

Lecture 1: ILP Formulations for the Graph Coloring Problem

Part 1.2: Graph Coloring

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Outline

- 1 Graph Coloring**
 - Definitions and Properties
 - Applications
 - Exact approaches for the graph coloring problem
- 2 ILP Formulations**
 - Assignment Model
 - Representatives Formulations
 - Partial Ordering based Models
 - Set Covering based Model
- 3 Computational Studies and Results**
- 4 Conclusion and Open Questions**

Graph Coloring Literature: Surveys

DISCLAIMER: There is a vast amount of literature concerning the graph coloring problem. Here, we can only discuss a tiny part of it!

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- T. Husfeldt: Graph colouring algorithms, Cambridge University Press, 2015
- R.M.R. Lewis: Guide to graph colouring: Algorithms and applications, Guide to Graph Colouring, 2016
- A.M.D. Lima and R. Carmo: Exact algorithms for the graph colouring problem. Revista de Informática Teórica e Aplicada, 2018
- E. Malaguti, and P. Toth: A survey on vertex coloring problems. International Transactions in Operational Research 2010

Graph Coloring Literature: Explicit ILP Models

- I. Mendez-Diaz and P. Zabala: A branch-and-cut algorithm for graph coloring, Discrete Applied Mathematics 2006
- M. Campêlo, M., R. Corrêa, R. and Frota, Y., Cliques, holes and the vertex coloring polytope, Information Processing Letters 2004
- M. Campêlo, V.A. Campos, and R.C. Corrêa: On the asymmetric representatives formulation for the vertex coloring problem, Discrete Applied Mathematics, 2008
- A. Jabrayilov and P. Mutzel: New integer linear models for the vertex coloring problem, LATIN 2018
- A. Jabrayilov and P. Mutzel: Strengthened partial-ordering based ILP models for the vertex coloring problem, arXiv 2022

Graph Coloring Literature: Independent Set based ILP Models

- A. Mehrotra and M.A. Trick: A column generation approach for graph coloring, *INFORMS Journal on Computing* 1996
- E. Malaguti, M. Monaci, and P. Toth: An exact approach for the vertex coloring problem, *Discrete Optimization* 2011
- S. Held, W. Cook, and E.C. Sewell: Maximum-weight stable sets and safe lower bounds for graph coloring, *IPCO* 2011
- S. Held, W. Cook, and E.C. Sewell: Safe lower bounds for graph coloring, *Mathematical Programming Computation* 2012
- W.-J. van Hoeve: Graph coloring with decision diagrams, *Mathematical Programming*, 2022

Some Graph Coloring Instances and Results

- D. Johnson, A. Mehrotra, and M. Trick: DIMACS graph coloring instances 2002, <https://mat.tepper.cmu.edu/COLOR02/>
- S. Gualandi and M. Chiarandini: Graph coloring instances 2017, <https://sites.google.com/site/graphcoloring/vertex-coloring>
- N. Tamura: CSP2SAT: GCP benchmark results, <https://csp2sat.gitlab.io/csp2sat/gcp/>

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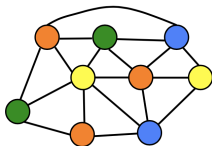
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Graph Coloring Problem

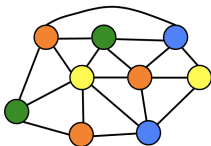


Definitions

Let $G = (V, E)$ be an undirected graph.

- A **(vertex-) (coloring)** of G is an assignment of one color to each vertex, so that adjacent vertices are colored differently.
- The smallest number of colors for a given graph G is called the **chromatic number** of G : $\chi(G)$.

Graph Coloring Problem



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The coloring problem: “Does G have a coloring with at most k colors?” (in short: “Is G k -colorable?”) is **NP-complete** [Karp 1972].

