

# Matheuristic Variants of DSATUR for the Vertex Coloring Problem

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Slides: TODO

# Notation

- $G = (V, E)$  is an undirected graph with vertex set  $V = \{v_1, \dots, v_n\}$  and edge set  $E$ , and denote  $I = \llbracket 1; n \rrbracket = [1; n] \cap \mathbb{Z}$ .
- An edge  $e = (v_i, v_j) \in E$  with  $i < j$  links the underlying vertices (*for VCP there is no sense to consider loop/multiple edges*).
- For  $i \in I$ ,  $\delta_i$  denotes the set of neighbors of vertex  $v_i$ , and  $d_i = |\delta_i| = |\{j \in I \mid \text{ngb}(v_i, v_j) = 1\}|$ , where  $\text{ngb}(v_i, v_j) = 1$  iff  $(v_i, v_j) \in E$ .<sup>2</sup>

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<sup>2</sup>ngb stands for “neighbor”.

# A $k$ -coloring of $G$

- It is an assignment of colors to vertices such that no two adjacent vertices share the same color.
- **VCP** consists in finding a  $k$ -coloring of  $G$  using the minimum number of colors  $k$  (the *chromatic number*  $\chi(G)$ ).
- A valid  $k$ -coloring ( $c$ ) fulfills:  $\forall i < j, (v_i, v_j) \in E \implies c_i \neq c_j$  where  $c_i = \llbracket 1; k \rrbracket$  is the color of  $v_i$ .

# A partial $k$ -coloring of $G$

- If a vertex may not be colored, we set  $c_i = -1$  s.t.  $c_i \in \llbracket 1; k \rrbracket \cup \{-1\}$
- A partial  $k$ -coloring  $(c)$  is *feasible* if  $\forall i < j, (v_i, v_j) \in E \implies c_i \neq c_j \vee c_i = c_j = -1$ .
- Given  $v_i$  the **saturation table**  $S_i$  is the set of colors assigned to its colored neighbors:  $S_i = \bigcup_{j \in \delta_i}^n \{c_j\} \setminus \{-1\}$ , and  $s_i = |S_i|$  is the **saturation degree**.
- A total order  $\succsim$  over  $V$  is defined as:  $v_i \succsim v_j \iff s_i > s_j \vee (s_i = s_j \wedge d_i \geq d_j)$

# Compact ILP Formulations<sup>3</sup>, feasible *iff* $\chi(G) \leq k$

|  |  |   |
|--|--|---|
| $z_{i,c} \in \{0, 1\}$<br>indicates if<br>vertex $v_i$ is<br>assigned color<br>$c$ . | $\min \sum_{c=1}^k y_c$  | The objective minimizes the<br>number of used colors.                         |
| $y_c \in \{0, 1\}$<br>indicates if<br>color $c$ is used<br>in the coloring.          | $s.t. \min \sum_{c=1}^k z_{i,c} = 1 \quad \forall i \in I$   | The 1st set ensures that each<br>vertex is assigned exactly one<br>color.     |
|  | $z_{i,c} + z_{j,c} \leq y_c \quad \forall (v_i, v_j) \in E,$<br>$\forall c \in \llbracket 1; k \rrbracket$ | The 2nd set ensures that<br>adjacent vertices do not share<br>the same color. |

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<sup>3</sup>Efficient formulations: extended column generation by Furini and Malaguti [1], and reduced formulation to MWSSP by Cornaz et al. [2].

# Observations

- Having an upper bound of the chromatic number as the initial value  $k$  (or simply  $k = |V|$ ) guarantees the optimality of the solution.
- The size of  $k$  strongly affects the performance of ILP solvers.
- Symmetries in the model (e.g., colors are permutable) enlarge the search space for Branch-and-Bound algorithms (the same solution can be represented in multiple ways)

# Representative ILP Model<sup>4</sup>, asymmetric and easily to LP-relax

$$\begin{aligned}
 & x_{i,i'} \in \{0, 1\}, \\
 & \quad \forall i, i' \in V \text{ s.t. } i \leq i', \\
 & \quad \text{indicates if} \\
 & \quad \text{vertices } v_i \text{ and } v_{i'} \text{ share the same color and} \\
 & \quad \quad i \text{ is the minimum index of its color class.} \\
 & \min_z \sum_{i=1}^n x_{i,i} \\
 & \text{s.t. } \sum_{i' \leq i} x_{i',i} \geq 1 \quad \forall i \in I \\
 & \quad x_{j,i} + x_{j,i'} \leq x_{j,j} \quad \forall (v_i, v_{i'}) \in E, \\
 & \quad \quad \forall j \leq i
 \end{aligned}$$

The objective **counts** the number of representative vertices (i.e., used colors).

The 1st set ensures either  $x_{i',i} = 1$  (it is **representative**) or its representative is a **previous** vertex  $i' < i$ .

