

# Maximum Independent Set Problem and Effective Methods for its Solving

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# Outline

- 1 Classic  $\mathcal{NP}$ -hard Problem
  - Complexity classes and their relations
  - MISP and MCP
- 2 Applications
- 3 An Exact Methods for MISP
  - Dynamic Programming
  - Bron-Kerbosh Algorithm
- 4 An Effective Methods for MISP
  - Variable Neighborhood Search (VNS)
  - Extended Simplicial Vertex Test (ESVT)

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# Class $\mathcal{NP}$

Let's consider any search-problem.

**Require:**  $I, S$  {an instance and a candidate solution}

**Ensure:** true/ false

Class  $\mathcal{NP}$  (non-deterministic polynomial-time) is a class of search-problems, for which there exist polynomial-time algorithms to check a candidate solutions.

# Class $\mathcal{P}$

Let's consider optimisation problem corresponding to some search-problem.

**Require:**  $I$  {an instance}

**Ensure:**  $\min S$  ( $\max S$ ) {an optimal solution}

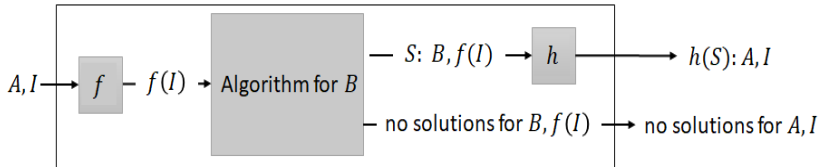
Class  $\mathcal{P}$  is a class of search-problems, for which there exist polynomial-time algorithms to solve corresponding optimization problems.

Are there search-problems, for which corresponding optimisation problems can't be solved by polynomial-time algorithm, that is

$$\mathcal{NP} \neq \mathcal{P}?$$

# Class $\mathcal{NPC}$

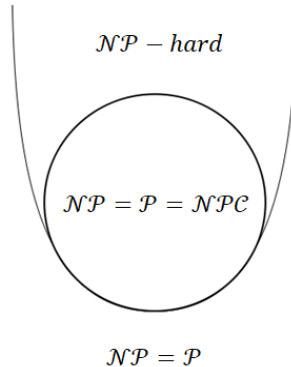
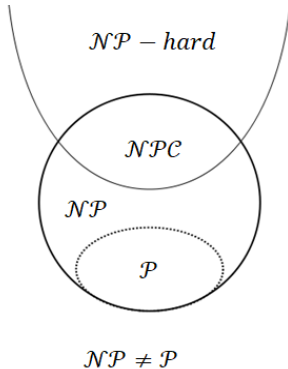
Let  $f$  and  $h$  be polynomial-time algorithms.



Class  $\mathcal{NPC}$  ( $\mathcal{NP}$ -complete) is a class of search-problems, to each of which one can reduced any problem from  $\mathcal{NP}$ .

## Class $\mathcal{NP}$ – hard

Class  $\mathcal{NP}$  – hard is a class of optimization problems, for which corresponding search-problems belong to  $\mathcal{NPC}$ .



# Outline

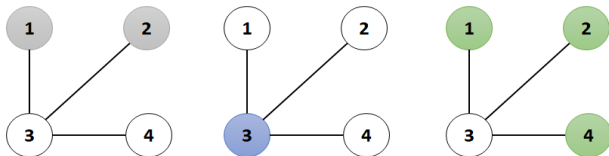
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# Independent Sets

Independent set (IS) of an undirected graph  $G = (V, E)$  is any empty induced subgraph, that is

$S, S \subseteq V$ ,  $S$  is an IS, if for all  $v, u \in S: (u, v) \notin E$ .



An IS:  $\{1, 2\}$ .

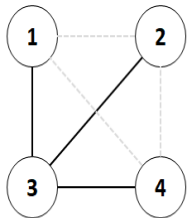
Maximal IS couldn't be extended by any vertex:  $\{3\}$ .

Maximum IS has the largest cardinality:  $\{1, 2, 4\}$ .

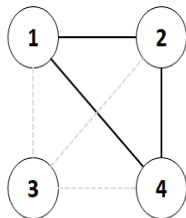
# Complementary Graph

A complementary graph  $\tilde{G} = (V, \tilde{E})$  has the same set of vertices and

$$(u, v) \in \tilde{E} \text{ if and only if } (u, v) \notin E .$$



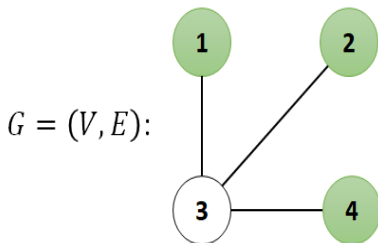
Initial graph  $G = (V, E)$ .