Properties

Notations

- Let $\omega(G)$ denote the **clique number** (size of largest clique),
- $\Delta(G)$ the **largest vertex degree** in G

The following holds

- $\chi(G) = 1 \Leftrightarrow$

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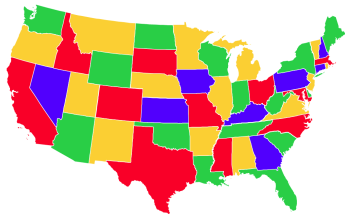
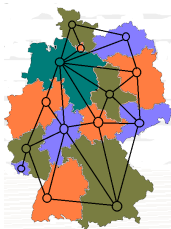
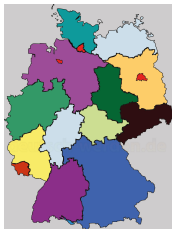
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- $\chi(G) \geq \omega(G)$
- $\chi(G) \leq \Delta(G) + 1$
- $\chi(G) \leq \Delta(G) \Leftrightarrow (G \text{ is not complete}) \text{ and } (G \text{ is not an odd cycle})$

Four Color Theorem

Four Color Conjecture by Francis Guthrie (1852)

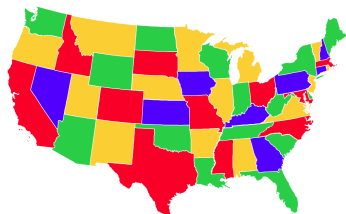
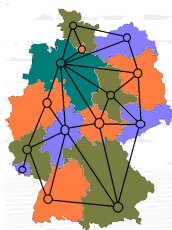
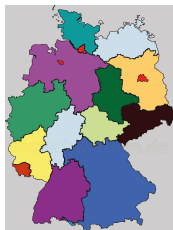
- Four colors are sufficient to color a map so that no two adjacent regions receive the same color.



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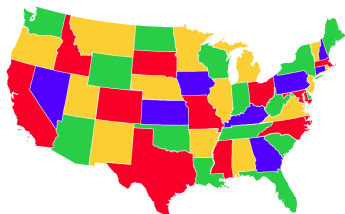
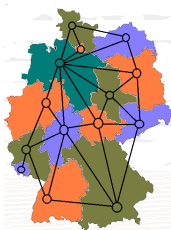
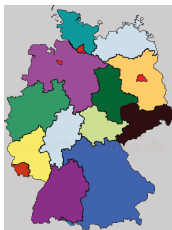
Four Color Theorem by Appel and Haken 1976

- Planar graphs are 4-colorable

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⇒ first generally accepted major **computer-aided proof**

Image sources: <https://de.wikiversity.org/>, <https://geoawesomeness.com>

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Remarks

- this was the first generally accepted major **computer-aided proof**
- Robertson, Sanders, Seymour, and Thomas (1997) found a simpler proof (also computer-aided)
- Gonthier used a general-purpose theorem-proving software Coq (2005)

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It is NP-complete to decide if a given planar graph is 3-colorable.

Strong Perfect Graph Theorem

Definitions

- A **perfect graph** is a graph G in which, for every induced subgraph F of G : $\omega(F) = \chi(F)$.
- G is a **Berge graph** if no **induced subgraph** of G is an odd cycle of length at least 5 or the **complement** of one.

Perfect graphs include, e.g., bipartite graphs, chordal graphs

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Strong Perfect Graph Conjecture by Claude Berge (1961)

- G is perfect $\Leftrightarrow G$ is Berge.

Strong Perfect Graph Theorem

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Strong Perfect Graph Theorem by Chudnovsky, Robertson, Seymour, Thomas (2002)

- Strong perfect graph conjecture holds.
- Every Berge graph either falls into one of five basic classes, or it has one of four different types of structural types of decomposition into simpler graphs (stronger conjecture by [Conforti, Cornuéjols, and Vuskovic](#) and [Cornuéjols, Robertson, Seymour, and Thomas](#)).

