

# Enumeration algorithms in graphs

Aurélie Lagoutte

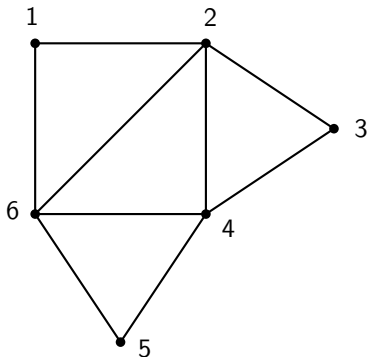
G-SCOP, Grenoble INP / Université Grenoble Alpes

*G@g project – January 2026*

# Enumeration : a typical example

**Input:** Graph  $G$

**Output :** The list of all **inclusion-wise maximal** stable sets of  $G$

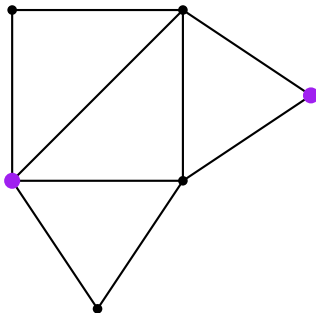


$\{1, 3, 5\}, \{1, 4\}, \{2, 5\}, \{3, 6\}$

# Focus on easy problems

**Input:** Graph  $G$

**Output :** one **inclusion-wise maximal** stable set.  $\in P$  (greedy)



Not to be confused with :

**Input:** Graph  $G$

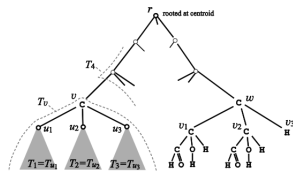
**Output :** a stable set of **maximum** size

NP-complete

# Enumerating in graphs : useful cases

- Graph databases : answer to a query
- Graph model is not exact : some solutions are *best* based on qualitative criteria, we have to examine them one by one
- Identify all problematic (or interesting !) patterns in a network

*Application fields: bioinformatics (phylogenetic trees), chemistry (molecule structure), complex system modeling, databases...*



*Diagram of a stereoisomer<sup>1</sup>*

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<sup>1</sup>[Comparison and Enumeration of Chemical Graphs](#), T. Akutsu, H. Nagamochi, *Comp. and Struct. Biotechnology Journal*

# Complexity for enumeration problems

In most cases : exponential number of solutions to output

(ex:  $3^{n/3}$  max. stable sets)

⇒ *Good complexity measure?*

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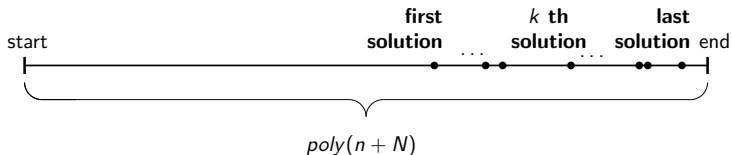
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## ① Output-polynomial

Input of size  $n$ ,  $N$  solutions to output.



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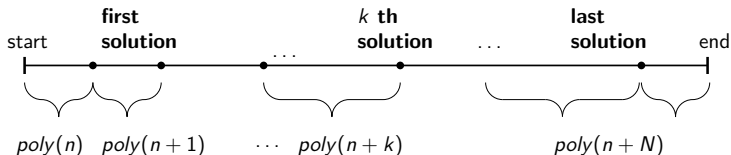
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- 1 Output-polynomial
- 2 Incremental polynomial

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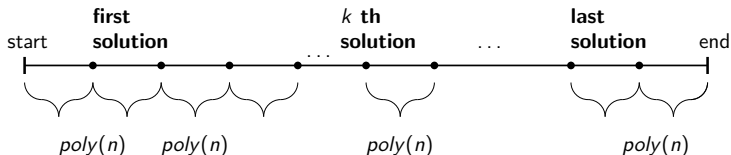
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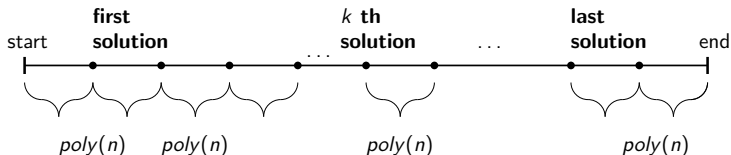
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poly space vs.  
exponential space

Input of size  $n$ ,  $N$  solutions to output.



# Interesting objects to enumerate

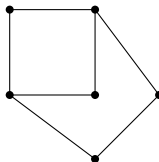
- ① Inclusion-wise minimal transversal of a hypergraph
- ② Inclusion-wise minimal dominating sets
- ③ Spanning trees
- ④ "Structured patterns" : inclusion-wise max. stable sets or cliques ...
- ⑤ **Inclusion-wise minimal " $\Pi$ -fixings" of a graph**
  - Completions, deletion, induced subgraphs of a graph ...  
... satisfying a given property  $\Pi$

# Minimal fixings

3 variants

We want to satisfy a given property  $\Pi$

Example :  $\Pi = C_4$ -free  
(contains no induced  $C_4$ )



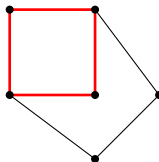
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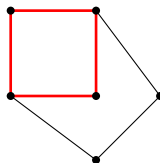


