



## **Contributions**

We present a new framework for performing data mining on dynamic networks in an on-top fashion. Existing subgraph mining algorithms on static graphs can be easily integrated into our framework. The efficient search for dynamic patterns inside static frequent sub-graphs is based on the idea of suffix trees.

## **Transformation into a Dynamic Graph**

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The three graphs on the left represent a time series of graphs with edge insertions and edge deletions over time. The graph on the right is a dynamic graph that summarizes all information represented by the time series by the use of so called existence strings.

## **Efficient Dynamic Frequent Subgraph Discovery**

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edge

edge (

edge

concatenation

concatenation

-



Two embeddings of a frequent subgraph. Matrix representation and common substring.

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3

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b

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3

<1:baɛ>

2

а

3

С

а

3

С

<2:aɛc>

<2:aɛc>

3

С

b

3

С

b

3

<3:cbε>



Dynamic pattern.

Algorithm 1 Dynamic frequent subgraph discovery

**Input:** All embeddings of one static frequent subgraph S for each embedding  $S_i$  of S do

1) Order the edges of  $S_i$  according to canonical labeling 2) Store their existence strings in this order as a matrix 3) Translate each column of the matrix  $M(S_i)$  into one character

4) Concatenate these characters into one string end for

Perform frequent substring discovery on all the resulting strings

**Output:** Dynamic patterns within embeddings of S

## **Results on Biological Yeast Data**

Dynamic network construction by integration of PPI data and a time series of yeast gene expression levels. Enumeration of all static frequent subof FANMOD graphs by the use [Wernicke, 2006]. The *p*-value of each static frequent subgraph is determined to asses the significance of the subgraph. Evolutionary conservation rate indicates the quality of the result.

Most frequent and longest dynamic patterns in subgraphs with 3 and 4 vertices.





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