(Harry Potter1, Sixth Sense)
13	0.4	(Patriot, Sixth Sense)
14	0.1	(LOTR2, Sixth Sense)
15	0.1	(LOTR, Sixth Sense)
16	0.2	(Green Mile, Sixth Sense)
17	0.6	(Patriot, Gladiator)
18	0.1	(LOTR, Gladiator)
19	0.1	(Braveheart, Gladiator)
20	0.1	(Green Mile, Gladiator)
21	0.1	(LOTR1, Harry Potter1)
22	0.2	(LOTR2, LOTR1)
23	0.1	(Green Mile, LOTR1)
24	0.1	(LOTR2, Harry Potter1)
25	0.1	(Harry Potter2, Harry Potter1)
26	0.1	(Green Mile, Harry Potter1)
27	0.1	(Braveheart, Patriot)
28	0.1	(LOTR2, Green Mile)
29	0.1	(Green Mile, LOTR)
30	0.4	(Patriot, Gladiator, Sixth Sense)
31	0.1	(LOTR, Gladiator, Sixth Sense)

32	0.1	(Green Mile, Gladiator, Sixth Sense)
33	0.1	(Harry Potter1, LOTR1, Sixth Sense)
34	0.1	(LOTR2, LOTR1, Sixth Sense)
35	0.1	(Green Mile, LOTR1, Sixth Sense)
36	0.1	(LOTR2, Harry Potter1, Sixth Sense)
37	0.1	(Harry Potter1, Green Mile, Sixth Sense)
38	0.1	(LOTR2, Green Mile, Sixth Sense)
39	0.1	(Green Mile, LOTR, Sixth Sense)
40	0.1	(Braveheart, Patriot, Gladiator)
41	0.1	(Green Mile, LOTR, Gladiator)
42	0.1	(LOTR2, LOTR1, Harry Potter1)
43	0.1	(Green Mile, LOTR1, Harry Potter1)
44	0.1	(LOTR2, Green Mile, LOTR1)
45	0.1	(LOTR2, Green Mile, Harry Potter1)
46	0.1	(Green Mile, LOTR, Gladiator, Sixth Sense)
47	0.1	(LOTR2, Harry Potter1, LOTR1, Sixth Sense)
48	0.1	(Harry Potter1, Green Mile, LOTR1, Sixth Sense)
49	0.1	(LOTR2, Green Mile, LOTR1, Sixth Sense)
50	0.1	(Harry Potter1, Green Mile, LOTR2, Sixth Sense)
51	0.1	(LOTR2, Green Mile, LOTR1, Harry Potter1)
52	0.1	(LOTR2, Green Mile, Sixth Sense, Harry Potter1

itemsets

support

In [9]: # 70% confidence
 rules=association_rules(frequent_itemsets,metric='lift',min_threshold=0.7)
 rules

Out[9]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage
0	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08
1	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08
2	(LOTR1)	(Sixth Sense)	0.2	0.6	0.1	0.500000	0.833333	-0.02
3	(Sixth Sense)	(LOTR1)	0.6	0.2	0.1	0.166667	0.833333	-0.02
4	(Harry Potter1)	(Sixth Sense)	0.2	0.6	0.1	0.500000	0.833333	-0.02
245	(LOTR2)	(Harry Potter1, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
246	(Green Mile)	(LOTR2, Harry Potter1, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
247	(Sixth Sense)	(LOTR2, Green Mile, LOTR1, Harry Potter1)	0.6	0.1	0.1	0.166667	1.666667	0.04
248	(Harry Potter1)	(LOTR2, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
249	(LOTR1)	(LOTR2, Green Mile, Harry Potter1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08

250 rows × 9 columns

In [10]: # Lift Ratio > 1 is a good influential rule in selecting the associated transacti
rules[rules.lift>1]

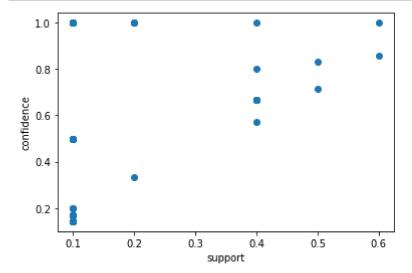
Out[10]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage
0	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08
1	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08
6	(Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04
7	(Sixth Sense)	(Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04
10	(LOTR)	(Sixth Sense)	0.1	0.6	0.1	1.000000	1.666667	0.04
245	(LOTR2)	(Harry Potter1, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
246	(Green Mile)	(LOTR2, Harry Potter1, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
247	(Sixth Sense)	(LOTR2, Green Mile, LOTR1, Harry Potter1)	0.6	0.1	0.1	0.166667	1.666667	0.04
248	(Harry Potter1)	(LOTR2, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
249	(LOTR1)	(LOTR2, Green Mile, Harry Potter1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08

236 rows × 9 columns

4

```
In [13]: # visualization of obtained rule
    plt.scatter(rules['support'], rules['confidence'])
    plt.xlabel('support')
    plt.ylabel('confidence')
    plt.show()
```



2. Association rules with 5% Support and 90% confidence

In [14]: # with 5% support
frequent_itemsets2=apriori(movie2,min_support=0.05,use_colnames=True)
frequent_itemsets2

Out[14]:

