

# 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  $\text{ngb}(v_i, v_j) = 1$  if  $(v_i, v_j) \in E$ .<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  $c_i = [\![1; k]\!]$  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  $\succcurlyeq$  over  $V$  is defined as:  $v_i \succcurlyeq v_j \iff s_i > s_j \vee (s_i = s_j \wedge d_i \geq d_j)$

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

$z_{i,c} \in \{0, 1\}$   
indicates if  
vertex  $v_i$  is  
assigned color  
 $c$ .

$y_c \in \{0, 1\}$   
indicates if  
color  $c$  is used  
in the coloring.

$$\min \sum_{c=1}^k y_c$$

$$s.t. \min \sum_{c=1}^k z_{i,c} = 1 \quad \forall i \in I$$

$$z_{i,c} + z_{j,c} \leq y_c \quad \forall (v_i, v_j) \in E, \\ \forall c \in \llbracket 1; k \rrbracket$$

The objective minimizes the number of used colors.

The 1st set ensures that each vertex is assigned exactly one color.

The 2nd set ensures that adjacent vertices do not share 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

$$x_{i,i'} \in \{0, 1\},$$

$\forall i, i' \in$

$V$  s.t.  $i \leq i'$ ,

indicates if

vertices  $v_i$  and  $v_{i'}$  share the same color and

$i$  is the minimum index of its color class.

$$\min_z \sum_{i=1}^n x_{i,i}$$

$$s.t. \sum_{i' \leq i} x_{i',i} \geq 1 \quad \forall i \in I$$

$$x_{j,i} + x_{j,i'} \leq x_{j,j} \quad \forall (v_i, v_{i'}) \in E, \\ \forall j \leq i$$

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  $x_{j,i} = 1 \Rightarrow x_{j,j} = 1$

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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  $\succcurlyeq$  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  $\succcurlyeq$ , 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  $\succsim$ , 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

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

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

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

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

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

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

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

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

- 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

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

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

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

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

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

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

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

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

- 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

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

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

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

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

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

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

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

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

- 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

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