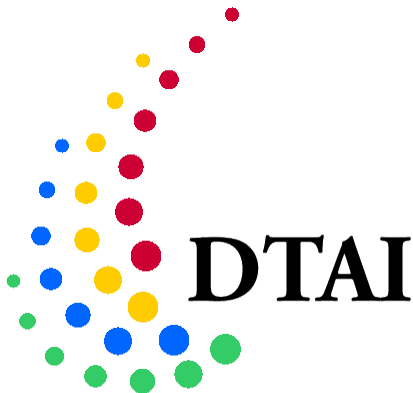


# Mining Large Single Networks under Subgraph Homomorphism

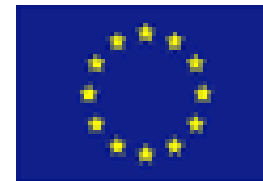
Mostafa H. Chehreghani

Jan Ramon

Thomas Fannes



DECLARATIVE LANGUAGES &  
ARTIFICIAL INTELLIGENCE



# Overview

- Introduction
- Problem definition and preliminaries
- Related work and motivation
- Our contributions and the proposed algorithm
- Conclusion

# Frequent Patterns

- Frequent patterns = pattern which occurs in a database more often than a user-defined threshold
- Two settings:
  - Transactional
  - **Single-network**
- Applications:
  - Web mining
  - Social network analysis
  - Biological & chemical interaction networks

# Problem Definition

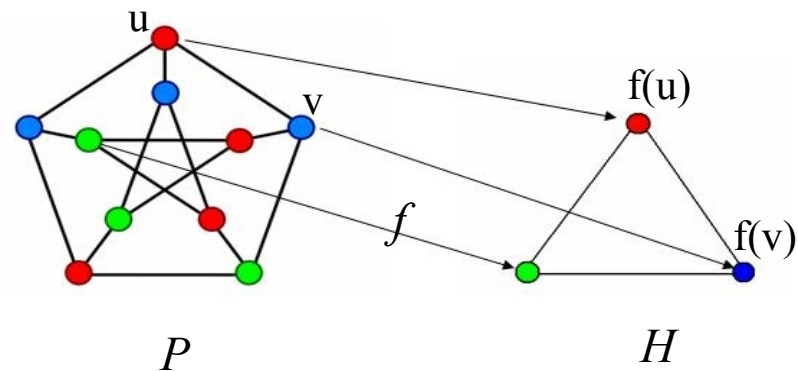
- Given:
  - a network graph  $H$
  - a pattern language  $L_p$
  - a matching operator  $\leq$
  - a threshold  $minsup \in R^+$
- Find (a **condensed representation** of) all **patterns** such that their frequency is at least  $minsup$

# Homomorphism

- **Graph homomorphism  $f$  from  $P$  to  $H$ :**
  - Label preserving
  - If  $u$  and  $v$  of  $P$  are adjacent in  $P$ , then  $f(u)$  and  $f(v)$  are adjacent in  $H$

**Subgraph Homomorphism:**

*Homomorphism from  $P$   
to (a subgraph of)  $H$*



- Subgraph homomorphism is easier than subgraph isomorphism
  - Polynomial algorithms for bounded treewidth graphs

# Related Work and Motivation

- Most approaches use any graph patterns
  - e.g. Kuramochi&Karypis ICDM'04
  - NP-hard under normal matching operators
    - **We will limit ourselves to bounded treewidth graphs**
      - **This is not a strong restriction**
- Most approaches use **subgraph isomorphism**
  - e.g. Zhu et. al., VLDB'11
  - Computationally expensive
  - A few methods use **subgraph homomorphism**
    - e.g. Dries&Nijssen, SDM12 (Only for trees)
    - e.g. J.Van den Bussche, (No antimonotonic pruning)

## Related Work and Motivation Cont.

- Matching operator  $\leq$ 
  - We use subgraph homomorphism
  - Candidate generation under homomorphism is challenging
    - Our solution: **root embedding equivalent classes**
- The frequency measure
  - Wang&Ramon, DMKD'13: s-measure: linear program
    - LP with one variable per embedding of pattern
    - Describes statistical power of the pattern
    - But: needs to construct **overlap graph** (exponential amount of embeddings)
    - We avoid overlap graph using bounded treewidth homomorphism!

