

```
In [1]: pip install apyori
```

Requirement already satisfied: apyori in /opt/conda/lib/python3.7/site-packages (1.1.2)

Note: you may need to restart the kernel to use updated packages.

```
In [2]: import pandas as pd
# load the bank transaction dataset
df = pd.read_csv('PatientRoute.csv')
# info and the first 10 transactions
print(df.info())
print(df.head(10))
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 6714 entries, 0 to 6713
Data columns (total 6 columns):
patient_id    6714 non-null int64
global_num    3571 non-null float64
date          6714 non-null object
location      6714 non-null object
latitude      6714 non-null float64
longitude     6714 non-null float64
dtypes: float64(3), int64(1), object(2)
memory usage: 314.8+ KB
None
```

	patient_id	global_num	date	location	latitude	longitude
0	1000000001	2.0	22/01/2020	Gyeonggi-do_Gimpo-si	37.615246	
1	1000000001	2.0	24/01/2020	Seoul_Jung-gu	37.567241	
2	1000000002	5.0	25/01/2020	Seoul_Seongbuk-gu	37.592560	
3	1000000002	5.0	26/01/2020	Seoul_Seongbuk-gu	37.591810	
4	1000000002	5.0	26/01/2020	Seoul_Seongdong-gu	37.563992	
5	1000000002	5.0	26/01/2020	Seoul_Seongbuk-gu	37.590330	
6	1000000002	5.0	26/01/2020	Seoul_Seongbuk-gu	37.589590	
7	1000000002	5.0	27/01/2020	Seoul_Seongbuk-gu	37.592057	
8	1000000002	5.0	27/01/2020	Seoul_Dongdaemun-gu	37.566262	
9	1000000002	5.0	28/01/2020	Seoul_Seongbuk-gu	37.591669	

	longitude
0	126.715632
1	127.005659
2	127.017048
3	127.016822
4	127.029534
5	127.015221
6	127.009766
7	127.018898
8	127.065815
9	127.018420

```
In [3]: df.location.value_counts()
```

```
Out[3]: Chungcheongnam-do_Cheonan-si    470
Seoul_Gangnam-gu                        293
Gangwon-do_Wonju-si                     275
Gyeongsangbuk-do_Yecheon-gun            257
Seoul_Guro-gu                           206
...
Gyeonggi-do_Hwaseong-si                  1
Jeollanam-do_Damyang-gun                  1
Busan_Yeongdo-gu                         1
Gyeongsangnam-do_Namhae-gun              1
Gyeonggi-do_Gwangju-si                   1
Name: location, Length: 174, dtype: int64
```

transaction - patient's route item - a single patient can have multiple rows in the dataset,

