

ParallelECLAT-ad

September 5, 2022

1 Advanced Tutorial on Implementing parallelECLAT Algorithm

In this tutorial, we will discuss two approaches to find frequent patterns in big data using parallelECLAT algorithm.

1. Advanced approach: Here, we generalize the basic approach by presenting the steps to discover frequent patterns using multiple minimum support values.

In this tutorial, we explain how the parallelECLAT (parallelECLAT) algorithm can be implemented by varying the minimum support values

Step 1: Import the parallelECLAT algorithm and pandas data frame

```
[1]: from PAMI.frequentPattern.pyspark import parallelECLAT as alg
import pandas as pd
```

Step 2: Specify the following input parameters

```
[2]: inputFile = 'temporal_T10I4D100K.csv'
seperator='\t'
numberWorkersCount=2
minimumSupportCountList = [100, 150, 200, 250, 300]
#minimumSupport can also specified between 0 to 1. E.g., minSupList = [0.005, 0.
↪0.006, 0.007, 0.008, 0.009]

result = pd.DataFrame(columns=['algorithm', 'minSup', 'patterns', 'runtime',
↪'memory'])
#initialize a data frame to store the results of parallelECLAT algorithm
```

Step 3: Execute the parallelECLAT algorithm using a for loop

```
[3]: algorithm = 'parallelECLAT' #specify the algorithm name
for minSupCount in minimumSupportCountList:
    obj = alg.parallelECLAT(iFile=inputFile,
↪minSup=minSupCount,numWorkers=numberWorkersCount, sep=seperator)
    obj.startMine()
    #store the results in the data frame
```

```
result.loc[result.shape[0]] = [algorithm, minSupCount, len(obj.  
↪getPatterns()), obj.getRuntime(), obj.getMemoryRSS()]
```

Setting default log level to "WARN".

To adjust logging level use `sc.setLogLevel(newLevel)`. For SparkR, use `setLogLevel(newLevel)`.

```
22/08/24 13:58:00 WARN NativeCodeLoader: Unable to load native-hadoop library  
for your platform... using builtin-java classes where applicable  
22/08/24 13:58:01 WARN Utils: Service 'SparkUI' could not bind on port 4040.  
Attempting port 4041.
```

```
100 <class 'int'>  
<class 'int'> 100
```

Frequent patterns were generated successfully using Parallel ECLAT algorithm
22/08/24 13:58:46 WARN Utils: Service 'SparkUI' could not bind on port 4040.
Attempting port 4041.

```
150 <class 'int'>  
<class 'int'> 150
```

Frequent patterns were generated successfully using Parallel ECLAT algorithm
22/08/24 13:59:20 WARN Utils: Service 'SparkUI' could not bind on port 4040.
Attempting port 4041.

```
200 <class 'int'>  
<class 'int'> 200
```

Frequent patterns were generated successfully using Parallel ECLAT algorithm
22/08/24 13:59:48 WARN Utils: Service 'SparkUI' could not bind on port 4040.
Attempting port 4041.

```
250 <class 'int'>  
<class 'int'> 250
```

Frequent patterns were generated successfully using Parallel ECLAT algorithm
22/08/24 14:00:11 WARN Utils: Service 'SparkUI' could not bind on port 4040.
Attempting port 4041.

```
300 <class 'int'>
<class 'int'> 300
```

Frequent patterns were generated successfully using Parallel ECLAT algorithm

```
[4]: print(result)
```

	algorithm	minSup	patterns	runtime	memory
0	parallelECLAT	100	27532	46.434464	223105024
1	parallelECLAT	150	19126	33.166129	224575488
2	parallelECLAT	200	13255	26.772719	224276480
3	parallelECLAT	250	7703	22.108692	223600640
4	parallelECLAT	300	4552	15.972808	223133696