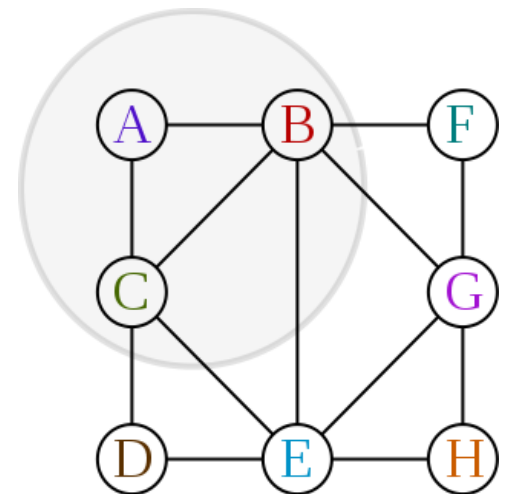
# A Summary of Our Contributions

- We consider the class of **rooted graphs**
  - We present an efficient method to generate them from data
- We present a new notion for compactly representing all frequent patterns
  - It gives a closure operator
- Two frequency counting settings:
  - Mining patterns with frequent root embeddings  
(= embeddings of the root of the pattern)
  - Mining s-measure-frequent patterns
    - Linear program to compute s-measure



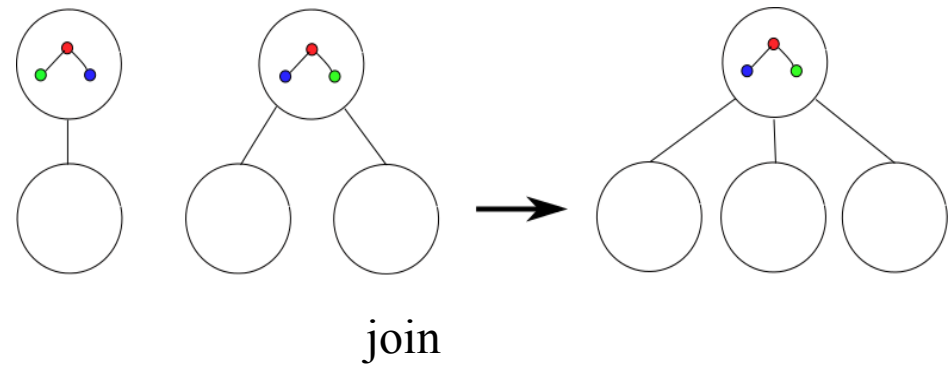
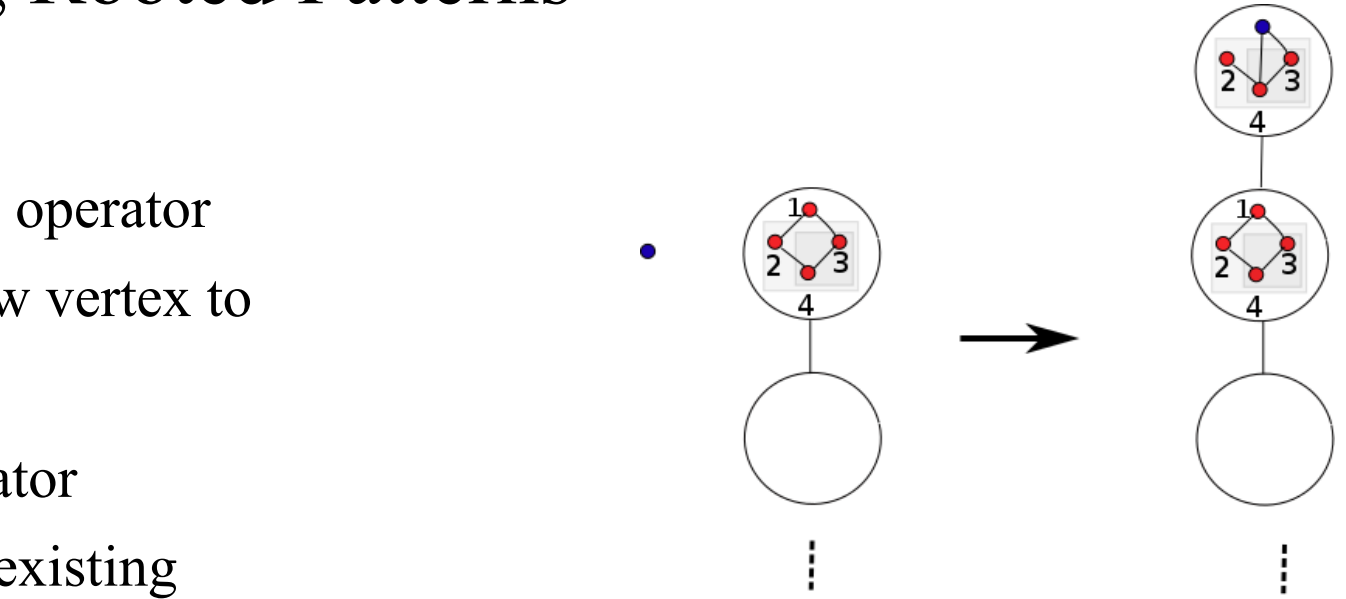
# Rooted Patterns and Root Embeddings