The 2nd set expresses the color incompatibility between adjacent vertices and  $x_{j,i} = 1 \implies x_{j,j} = 1$

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<sup>4</sup>A vertex is representative of its color class if it has the minimum index among the vertices sharing the same color.

# Standard DSATUR Algorithm

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**Algorithm 1: Standard DSATUR algorithm**

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**Input:**  $G = (V, E)$  a non-empty and non-oriented graph

**Initialization:**

define partial coloring  $c$  with  $c_i := -1$  for all  $i \in I$

define saturation table  $S$  with  $S_i := \emptyset$  for all  $i \in I$

initialize set  $U := V$ , and color  $k := 0$

**while**  $U \neq \emptyset$

**find**  $u \in U$ , a maximum of  $\succsim$  in  $U$ .

**if**  $|S_u| = k$  **then**  $k := k + 1$  // a new color is added

**compute**  $c_i := \min S_u$  // assign color to  $u$

    remove  $u$  from  $U$

**for all**  $i \in \delta_u \cap U$ ,  $S_i = S_i \cup \{c_i\}$  // update saturation

**end while**

**return** color  $k$  and  $(c)$  a  $k$ -coloring of  $G$

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- DSATUR is an **adaptive greedy heuristic** proposed by Brélaz [3], which colors vertices iteratively.
- Selection of the uncolored vertex to color is given with order  $\succsim$ , maximizing first the saturation degree and secondly the degree.
- Coloring a new vertex updates saturation, the iteration order of vertices is thus adaptive.



# DASTUR Matheuristic Variants<sup>5</sup>

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<sup>5</sup>N. Dupin, “Matheuristic Variants of DSATUR for the Vertex Coloring Problem,” in *Metaheuristics 2024* [[4](#)]

# Initialization

Defining an initial partial coloring and computing the saturation table for the uncolored vertices, **before** starting the main DSATUR iterations.

*Variants:*

1. maxDeg: color the vertex with the maximum degree — equivalent to standard DSATUR by definition of  $\succ$ , it would suffer from many ties;
2. col- $n$ : consider  $n$  vertices having the maximum degree and color them solving a representative ILP model for the *induced* subgraph — more depth pre-processing, it tries to prevent erroneous decisions in the initial steps of DSATUR;
3. clq: find a maximum clique<sup>6</sup> and color it with different colors — an exact pre-processing (not heuristic), it leads to a better initial saturation table  $S$  for the uncolored vertices;
4. clq-col- $n$ : combine clq and col- $n$  — best of both worlds.

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<sup>6</sup>It is NP-hard, an heuristic can be used.

# Local Optimization with Larger Neighborhoods

Let  $(c)$  be a partial  $k$ -coloring, where  $k$  is the number of colors used until now.

- $C = \{i \in I \mid c_i > 0\}$  is the set of colored vertices in  $(c)$ .
- $U \subset \{i \in I \mid c_i = -1\}$  is a subset of uncolored vertices in  $(c)$ .

We want to define an ILP formulation to **assign** a color to each vertex  $u \in U$  while **preserving** the colors of vertices in  $C$ .

An **hybrid** formulation of **assignment**-based and **representative**-based formulations is used.

# Matheuristic DSATUR Formulation

$$\min_z \sum_{u \in U} x_{u,u}$$

$$s.t. \quad z_{i,l} + z_{i',l} \leq 1$$

$$x_{u,i} + x_{u,i'} \leq x_{u,u}$$

$$\sum_{i' \in U: i' \leq i} x_{i',i} + \sum_{l \in K_u} z_{i,l} \geq 1 \quad \forall u \in U$$

$$\forall (v_i, v_{i'}) \in E_U,$$

$$\forall l \in \llbracket 1; k \rrbracket$$

$$\forall (v_i, v_{i'}) \in E_U,$$

$$\forall u \in U, u \leq i$$

$$\forall u \in U$$

- Binary variables  $x_{u,u'}$  are defined only for  $u \leq u' \in U$ , when considering  $E_U = \{(v_u, v_{u'})\}_{u < u' \in U} \subset E$ .
- Binary variables  $z_{u,l}$ , to **assign previous colors**, are defined for  $u \in U$  and  $l \in \llbracket 1; k \rrbracket$  s.t. no neighbor  $u$  has color  $l$  in  $(c)$  – i.e., for all  $u \in U$  and  $l \in K_u$ , where  $K_u = \{l \in \llbracket 1; k \rrbracket \mid \forall i \in C, c_i = l \implies \text{ngb}(i, j) = 0\}$

# Matheuristic DSATUR Formulation

$$\min_z \sum_{u \in U} x_{u,u}$$

$$s.t. \quad z_{i,l} + z_{i',l} \leq 1$$

$$x_{u,i} + x_{u,i'} \leq x_{u,u}$$

$$\sum_{i' \in U: i' \leq i} x_{i',i} + \sum_{l \in K_u} z_{i,l} \geq 1$$

$$\forall (v_i, v_{i'}) \in E_U,$$

$$\forall l \in \llbracket 1; k \rrbracket$$

$$\forall (v_i, v_{i'}) \in E_U,$$

$$\forall u \in U, u \leq i$$

$$\forall u \in U$$

- It is **assignment**-based for variables  $z_{u,l}$ , ensuring that vertices in  $U$  are assigned either a previous color  $l$  in  $K_u$  or share the color with another vertex in  $U$ .
- It is **representative**-based for variables  $x_{i,i'}$ , ensuring that vertices in  $U$  sharing the same color have a representative vertex with the minimum index.