and each of the row represents the location he/she visited.

```
In [4]: df.location.unique()
```

```
Out[4]: array(['Gyeonggi-do_Gimpo-si', 'Seoul_Jung-gu', 'Seoul_Seongbuk-gu',
'Seoul_Seongdong-gu', 'Seoul_Dongdaemun-gu', 'Seoul_Jungnang-gu',
'Seoul_Gangnam-gu', 'Seoul_Jongno-gu', 'Gyeonggi-do_Goyang-si',
'Gyeonggi-do_Seongnam-si', 'Seoul_Songpa-gu', 'Incheon_Yeonsu-gu',
'Seoul_Seodaemun-gu', 'Seoul_Mapo-gu', 'Gyeonggi-do_Uijeongbu-si',
'Gyeonggi-do_Dongducheon-si', 'Incheon_Jung-gu',
'Seoul_Eunpyeong-gu', 'Daegu_Nam-gu', 'Seoul_Seochon-gu',
'Gyeonggi-do_Gwacheon-si', 'Seoul_Gwangjin-gu', 'Seoul_Guro-gu',
'Daegu_Jung-gu', 'Seoul_Dobong-gu', 'Seoul_Dongjak-gu',
'Seoul_Yongsan-gu', 'Seoul_Yeongdeungpo-gu',
'Gyeonggi-do_Icheon-si', 'Gyeonggi-do_Suwon-si',
'Gyeonggi-do_Pyeongtaek-si', 'Gyeonggi-do_Paju-si',
'Gyeongsangbuk-do_Cheongdo-gun', 'Seoul_Geumcheon-gu',
'Seoul_Gwanak-gu', 'Daejeon_Yuseong-gu', 'Seoul_Gangdong-gu',
'Seoul_Nowon-gu', 'Seoul_Yangcheon-gu', 'Daegu_Dong-gu',
'Seoul_Gangseo-gu', 'Daegu_Dalseo-gu',
'Gyeongsangnam-do_Tongyeong-si', 'Incheon_Namdong-gu',
'Daejeon_Seo-gu', 'Gyeonggi-do_Anyang-si', 'Gyeonggi-do_Gunpo-si',
'Jeollanam-do_Suncheon-si', 'Seoul_Gangbuk-gu', 'Jeju-do_Jeju-si',
'Gyeonggi-do_Yongin-si', 'Gyeonggi-do_Gwangju-si', 'Daegu_Buk-gu',
'Gyeonggi-do_Namyangju-si', 'Gyeonggi-do-Ansan-si',
'Gyeonggi-do_Bucheon-si', 'Gyeonggi-do_Hanam-si',
'Incheon_Gyeyang-gu', 'Gyeonggi-do_Siheung-si',
'Gyeonggi-do_Yangpyeong-gun', 'Chungcheongbuk-do_Cheongju-si',
'Gyeonggi-do_Yeoncheon-gun', 'Jeollabuk-do_Gunsan-si',
'Chungcheongbuk-do_Yeongdong-gun', 'Gyeongsangnam-do_Jinju-si',
'Jeollanam-do_Gwangyang-si', 'Jeollabuk-do_Jeonju-si',
'Gyeonggi-do_Gwangmyeong-si', 'Jeollanam-do_Muan-gun',
'Chungcheongbuk-do_Chungju-si', 'Busan_Dongnae-gu',
'Busan_Suyeong-gu', 'Busan_Haeundae-gu', 'Busan_Yeonje-gu',
'Busan_Dong-gu', 'Busan_Busanjin-gu', 'Busan_Seo-gu',
'Busan_Geumjeong-gu', 'Gwangju_Gwangsan-gu', 'Gwangju_Seo-gu',
'Gwangju_Buk-gu', 'Busan_Buk-gu', 'Busan_Nam-gu', 'Busan_Saha-gu',
'Daegu_Seo-gu', 'Busan_Jung-gu', 'Busan_Gangseo-gu',
'Incheon_Seo-gu', 'Busan_Gijang-gun',
'Gyeongsangbuk-do_Gyeongju-si', 'Gyeongsangnam-do_Yangsan-si',
'Busan_Sasang-gu', 'Ulsan_Nam-gu', 'Daegu_Dalseong-gun',
'Gyeongsangnam-do_Gimhae-si', 'Jeollanam-do_Gurye-gun',
'Gyeonggi-do_Anseong-si', 'Gyeongsangnam-do_Haman-gun',
'Daegu_Suseong-gu', 'Gwangju_Nam-gu', 'Jeollanam-do_Naju-si',
'Gwangju_Dong-gu', 'Jeollanam-do_Damyang-gun',
'Jeollanam-do_Goheung-gun', 'Jeollanam-do_Boseong-gun',
