### DS/CMPSC 410 MiniProject #1

#### Spring 2021

**Instructor: John Yen** 

#### TA: Dongkuan Xu, Rupesh Prajapati

#### **Learning Objectives**

- Be able to identify frequent 2 port sets and 3 port sets that are scanned by scanners in the Darknet dataset
- Be able to improve the frequent port set mining algorithm by adding suitable filtering
- Be able to improve the performance of frequent port set mining by suitable reuse of RDD, together with appropriate persist and unpersist on the reused RDD.

#### **Total points: 100**

- Exercise 1: 10 points
- Exercise 2: 10 points
- Exercise 3: 20 points
- Exercise 4: 20 points
- Exercise 5: 10 points
- Exercise 6: 30 points (run spark-submit on a large Dataset)

#### Submit the following items for this mini project deliverable:

- Completed Jupyter Notebook (including answers to Exercise 9.1 to 9.5; in HTML or PDF format)
- The python file (.py) used for spark-submit
- The output file that contains counts of 2-port sets and 3-port sets.

from pyspark.ml.clustering import KMeans

• The log file of spark-submit that shows the CPU time for completing the spark-submit job.

#### Due: midnight, April 2, 2021

ndexToString

```
In [1]: import pyspark
import csv
import pandas as pd

In [2]: from pyspark import SparkContext
from pyspark.sql import SparkSession
from pyspark.sql.types import StructField, StructType, StringType, LongType
from pyspark.sql.functions import col, column
from pyspark.sql.functions import expr
from pyspark.sql.functions import split
from pyspark.sql import Row
from pyspark.ml import Pipeline
from pyspark.ml.feature import OneHotEncoder, StringIndexer, VectorAssembler, I
```

```
In [3]: ss = SparkSession.builder.master("local").appName("Lab9 FrequentPortSets").get0
    rCreate()
```

# Exercise 9.1 (10 points)

- Complete the path below for reading "sampled\_profile.csv" you downloaded from Canvas, uploaded to Lab9 folder. (5 points)
- Fill in your Name (5 points):

```
In [4]: Scanners_df = ss.read.csv("/storage/home/kky5082/ds410/Lab9/sampled_profile.cs
v", header= True, inferSchema=True )
```

# We can use printSchema() to display the schema of the DataFrame Scanners\_df to see whether it was inferred correctly.

```
In [5]: | Scanners df.printSchema()
        root
         |-- _c0: integer (nullable = true)
          -- id: integer (nullable = true)
          -- numports: integer (nullable = true)
          |-- lifetime: double (nullable = true)
          -- Bytes: integer (nullable = true)
          |-- Packets: integer (nullable = true)
          -- average_packetsize: integer (nullable = true)
          -- MinUniqueDests: integer (nullable = true)
          -- MaxUniqueDests: integer (nullable = true)
          -- MinUniqueDest24s: integer (nullable = true)
          |-- MaxUniqueDest24s: integer (nullable = true)
          -- average lifetime: double (nullable = true)
          |-- mirai: boolean (nullable = true)
          -- zmap: boolean (nullable = true)
          |-- masscan: boolean (nullable = true)
          |-- country: string (nullable = true)
          -- traffic_types_scanned_str: string (nullable = true)
          |-- ports scanned str: string (nullable = true)
          -- host_tags_per_censys: string (nullable = true)
         |-- host services per censys: string (nullable = true)
```

# Part A Transforrm the feature "ports\_scanned\_str" into an array of ports.

The original value of the column is a string that connects all the ports scanned by a scanner. The different ports that are open by a scanner are connected by dash "-". For example, "81-161-2000" indicates the scanner has scanned three ports: port 81, port 161, and port 2000. Therefore, we want to use split to separate it into an array of ports by each scanner. This transformation is important because it enables the identification of frequent ports scanned by scanners.