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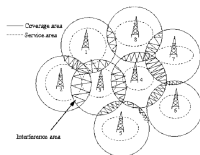
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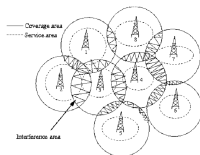
Applications



	8		2	6			
1	6						
6	1			4			
	9					3	
						7	5
							1
			7				8
		7		3			

- map coloring
- frequency assignment of telecommunication networks
- computer register allocation problem

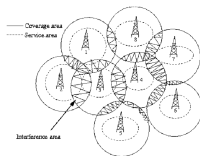
Applications



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Applications



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					7		5
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			7			8	
		7	3				

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- time tabling
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- map labelling
- Sudoku

Sudoku

Sudoku Task

	8		2	6				
1	6							
6	1			4				
	9					3		
						7		5
							1	
				7			8	
		7		3				

Sudoku

Sudoku Task

Complete a given 9×9 matrix, so that in each row, in each column, and in each of the marked 3×3 squares each of the numbers $1, 2, \dots, 9$ occurs exactly once.

	8		2	6				
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				7			8	
		7		3				

A (correct) Sudoku has exactly one feasible solution.

Sudoku: Formulation as graph coloring problem

Sudoku: Formulation as graph coloring problem

The vertices of the graph correspond to the cells. For each pair of vertices within the same row, the same column, or the same marked square there is an edge.

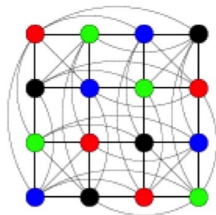


Abbildung: small example with 2x2 squares

Coloring task: Find a coloring with 9 colors (here: 4 colors) that are consistent with the given partial coloring.

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Exact approaches for the graph coloring problem

- Dynamic programming approaches
- Branch-and-bound based enumeration
- SAT based approaches
- ILP approaches

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- explicit coloring models, e.g., assignment, partial-ordering based model
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In the following: Let G be a connected graph with at least 3 vertices and H be an upper bound of $\chi(G)$.

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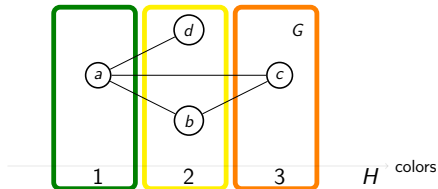
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Assignment-based ILP model

Binary variables

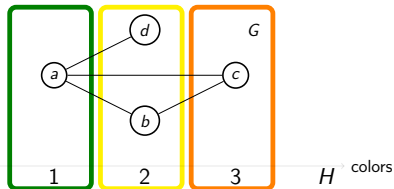
- For each vertex $v \in V$ and color $i \in 1, \dots, H$ we introduce assignment variables $x_{v,i} = \begin{cases} 1 & v \text{ is assigned to color } i \\ 0 & \text{otherwise.} \end{cases}$



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- For each color $i \in 1, \dots, H$ we introduce $w_i = \begin{cases} 1 & \text{color } i \text{ is used} \\ 0 & \text{otherwise} \end{cases}$



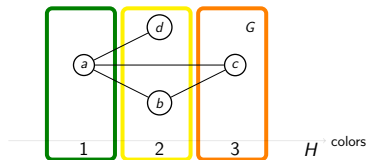
Assignment model

$$\min \sum_{1 \leq i \leq H} w_i$$

$$s.t. \quad \sum_{i=1}^H x_{v,i} = 1 \quad \forall v \in V \quad (1)$$

$$x_{u,i} + x_{v,i} \leq w_i \quad \forall (u, v) \in E, i = 1, \dots, H \quad (2)$$

$$x_{v,i}, w_i \in \{0, 1\} \quad \forall v \in V, i = 1, \dots, H \quad (3)$$



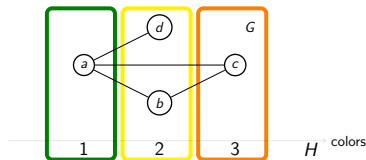
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- number of variables: $H + H|V|$ and constraints: $|V| + H|E|$

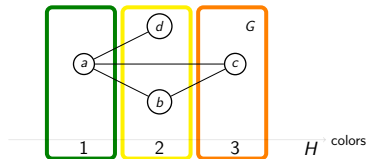
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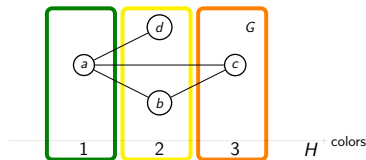
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- number of variables: $H + H|V|$ and constraints: $|V| + H|E|$
- **simple and easy** to use, easily adaptable for problem variations
- **problem**: many optimal solutions that are **symmetric** to each other
- \Rightarrow problem for branch-and-bound approaches

