```
Importing essential modules
```

```
In [1]:
```

```
import findspark
findspark.init()
```

In [2]:

```
from pyspark.sql.functions import split
```

In [3]:

```
from pyspark.ml.fpm import FPGrowth
from pyspark import SparkContext, since
```

In [4]:

```
from pyspark.sql import SQLContext as sc
```

In [5]:

```
from pyspark.context import SparkContext
from pyspark.sql.session import SparkSession
```

#### Creating a spark context and starting a session

```
In [6]:
```

```
sc = SparkContext.getOrCreate()
spark = SparkSession(sc)
```

### Reading the input data

```
In [7]:
```

```
lines = sc.textFile("F:\Docs\Big data\Assignment\Assignmnet 4\Dataset\kosarak.dat")
```

Converting the input data into id and transactions as a list

# In [8]:

```
data = []
i = 0
for line in lines.collect():
    data.append((i,list(set(map(lambda x : int(x) ,str(line).split())))))
    i = i + 1
print("The total number of transactions are : ",len(data))
```

The total number of transactions are : 990002

### Creating a sparks dataframe

```
In [17]:
```

```
df = spark.createDataFrame(data, ["id", "items"])
```

```
In [18]:
```

```
print(df.show())
```

```
+---+
| id|
               items
+---+
      [1, 2, 3]|
 0 1
| 0|
                [1]|
         [4, 5, 6, 7]|
| 0|
1 01
              [8, 1]|
  0 |
              [9, 10]|
0|[6, 11, 12, 13, 1...|
  0| [1, 3, 7]|
             [17, 18]|
  0|[6, 11, 19, 20, 2...|
 0| [1, 3, 25]|
0 |
              [26, 3]|
  0|[32, 33, 34, 3, 3...|
  0| [38, 2, 6]|
 0|[1, 3, 6, 7, 11, ...|
 0| [3, 52, 53, 6]|
  0 |
        [1, 55, 54, 6]|
0|[64, 6, 11, 56, 5...|
  0 1
                 [3]|
 0|[65, 1, 66, 67, 6...|
| 0| [1, 11, 69, 6]|
+---+--
only showing top 20 rows
```

None

Aliasing the inbuild FPGrowth function as fpgrwoth and setting the minimum support to be 0.05 (that is around 50000) and min confidence as 0.75

```
In [19]:
```

```
fpGrowth = FPGrowth(itemsCol="items", minSupport=0.05, minConfidence=0.75)
```

### Fitting our model

```
In [20]:
```

```
model = fpGrowth.fit(df)
```

# Printing frequent item list

#### In [21]:

```
print("The frequent Item set is : ")
model.freqItemsets.show()
```

```
The frequent Item set is :
+----+
items| freq|
+------
          [148]| 69922|
       [148, 11]| 55759|
    [148, 11, 6]| 55230|
      [148, 218]| 58823|
   [148, 218, 11]| 50098|
|[148, 218, 11, 6]| 49866|
   [148, 218, 6]| 56838|
         [148, 6]| 64750|
             [6]|601374|
             [3]|450031|
           [3, 6]|265180|
            [55] | 65412|
            [11]|364065|
         [11, 3]|161286|
       [11, 3, 6]|143682|
         [11, 6]|324013|
             [1]|197522|
         [1, 11]| 91882|
```

#### Printing association rules

#### In [22]:

```
print("The Association rule is : ")
model.associationRules.show()
```

The Association rule is :

```
+-----
   antecedent|consequent| confidence|
                                                       lift.l
  ______
     [7]| [6]| 0.847085088264402|1.3944998146776124|
  [148, 11, 6]| [218]|0.9028788701792504|10.088849491356443|

[148, 6]| [11]|0.852972972972912.3194895120079906|
      [148, 6]| [218]|0.8778069498069498| 9.808693603950202|
218. 1111
                  [6]|0.9953690766098447|1.6386098776832714|
[6]|0.8767127926138287| 1.443273932882492|
|[148, 218, 11]|
       [218]|
       [7, 11]|
                    [6] | 0.9782913410659845 | 1.6104959380319184 |
     [148, 11]|
                    [6] | 0.9905127423375598 | 1.6306152177175417 |
     [148, 11]|
                  [218]|0.8984737889847378| 10.03962671891542|
                  [11]|0.7585246569759544|2.0626561945133663|
        [7, 6]|
         [11]|
                     [6] | 0.8899866782030681 | 1.4651258474666247 |
   [148, 218]|
                   [11]|0.8516736650629856|2.3159563038459776|
                    [6] | 0.9662546962922667 | 1.5906808106747825 |
    [148, 218]|
     [218, 11]| [148]|0.8125405475541715|11.504487388228668|
                   [6]|0.9833592837680031| 1.618838954874821|
     [218, 11]|
         [27]|
                     [6] | 0.8237169711925029 | 1.3560304384867325 |
       [11, 3]|
                     [6] | 0.8908522748409657 | 1.466550821613681 |
        [148]|
                    [11]|0.7974457252366923| 2.168494260299056|
         [148]|
                  [218] | 0.8412659820943337 | 9.400381552691421 |
                   [6] | 0.9260318640771145 | 1.5244646384780045 |
        [148]|
  ------
```

only showing top 20 rows

### Showing the predicted consequents if any for each transaction

#### In [23]:

```
print("Examining the input items against all the association rules and summarize the consequents a
s prediction")
model.transform(df).show()
```

Examining the input items against all the association rules and summarize the consequents as prediction

```
______
               items|prediction|
+--+----
1 01
      [1, 2, 3]| []|
                [1]|
  01
                           []
          [4, 5, 6, 7]|
                         [11]|
  0 |
              [8, 1]|
 0 1
                          []
              [9, 10]|
                           []
 0|[6, 11, 12, 13, 1...|
                           []
                         [6]|
 0| [1, 3, 7]|
0 |
             [17, 18]|
0|[6, 11, 19, 20, 2...|
0 |
           [1, 3, 25]
                           []|
 0 |
             [26, 3]|
 0|[32, 33, 34, 3, 3...|
                           []
  0| [38, 2, 6]|
0|[1, 3, 6, 7, 11, ...|
[][
                            []|
  0| [3, 52, 53, 6]|
0| [1, 55, 54, 6]|
                           []
                           []
  0|[64, 6, 11, 56, 5...|
                           []
  01
                 [3]|
                            [][
  NI 1 66 67 6 1
```