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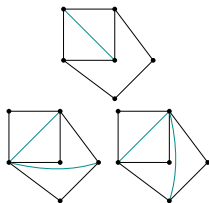
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Fixing by  
adding edges

**Min.  $\Pi$ -completion**



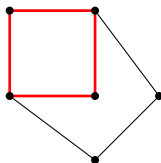
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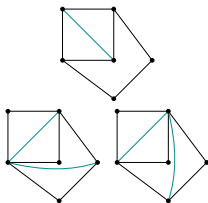
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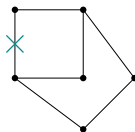
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**Min.  $\Pi$ -completion**



Fixing by  
deleting edges

**Min.  $\Pi$ -deletion**



+ 3 others

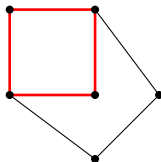
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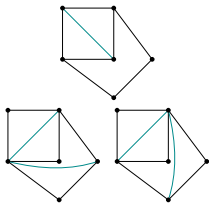
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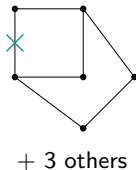
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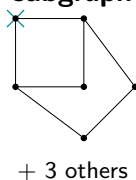
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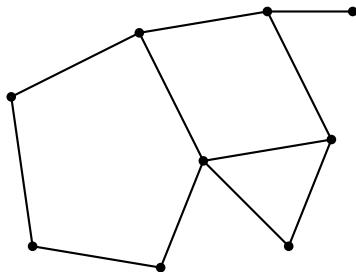
Fixing by  
deleting vertices  
**Max.  $\Pi$ -induced  
subgraph**



# Chordal completion

*Chordal completion* of a graph  $G$  : a completion of  $G$  that is chordal (no chordless cycle of length  $\geq 4$ ).

A chordal completion of  $G$  is also called a *triangulation* of  $G$  or sometimes a *fill-in*.

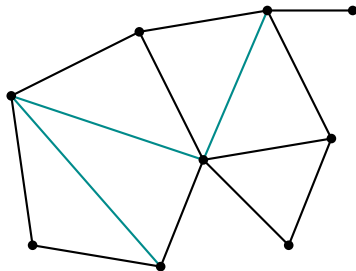




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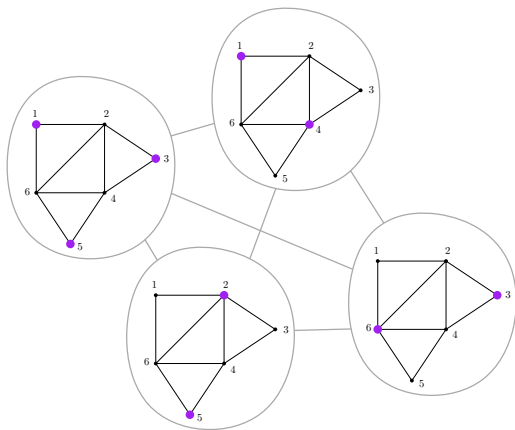


# Algorithmic methods to enumerate

# General principle of an enumeration algorithm

- Metagraph of solutions : traversal of this metagraph

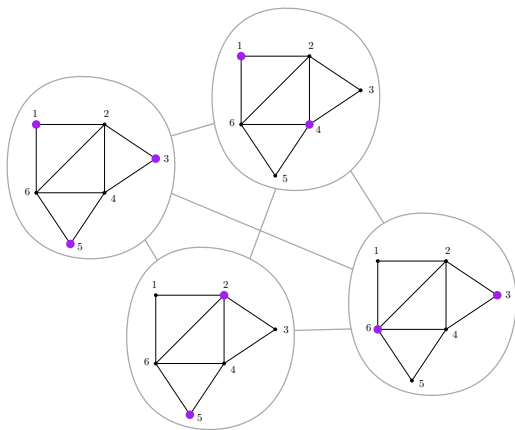
Example with  
maximal stable sets  
Solution metagraph



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- outputting each solution once

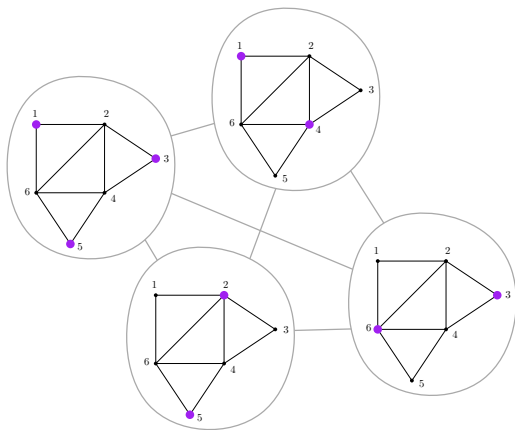
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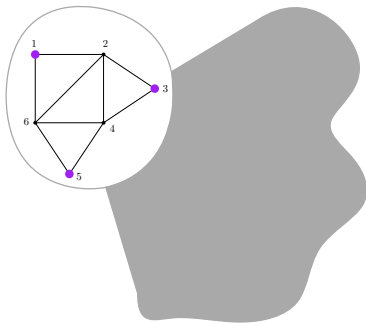
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Three classical methods :

Goal : get polynomial delay

+ poly space for 1 et 2 (and sometimes 3)