'Jeollanam-do_Hwasun-gun', 'Incheon_Bupyeong-gu',
'Incheon_Dong-gu', 'Incheon_Michuhol-gu',
'Chungcheongbuk-do_Eumseong-gun', 'Gangwon-do_Gangneung-si',
'Gangwon-do_Jeongseon-gun', 'Chungcheongnam-do_Seosan-si',
'Chungcheongnam-do_Asan-si', 'Gangwon-do_Samcheok-si',
'Daejeon_Dong-gu', 'Daejeon_Jung-gu', 'Ulsan_Jung-gu',
'Ulsan_Ulju-gun', 'Ulsan_Buk-gu', 'Gyeongsangnam-do_Uiryeong-gun',
'Gyeongsangbuk-do_Gyeongsan-si', 'Ulsan_Dong-gu',
'Gyeongsangbuk-do_Gimcheon-si', 'Gyeonggi-do_Guri-si',
'Gyeongsangnam-do_Sacheon-si', 'Gangwon-do_Wonju-si',
'Gangwon-do_Donghae-si', 'Gyeonggi-do_Yeouju-si',
'Gyeongsangnam-do_Changwon-si', 'Gyeongsangbuk-do_Chilgok-gun',
'Gyeongsangbuk-do_Bonghwa-gun', 'Gangwon-do-Taebaek-si',
'Chungcheongbuk-do_Jeungpyeong-gun',
'Chungcheongbuk-do_Jincheon-gun', 'Chungcheongbuk-do_Goesan-gun',
'Gyeonggi-do_Hwaseong-si', 'Gangwon-do_Sokcho-si',
'Chungcheongnam-do_Gyeryong-si', 'Chungcheongnam-do_Cheonan-si',
'Busan_Yeongdo-gu', 'Chungcheongnam-do_Taejeon-gun',
'Chungcheongnam-do_Gongju-si', 'Chungcheongnam-do_Dangjin-si',
'Gyeongsangbuk-do_Yecheon-gun', 'Chungcheongnam-do_Seochon-gun',
'Chungcheongnam-do_Nonsan-si', 'Jeollabuk-do_Iksan-si',
'Jeollanam-do_Yeosu-si', 'Jeollanam-do_Gangjin-gun',
```

```
'Jeollanam-do_Mokpo-si', 'Gyeongsangbuk-do_Uljin-gun',
'Gyeongsangbuk-do_Yeongcheon-si', 'Gangwon-do_Pyeongchang-gun',
'Gyeongsangbuk-do_Cheongsong-gun', 'Chungcheongnam-do_Boryeong-si',
'Gyeongsangnam-do_Changnyeong-gun', 'Gyeongsangbuk-do_Pohang-si',
'Gyeongsangbuk-do_Gumi-si', 'Gyeongsangbuk-do_Andong-si',
'Gyeongsangnam-do_Geochang-gun', 'Gyeongsangbuk-do_Sangju-si',
'Gyeongsangbuk-do_Yeongju-si', 'Chungcheongbuk-do_Danyang-gun',
'Gyeongsangbuk-do_Mungyeong-si', 'Gangwon-do_Chuncheon-si',
'Jeollanam-do_Jangheung-gun', 'Gyeongsangbuk-do_Goryeong-gun',
'Gyeongsangnam-do_Hapcheon-gun', 'Gyeongsangnam-do_Namhae-gun',
'Gyeongsangnam-do_Geoje-si', 'Gyeonggi-do_Osan-si',
'Gyeongsangnam-do_Miryang-si', 'Gyeongsangnam-do_Goseong-gun'],
dtype=object)
```

```
In [5]: df.patient_id.value_counts()
```

```
Out[5]: 10000000417    45
        3009000014    42
        1400000021    38
        3009000013    37
        6016000012    36
        ..
        6100000067     1
        6100000079     1
        6100000045     1
        6100000029     1
        1100000017     1
        Name: patient_id, Length: 1211, dtype: int64
```

```
In [6]: print(df.patient_id.unique())
```

```
[10000000001 10000000002 10000000003 ... 61000000088 61000000089 61000000090]
```