Assignment model (extended)

In order to overcome the symmetry, Mendez-Diaz and Zabala (2006) have suggested to add the following constraints:

$$\min \sum_{1 \leq i \leq H} w_i$$

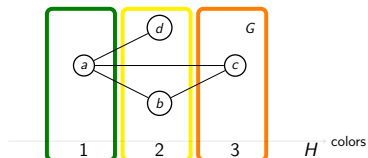
$$\text{s.t.} \quad \sum_{i=1}^H x_{v,i} = 1$$

$$x_{u,i} + x_{v,i} \leq w_i$$

$$w_i \leq \sum_{v \in V} x_{v,i}$$

$$w_i \leq w_{i-1}$$

$$x_{v,i}, w_i \in \{0, 1\}$$



$$\forall v \in V \quad (4)$$

$$\forall (u, v) \in E, i = 1, \dots, H \quad (5)$$

$$i = 1, \dots, H \quad (6)$$

$$i = 2, \dots, H \quad (7)$$

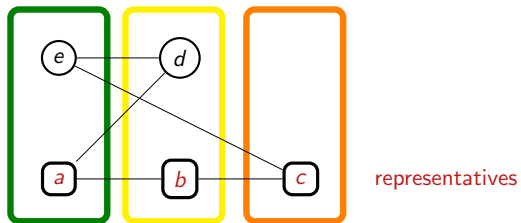
$$\forall v \in V, i = 1, \dots, H \quad (8)$$

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Representatives Formulation

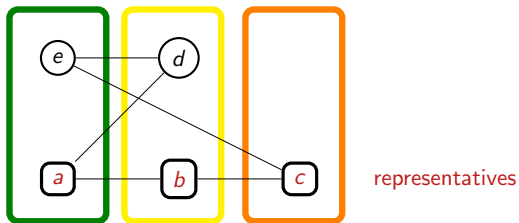
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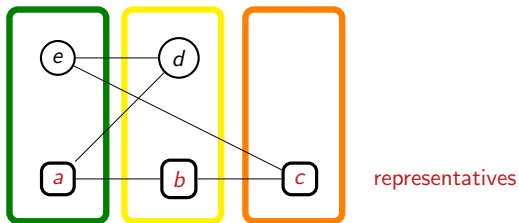
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Representatives Formulation

Binary variables

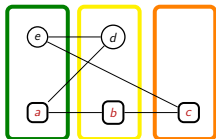
- For each vertex $u \in V$ we introduce a variable $x_{uu} = \begin{cases} 1 & u \text{ is a representative} \\ 0 & \text{otherwise.} \end{cases}$
- For each pair of non-adjacent vertices $u, v \in V$ we introduce $x_{uv} = \begin{cases} 1 & u \text{ represents the color of } v \\ 0 & \text{otherwise.} \end{cases}$



- For example: $x_{a,e} = 1$, $x_{c,c} = 1$, $x_{a,b} = 0$, ...

Representatives Formulation

$$\begin{aligned}
 \min \quad & \sum_{u \in V} x_{uu} \\
 \text{s.t.} \quad & \sum_{u \in \bar{N}(v) \cup v} x_{uv} \geq 1 \quad \forall v \in V
 \end{aligned} \tag{9}$$



representatives

$$x_{uv} + x_{uw} \leq x_{uu} \quad \forall u \in V, \forall (v, w) \in E \text{ and } v, w \in \bar{N}(u) \tag{10}$$

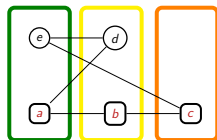
$$x_{uu} \in \{0, 1\} \quad \forall u \in V \tag{11}$$

$$x_{uv} \in \{0, 1\} \quad \forall \text{ non-adjacent pairs } u, v. \tag{12}$$

- $\bar{N}(v)$ is the set of vertices not adjacent to v

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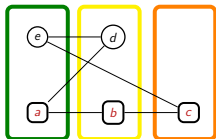
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- number of variables: $|(V \times V) \setminus E|$ and constraints: up to $|V| + |V||E|$