① *Flashlight search or Binary partition*

[Read, Tarjan '75]

② *Reverse search*

[Avis, Fukuda '96]

③ *Proximity Search*

[Conte, Uno '19]

[Conte, Grossi, Marino, Uno, Versari, '21]

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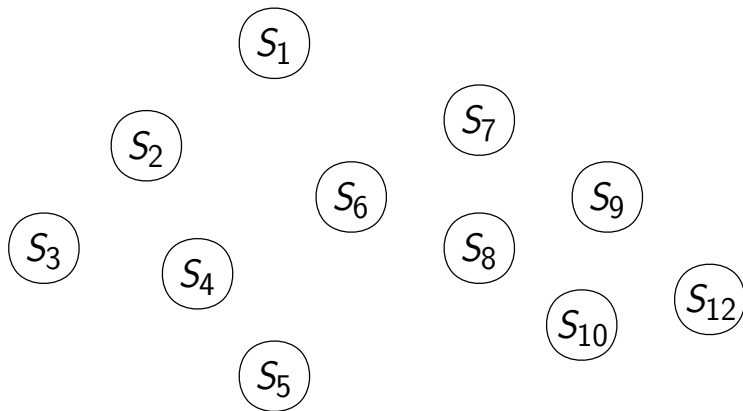
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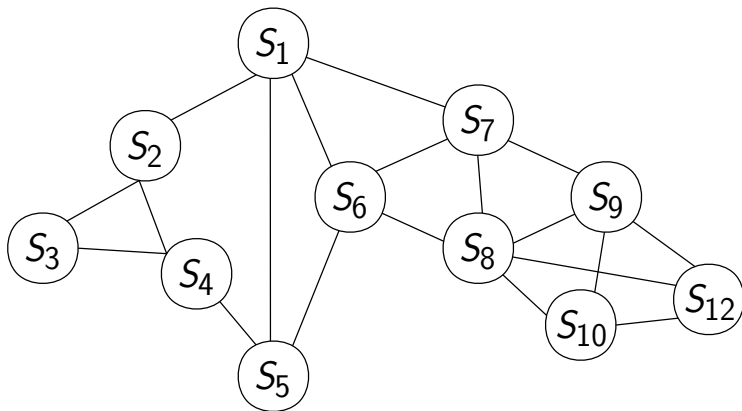
# Reverse search

