General Graph Refinement with Polynomial Delay

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Graph enumeration Earlier work

Introduction - Enumeration

Pattern Mining:

► Given a language L and an (often 'anti-monotonic') predicate *interesting*, list all interesting patterns.

Many pattern mining algorithms perform essentially two tasks:

- Generating candidate patterns
- Checking interestingness of patterns (e.g. counting frequency)

Graph enumeration Earlier work

Introduction - Enumeration

This paper deals with the first step:

- ► Given a (graph) language *L*
- ► Enumerate all (graph) patterns p ∈ L from small to large
- ... allowing for suitable pruning
- ... and do not generate duplicates

In other words, we assume the predicate "interesting" to be evaluable efficiently.

Avoiding duplicates is non-trivial: for graphs, isomorphism is not known to be polynomial.

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Introduction - Enumeration

Complexity measures:

- Polynomial time total running time bounded by polynomial in input
- Polynomial delay time needed for generating next solution is bounded by polynomial in size of input
- Incremental polynomial time time needed for generating next solution is bounded by polynomial in size of input and output so far
- Output polynomial time total running time bounded by polynomial in input + output.

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Introduction - Enumeration

Applications of graph enumerating:

- Pattern mining (e.g. candidate generation)
- Combinatorics (e.g. counting)

Search

Graph enumeration Earlier work

Introduction - Earlier work

Earlier work

- Data mining and existing enumeration algorithms
- 'Simple' cases
- Graph listing (L.A. Goldberg)

Introduction

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Introduction - Earlier work

Literature:

- Systems:
 - ▶ gSpan, AGM, ...
 - McKay's Nauty
- All devote much attention to efficient candidate generation
- Methods: Canonical forms, Joining operators, ...
- But none of them proves polynomial delay.

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Special cases:

- Itemsets: Apriori runs with polynomial delay
- Free trees (Wright'86)
- Outerplaner graphs: (Horvath & Ramon'06)

▶

Graph enumeration Earlier work

Introduction - Earlier work

L.A.Goldberg's graph listing work

- One can enumerate all graphs of size at most n without duplicates with delay O(n⁶).
 - Relies on "most graphs are easy"
- If p is an allmost sure property, then one can enumerate all graphs that satisfy p (in the original paper including duplicates) with polynomial delay. Every FOL property is either allmost always true or allmost always false.
- For pattern mining, typically not all patterns are interesting. In particular, the interesting case is when there are only few interesting patterns.

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Contribution

Given:

- a language \mathcal{L} of graphs
- ► a dense refinement schema for L
- an anti-monotonic constraint *interesting* (that can be evaluated efficiently)

we enumerate

- all interesting patterns
- with polynomial delay

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Better?

- More general than L.A.Goldberg (*L* or *interesting* can produce only 'difficult' graphs)
- More general than specialized methods
- More efficient (better proven assymptotic complexity) than gSpan, AGM, ...



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Refinement description

Definition A refinement description is a pair (V(r), E(r)) of vertices and edges.



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Refinement schema

Definition

A refinement schema is a pair (ρ^+, ρ^-) of functions from graphs onto sets of refinement descriptions. Then,

- r ∈ ρ⁺(g) are the downward refinement descriptions of g and g + r are (downward refinement / specialisation / supergraph)
- r ∈ ρ⁻(g) are the upward refinement descriptions of g and g − r are (upward refinement / generalisation / subgraph)
- ► Consistent: $r \in \rho^+(g)$ iff $r \in \rho^-(g + r)$
- ► Isomorphism-invariant= If $r \in \rho^+(g)$ and $g \equiv_{\varphi} h$, then $\varphi(r) \in \rho^+(h)$.

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Refinement schema (ex)

Connected graphs:

Combining refinements

↑ is an operator 'lifting' an upward refinement. In particular, if $r_1, r_2 \in \rho^-(g)$ and removing r_2 from $g - r_1$ makes sense, $r_2 \uparrow^g r_1$ is the 'lifted' version of r_2 which can be applied to $g - r_1$.

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Combining refinements



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Graph of parents $\overline{GOP_{\rho^-,\uparrow}}(g)$



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Dense refinement schema

Definition

A dense refinement schema is a triple $(\rho^+, \rho^-, \uparrow)$ s.t.

- (ρ^+, ρ^-) is a refinement schema
- ▶ The graph $GoP_{\rho^-,\uparrow}(g)$ is connected.

This requirement has far-reaching implications, but still allows for dense refinement schemas for a wide range of graph classes.

Lemma

For every dense refinement schema, one can define a size function $|\cdot|$ such that for every graph g and $r \in \rho^+(g), |g + r| = |g| + 1$. J. Ramon & S. Nijssen (K.U.Leuven) General Graph Befinement MLG. August 2007

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Complexity theorem

Theorem

Consider a dense refinement schema (ρ^+ , ρ^- , \uparrow). If the following conditions hold:

- ▶ |ρ⁺(g)| and |ρ⁻(g)| are O(|g|)
- For every $r \in \rho^+(g)$, $|r \cap g|$ is bounded by a constant
- The refinements r have an easy structure or |r| is bounded by a constant.

Then, our algorithm runs with polynomial delay

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Data structure

- Our algorithm builds a data structure storing one representative for every isomorphism class.
- This may take exponential space in the input size (if the number of solutions is exponential), but for mining purposes this is not to be expected
- Given any graph, the data structure can be used to search the representative in polynomial time.

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Key idea of algorithm

- We construct a candidate pattern from a parent.
- Since the refinement schema is dense, we can hop from one parent to the next one through grandparents.
- In this way we can avoid to generate children from different parents that are isomorphic.
- Finally, we can avoid isomorphic children from one single parent by (incrementally) computing automorphism groups.

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What can we enumerate?

- 'Monotone' classes
- Hereditary classes with bounded degree
- Classes restricted to connected graphs

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'Monotone'¹ classes

interesting is monotone iff for every graph *G* such that *interesting*(*G*), and for every subgraph $S \leq G$, *interesting*(*S*).

- Minimal (efficient) frequency constraints under subgraph isomorphism.
- Maximal vertex & edge counts
- Maximal degree, treewidth, ...
- Forbidden subgraphs and minors

¹ : in different communities, monotone and anti-monotone are reversed

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Corollary

Let interesting be an antimonotonic predicate on the set of all (connected) graphs. Then, we can list all interesting graphs g with delay $O(|V(g)|^5)$.

Compare with L.A.Goldberg: enumerates all graphs with delay $O(|V(g)|^6)$

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Hereditary classes with bounded degree

interesting is hereditary iff for every graph *G* such that *interesting*(*G*), and for every **induced** subgraph $S \leq_i G$, *interesting*(*S*).

- Minimal (efficient) frequency constraints under induced subgraph isomorphism.
- Maximal degree, treewidth, edge count, vertex count
- ► Forbidden induced subgraphs, e.g. claw-free graphs (graphs not containing an induced claw K_{1,3})

▶ . . .

Connected graphs

We can combine both previous examples with the constraint that the graphs should be connected. (even though e.g. connectedness is not closed under taking subgraphs).

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Conclusions

We proposed:

- A (more) general method
- to list all interesting graphs
- of a wide range of graph classes (dense refinement schema needed)
- with polynomial delay

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Open problems

Theory:

- Enumerate graphs under induced subgraph isomorphism (e.g. claw-free graphs, no degree bound).
- How about homomorphism (aka theta-subsumption in ILP) ? (some negative results are already known).
- Larger refinement steps ? (e.g. for closed pattern mining)

Practice:

- Can a canonical form help?
- Experiments?

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Questions or comments?

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