	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	0.1	(Harry Potter2)
7	0.1	(LOTR)
8	0.1	(Braveheart)
9	0.2	(Green Mile)
10	0.5	(Gladiator, Sixth Sense)
11	0.1	(LOTR1, Sixth Sense)
12	0.1	(Harry Potter1, Sixth Sense)
13	0.4	(Patriot, Sixth Sense)
14	0.1	(LOTR2, Sixth Sense)
15	0.1	(LOTR, Sixth Sense)
16	0.2	(Green Mile, Sixth Sense)
17	0.6	(Patriot, Gladiator)
18	0.1	(LOTR, Gladiator)
19	0.1	(Braveheart, Gladiator)
20	0.1	(Green Mile, Gladiator)
21	0.1	(LOTR1, Harry Potter1)
22	0.2	(LOTR2, LOTR1)
23	0.1	(Green Mile, LOTR1)
24	0.1	(LOTR2, Harry Potter1)
25	0.1	(Harry Potter2, Harry Potter1)
26	0.1	(Green Mile, Harry Potter1)
27	0.1	(Braveheart, Patriot)
28	0.1	(LOTR2, Green Mile)
29	0.1	(Green Mile, LOTR)
30	0.4	(Patriot, Gladiator, Sixth Sense)
31	0.1	(LOTR, Gladiator, Sixth Sense)
32	0.1	(Green Mile, Gladiator, Sixth Sense)

	support	itemsets
33	0.1	(Harry Potter1, LOTR1, Sixth Sense)
34	0.1	(LOTR2, LOTR1, Sixth Sense)
35	0.1	(Green Mile, LOTR1, Sixth Sense)
36	0.1	(LOTR2, Harry Potter1, Sixth Sense)
37	0.1	(Harry Potter1, Green Mile, Sixth Sense)
38	0.1	(LOTR2, Green Mile, Sixth Sense)
39	0.1	(Green Mile, LOTR, Sixth Sense)
40	0.1	(Braveheart, Patriot, Gladiator)
41	0.1	(Green Mile, LOTR, Gladiator)
42	0.1	(LOTR2, LOTR1, Harry Potter1)
43	0.1	(Green Mile, LOTR1, Harry Potter1)
44	0.1	(LOTR2, Green Mile, LOTR1)
45	0.1	(LOTR2, Green Mile, Harry Potter1)
46	0.1	(Green Mile, LOTR, Gladiator, Sixth Sense)
47	0.1	(LOTR2, Harry Potter1, LOTR1, Sixth Sense)
48	0.1	(Harry Potter1, Green Mile, LOTR1, Sixth Sense)
49	0.1	(LOTR2, Green Mile, LOTR1, Sixth Sense)
50	0.1	(Harry Potter1, Green Mile, LOTR2, Sixth Sense)
51	0.1	(LOTR2, Green Mile, LOTR1, Harry Potter1)
52	0.1	(LOTR2, Green Mile, Sixth Sense, Harry Potter1

In [15]: # 90% confidence
 rules2=association_rules(frequent_itemsets2,metric='lift',min_threshold=0.9)
 rules2

Out[15]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage
0	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08
1	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08
2	(Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04
3	(Sixth Sense)	(Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04
4	(LOTR)	(Sixth Sense)	0.1	0.6	0.1	1.000000	1.666667	0.04
233	(LOTR2)	(Harry Potter1, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
234	(Green Mile)	(LOTR2, Harry Potter1, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
235	(Sixth Sense)	(LOTR2, Green Mile, LOTR1, Harry Potter1)	0.6	0.1	0.1	0.166667	1.666667	0.04
236	(Harry Potter1)	(LOTR2, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
237	(LOTR1)	(LOTR2, Green Mile, Harry Potter1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08

238 rows × 9 columns

In [16]: # Lift Ratio > 1 is a good influential rule in selecting the associated transacti
rules2[rules2.lift>1]

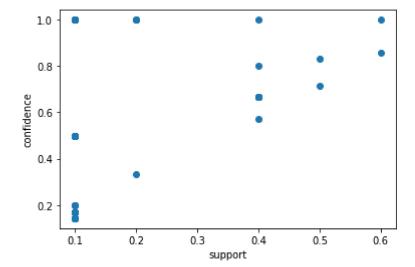
Out[16]:

	antecedents	consequents	antecedent support	consequent support	support	confidence	lift	leverage
0	(Gladiator)	(Sixth Sense)	0.7	0.6	0.5	0.714286	1.190476	0.08
1	(Sixth Sense)	(Gladiator)	0.6	0.7	0.5	0.833333	1.190476	0.08
2	(Patriot)	(Sixth Sense)	0.6	0.6	0.4	0.666667	1.111111	0.04
3	(Sixth Sense)	(Patriot)	0.6	0.6	0.4	0.666667	1.111111	0.04
4	(LOTR)	(Sixth Sense)	0.1	0.6	0.1	1.000000	1.666667	0.04
233	(LOTR2)	(Harry Potter1, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
234	(Green Mile)	(LOTR2, Harry Potter1, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
235	(Sixth Sense)	(LOTR2, Green Mile, LOTR1, Harry Potter1)	0.6	0.1	0.1	0.166667	1.666667	0.04
236	(Harry Potter1)	(LOTR2, Green Mile, LOTR1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08
237	(LOTR1)	(LOTR2, Green Mile, Harry Potter1, Sixth Sense)	0.2	0.1	0.1	0.500000	5.000000	0.08

236 rows × 9 columns

4

```
In [17]: # visualization of obtained rule
    plt.scatter(rules2['support'],rules2['confidence'])
    plt.xlabel('support')
    plt.ylabel('confidence')
    plt.show()
```



```
In [ ]:
```