Solution space



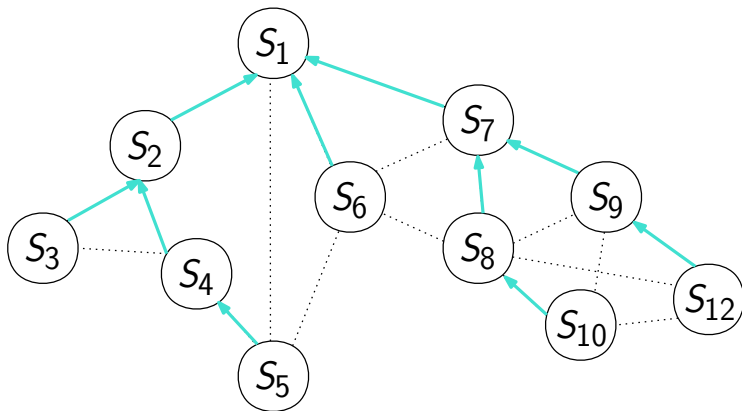
# Reverse search

## Solution metagraph



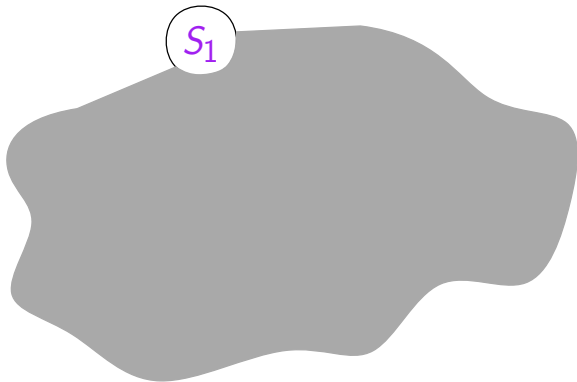
# Reverse search

Solution tree

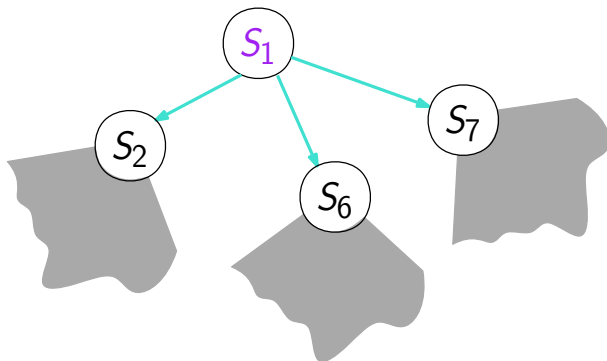


## *Reverse search*

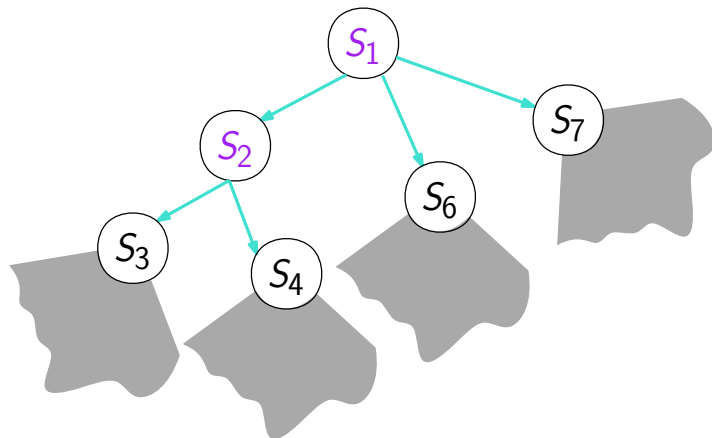
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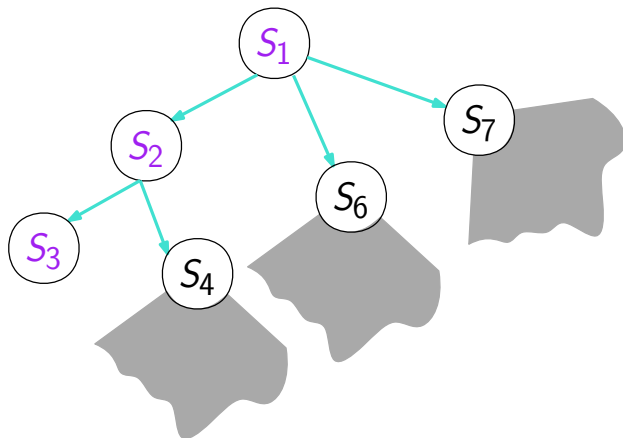
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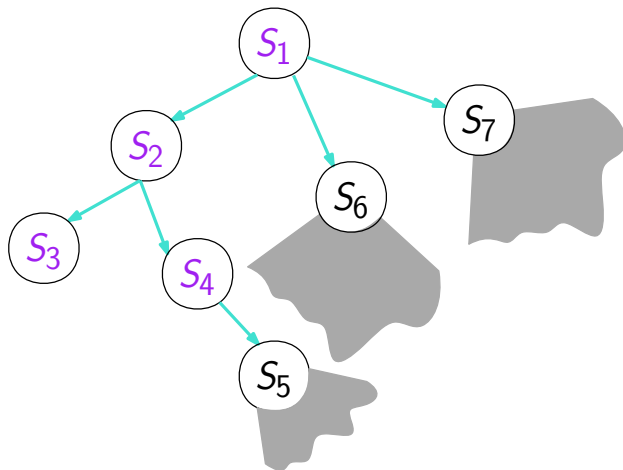
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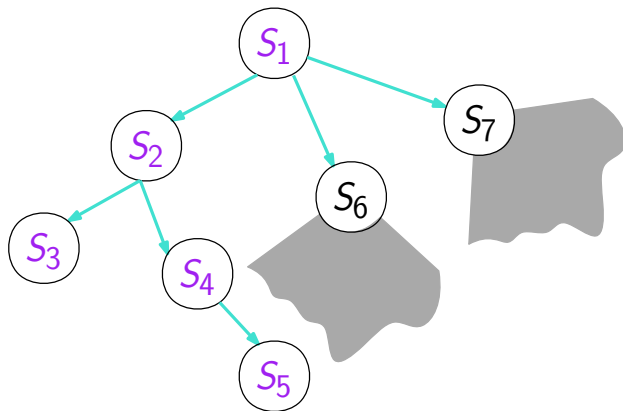


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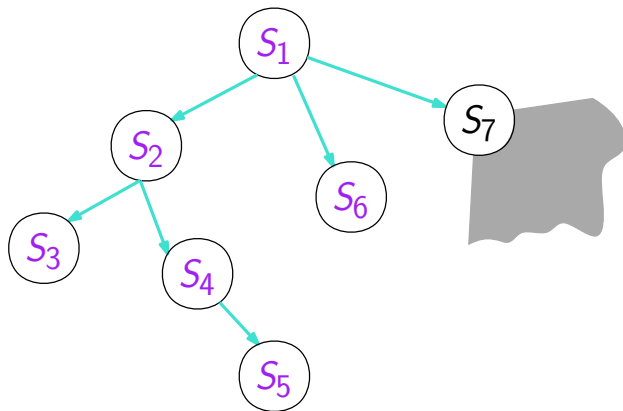




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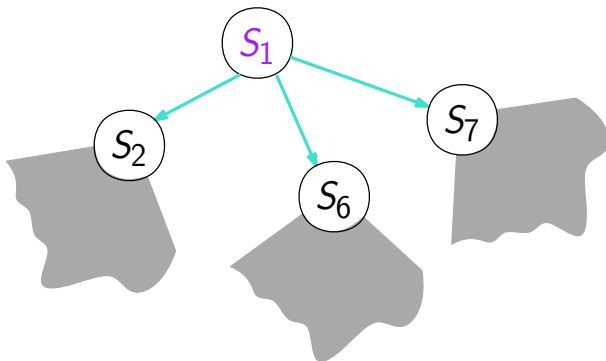
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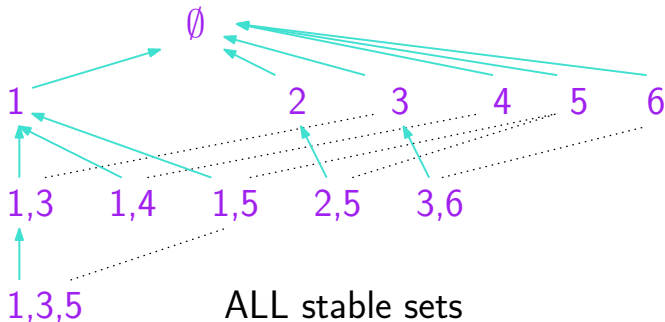
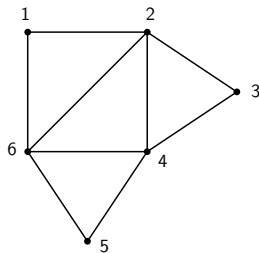
## Reverse search