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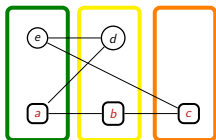
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- simple and compact

Representatives Formulation

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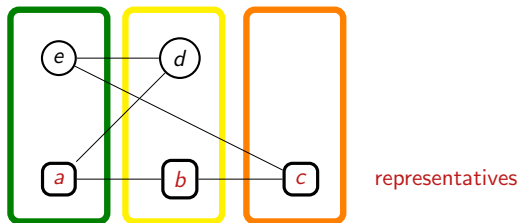
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- $\bar{N}(v)$ is the set of vertices not adjacent to v
- number of variables: $|(V \times V) \setminus E|$ and constraints: up to $|V| + |V||E|$
- **simple and compact**
- **problem:** symmetry, since any of the vertices can be representative

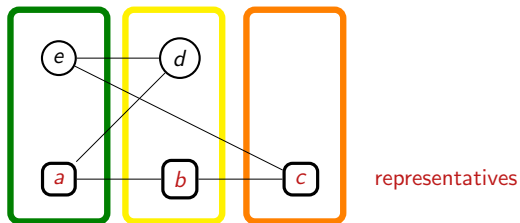
Asymmetric Representatives Formulation

- Let $v_1, v_2, \dots, v_i, \dots, v_j, \dots, v_{|V|}$ be a linear ordering of the vertices



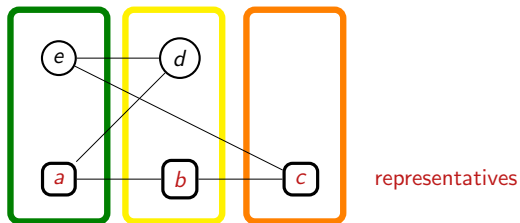
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- variable for every **ordered pair** of non-adjacent vertices,



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- \Rightarrow reduces symmetry and number of variables



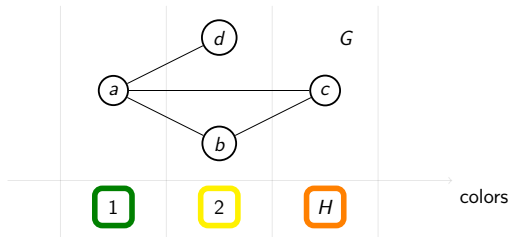
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 - Definitions and Properties
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 - Set Covering based Model
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Partial Ordering based Model

Binary variables

- Let $1, 2, \dots, H$ be a linear ordering of the colors



Partial Ordering based Model

Binary variables

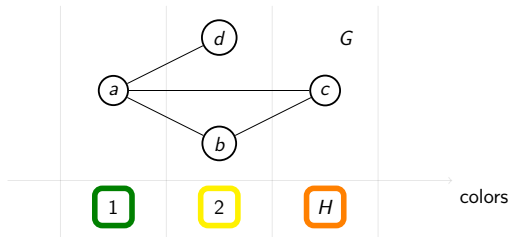
- Let $1, 2, \dots, H$ be a linear ordering of the colors
- For each vertex $v \in V$ and color $i = 1, \dots, H$ introduce variables

$$l_{v,i} = \begin{cases} 1 & \text{color}(v) < i \\ 0 & \text{otherwise.} \end{cases} \quad \text{and} \quad r_{i,v} = \begin{cases} 1 & i < \text{color}(v) \\ 0 & \text{otherwise} \end{cases}$$

- Example:

$$l_{a,2} = 1, \quad r_{2,a} = 0$$

$$l_{b,2} = 0, \quad r_{2,b} = 0$$



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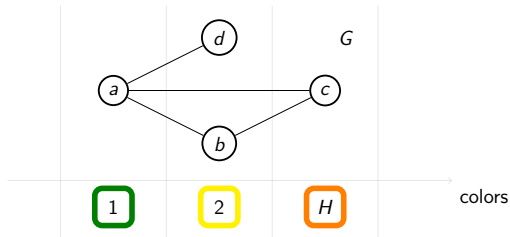
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Connection to assignment variables: $x_{v,i} = 1 - (l_{v,i} + r_{i,v})$