As we are looking to generate association rules from locations visited by each patient, we need to group the patients and then generate list of places visited

```
In [7]: # group by account, then list all services
transactions = df.groupby(['patient_id'])['location'].apply(list)
print(transactions.head(5))
```

```
patient_id
10000000001    [Gyeonggi-do_Gimpo-si, Seoul_Jung-gu]
10000000002    [Seoul_Seongbuk-gu, Seoul_Seongbuk-gu, Seoul_S...
10000000003    [Seoul_Jongno-gu, Seoul_Jongno-gu]
10000000004    [Seoul_Jungnang-gu]
10000000005    [Seoul_Jungnang-gu]
Name: location, dtype: object
```

Run the apriori model with min_support of 0.05

```
In [8]: from apyori import apriori

transaction_list = list(transactions)
results = list(apriori(transaction_list, min_support=0.05))

def convert_apriori_results_to_pandas_df(results):
    rules = []

    for rule_set in results:
        for rule in rule_set.ordered_statistics:
            # items_base = left side of rules, items_add = right side
            # support, confidence and lift for respective rules
            rules.append(['', '.join(rule.items_base), '', '.join(rule.items_add)
```

```
rule_set.support, rule.confidence, rule.lift])
```

```
# typecast it to pandas df
return pd.DataFrame(rules, columns=['Left_side', 'Right_side', 'Support',
```

```
result_df = convert_apriori_results_to_pandas_df(results)
# sort all acquired rules descending by lift
result_df = result_df.sort_values(by='Lift', ascending=False)
print(result_df.head(10))
```

	Left_side	Right_side	Support	Confidence	Lift
0		Busan_Yeonje-gu	0.060281	0.060281	1.0
1	Chungcheongnam-do_Cheonan-si		0.078448	0.078448	1.0
2		Daegu_Jung-gu	0.061932	0.061932	1.0
3		Incheon_Jung-gu	0.123865	0.123865	1.0
4		Seoul_Dongjak-gu	0.085879	0.085879	1.0
5		Seoul_Gangnam-gu	0.086705	0.086705	1.0
6		Seoul_Guro-gu	0.066887	0.066887	1.0
7		Seoul_Jung-gu	0.057803	0.057803	1.0
8		Seoul_Jungnang-gu	0.075970	0.075970	1.0
9		Seoul_Yangcheon-gu	0.052023	0.052023	1.0

With min_support 0.02

```
In [9]: from apyori import apriori

# type cast the transactions from pandas into normal list format and run apriori
transaction_list = list(transactions)
results = list(apriori(transaction_list, min_support=0.02))

result_df = convert_apriori_results_to_pandas_df(results)
result_df = result_df.sort_values(by='Lift', ascending=False)
print(result_df.head(10))
```

	Left_side	Right_side	Support	Confidence	Lift
37	Seoul_Songpa-gu	Incheon_Jung-gu	0.020644	0.454545	3.669697
36	Incheon_Jung-gu	Seoul_Songpa-gu	0.020644	0.166667	3.669697
34	Seoul_Gangnam-gu	Incheon_Jung-gu	0.030553	0.352381	2.844889
33	Incheon_Jung-gu	Seoul_Gangnam-gu	0.030553	0.246667	2.844889
27		Seoul_Seodaemun-gu	0.022296	0.022296	1.000000
21		Seoul_Gwanak-gu	0.034682	0.034682	1.000000
22		Seoul_Jongno-gu	0.034682	0.034682	1.000000
23		Seoul_Jung-gu	0.057803	0.057803	1.000000
24		Seoul_Jungnang-gu	0.075970	0.075970	1.000000
25		Seoul_Mapo-gu	0.025599	0.025599	1.000000

With min_support 0.003

```
In [10]: from apyori import apriori

# type cast the transactions from pandas into normal list format and run apriori
transaction_list = list(transactions)
results = list(apriori(transaction_list, min_support=0.003))

result_df = convert_apriori_results_to_pandas_df(results)
result_df = result_df.sort_values(by='Lift', ascending=False)
print(result_df.head(10))
```

	Left_side	Right_side
567		Incheon_Yeonsu-gu
568	Chungcheongnam-do_Cheonan-si, Gyeongsangbuk-do...	
569	Incheon_Yeonsu-gu, Chungcheongnam-do_Cheonan-si	

```

566 Gyeongsangbuk-do_Andong-si
281 Gyeongsangnam-do_Hapcheon-gun
282 Gyeongsangnam-do_Uiryeong-gun
257 Gwangju_Dong-gu
258 Gwangju_Nam-gu
255 Gwangju_Dong-gu
254 Gwangju_Buk-gu