To have *Reverse search* run in poly delay and space :

Generate in poly time and space the *children* of a solution  
(each solution must have **a single** father)

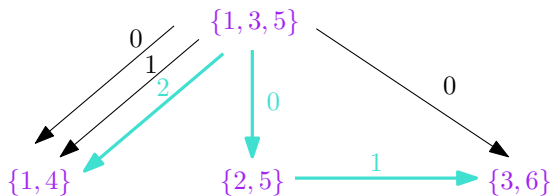
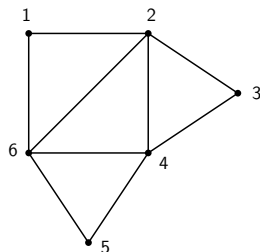


# Reverse Search



ALL stable sets

# Reverse Search for Maximal Stable Sets



$S \xrightarrow{i} S'$  if  $S \cap \{v_1, \dots, v_i\} = S' \cap \{v_1, \dots, v_i\}$  and  $S$  is the lexicographically smallest among all solutions containing  $S' \cap \{v_1, \dots, v_i\}$

$P$  is the father of  $S$  if  $P \xrightarrow{i} S$  and there is no arc to  $S$  indexed  $> i$

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Designed for enumerating *inclusion-wise maximal* (or min.) solutions to a problem.

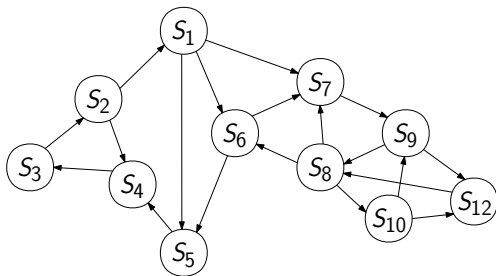
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Designed for enumerating *inclusion-wise maximal* (or min.) solutions to a problem.

Recursive depth-first search  
(rather classical) with conditions :

- generate **neighbors** of a vertex in poly time  
→ *poly degree*

Solution metagraph





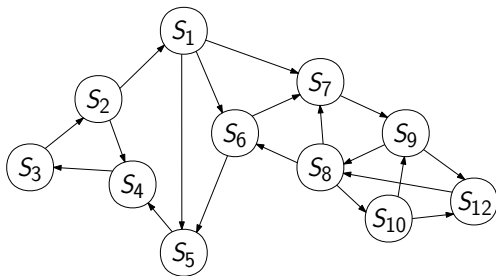
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→ *exponential space* :-)

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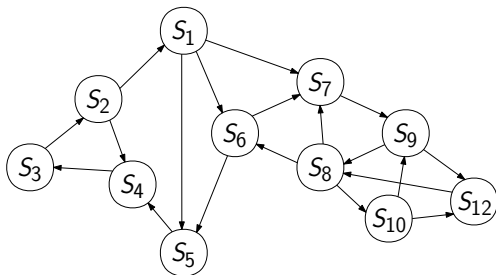
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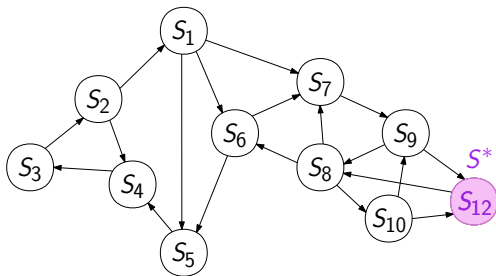
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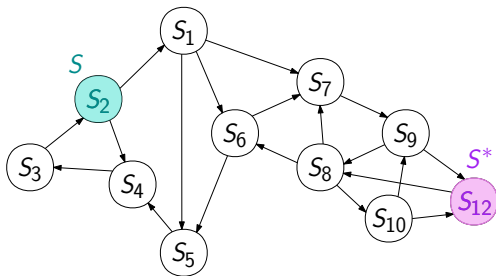
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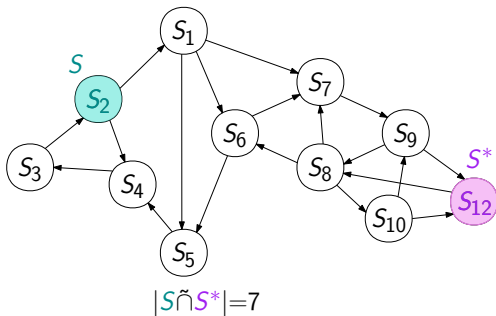
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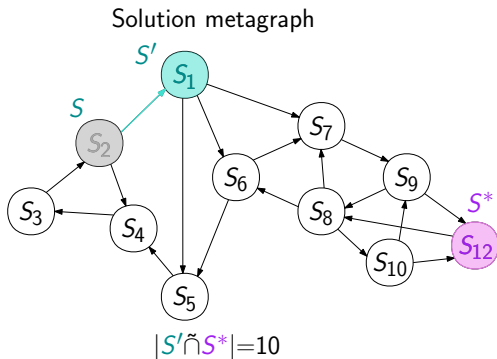