- Example:

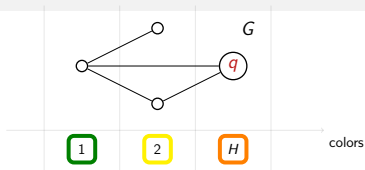
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Partial Ordering based Model

$$\min 1 + \sum_{1 \leq i \leq H} r_{i,q}$$



$$\text{s.t. } l_{v,1} = r_{H,v} = 0 \quad \forall v \in V \quad (13)$$

$$r_{i-1,v} - r_{i,v} \geq 0 \quad \forall v \in V, i = 2, \dots, H \quad (14)$$

$$r_{i-1,v} + l_{v,i} = 1 \quad \forall v \in V, i = 2, \dots, H \quad (15)$$

$$(r_{i,u} + l_{u,i}) + (r_{i,v} + l_{v,i}) \geq 1 \quad \forall uv \in E, i = 1, \dots, H \quad (16)$$

$$r_{i,q} - r_{i,v} \geq 0 \quad \forall v \in V, i = 1, \dots, H-1 \quad (17)$$

$$r_{i,v}, l_{v,i} \in \{0, 1\} \quad \forall v \in V, i = 1, \dots, H$$

q is an arbitrary vertex chosen from V which we will fix to the largest occupied color class

Jabrayilov and Mutzel 2018

Partial Ordering based Model

- Number of variables: $2H|V|$ and constraints: up to $3H|V| + H|E|$

Jabayilov and Mutzel 2018, Jabrayilov and Mutzel 2022

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- Number of variables: $2H|V|$ and constraints: up to $3H|V| + H|E|$
- Equations can be used to reduce the model.
- The reduced model has $(H - 1)|V|$ variables only.

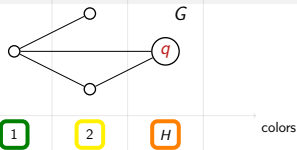
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Partial Ordering based Model

- Number of variables: $2H|V|$ and constraints: up to $3H|V| + H|E|$
- Equations can be used to reduce the model.
- The reduced model has $(H - 1)|V|$ variables only.
- Model can be strengthened further \Rightarrow POP2

Jabayilov and Mutzel 2018, Jabrayilov and Mutzel 2022

Partial Ordering based Model POP2

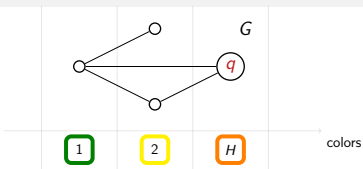


$$\begin{array}{ll}
 \min & 1 + \sum_{1 \leq i \leq H} r_{i,q} \\
 \text{s.t.} & r_{H,v} = 0 \quad \forall v \in V \\
 & r_{i-1,v} - r_{i,v} \geq 0 \quad \forall v \in V, i = 2, \dots, H \\
 & r_{i,u} + r_{i,v} \geq 2 - r_{1,q} \quad \forall uv \in E \\
 & (r_{i-1,u} - r_{i,u}) + (r_{i-1,v} - r_{i,v}) \leq r_{i-1,q} \quad \forall uv \in E, i = 2, \dots, H \\
 & r_{i,q} - r_{i,v} \geq 0 \quad \forall v \in V, i = 1, \dots, H-1 \\
 & r_{i+1,q} - r_{i,v} \geq 0 \quad \forall v \in N(q), i = 1, \dots, H-1 \\
 & r_{i,v}, l_{v,i} \in \{0, 1\} \quad \forall v \in V, i = 1, \dots, H
 \end{array}$$

with growing density, the POP models have more non-zero elements than that of ASS \Rightarrow Hybrid model