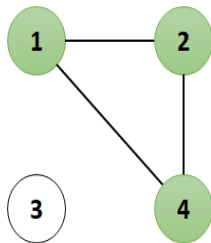
Complementary graph  $\tilde{G} = (V, \tilde{E})$ .

# Maximum Clique

Maximum clique (MC) of an undirected graph  $G = (V, E)$  is the largest (by cardinality) complete subgraph.



$\tilde{G} = (V, \tilde{E})$ :



$$\text{MIS}(G) = \text{MC}(\tilde{G}) .$$

# MIS Problem

## Class $\mathcal{NP}$

**Require:**  $G = (V, E), S \subseteq V$ .

**Ensure:** Is the set  $S$  an IS of the initial graph  $G$  (yes/ no)?

## Class $\mathcal{NPC}$

**Require:**  $G = (V, E), k \in \mathbb{Z}$ .

**Ensure:** Is there in the initial graph  $G$  an IS  $S$  of the size  $k$  (yes/ no)?

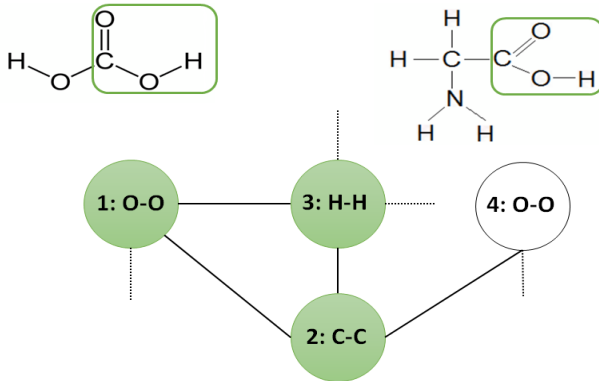
## Class $\mathcal{NP}$ -hard

**Require:**  $G = (V, E)$ .

**Ensure:** MIS  $S$  of the initial graph  $G$  (optimal solution).

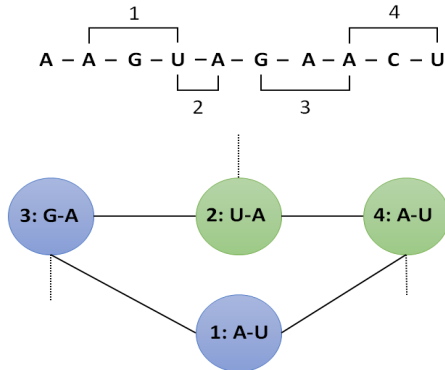
# Chemistry

MC in a special graph is a maximum common substructures of molecules.



# Bioinformatics

MC in a special graph is a secondary structure of RNA with the largest number of paired bases.



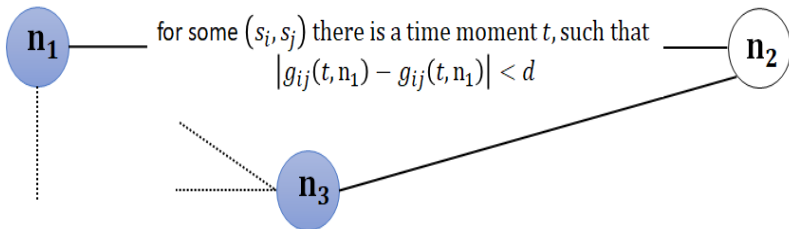
# Transportation Management

Let  $\vec{G} = (S, E)$  be a railway network and let

$$N = \{n: (s_1, s_2, g_{12}(t, n)), (s_2, s_3, g_{23}(t, n)), \dots\} \text{ —}$$

be the set of potential routes for trains moving.

Any IS of conflict-graph  $G = (N, \mathcal{E})$  is the set of feasible schedules for execution of transportations plan.



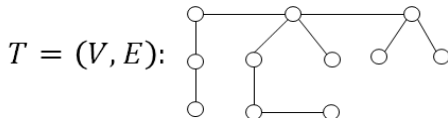
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# Polynomial class of MISP

A tree is an undirected connected acyclic graph.