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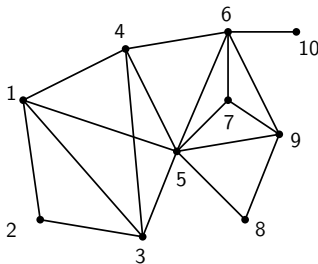
# Enumerate chordal fixings

**Chordal** graph : no chordless cycle of length  $\geq 4$ .

### Simplicial vertex

Any chordal graph has a **simplicial vertex** : its neighborhood is a clique.

$\Rightarrow$  **Perfect elimination ordering** : remove simplicial vertices one by one.



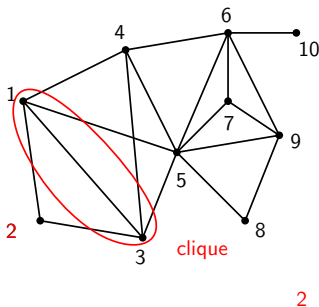


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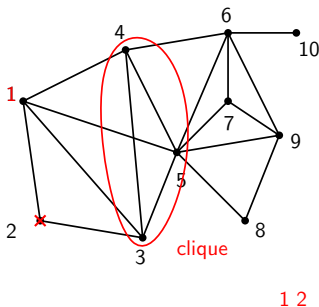


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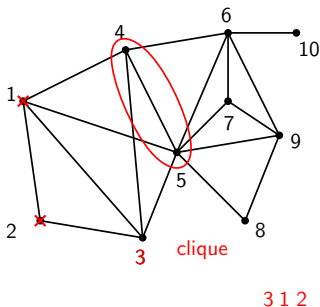


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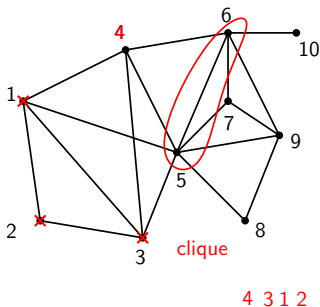


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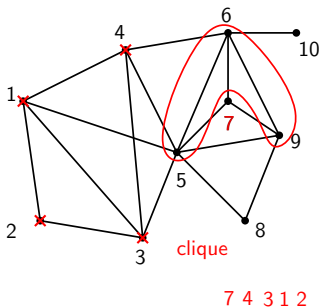


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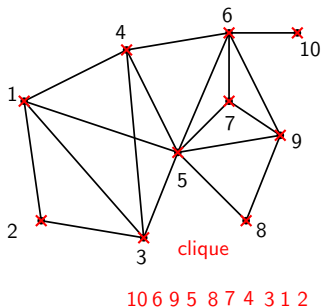


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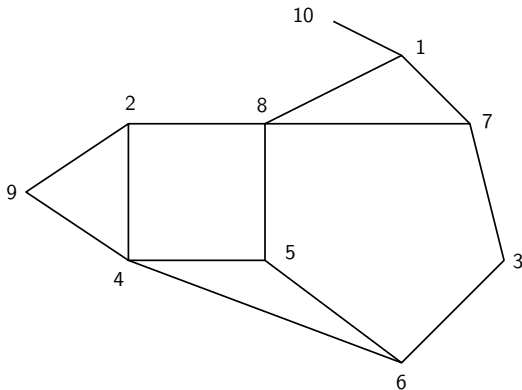
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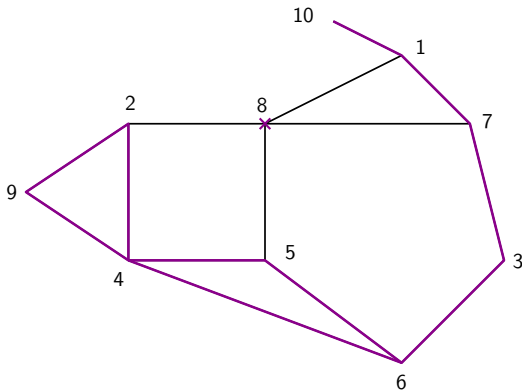
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*Goal* : enumerate all **induced subgraphs** of  $G$  that are **chordal**  
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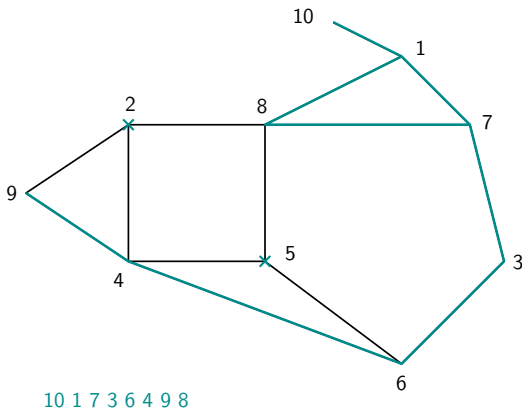