Hybrid Partial Ordering based Model POPH2

Introducing assignment variables:



$$\min \quad 1 + \sum_{1 \leq i \leq H} r_{i,q}$$

$$\text{s.t.} \quad r_{H,v} = 0$$

$$x_{v,1} = 1 - r_{1,v} \quad \forall v \in V$$

$$x_{v,i} = r_{i-1,v} - r_{i,v} \quad \forall v \in V, i = 2, \dots, H$$

$$x_{u,1} + x_{v,1} \leq r_{1,q} \quad \forall uv \in E$$

$$x_{u,i} + x_{v,i} \leq r_{i-1,q} \quad \forall uv \in E, i = 2, \dots, H$$

$$r_{i,q} - r_{i,v} \geq 0 \quad \forall v \in V, i = 1, \dots, H$$

$$r_{i+1,q} - r_{i,v} \geq 0 \quad \forall v \in N(q), i = 1, \dots, H - 1$$

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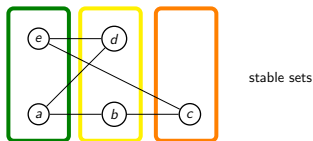
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Set Covering based Model

Observation

- The vertices of the same color build a **stable** (independent) set.
- **Idea:** Find a small **set of stable sets** S_1, \dots, S_k , that **cover** all vertices, i.e., $S_1 \cup \dots \cup S_k = V$.



Mehrotra and Trick 1996, Malaguti, Monaci and Toth 2011, Held, Cook and Sewell 2011, Held, Cook, and Sewell 2012

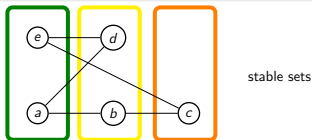
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Binary variables

- Let S be the set of all maximal stable sets. For each $s \in S$ define a variable $x_s = \begin{cases} 1 & \text{stable set } s \text{ is selected} \\ 0 & \text{otherwise.} \end{cases}$



Mehrotra and Trick 1996, Malaguti, Monaci and Toth 2011, Held, Cook and Sewell 2011, Held, Cook, and Sewell 2012

Set covering based Model

$$\min \sum_{s \in S} x_s$$

$$s.t. \sum_{s \in S: v \in s} x_s \geq 1$$

$$\forall v \in V$$

$$x_s \in \{0, 1\}$$

$$\forall s \in S.$$

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- Number of variables: **exponential**; number of constraints: $|V|$

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- Number of variables: **exponential**; number of constraints: $|V|$
- \Rightarrow **Branch-and-price** algorithms necessary for computation (column generation)
- previous models have polynomial size, and can be computed directly (branch-and-cut)

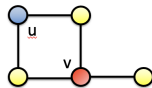
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Preprocessing

The following techniques are well known:

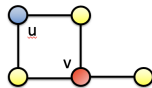


Preprocessing: domination

- vertex u is **dominated** by another vertex v , if $N(u) \subseteq N(v)$.
- $\Rightarrow u$ can be deleted
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Preprocessing: precoloring

- Fix variables in ASS and POP models by **precoloring a clique Q** with $\max(|Q|H + |\delta(Q)|)$.
- Choose q in POP models from the clique, and color remaining vertices in Q with colors $1, \dots, |Q| - 1$.

Implementation

Heuristically compute small H

- small upper bound H leads to smaller number of variables in ASS and POP models
- we use heuristics in `networkx`

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Implementation

- Gurobi-python API
- branching priority option: high degree

Implementation by A. Jabrayilov

Computational Study 2018: Simple Methods

Test Set Up

- Gurobi 6.5 single-threadedly on Intel Xeon E5-2640, 2.60GHz with 64 GB of memory, Ubuntu Linux 14.04

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Test instances

- hardest instances (68) from the DIMACS benchmark set and the GPIA graphs: $(|V|, |E|) \leq (10\,000, 990\,000)$

