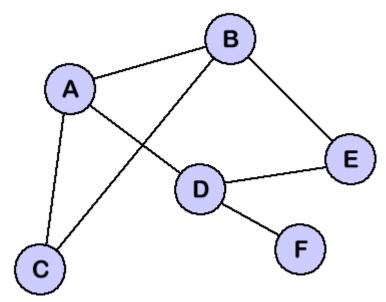
### Karel Vaculík Mining Graph Data

### Outline

- Introduction
- Application domains
- Graph Mining Algorithms

### Introduction

- Graph: *G* = (*V*, *E*)
  - V ... set of nodes,
  - $E \subseteq V \times V...$  set of edges



### Introduction

• Graph: *G* = (*V*, *E*)

. . .

- V ... set of nodes,
- $E \subseteq V \times V...$  set of edges,
- $w: E \rightarrow \mathbb{R}$  ... weight function,
- $\mu: V \rightarrow Lv \dots$  node labeling function,
- $v: E \rightarrow L_E$  ... edge labeling function,

### **Application domains**

- Chemical data analysis
- Computational biology
- Social networking
- Web link analysis
- Computer networks

## **Main Graph Mining Algorithms**

- Clustering
- Classification
- Frequent pattern / substructure mining

### Considerations

- Data properties
  - One large graph vs. set of (smaller) graphs (also transactions)
  - Size
    - Streaming of massive graphs (they are too large to fit in the main memory and random access is slow in large capacity storage devices)
  - Static vs. dynamic

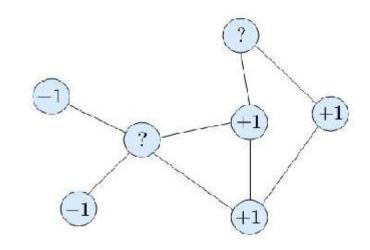
# **Clustering Algorithms**

#### Node clustering

- Based on distance functions for nodes
- Related to minimum cut (polynomially solvable) and graph partitioning (NP-hard) problems
- Applications: determining dense regions (=> summarization, dimensinality reduction), ...
- Graph clustering
  - Based on structural behavior
  - Applications: molecular biology, chemical graphs, XML data, ...

### **Classification Algorithms**

#### Node classification



#### Graph classification

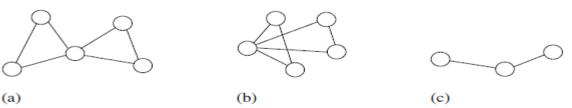


- Pattern ≈ subgraph
- Single-graph setting
  - Frequency: number of pattern occurrences in the single graph.
  - Examples of algorithms: SUBDUE, SEuS, GREW, SIGRAM, GBI
  - Not discussed further
- Graph-transaction setting
  - Frequency (of a pattern): number of graph transactions in which the pattern occurs

### **Apriori-like algorithms**

- Basically two steps:
  - Generation of frequent substructure candidates
    - Based on adding nodes, edges or paths
  - Checking the frequencies of candidates
- Examples of algorithms : AGM, FSG

Checking the frequencies of candidates:(Sub)graph isomorphism



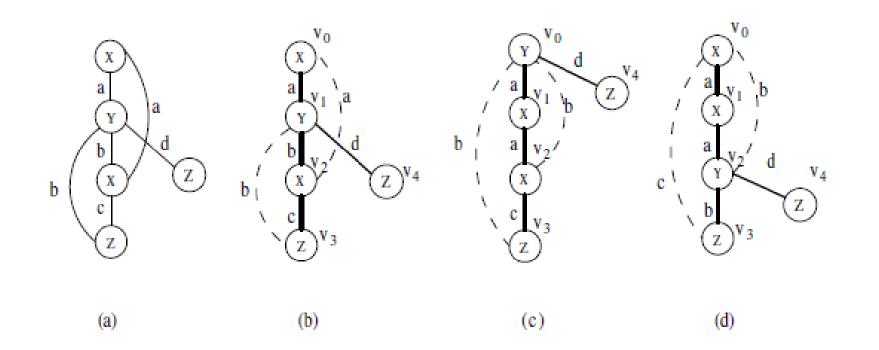
- Canonical labeling
- Unique code for the set of graphs with the same topological structure and the same labeling
  Both problems are not known to be either in P or in NP-complete → relaxed problems

#### Pattern growth algorithms

- gSpan
- Gaston
- ...

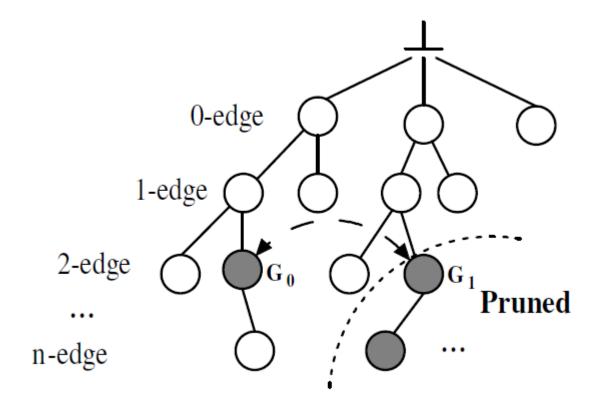
#### gSpan

- Without candidate generation
- Minimum DFS code as canonical label



### gSpan

- Without candidate generation
- Minimum DFS code as canonical label
- DFS lexicographic orderering on DFS codes
  → DFS code tree



### gSpan

- Without candidate generation
- Minimum DFS code as canonical label
- DFS lexicographic orderering on DFS codes
  → DFS code tree
- Searching frequent patterns: traversing DFS code tree

### References

- Diane J. Cook, Lawrence B. Holder. *Mining* graph data. John Wiley and Sons, 2007.
- Charu C. Aggarwal, Haixun Wang. Managing and Mining Graph Data. Springer, 2010
- X. Yan and J. Han. gSpan: Graph-based substructure pattern mining. In Proceedings of 2002 IEEE International Conference on Data Mining (ICDM), pp. 721–724, 2002.