10 1 7 3 6 4 9 5 2



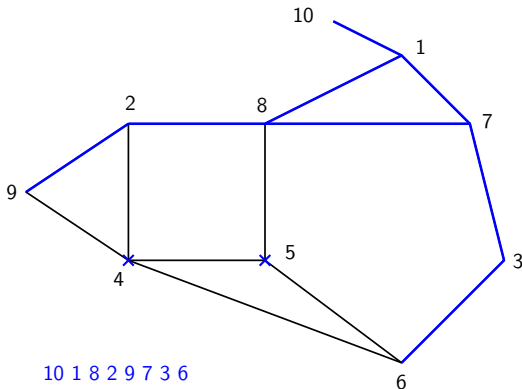
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*Goal* : enumerate all **induced subgraphs** of  $G$  that are **chordal**  
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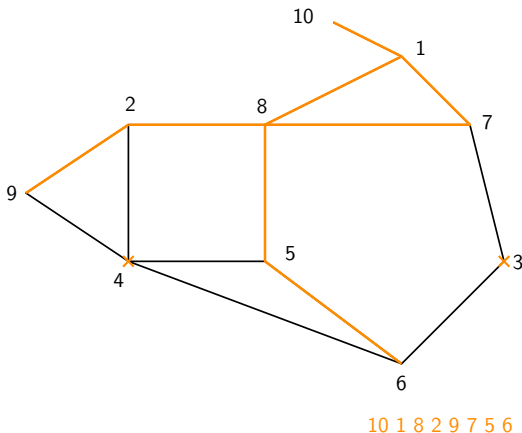
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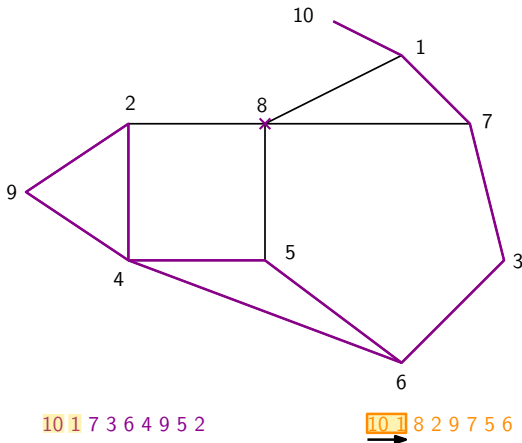
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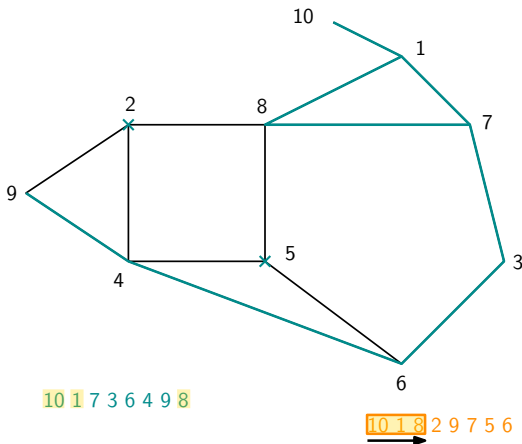
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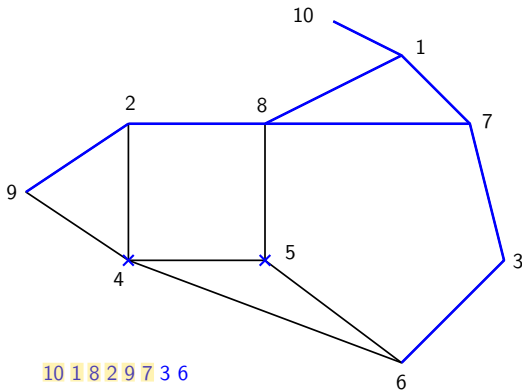
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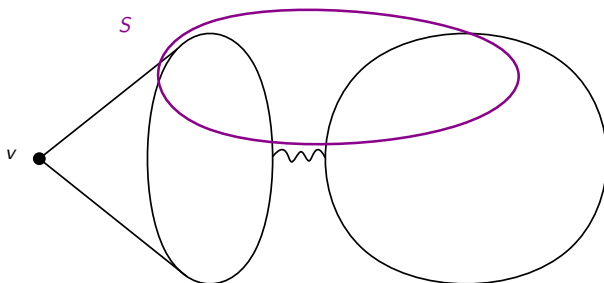
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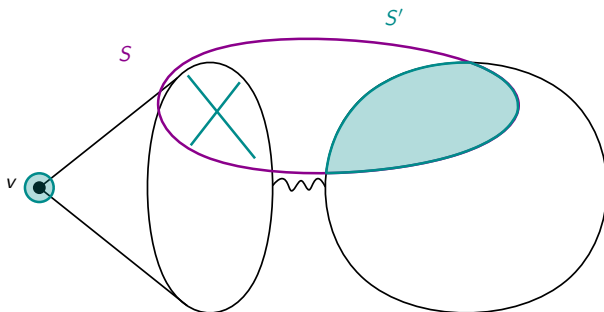
10 1 8 2 9 7 3 6