Computational Study 2018: Simple Methods

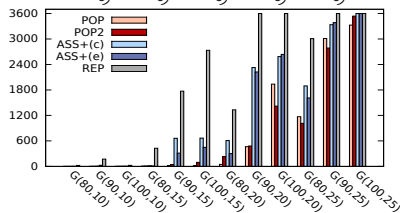
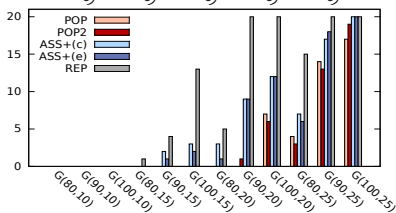
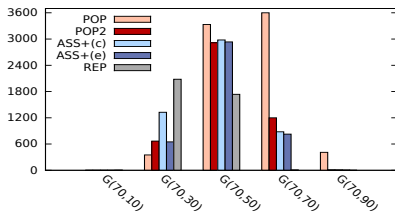
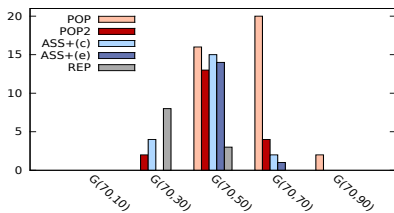
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Test instances

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- 340 randomly generated Erdős-Rényi graphs $G(n, p)$ (20 each):
 $n = |V| = 70, 80, 90, 100$ and edge probability $p \in \{0.1, 0.25\}$

Results for random graphs $G(n, p) \leq G(100, 0.25)$



no. of unsolved instances

avg. time [sec]

Experiments conducted by A. Jabrayilov 2018

Results for 68 DIMACS instances 2018

- **COV**: Set-Covering model [Malaguti et al. 2011]
- taken from paper, processor was $\approx 1.3\times$ slower

	REP	ASS	POPH	COV
solved	15	21	25	25
avg. time [sec]	170	171	182	1196

Experiments conducted by A. Jabrayilov 2018

Experimental Results 2022:

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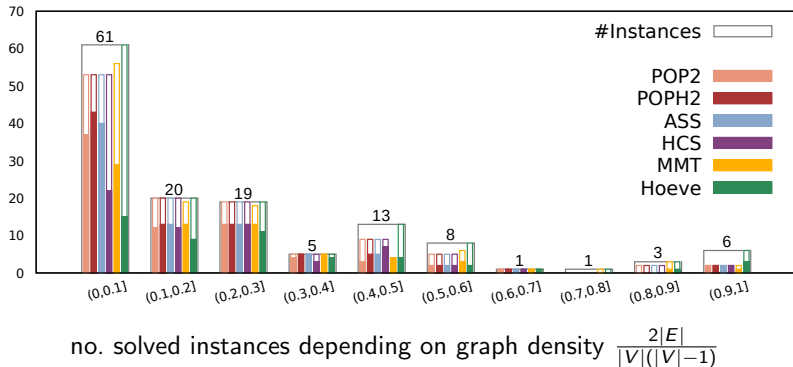
Test instances

- **all instances with $|E| \leq 100\,000$** from the DIMACS benchmark set (in total **116** from 137):

Experiments conducted by A. Jabrayilov 2022

Results for 116 DIMACS instances 2022

- **HCS**: Set covering model, code by Held, Cook, Sewell, 2012
- **MMT**: Set covering model, from paper by Malaguti, Monaci, Toth, 2011, processor was $\approx 1.3\times$ slower
- **Hoeve**: OBDD approach, from paper by van Hoeve 2022



Experiments conducted by A. Jabrayilov 2022

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- On the **sparse instances** with density ≤ 0.1 , **POPH2** dominates the set covering approaches **HCS** and **MMT**.
- This is no more true if the density increases.

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- On the **sparse instances** with density ≤ 0.1 , **POPH2** dominates the set covering approaches **HCS** and **MMT**.
- This is no more true if the density increases.
- The representative models **REP** and **AREP** dominate **ASS** and **POP2H** for graphs with density larger than 0.5
- **POPH2** was the only model that solved all five DIMACS GPIA graphs (estimation of sparse Jacobian matrix problem). **POPH2** is the first ILP model solving instance **abb313GPIA** ($n = 1555$, $|E| = 53356$).

Open Questions

Many open questions

- How do these models compare to **SAT-based** approaches?

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