Class  $\mathcal{P}$

**Require:**  $T = (V, E), k \in \mathbb{Z}$ .

**Ensure:** Is there in the initial tree  $T$  an IS  $S$  of the size  $k$  (yes/no)?

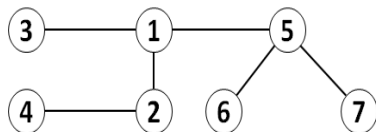
# Dynamic Programming for MISP

Function MIS.Tr ( $v$ )

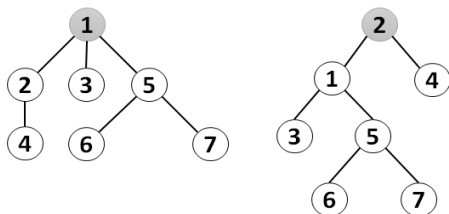
```
1: if  $I[v] \neq 0$  then  
2:   return  $I[v]$   
3: else  
4:    $child.sum = 0$   
    $g.child.sum = 0$   
5:   for  $i=1$  to  $child.num$  do  
6:      $child.sum \leftarrow child.sum + mis.tree(child[i])$   
7:   for  $i=1$  to  $g.child.num$  do  
8:      $g.child.sum \leftarrow g.child.sum + MIS.Tr(child[i])$   
9:    $I[v] = \max\{child.sum; g.child.sum + 1\}$   
   return  $I[v]$ 
```

# Dynamic Programming

Let a tree  $T = (V, E)$  be given.



One can choose any vertex of an initial tree as a root.



# Dynamic Programming

MIS.Tr (1):

$$I[1] = \max\{I[2] + I[3] + I[5]; 1 + I[4] + I[6] + I[7]\}$$

$$I[2] = \max\{I[4]; 1\}$$

$$I[4] = 1$$

$$I[2] = 1$$

$$I[3] = 1$$

$$I[5] = \max\{I[6] + I[7]; 1\}$$

$$I[6] = 1$$

$$I[7] = 1$$

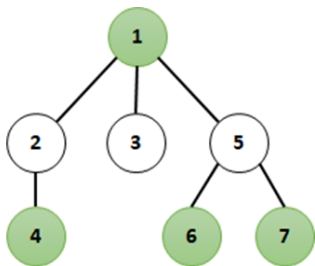
$$I[5] = 2$$

$$I[4] = 1$$

$$I[6] = 2$$

$$I[7] = 1$$

$$I[1]=4$$




$$O(n + m)$$

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# Bron-Kerbosh Algorithm

 Bron C, Kerbosh J. (1973) Algorithm 457 — Finding all cliques of an undirected graph, Comm. of ACM, 16, pp. 575–577.

M: = the current independent set of vertices;

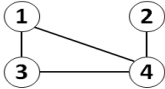
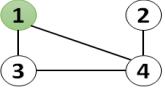
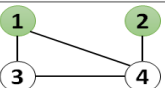
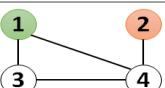
K: = the set of candidates which could be included in MIS under construction;

P: = the set of excluded vertices.

# Bron-Kerbosh Algorithm

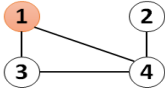
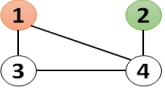
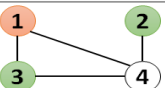
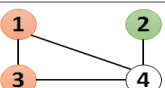
```
1: while  $K \neq \{\}$  or  $M \neq \{\}$  do  
2:   if  $K \neq \{\}$  then  
3:      $v \leftarrow K.\text{first}$   
     push  $v, M, K, P$   
      $M \leftarrow M + \{v\}$   
      $K \leftarrow K - \mathcal{N}(v) - \{v\}$   
      $P \leftarrow P - \{v\}$   
4:   else  
5:     return  $M$   
6:   pop  $v, M, K, P$   
      $K \leftarrow K - \{v\}$   
      $P \leftarrow P + \{v\}$ 
```

# Bron-Kerbosh Algorithm

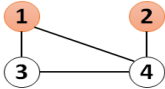
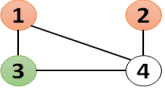
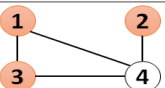
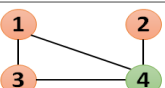
	M	K	P	v		
1	{ }	1234	{ }	1	$K \neq \{ \};$ $v = K.first$	
2	1	2	{ }	2	push M, K, P; $K \neq \{ \};$ $v = K.first$	
3	12	{ }	{ }		push M, K, P; $K = \{ \};$ ► $M = \{12\}$	
4	1	{ }	2		pop M, K, P (go to 2); $K = \{ \};$ ► $M = \{1\}$	



