# Mining Local Periodic Patterns in Temporal Databases

### What is local periodic pattern mining?

Local periodic pattern mining aims to discover all interesting patterns in a temporal database that have **periodicity** no greater than the user-specified **maximum periodicity** (**maxPer**) constraint, **time interval of occurence** no greater than user-specified **maximum period of spillovers** (**maxSoPer**) constraint and **minDur** is no less than **minimum duration** (**minDur**). The **minDur** controls the minimum duration that a pattern is reocurring.

Research paper: Fournier Viger, Philippe & Yang, Peng & Rage, Uday & Ventura, Sebastian & Luna, José María. (2020). Mining Local Periodic Patterns in a Discrete Sequence. Information Sciences. 544. 10.1016/j.ins.2020.09.044.

#### What is a temporal database?

A temporal database is a collection of transactions at a particular timestamp, where each transaction contains a timestamp and a set of items.

A hypothetical temporal database containing the items **a**, **b**, **c**, **d**, **e**, **f**, **and g** as shown below

TS	Transactions
1	a b c g
2	b c d e
3	a b c d
4	a c d f
5	a b c d g
6	c d e f
7	a b c d
8	a e f
9	a b c d
10	b c d e

**Note:** Duplicate items must not exist in a transaction.

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#### Acceptable format of temporal databases in PAMI

Each row in a temporal database must contain timestamp and items.

1abcg
2bcde
3abcd
4acdf
5abcdg
6cdef
7abcd
8aef
9abcd

### Understanding the statisctics of database

To understand about the database. The below code will give the detail about the transactional database.

- Total number of transactions (Database size)
- Total number of unique items in database
- Minimum lenth of transaction that existed in database
- Average length of all transactions that exists in database
- Maximum length of transaction that existed in database
- Minimum periodicity exists in database
- Average periodicity exists in database
- Maximum periodicity exists in database
- Standard deviation of transaction length
- Variance in transaction length
- Sparsity of database

#### The sample code

```
In [ ]: import PAMI.extras.dbStats.temporalDatabaseStats as stats
    obj = stats.temporalDatabaseStats('sampleInputFile.txt', ' ')
    obj.run()
    obj.printStats()
```

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## What is the input to local periodic pattern mining algorithms

Algorithms to mine the local periodic patterns requires temporal database, maxPer, maxSoPer and minDur (specified by user).

- Temporal database is accepted following formats:
  - String: E.g., 'temporalDatabase.txt'
  - URL: E.g., https://u-aizu.ac.jp/~udayrage/datasets/transactionalDatabases/transactional\_T10
  - DataFrame. Please note that dataframe must contain the header titled 'TS' and 'Transactions'
- maxPer should be mentioned in
  - count (beween 0 to length of database)
  - **[**0, 1]
- maxSoPer should be mentioned in
  - count (beween 0 to length of database)
  - **[**0, 1]
- minDur should be mentioned in
  - count (beween 0 to length of database)
  - **[**0, 1]
- seperator default seperator is '\t' (tab space)

## How to run the local periodic pattern algorithm in terminal

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- Download the PAMI source code from github.
- Unzip the PAMI source code folder and enter into local periodic pattern folder.
- Enter into localPeriodicPattern folder
- Enter into a basic folder of your choice and execute the following command on terminal.

syntax: python3 algorithmName.py <path to the input file> <path to
the output file> <maxPer> <maxSoPer> <minDur> <seperator>

## Sample command to execute the LPPGrowth algorithm in localPeriodicPattern/basic folder

```
python3 LPPGrowth.py inputFile.txt outputFile.txt 3 4 2 ' '
```

## How to implement the LPPGrowth algorithm by importing PAMI package

Import the PAMI package executing: pip3 install PAMI

- Install the PAMI package from the PYPI repository by executing the following command: pip3 install PAMI
- Run the below sample code by making necessary changes

The localPeriodicPatterns.txt file contains the following patterns (*format:* pattern:support):!cat localPeriodicPatterns.txt

```
In [5]: !cat localPeriodicPatterns.txt
```

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```
d: {(2, 18)}
('d', 'c'): {(2, 10)}
('d', 'c', 'b'): {(2, 10)}
('d', 'c', 'b', 'a'): {(3, 10)}
('d', 'c', 'a'): {(3, 10)}
('d', 'b'): {(2, 10)}
('d', 'b', 'a'): {(3, 10)}
('d', 'b', 'a'): {(3, 10)}
c: {(1, 10)}
('c', 'b'): {(1, 10)}
('c', 'b', 'a'): {(1, 10)}
('c', 'a'): {(1, 10)}
b: {(1, 10)}
('b', 'a'): {(1, 10)}
a: {(1, 9)}
```

The dataframe containing the patterns is shown below:

#### In [6]: df

Out[6]:		Patterns	PTL
	0	d	{(2, 18)}
	1	(d, c)	{(2, 10)}
	2	(d, c, b)	{(2, 10)}
	3	(d, c, b, a)	{(3, 10)}
	4	(d, c, a)	{(3, 10)}
	5	(d, b)	{(2, 10)}
	6	(d, b, a)	{(3, 10)}
	7	(d, a)	{(3, 10)}
	8	С	{(1, 10)}
	9	(c, b)	{(1, 10)}
	10	(c, b, a)	{(1, 10)}
	11	(c, a)	{(1, 10)}
	12	b	{(1, 10)}

13

14

(b, a) {(1, 10)}

{(1, 9)}

а

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