```

		Right_side	Support	Confidence
567	Chungcheongnam-do_Cheonan-si,Gyeongsangbuk-do_...		0.003303	0.333333
568	Incheon_Yeonsu-gu		0.003303	1.000000
569	Gyeongsangbuk-do_Andong-si		0.003303	1.000000
566	Incheon_Yeonsu-gu,Chungcheongnam-do_Cheonan-si		0.003303	0.266667
281	Gyeongsangnam-do_Uiryeong-gun		0.003303	0.800000
282	Gyeongsangnam-do_Hapcheon-gun		0.003303	0.307692
257	Gwangju_Nam-gu		0.005780	0.636364
258	Gwangju_Dong-gu		0.005780	0.538462
255	Gwangju_Buk-gu		0.003303	0.363636
254	Gwangju_Dong-gu		0.003303	0.500000

	Lift
567	100.916667
568	100.916667
569	80.733333
566	80.733333
281	74.523077
282	74.523077
257	59.279720
258	59.279720
255	55.045455
254	55.045455

```
In [11]: result_df.loc[result_df['Left_side'] == 'Daegu_Buk-gu']
```

	Left_side	Right_side	Support	Confidence	Lift
579	Daegu_Buk-gu	Daegu_Jung-gu, Daegu_Nam-gu	0.004129	0.128205	17.250712
572	Daegu_Buk-gu	Daegu_Jung-gu, Daegu_Dong-gu	0.003303	0.102564	11.291375
206	Daegu_Buk-gu	Daegu_Nam-gu	0.004129	0.128205	10.350427
212	Daegu_Buk-gu	Gyeongsangbuk-do_Pohang-si	0.009083	0.282051	8.758054
203	Daegu_Buk-gu	Daegu_Jung-gu	0.015690	0.487179	7.866325
200	Daegu_Buk-gu	Daegu_Dong-gu	0.003303	0.102564	5.175214
209	Daegu_Buk-gu	Gyeongsangbuk-do_Gyeongsan-si	0.003303	0.102564	3.268556

As we would like to find out 10 routes for positive patients who visited Daegu_Buk-gu, setting up min_support with 0.003 can't meet the requirement, we then reduce the min_support further.

With min_support 0.002

```
In [20]: from apyori import apriori

# type cast the transactions from pandas into normal list format and run apriori
transaction_list = list(transactions)
results = list(apriori(transaction_list, min_support=0.002))

result_df = convert_apriori_results_to_pandas_df(results)
```

```
result_df = result_df.sort_values(by='Lift', ascending=False)
print(result_df)
```

```

Left_side \
1283 Daegu_Jung-gu, Daegu_Nam-gu, Daegu_Seo-gu
1269 Daegu_Buk-gu, Daegu_Seo-gu
1275 Seoul_Jung-gu, Daegu_Nam-gu
1270 Seoul_Jung-gu, Daegu_Buk-gu
1222 Daegu_Buk-gu, Daegu_Seo-gu
...
648 Seoul_Gangnam-gu
690 Seoul_Guro-gu
691 Seoul_Jungnang-gu
348 Daegu_Jung-gu
349 Seoul_Gangnam-gu

Right_side Support Confidence \
1283 Seoul_Jung-gu, Daegu_Buk-gu 0.002477 1.000000
1269 Seoul_Jung-gu, Daegu_Jung-gu, Daegu_Nam-gu 0.002477 1.000000
1275 Daegu_Jung-gu, Daegu_Buk-gu, Daegu_Seo-gu 0.002477 1.000000
1270 Daegu_Jung-gu, Daegu_Nam-gu, Daegu_Seo-gu 0.002477 1.000000
1222 Seoul_Jung-gu, Daegu_Nam-gu 0.002477 1.000000
...
648 Seoul_Jung-gu 0.002477 0.028571
690 Seoul_Jungnang-gu 0.002477 0.037037
691 Seoul_Guro-gu 0.002477 0.032609
348 Seoul_Gangnam-gu 0.002477 0.040000
349 Daegu_Jung-gu 0.002477 0.028571

Lift
1283 403.666667
1269 403.666667
1275 403.666667
1270 403.666667
1222 403.666667
...
648 0.494286
690 0.487520
691 0.487520
348 0.461333
349 0.461333