# Bron-Kerbosh Algorithm

	M	K	P	v		
5	{ }	234	1	2	pop M, K, P (go to 1); K ≠ { }; v=K.first	
6	2	3	1	3	push M, K, P; K ≠ { }; v=K.first	
7	23	{ }	1		push M, K, P; K = { }; ► M={23}	
8	2	{ }	13		pop M, K, P (go to 6); K = { }; ► M={2}	


# Bron-Kerbosh Algorithm

	M	K	P	v		
9	{ }	34	12	3	pop M, K, P (go to 5); K ≠ { }; v=K.first	
10	3	{ }	12		push M, K, P; K = { }; ► M={3}	
11	{ }	4	123	4	pop M, K, P (go to 9); K ≠ { }; v=K.first	
12	4	{ }	123		push M, K, P; K = { }; ► M={4}	

# Bron-Kerbosh Algorithm

	M	K	P	v		
13	{ }	{ }	1234		pop M, K, P (go to 11); K = { }; M = { }	<pre> graph TD     1 --- 3     1 --- 4     3 --- 4             </pre>
					STOP	$O(3^{n/3})$

All ISs of initial graph:  $\{1\}, \{2\}, \{3\}, \{4\}, \{1, 2\}, \{2, 3\}$ .

-  E. Tomita, A. Tanaka, H. Takahashi (2006) The worst-case time complexity for generating all max cliques and computation experiments. Theoretical Computer Sciences, Vol. 363, Issue 1, pp. 28–42.

# Computational Results

$n$ (number)	density (%)	$\alpha(G)$ (number)	$t$ (sec)
50	20	16	50,5
50	50	11	3,6
50	70	5	0,03
100	20	20	$> 60 \cdot 60$
100	50	15	$7,2 \cdot 60$
100	70	6	0,3
150	20	23	$> 60 \cdot 60$
150	50	17	$> 60 \cdot 60$
150	70	6	1,5

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## Steps of the Basic VNS

**Initialization.** Select  $\mathcal{N}_k, k = 1, \dots, k_{max}$ ; find  $X$ ; choose a stopping condition.

**Repeat** until the «stopping condition».

- 1  $k \leftarrow 1$
- 2 **Repeat** until  $k = k_{max}$ .
  - 1 **Shaking.**  $X'$  at random from  $\mathcal{N}_k(X)$ .
  - 2 **Local Search.**  $X''$  obtained with  $X'$  as initial solution.
  - 3 **Move or Not.**
    - if**  $X''$  is better than  $X$  **then**
      - $X \leftarrow X''$
      - $k \leftarrow 1$
    - else**
      - $k \leftarrow k + 1$

## Basic VNS

Let  $G = (V, E)$  be given and

$C \subseteq V$ : any clique of  $G$ ;

$S = C$ : independent set of  $\tilde{G}$ ;

$T = V - S$ : transversal of  $\tilde{G}$ .

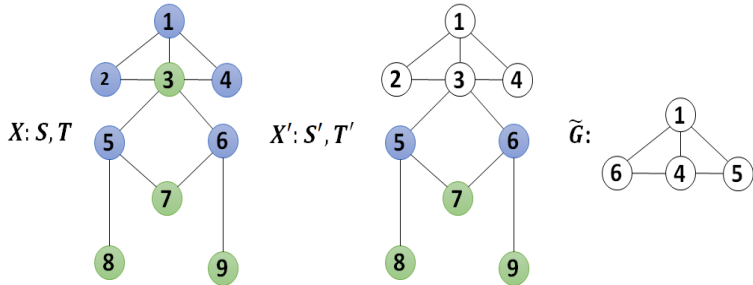
A transversal  $T_{opt}$  has a minimum cardinality and covers all vertices from  $S$ .

$$\mathcal{N}_k(X) = \{X' : \rho(X, X') = k\},$$

where  $\rho(X, X')$  is the Hamming distance between corresponding 0-1 arrays for  $C$  and  $C'$ .

# Shaking

Let  $X$  and  $k = 1$  are given.

