

Matheuristic Variants of DSATUR for the Vertex Coloring Problem

Federico Bruzzone,¹ PhD Candidate

Milan, Italy – 4 December 2025



¹ADAPT Lab – Università degli Studi di Milano,
Website: federicobruzzone.github.io,
Github: github.com/FedericoBruzzone,
Email: federico.bruzzone@unimi.it
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 $i \in I$, δ_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 $(v_i, v_j) \in E$.²

²ngb 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 Formulations³, 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.

³Efficient 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 Model⁴, 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.

$$\begin{aligned} & \min_z \sum_{i=1}^n x_{i,i} \\ \text{s.t. } & \sum_{i' \leq i} x_{i',i} \geq 1 \quad \forall i \in I \\ & x_{j,i} + x_{j,i'} \leq x_{j,j} \quad \forall (v_i, v_{i'}) \in E, \\ & \quad \forall j \leq i \end{aligned}$$

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

The 1st set ensures either $x_{i,i} = 1$ or v_i shares the color with a vertex $v_{i'}$ with $i' < i$.

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

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

Standard DSATUR Algorithm

Algorithm 1: Standard DSATUR algorithm

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

- 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 Variants⁵

⁵N. 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 clique⁶ 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.

⁶It 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

Algorithm 2: Matheuristic DSATUR variants

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

- TODO
- TODO
- 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.