```

[1292 rows x 5 columns]

```
In [18]: print(result_df.head(5))
```

```

Left_side \
1283 Daegu_Jung-gu, Daegu_Nam-gu, Daegu_Seo-gu
1269 Daegu_Buk-gu, Daegu_Seo-gu
1275 Seoul_Jung-gu, Daegu_Nam-gu
1270 Seoul_Jung-gu, Daegu_Buk-gu
1222 Daegu_Buk-gu, Daegu_Seo-gu

Right_side Support Confidence \
1283 Seoul_Jung-gu, Daegu_Buk-gu 0.002477 1.0
1269 Seoul_Jung-gu, Daegu_Jung-gu, Daegu_Nam-gu 0.002477 1.0
1275 Daegu_Jung-gu, Daegu_Buk-gu, Daegu_Seo-gu 0.002477 1.0
1270 Daegu_Jung-gu, Daegu_Nam-gu, Daegu_Seo-gu 0.002477 1.0
1222 Seoul_Jung-gu, Daegu_Nam-gu 0.002477 1.0

Lift
1283 403.666667
1269 403.666667
1275 403.666667
1270 403.666667
1222 403.666667

```

Top 5 frequently occurring rules are as follows:

1. Daegu_Jung-gu, Daegu_Seo-gu, Daegu_Nam-gu ==> Daegu_Buk-gu, Seoul_Jung-gu
2. Daegu_Buk-gu, Daegu_Seo-gu ==> Daegu_Jung-gu, Daegu_Nam-gu, Seoul_Jung-gu
3. Daegu_Nam-gu, Seoul_Jung-gu ==> Daegu_Buk-gu, Daegu_Jung-gu, Daegu_Seo-gu
4. Daegu_Buk-gu, Seoul_Jung-gu ==> Daegu_Jung-gu, Daegu_Seo-gu, Daegu_Nam-gu
5. Daegu_Buk-gu, Daegu_Seo-gu ==> Daegu_Nam-gu, Seoul_Jung-gu

In [16]:

```
daegu = result_df.loc[result_df['Left_side'] == 'Daegu_Buk-gu']
daegu.head(10)
```

Out[16]:

	Left_side	Right_side	Support	Confidence	Lift
968	Daegu_Buk-gu	Seoul_Jung-gu, Daegu_Seo-gu	0.002477	0.076923	31.051282
961	Daegu_Buk-gu	Seoul_Jung-gu, Daegu_Nam-gu	0.002477	0.076923	31.051282
1202	Daegu_Buk-gu	Daegu_Jung-gu, Seoul_Jung-gu, Daegu_Seo-gu	0.002477	0.076923	31.051282
1217	Daegu_Buk-gu	Seoul_Jung-gu, Daegu_Nam-gu, Daegu_Seo-gu	0.002477	0.076923	31.051282
1262	Daegu_Buk-gu	Seoul_Jung-gu, Daegu_Jung-gu, Daegu_Nam-gu, Daegu...	0.002477	0.076923	31.051282
1172	Daegu_Buk-gu	Daegu_Jung-gu, Daegu_Nam-gu, Daegu_Seo-gu	0.002477	0.076923	31.051282
1187	Daegu_Buk-gu	Daegu_Jung-gu, Daegu_Nam-gu, Seoul_Jung-gu	0.002477	0.076923	31.051282
954	Daegu_Buk-gu	Daegu_Nam-gu, Daegu_Seo-gu	0.002477	0.076923	23.288462
940	Daegu_Buk-gu	Daegu_Jung-gu, Daegu_Seo-gu	0.002477	0.076923	18.630769
933	Daegu_Buk-gu	Daegu_Jung-gu, Daegu_Nam-gu	0.004129	0.128205	17.250712

We are not able to get at least 10 routes from "Daegu_Buk-gu". So, we set min_support=0.002.

As it can be observed, we will be able to get at least 10 common routes from "Daegu_Buk-gu" with min_support=0.002.