- A rooted graph  $P^X$ , is a graph  $P$  where the set  $X \subseteq V(P)$ ; distinguished.
- Let  $H$  be a database graph
- Let  $\varphi$  be a subgraph homomorphism mapping from  $P$  to  $H$
- $\varphi|_X$ :  $\varphi$  restricted to the vertices in  $X$
- $\varphi|_X$  is called a *root embedding* of  $P^X$  in  $H$
- Two rooted graphs are **equivalent under root embedding** iff they have the same set of root embeddings



# Generating Rooted Patterns

- The **extension** operator
  - Adds a new vertex to a pattern
- The **join** operator
  - Joins two existing patterns



extension

join

# Closed Pattern

- $f$ : maps a root embedding equivalence class  $eq$  to a finite set  $\{P_1^{X_1}, \dots, P_n^{X_n}\} \subseteq eq$  which contains all rooted cores of  $eq$
- 
- $P^X$  is defined as  $\rho_{join}(\dots \rho_{join}(\rho_{join}(P_1^{X_1}, P_2^{X_2}), P_3^{X_3}), \dots, P_n^{X_n})$
- The operator  $\sigma_f$  maps every member of  $eq$  to  $P^X$
- $P^X$  is a **closed** pattern
- $\sigma_f$  is a **closure** operator
  - It is *extensive*, *increasing* and *idempotent*

## s-measure

- Let  $P^X$  be a rooted pattern and  $H$  be a database graph
- To every embedding  $\varphi$  of  $P^X$  in  $H$  a weight  $\omega_\varphi$  is assigned
- **Feasible** assignment:
  - $\forall \varphi \in Emb(P^X, H) : \omega_\varphi \geq 0$
  - $\forall v \in V(H), u \in V(P^X) : \sum_{\varphi \in Emb(P^X, H) | \varphi(u)=v} \omega_\varphi \leq 1$
- **s-measure: minimum feasible assignment**
- Can be computed efficiently for **rooted graphs** when matching operator is **subgraph homomorphism**
  - Without forming overlap graph

# Conclusion

- A new class of patterns: rooted patterns
- Mining patterns with frequent root embeddings
- Mining patterns with minimal s-measure
- A new notion for compactly representing all frequent patterns under homomorphism
  - It gives a closure operator