### The original value of the column "ports\_scanned\_str"

only showing top 30 rows

```
In [6]: | Scanners df.select("ports scanned str").show(30)
              ports scanned str
               . - - - - - - - - - - - - +
                            13716
                    17128-17136|
                            35134I
                            17140|
                             545941
                             17130 l
                             54594
                            37876
                            17142|
          | 17128 - 17130 - 17132 . . . |
                            54594|
                            12941
                            301881
          |23-80-81-1023-232...|
                             54594|
          | 17128 - 17132 - 17136 . . . |
                            17136 l
                            54594|
                             17134|
                               445|
                             34226|
                            17130 l
                            17134|
                        137-17130|
                            171421
                            171421
          | 17128 - 17130 - 17132 . . . |
                                231
                             545941
                            54594|
```

Convert the Column 'ports_scanned_str' into an Array of scanned by each scanner (row)	ports

```
In [7]: Scanners_df2=Scanners_df.withColumn("Ports_Array", split(col("ports_scanned_st
r"), "-") )
Scanners_df2.persist().show(10)
```

```
id|numports|lifetime|Bytes|Packets|average packetsize|MinUniqueDe
sts|MaxUniqueDests|MinUniqueDest24s|MaxUniqueDest24s|average lifetime|mirai| zm
ap|masscan|country|traffic_types_scanned_str|
                                 ports scanned str|host tags per
_censys|host_services per censys|
                             Ports Arrayl
|1645181|1645181|
                 1|
                      0.0
          1|
                      1|
                                             0.0|false|false
                                  1|
 false|
          BR |
                            15|
                                         13716
                              [13716]|
null|
                 null|
|2091467|2091467|
                 2 | 199.84
                          752|
                                12|
                                              62|
                                            66.61|false|false
          11
                      1|
                                  11
| false|
                                     17128 - 17136 |
          CN|
                          11-16|
                 null|
                         [17128, 17136]|
null
                                              60 I
| 888618| 888618|
                 1|
                      0.0
                           60|
                      1|
                                             0.0|false|false
1|
          1|
                                  1|
                            15|
                                         35134|
  false|
         USI
null|
                 nullI
                              [35134]|
|1512937|1512937|
                 1 | 793.37 | 1561 |
                                23|
                                              67 I
                                           396.69|false|false
                      1|
                                  1|
  falsel
          JP|
                                         17140|
                          11-16|
null|
                 null|
                              [17140]|
| 654939| 654939|
                 1|
                    48.69
                          571|
                                             1901
                                 3|
                                  2|
                                            24.34|false|false
1|
          2|
                      1|
  falsel
          TRI
                            16|
                                         54594|
                              [54594]|
null|
                 8081|
       73109|
                 1 | 1056.83 | 1924 |
  73109|
                                              74|
                                26|
                                          1056.83|false|false
1|
          1|
                      1|
                                  1|
 false|
          CN|
                            11|
                                         17130|
null|
                 null|
                              [17130]|
| 923577| 923577|
                 1 348.42
                          465|
                                 3|
                                             155 l
                                           174.21|false|false
1|
          2|
                      1|
                                  2|
 false|
          J0|
                            16|
                                         54594|
                              [54594]|
                 null|
null|
                                              60|
                      0.0
|1349341|1349341|
                 1|
                           60|
                                 1|
1|
          11
                      1|
                                  1|
                                             0.0|false|false
                                         378761
 false|
          US I
                            15 l
null|
                 null|
                              [37876]|
|1959916|1959916|
                 1 | 814.15 | 1631 |
                                24|
                                              67|
                                           407.07|false|false
1|
          11
                      1|
                                  1|
| false|
                          11-16|
                                         17142|
          TWI
null|
                 null|
                              [17142]|
| 565394| 565394|
                 7| 2505.48| 5422|
                                              761
                                71|
1|
                                           119.31|false|false
          3|
                      1|
                                  3|
  falsel
          CNI
                          11-16|17128-17130-17132...|
null|
                 null|[17128, 17130, 17...|
-----+
```