10 1 8 2 9 7 5 6  
→

# Neighbors in the solution metagraph

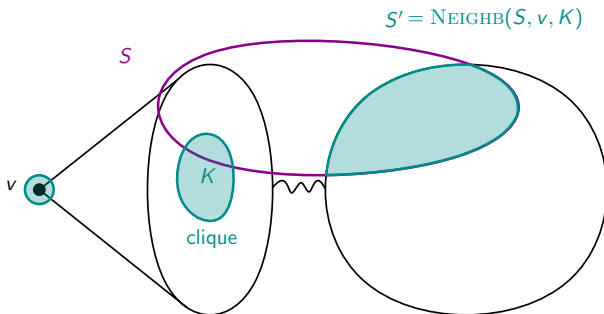


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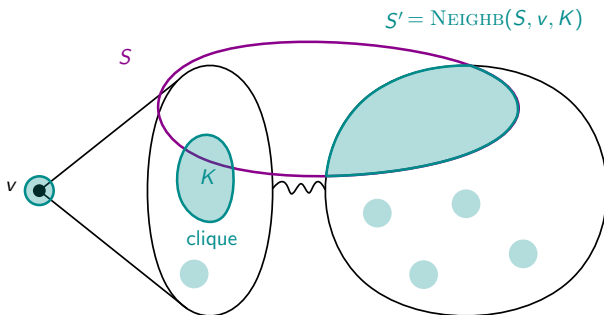




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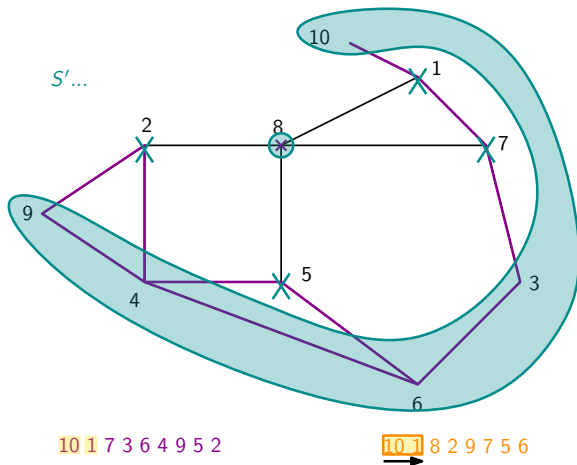


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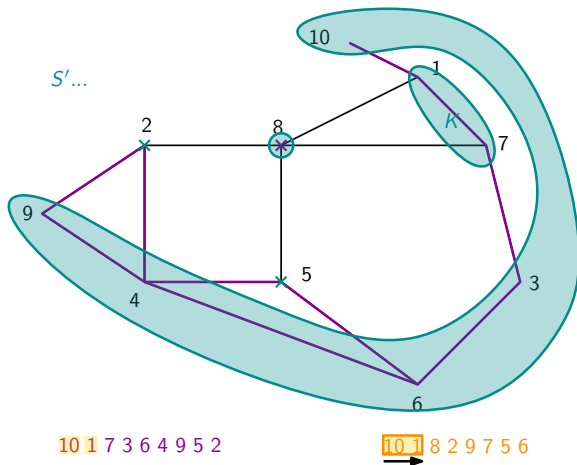




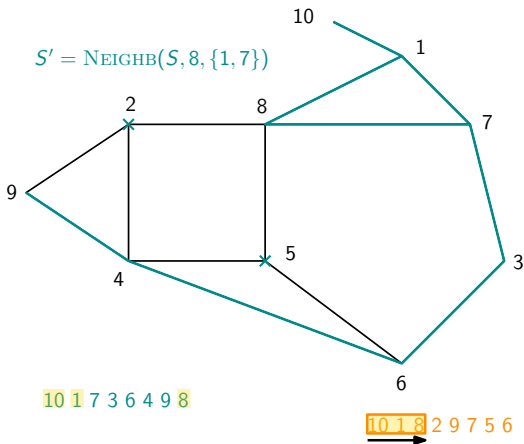
# Neighbors in the solution metagraph : example



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# Chordal fixings : recap

We have all the ingredients for Proximity Search to work :

- ① Proximity between two solutions  $S$  and  $S^*$  is defined
- ②  $Neighbors(S)$  is computable in polytime
- ③ Lemma proving that we can always find a neighbor with higher proximity with a target solution, thanks to the perfect elimination ordering (not shown here)

→ Enumeration of chordal fixings by inclusion-wise min. deletion of vertices in polynomial delay

# Other type of fixings : Chordal completions

→ Enumerating minimal triangulations of a graph ?



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Theorem [Brosse, Limouzy, Mary, 2021<sup>+</sup>]

There exists a polynomial delay polynomial space algorithm to enumerate all inclusion-wise minimal chordal completion of a graph  $G$  given in input.

→ *Proximity Search* with careful arguments (to get polynomial space in particular)

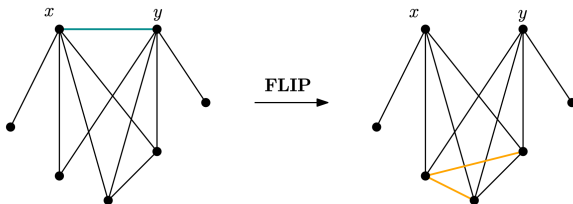
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Removing edge  $xy \rightarrow$  Common neighb. turned into clique

# Thank you