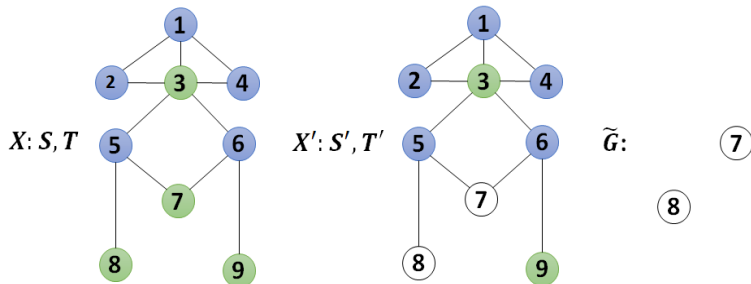


$X' \leftarrow \text{Shaking}(X, 1),$   
 $X'' \leftarrow \text{Local Search}(\tilde{G}, X').$



# Shaking

Let  $X$  and  $k = 2$  are given.



$X' \leftarrow \text{Shaking}(X, 2),$   
 $X'' \leftarrow \text{Local Search}(\tilde{G}, X').$

# Local Search

## 1 Solution Test

if  $\tilde{V} = \{\}$  then  
stop

## 2 Add Step

$v \leftarrow \min \text{degree}(\tilde{G})$  {using some tie-breaking rule}  
 $C \leftarrow C + \{v\}$   
 $T \leftarrow T + \{u: (u, v) \in \tilde{E}\}$

## 3 Subproblem Reduction

$\tilde{V} \leftarrow \tilde{V} - \{v, u\}$   
 $\tilde{E} \leftarrow \tilde{E} - \{(w, u)\}$  { $e \in \tilde{E}$  if and only if both endpoints  
belong to  $\tilde{V}$ }  
return to **Solution Test**

# Greedy Selection Rule

	$C$	$T$	$v$	$\tilde{V}$	$\tilde{E}$
	{ }	{ }	8	123456789	(1,2), (1,3), (1,4), (2,1), (2,3) (3,1), (3,2), (3,4), (3,5), (3,6) (4,1), (4,3) (5,3), (5,7), (5,8) (6,3), (6,7), (6,9) (7,5), (7,6) (8,5) (9,6)
	8	5			

# Greedy Selection Rule

	$C$	$T$	$v$	$\tilde{V}$	$\tilde{E}$
	8	5	7	1234679	(1,2), (1,3), (1, 4), (2, 1), (2, 3) (3, 1), (3, 2), (3, 4), (3, 6) (4, 1), (4, 3) (6, 3), (6, 7), (6, 9) (7, 6) (9, 6)
	87	56			

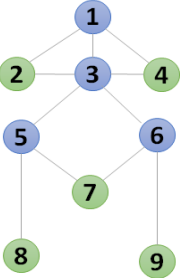
# Greedy Selection Rule

	$C$	$T$	$v$	$\tilde{V}$	$\tilde{E}$
	<b>87</b>	<b>56</b>	<b>9</b>	<b>12349</b>	<b>(1,2), (1,3), (1, 4)</b> <b>(2, 1), (2, 3)</b> <b>(3, 1), (3, 2), (3, 4)</b> <b>(4, 1), (4, 3)</b>
	<b>879</b>	<b>56</b>			

# Greedy Selection Rule

	$C$	$T$	$v$	$\tilde{V}$	$\tilde{E}$
	879	56	4	1234	(1,2), (1,3), (1, 4) (2, 1), (2, 3) (3, 1), (3, 2), (3, 4) (4, 1), (4, 3)
	8794	5613			

# Greedy Selection Rule

	$C$	$T$	$v$	$\tilde{V}$	$\tilde{E}$
	8794	5613	2	2	(1,2), (1,3), (1, 4) (2, 1), (2, 3) (3, 1), (3, 2), (3, 4) (4, 1), (4, 3)
	87942	5613			
	87942	5613		{ }	{ }
	stop				

# VNS for MCP

## Initialization

Neighborhood structures:  $k_{max} = 3$ ;

Initial solution  $X_0$ :  $C = \{\}$ ,  $T = \{\}$ ;

Stopping condition: 2 iterations between 2 improvements.

## Shaking

Random tie-breaking rule.

## Move or Not

Local solution  $X''$  is better than current solution  $X$  if and only if

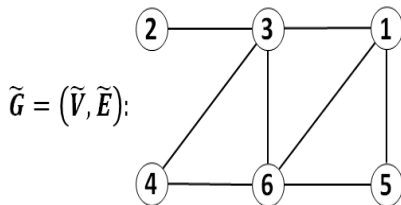
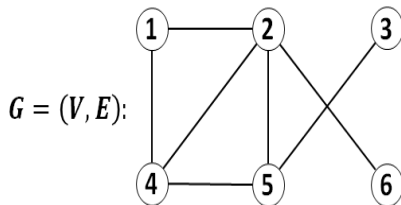
$$|C(X'')| > |C(X)| ,$$

where  $C(X)$  and  $C(X'')$  are the corresponding cliques.