# For Mining Frequent Port Sets being scanned, we only need the column Ports\_Array

# Because each port number in the Ports\_Array column for each row occurs only once, we can count the total occurance of each port number through flatMap.

```
In [10]: Ports list RDD = Ports Scanned RDD.map(lambda row: row[0] )
In [11]: Ports list RDD.persist()
Out[11]: PythonRDD[27] at RDD at PythonRDD.scala:53
In [12]: Ports list2 RDD = Ports Scanned RDD.flatMap(lambda row: row[0])
In [13]: Port count RDD = Ports list2 RDD.map(lambda \times (x, 1))
         Port count RDD.take(2)
Out[13]: [('13716', 1), ('17128', 1)]
         Port count total RDD = Port count RDD.reduceByKey(lambda x,y: x+y, 1)
In [14]:
         Port count total RDD.persist().take(5)
Out[14]: [('13716', 14),
          ('17128', 31850),
          ('17136', 31617),
          ('35134', 13),
          ('17140', 31865)]
In [15]: | Port count total RDD.count()
Out[15]: 65536
         Sorted Count Port RDD = Port count total RDD.map(lambda x: (x[1], x[0])).sortBy
In [16]:
         Key( ascending = False)
```

```
Sorted_Count_Port_RDD.persist().take(50)
Out[17]: [(32014,
                   '17132'),
           (31865, '17140'),
           (31850, '17128'),
           (31805, '17138'),
                    '17130'),
           (31630,
                   '17136'),
           (31617,
                    '23'),
           (29199,
                   '445'),
           (25466,
           (25216, '54594'),
                    '17142'),
           (21700,
           (21560,
                   '17134'),
           (15010, '80'),
           (13698, '8080'),
           (8778, '0'),
                   '2323'),
           (6265,
           (5552,
                   '5555'),
           (4930, '81'),
                   '1023'),
           (4103,
                   '52869'),
           (4058,
                   '8443'),
           (4012,
                   '49152'),
           (3954,
                   '7574'),
           (3885,
                   '37215'),
           (3874,
                   '34218'),
           (3318,
           (3279,
                   '34220'),
           (3258,
                   '33968'),
                   '34224'),
           (3257,
           (3253,
                   '34228'),
                   '33962'),
           (3252,
           (3236,
                   '33960'),
                  '33964'),
           (3209,
                   '34216'),
           (3179,
           (3167,
                   '34226'),
                   '33970'),
           (3155,
                   '33972'),
           (3130,
           (2428,
                   '50401'),
           (1954,
                   '34222'),
           (1921,
                   '34230'),
                   '33966'),
           (1919,
                  '33974'),
           (1819,
           (1225, '3389'),
           (1064, '1433'),
           (885, '22'),
           (878, '5353'),
           (604, '21'),
           (594, '8291'),
                 '8728'),
           (554,
           (512, '443'),
           (382, '5900'),
           (330, '8000')]
```

The value of the threshold below should be identical to your choice of threshold for Exercise 9.3

```
In [18]: threshold = 1000
    Filtered_Sorted_Count_Port_RDD= Sorted_Count_Port_RDD.filter(lambda x: x[0] > t
    hreshold)
    Filtered_Sorted_Count_Port_RDD.persist().count()

Out[18]: 42
In [19]: Top_Ports = Filtered_Sorted_Count_Port_RDD.map(lambda x: x[1]).collect()
In [20]: Top_1_Port_count = len(Top_Ports)
In [21]: print(Top_1_Port_count)
    42
```

# Exercise 9.2 (10 points)

Compute the total number of scanners in Ports\_list\_RDD with the total number of scanners that scan more than one port. What is the impact of this filter on the size of the RDD? Complete the following code to find out the answers. Then, fill the answer in the cell marked as Answer to Exercise 9.2.

#### Answer to Exercise 9.2

- Original number of scanners: ## 227062
- Number of scanners that scan more than one port: ## 73663
- Impact of the filtering on the size of filtered scanners: ## The size flitered scanners is 1/3 of the original scanners

