

Mining High-Utility Frequent Patterns in Utility Databases

What is High-Utility Frequent pattern mining?

High utility frequent pattern mining aims to discover all the patterns with **utility** of pattern is no less than user-specified **minimum utility** (*minutil*) and **support** is no less than user-specified **minimum support** (*minSup*).

What is the utility database?

A utility database is a collection of transaction, where each transaction contains a set of items and a positive integer called **internal utility** respectively. And each unique item in database is also associated with another positive number called **external utility**.

A hypothetical utility database with items **a, b, c, d, e, f and g** and its **internal utility** is shown below at right side and items with its **external utility** is presented at left side.

Transactions	Item	Profit
(a,2) (b,3) (c,1) (g,1)	a	4
(b,3) (c,2) (d,3) (e,2)	b	3
(a,2) (b,1) (c,3) (d,4)	c	6
(a,3) (c,2) (d,1) (f,2)	d	2
(a,3) (b,1) (c,2) (d,1) (g,2)	e	5
(c,2) (d,2) (e,3) (f,1)	f	2
(a,2) (b,1) (c,1) (d,2)	g	3
(a,1) (e,2) (f,2)		
(a,2) (b,2) (c,4) (d,2)		
(b,3) (c,2) (d,2) (e,2)		

Note: Duplicate items must not exist in a transaction.

Acceptable format of utility databases in PAMI

Each row in a utility database must contain only items, total sum of utilities and utility values.

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

Understanding the statistics 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 length of transaction that existed in database
- Average length of all transactions that exists in database
- Maximum length of transaction that existed in database
- Minimum utility value exists in database
- Average utility exists in database
- Maximum utility exists in database
- Standard deviation of transaction length
- Variance in transaction length
- Sparsity of database

The sample code

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

What is the input to high-utility frequent pattern mining algorithms

Algorithms to mine the high-utility patterns requires utility database, minUtil (specified by user).

- Utility database can be provided in following formats:

- String : E.g., 'utilityDatabase.txt'
- URL : E.g., https://u-aizu.ac.jp/~udayrage/datasets/transactionalDatabases/transactional_T10
- In DataFrame format (dataframe variable with heading Transactions, Utilities and TransactionUtility)

- **minUtil**
specified in

- [0, 1]

- **minSup**
specified in

- **count (between 0 to length of a database)** or
- [0, 1]

- **seperator**
default seperator is '\t' (tab space)

How to run the high-utility frequent pattern algorithm in terminal

- Download the code from github.
- Navigate to PAMI folder where you downloaded the file.
- Go to highUtilityFrequentPattern/basic folder

And execute the following command on terminal.

```
python3 algorithmName.py <path to input file> <path of output file>
<minUtil> <minSup> <seperator>
```

Sample command to execute the EFIM algorithm in highUtilityPattern/basic folder

```
python3 HUFIM.py inputFile.txt outputFile.txt $20$ $5$ ' '
```

How to implement the HUFIM algorithm by importing PAMI package

Import the PAMI package executing: **pip3 install PAMI**

```
In [ ]: import PAMI.highUtilityFrequentPatterns.basic.HUFIM as alg

iFile = 'sample_Input.txt' #specify the input transactional database <br>
minUtil = 20 #specify the minSupvalue <br>
minSup = 5 #specify the minSupvalue <br>
seperator = ' ' #specify the seperator. Default seperator is tab space. <br>
oFile = 'utilityfrequentPatterns.txt' #specify the output file name<br>

obj = alg.HUFIM(iFile, minUtil, minSup, seperator) #initialize the algori
obj.startMine() #start the mining process <br>
obj.savePatterns(oFile) #store the patterns in file <br>
df = obj.getPatternsAsDataFrame() #Get the patterns discovered into a
obj.printStats() #Print the statistics of mining pro
```

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

```
In [3]: !cat utilityfrequentPatterns.txt
```

```
c d : 35:8
c d a : 34:5
c d b : 39:6
c a : 27:6
c a b : 30:5
c b : 29:7
d a : 22:5
d b : 25:6
```

```
In [4]: df
```

```
Out[4]:
```

	Patterns	Utility:Support
0	c d	35:8
1	c d a	34:5
2	c d b	39:6
3	c a	27:6
4	c a b	30:5
5	c b	29:7
6	d a	22:5
7	d b	25:6

	Patterns	Utility:Support
0	c d	35:8
1	c d a	34:5
2	c d b	39:6
3	c a	27:6
4	c a b	30:5
5	c b	29:7
6	d a	22:5
7	d b	25:6