## VNS for MCP

Let's consider an arbitrary graph  $G = (V, E)$  and corresponding complementary graph  $\tilde{G} = (\tilde{V}, \tilde{E})$ .



# VNS for MCP

VNS step		$C$	$T$	$\tilde{V}$	$\tilde{E}$
Initialization	$X$	{ }	{ }	123456	(1,3) (1,5) (1,6) (2,3) (3,1) (3,2) (3,4) (3,6) (4,3) (4,6) (5,1) (5,6) (6,1) (6,3) (6,4) (6,5)
k=1					
Shaking ( $X,1$ )	$X'$	3	1246	5	{ }
Local Search ( $X'$ )	$X''$	35	1246	{ }	{ }
<b>Move</b> or not	$X$	35	1246		
k=1					

# VNS for MCP

VNS step		$C$	$T$	$\tilde{V}$	$\tilde{E}$
Shaking ( $X,1$ )	$X'$	5	16	234	(2,3) (3,2) (3,4) (4,3)
Local Search ( $X'$ )		52	163	4	{ }
	$X''$	524	163	{ }	{ }
<b>Move</b> or not	$X$	524	163		
<b>k=1</b>					

# VNS for MCP

VNS step		$C$	$T$	$\tilde{V}$	$\tilde{E}$
	$X$	524	163		
Shaking ( $X,1$ )	$X'$	24	63	15	(1,5) (5,1)
Local Search ( $X'$ )	$X''$	241	635	{ }	{ }
Move or <b>not</b>	$X$	524	163		
<b>k=2</b>					

# VNS for MCP

VNS step		$C$	$T$	$\tilde{V}$	$\tilde{E}$
Shaking ( $X, 2$ )	$X'$	2	3	1456	(1,5) (1,6) (4,6) (5,1) (5,6) (6,1) (6,4) (6,5)
Local Search ( $X'$ )		24	36	15	(1,5) (5,1)
	$X''$	245	361	{ }	{ }
Move or <b>not</b>	$X$	524	163		
<b>k=3</b>					
<b>1-st iteration between 2 improvements</b>					

# VNS for MCP

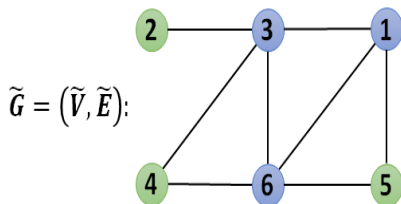
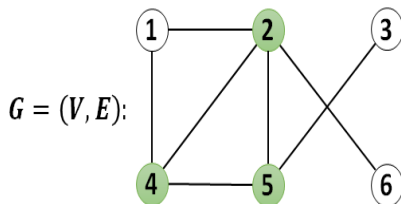
VNS step		$C$	$T$	$\tilde{V}$	$\tilde{E}$
<b>k=1</b>					
	$X$	524	163		
Shaking ( $X,1$ )	$X'$	54	136	2	{ }
Local Search ( $X'$ )	$X''$	542	136	{ }	{ }
Move or <b>not</b>	$X$	524	163		
<b>k=2</b>					

# VNS for MCP

VNS step		$C$	$T$	$\tilde{V}$	$\tilde{E}$
Shaking ( $X,2$ )	$X'$	4	36	125	(1,5) (5,1)
Local Search ( $X'$ )		42	36	15	(1,5) (5,1)
	$X''$	421	365	{ }	{ }
Move or <b>not</b>	$X$	524	163		
<b>k=3</b>					
<b>2-d iteration between 2 improvements</b>					
<b>STOP</b>					

# VNS for MCP

MC, IS and transversal of the initial and corresponding complementary graphs found by VNS.





# Outline

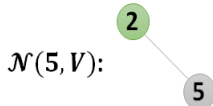
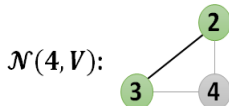
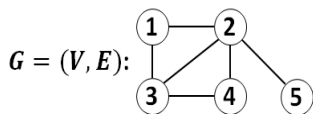
- 1 Classic  $\mathcal{NP}$ -hard Problem
  - Complexity classes and their relations
  - MISP and MCP
- 2 Applications
- 3 An Exact Methods for MISP
  - Dynamic Programming
  - Bron-Kerbosh Algorithm
- 4 An Effective Methods for MISP
  - Variable Neighborhood Search (VNS)
  - Extended Simplicial Vertex Test (ESVT)

# Simplicial Vertex Test (SVT)

A vertex  $v$ ,  $v \in V$  of a graph  $G = (V, E)$  is a simplicial vertex, if its neighborhood is a complete subgraph.