# Exercise 9.3 (20 points)

- Choose a threshold (suggest a number between 500 and 1000) (5 points)
- Complete the following code for finding 2 port sets (7 points)
- Add suitable persist and unpersist to suitable RDD (8 points)

```
In [26]: | multi Ports list RDD.take(2)
Out[26]: [['17128', '17136'],
          ['17128', '17130', '17132', '17134', '17136', '17138', '17140']]
In [29]:
         # Initialize a Pandas DataFrame to store frequent port sets and their counts
         Freq Port Sets df = pd.DataFrame( columns= ['Port Sets', 'count'])
         # Initialize the index to the Freq Port Sets df to 0
         index = 0
         # Set the threshold for Large Port Sets to be 100
         threshold = 1000
         multi_Ports_list_RDD.persist()
         for i in range(0, Top 1 Port count-1):
             Scanners port i RDD = multi Ports list RDD.filter(lambda x: Top Ports[i] in
         X)
             Scanners port i RDD.persist()
             for j in range(i+1, Top 1 Port count-1):
                 Scanners port i j RDD = Scanners port i RDD.filter(lambda x: Top Ports
         [j] in x)
                 two ports count = Scanners port i j RDD.count()
                  if two ports count > threshold:
                      Freq Port Sets df.loc[index]=[ [Top Ports[i], Top Ports[j]], two po
         rts count]
                      index = index +1
             Scanners port i RDD.unpersist()
In [30]: | print(Freq Port Sets df)
                   Port Sets
                              count
         0
               [17132, 17140]
                               16317
         1
              [17132, 17128]
                               16279
         2
              [17132, 17138]
                               16299
         3
              [17132, 17130]
                               16336
         4
              [17132, 17136]
                               16148
                                 . . .
              [33960, 33964]
         259
                                1005
              [33960, 34226]
         260
                                1034
         261
              [33964, 34226]
                                1004
              [34216, 34226]
         262
                                1015
         263 [34226, 33972]
                                1023
         [264 rows x 2 columns]
In [31]: | tri Ports list RDD=multi Ports list RDD.filter(lambda x: len(x)>2)
In [32]: | tri Ports list RDD.count()
Out[32]: 49549
In [33]: | index
Out[33]: 264
```

# Exercise 9.5 (20 points)

- Use the same threshold as Exercise 9.4 (5 points)
- Complete the following code to find frequent 3 port sets (7 points)

In [36]: # Set the threshold for Large Port Sets to be 100

• Add persist and unpersist to suitable RDD (8 points)

```
threshold = 1000
         tri Ports list RDD.persist()
         for i in range(0, Top 1 Port count-1):
             Scanners_port_i_RDD = tri_Ports_list_RDD.filter(lambda x: Top Ports[i] in x
             Scanners port i RDD.persist()
             for j in range(i+1, Top 1 Port count-1):
                 Scanners port i j RDD = Scanners port i RDD.filter(lambda x: Top Ports[
         j] in x)
                 two_ports_count = Scanners_port_i_j_RDD.count()
                 Scanners port i j RDD.persist()
                 if two_ports_count > threshold:
                     Scanners port i RDD.unpersist()
                     for k in range(j+1, Top 1 Port count -1):
                         Scanners port i j k RDD = Scanners port i j RDD.filter(lambda x
         : Top Ports[k] in x)
                         Scanners_port_i_j_RDD.unpersist()
                         three ports count = Scanners port i j k RDD.count()
                         if three ports count > threshold:
                              Freq Port Sets df.loc[index] = [ [Top Ports[i], Top Ports[j
         ], Top Ports[k]], three ports count]
                              index = index + 1
                              # print("Ports: ", Top Ports[i], ", ", Top Ports[j], ", ",
         Top_Ports[k], ": Count ", three_ports_count)
In [41]: | Freq Port Sets DF = ss.createDataFrame(Freq Port Sets df)
```

# Exercise 9.5 (10 points)

Complete the following code to save your frequent 2 port sets and 3 port sets in an output file.

```
In [44]: output_path = "/storage/home/kky5082/ds410/Lab9/output"
Freq_Port_Sets_DF.rdd.saveAsTextFile(output_path)
```

# Exercise 9.6 (30 points)

- Remove .master("local") from SparkSession statement
- Change the input file to "/gpfs/scratch/juy1/Day\_2020\_profile.csv"
- Change the output file to a different directory from the one you used in Exercise 9.4
- · Export the notebook as a .py file
- Run spark-submit on ICDS Roar (following instructions on Canvas)