# Matheuristic DSATUR Formulation

$$\min_z \sum_{u \in U} x_{u,u}$$

$$s.t. \quad z_{i,l} + z_{i',l} \leq 1$$

$$x_{u,i} + x_{u,i'} \leq x_{u,u}$$

$$\sum_{i' \in U: i' \leq i} x_{i',i} + \sum_{l \in K_u} z_{i,l} \geq 1$$

$$\forall (v_i, v_{i'}) \in E_U,$$

$$\forall l \in \llbracket 1; k \rrbracket$$

$$\forall (v_i, v_{i'}) \in E_U,$$

$$\forall u \in U, u \leq i$$

$$\forall u \in U$$

- The 1st set ensures that adjacent vertices in  $U$  do not share the same **existing** color  $l$ .
- The 2nd set ensures that two adjacent vertices in  $U$  cannot share the same representative color.
- The 3rd set ensures,  $\forall i \in U$ , that either it receives a **previous** color  $l$  in  $K_u$  or it receives a **new** color represented by another vertex  $i'$  in  $U$  with  $i' \leq i$ .

# Matheuristic DSATUR Algorithm

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**Algorithm 2: Matheuristic DSATUR variants**

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**Input:**  $G = (V, E)$  a non-empty and non-oriented graph

**Parameters:**

- an initialization strategy  $\mathcal{S}$  (from Sect. 3.1) ;
- $o \in \mathbb{N}$ ,  $o > 1$  ;
- $r \in \mathbb{N}$ .

**Initialization:**

initialize colored set  $C$ , and color  $k$  with strategy  $\mathcal{S}$ .

initialize  $W := V \setminus C$ .

update partial coloring  $c$  and saturation table  $S$  with strategy  $\mathcal{S}$ .

**while**  $W \neq \emptyset$

**sort**  $W$  with order  $\succsim$ .

**define**  $U_1$  as the  $o$  first elements after sorting.

**define**  $U_2$  as the elements of rank  $o + 1$  and  $\min(|W|, o + r)$  after sorting.

**solve** ILP (15) with  $C$  and  $U = U_1 \cup U_2$ .

$k := k + OPT$  where  $OPT$  is the optimal value of the last ILP.

**if**  $o + r \leq |W|$  **then**  $U_1 = U$  **end if**

**set**  $W := W \setminus U_1$

**assign** colors  $c_u$  of the ILP for  $u \in U_1$

**end while**

**return** color  $k$  and  $(c)$  a  $k$ -coloring of  $G$

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- $\mathcal{S}$  for *initialization* induces  $k$ ,  $C$ ,  $S$ ,  $c$ , and  $W$ .
- Simultaneously colors  $o$  vertices solving the **matheuristic DSATUR ILP** formulation (the standard DSATUR have  $o = 1$  and  $r = 0$ ).
- Having  $r > 0$  ensures **more depth** in the local search and the possibility to **reoptimize** in later iterations (set  $W := W \setminus U_1$ ).
- $U_2$  helps the ILP in having context when coloring **critical** vertices  $U_1$ .<sup>7</sup>
- $o + r \geq W$  holds in the last iteration, and  $U_1 = U = W$  ensures both termination ( $W \setminus U_1 = \emptyset$ ) and efficiency (no useless re-optimization—i.e., recoloring  $r$  vertices).

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<sup>7</sup> $o + r$  should be fine-tuned according to the ILP solver capabilities and instance features.

# Dual Bounds

TODO



**Thank You!**

# Bibliography

- [1] F. Furini and E. Malaguti, “Exact weighted vertex coloring via branch-and-price,” *Discrete Optimization*, vol. 9, no. 2, pp. 130–136, 2012.
- [2] D. Cornaz, F. Furini, and E. Malaguti, “Solving vertex coloring problems as maximum weight stable set problems,” *Discrete Applied Mathematics*, vol. 217, pp. 151–162, 2017.
- [3] D. Brélaz, “New methods to color the vertices of a graph,” *Commun. ACM*, vol. 22, no. 4, pp. 251–256, Apr. 1979.
- [4] N. Dupin, “Matheuristic Variants of DSATUR for the Vertex Coloring Problem,” in *Metaheuristics*, M. Sevaux, A.-L. Olteanu, E. G. Pardo, A. Sifaleras, and S. Makboul, Eds., Cham: Springer Nature Switzerland, 2024, pp. 96–111.