We denote with  $k[v]$  a number of vertices in a neighborhood of simplicial vertex  $v$ , that is

$$k[v] = |\mathcal{N}(v, V)|, \langle \mathcal{N}(v, V) \rangle_G \text{ — complete subgraph .}$$



$$k[4] = 2;$$

$$k[5] = 1.$$

# SVT for MISP

**Require:**  $G = (V, E)$

**Ensure:**  $S, S \subseteq V$

1:  $V_0 \leftarrow V$

$S \leftarrow \{\}$

$E_0 \leftarrow E$

2: **if**  $V_0 = \{\}$  **then**

3:     **stop**

      return  $S$

4: **for all**  $v \in V_0$  **do**

5:     **if**  $v$  — is a simplicial vertex **then**

6:          $S \leftarrow S + \{v\}$

$V_0 \leftarrow V_0 - \{v\} - \mathcal{N}(v, V_0)$  {neighborhood of  $v$  in a subgraph induced by  $V_0$ }

$E_0 \leftarrow E_0 - \{(u, v)\} - \{(u, w) : w \in \mathcal{N}(v, V_0)\}$

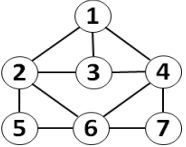
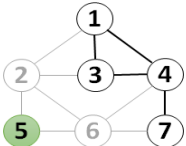
          go to step 2

7:     **else**

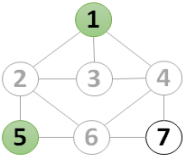
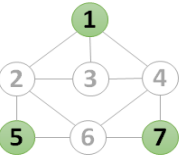
8:         **stop**

          return  $S, V_0$

# SVT for MISP

$G_0$	Step	$S$	$V_0$	$E_0$
	Initialization	{ }	123456	(1,2) (1,3) (1,4) (2,1) (2,3) (2,5) (2,6) (3,1) (3,2) (3,4) (4,1) (4,3) (4,6) (4,7) (5,2) (5,6) (6,2) (6,4) (6,5) (6,7) (7,4) (7,6)
	Search	5	1347	(1,3) (1,4) (3,1) (3,4) (4,1) (4,3) (4,7) (7,4)

# SVT for MISP

$G_0$	Step	$S$	$V_0$	$E_0$
	Search	51	7	{ }
	Search	517	{ }	{ }
<b>STOP</b>				

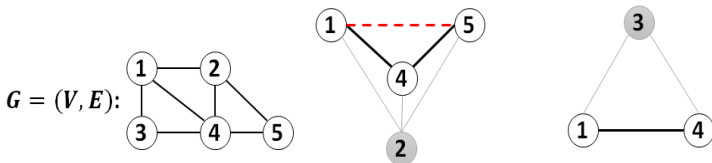
# Extended SVT (ESVT)

## Proposition

If for  $S$  found by SVT  $V_0 = \{\}$ , than  $S$  — MIS of  $G = (V, E)$ .

Let's denote with  $m[v]$  a number of edges missed in a neighborhood of this vertex to be a complete subgraph, that is

$$m[v] = C_{k[v]}^2 - |E \cup E(\langle \mathcal{N}(v, V) \rangle_G)| .$$



$$k[2] = 3, m[2] = 1 \quad k[3] = 2, m[3] = 0$$

# ESVT

**Require:**  $G = (v, E)$

**Ensure:**  $S \subseteq V, Est$

1:  $V_0 \leftarrow V$

$S \leftarrow \{\}$

$Est \leftarrow 0$

2: **while**  $V_0 \neq \{\}$  **do**

3: **for all**  $v \in V_0$  **do**

4:     calculate  $k[v]$

      calculate  $m[v]$

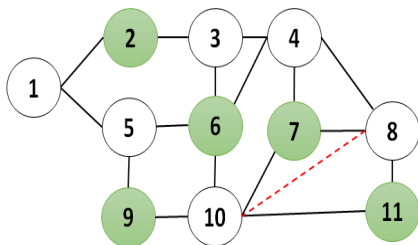
5:  $v_0 \leftarrow MinMaxParam(v)$

$S \leftarrow S + \{v_0\}$

$Est \leftarrow Est + m[v_0]$

$V_0 \leftarrow V_0 - \{v_0\} - \mathcal{N}(v_0, V_0)$

# ESVT for MISP



1	2	3	4	5	6	7	8	9	10	11	S	Est
[2,1]	[2,1]	[3,2]	[4,4]	[3,3]	[4,5]	[3,2]	[3,2]	[2,1]	[4,6]	[2,1]	11	1
[2,1]	[2,1]	[3,2]	[3,2]	[3,3]	[3,2]	[2,1]		[1,0]			11,9	1
[1,0]	[2,1]	[3,2]	[3,2]		[2,0]	[2,1]					11,9,6	1
[1,0]	[1,0]					[0,0]					11,9,6,2	1
						[0,0]					11,9,6,2,7	1