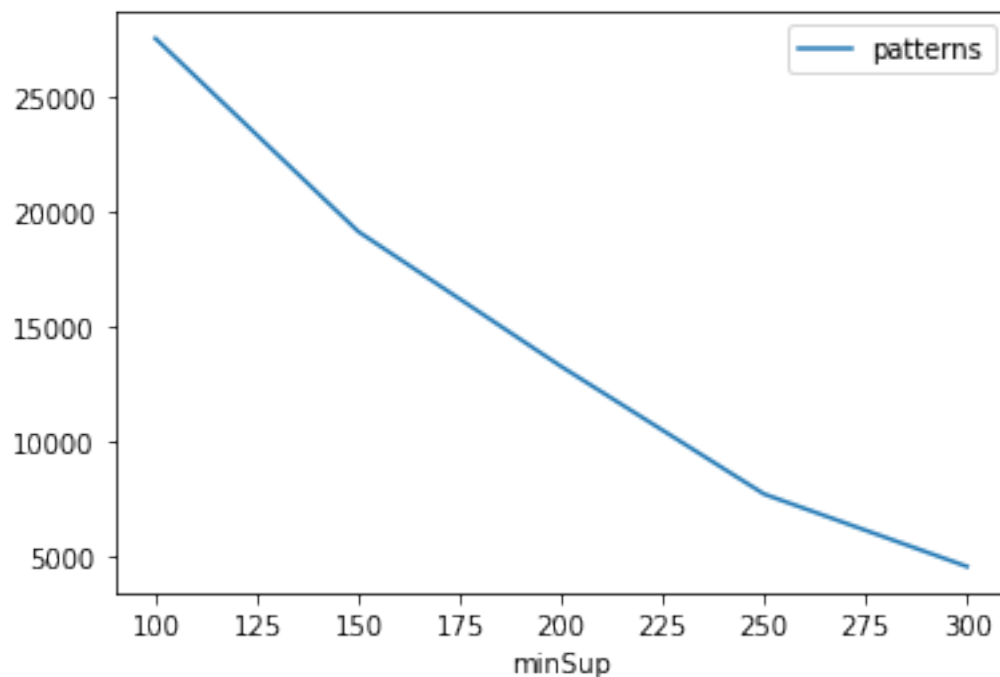
Step 5: Visualizing the results

Step 5.1 Importing the plot library

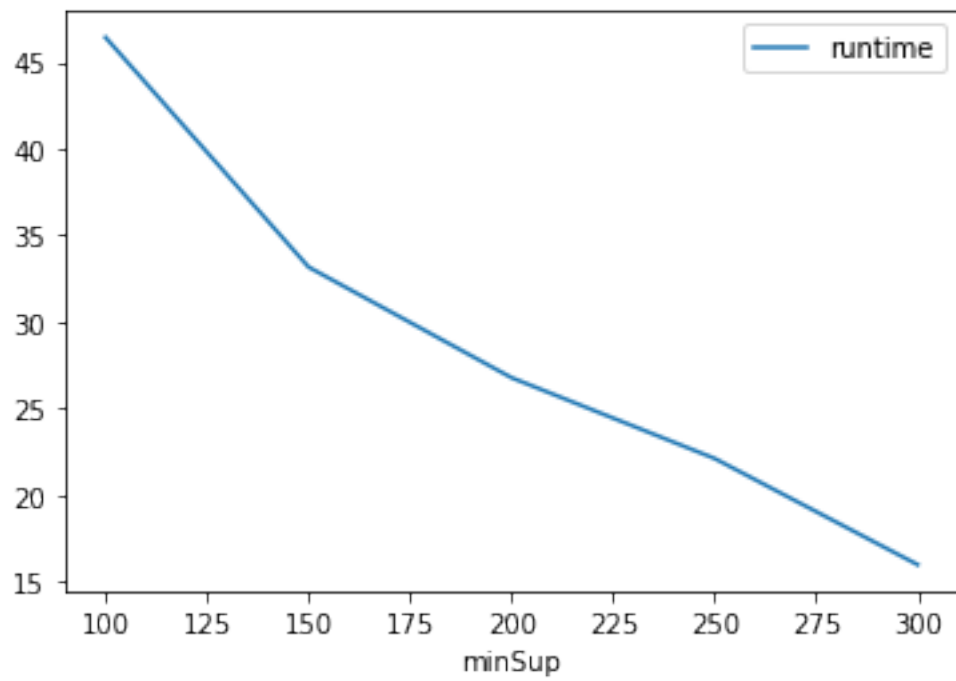
```
[5]: from PAMI.extras.graph import plotLineGraphsFromDataFrame as plt
```

Step 5.2. Plotting the number of patterns

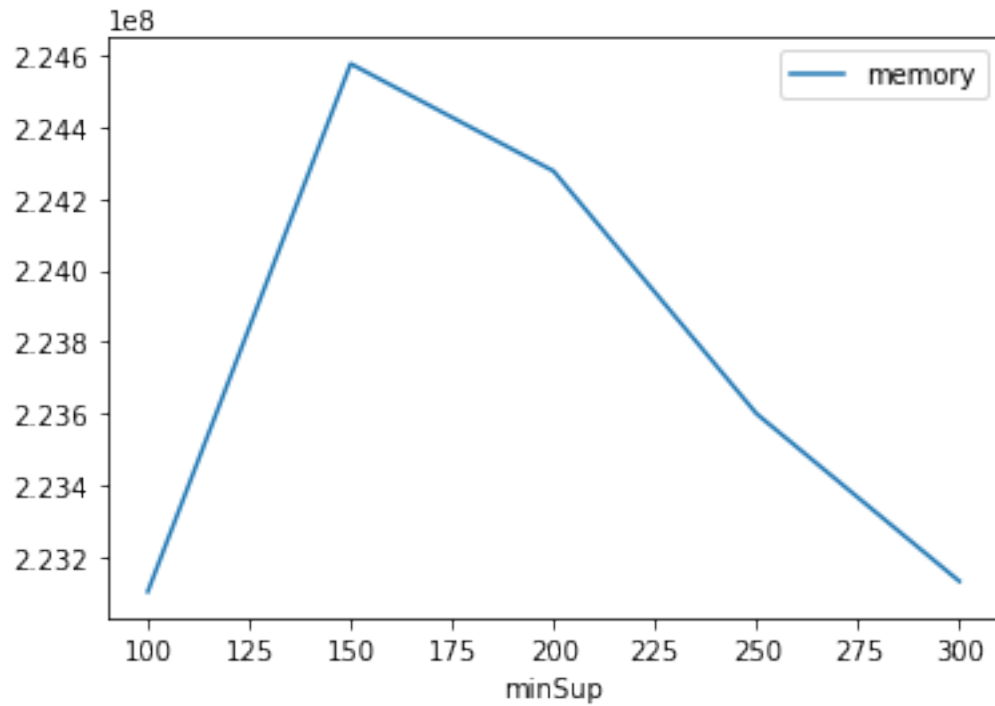
```
[6]: ab = plt.plotGraphsFromDataFrame(result)
      ab.plotGraphsFromDataFrame() #drawPlots()
```



Graph for No Of Patterns is successfully generated!



Graph for Runtime taken is successfully generated!



Graph for memory consumption is successfully generated!

1.0.1 Step 6: Saving the results as latex files

```
[7]: from PAMI.extras.graph import generateLatexFileFromDataFrame as gdf
gdf.generateLatexCode(result)
```

Latex files generated successfully