# VNS with SVT

**Initialization**  $X \leftarrow SVT(X_0)$ ;

**Repeat** until the **Stopping condition**.

①  $k \leftarrow 1$

② **Repeat** until  $k = k_{max}$ .

① **Shaking** using random tie-breaking rule.

② **Local Search**

$X'' \leftarrow SVT(X')$

③ **Move or Not**

# Computational Results (VNS)

	GA1		GA2		VNS		BR
	C	Time	C	Time	C	Time	C
<b>MANN_a27</b>	126	2,55	126	8	126	0,07	126*
<b>MANN_a45</b>	343	34,38	343,59	383	344,5	20,79	345*
<b>brock200_2</b>	11	0,30	9,92	17	11,3	20,79	12*
<b>brock200_4</b>	16	0,25	16,04	24	16,9	6,8	17*
<b>brock400_2</b>	24	2,57	24	79	27,4	57,19	29*
<b>brock400_4</b>	24	1,59	24,13	12	33	36,90	33*
<b>brock800_2</b>	19	4,77	18,47	47	21	11,78	21
<b>brock800_2</b>	19	10,28	19,05	550	21	11,78	21
<b>hamming8-4</b>	16	0,33	16	1	16	0,01	16*
<b>hamming10-4</b>	33	15,68	39,18	46	40	0,26	40
<b>keller4</b>	11	0,18	11	1	11	0,01	11*
<b>keller5</b>	25	11,96	25,76	54	27	0,38	27

# Computational Results (VNS)

	GA1		GA2		VNS		BR
	C	Time	C	Time	C	Time	C
<b>p_hat300-1</b>	8	0,63	7,69	2	8	0,02	8*
<b>p_hat-2</b>	25	0,93	25	1	25	0,01	25*
<b>p_hat300-3</b>	36	0,37	35,85	1	36	0,07	36*
<b>p_hat700-1</b>	9	4,74	9,45	180	11	0,52	11*
<b>p_hat700-2</b>	44	11,34	44	3	44	0,05	44*
<b>p_hat700-3</b>	62	4,32	62	3	62	0,06	62
<b>p_hat1500-1</b>	10	12,53	10,10	107	12	415,68	12*
<b>p_hat1500-2</b>	59	56	65	10	65	76,3	65
<b>p_hat1500-3</b>	92	56,24	93,44	860	94	0,58	94



C. Aggarwal, Orlin, R. Tai, Optimized crossover for the maximum independent set proble. Oper. Res. 45 (1997) 226-234.

# Computational Results (ESVT)

	IncMaxCLQ	BBMCX	ESVT		BR
	Time	Time	C	Time	C
MANN_a81			1099	696,870972	1100*
brock400_2	259,9	132	22	0,024	29
brock400_4	197,7	20,2	22	0,04	33
brock800_2		3568	18	0,19	24
brock800_4		2532	17	0,191	26
hamming8-4			16	0,011	16
p_hat300-1			8	0,038	8
p_hat300-2			25	0,029	25
p_hat300-3	0,87	0,531	34	0,015	36

## Computational Results (ESVT)

	IncMaxCLQ	BBMCX	ESVT		BR
	Time	Time	C	Time	C
<b>p_hat700-1</b>			8	0,363	11
<b>p_hat700-2</b>	1,25	1,42	43	0,691	44
<b>p_hat700-3</b>	357,4	718	61	0,144	62*
<b>C125.9</b>			33	0,001	34*
<b>C250.9</b>	333,2	1144	42	0,005	44*
<b>DSJC1000_5</b>	261	211	13	0,529	15



Li C et al. Combining MaxSAT Reasoning and Incremental Upper Bound for the Maximum Clique Problem.



San Segundo P et al. Infra-automated Bound for the exact Clique. *Comp. Oper. Research.* 2015. 64, pp.293-303.

THANK YOU FOR ATTENTION!!