

Algorithms and Experiments for the Approximate Balanced Coloring Problem

SIAM DS17

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Joint work with

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May 24, 2017

Overview

- Describe **balanced coloring**
 - Finds synchronous clusters of nodes
 - Algorithm of Belykh and Hasler (2011) finds the **largest** synchronous clustering of nodes
 - How do we generalize this?

Premise

Balanced
Coloring

OR approach

Overview

An Integer program

Future work

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- Describe **balanced coloring**
 - Finds synchronous clusters of nodes
 - Algorithm of Belykh and Hasler (2011) finds the **largest** synchronous clustering of nodes
 - How do we generalize this?
- Apply **integer programming** to problem
 - A technique from **operations research**, and, more specifically, **mathematical programming**
 - Provides a framework to systematically model variations of a given problem
- Cross disciplinary fun with operations research!
 - **Disclaimer:** I'm not a dynamical systems expert
 - Optimization, machine learning, (big) data analytics, stochastics, control, etc
 - **Talk goal: breaking the DS/OR language barrier**

Premise

Balanced
Coloring

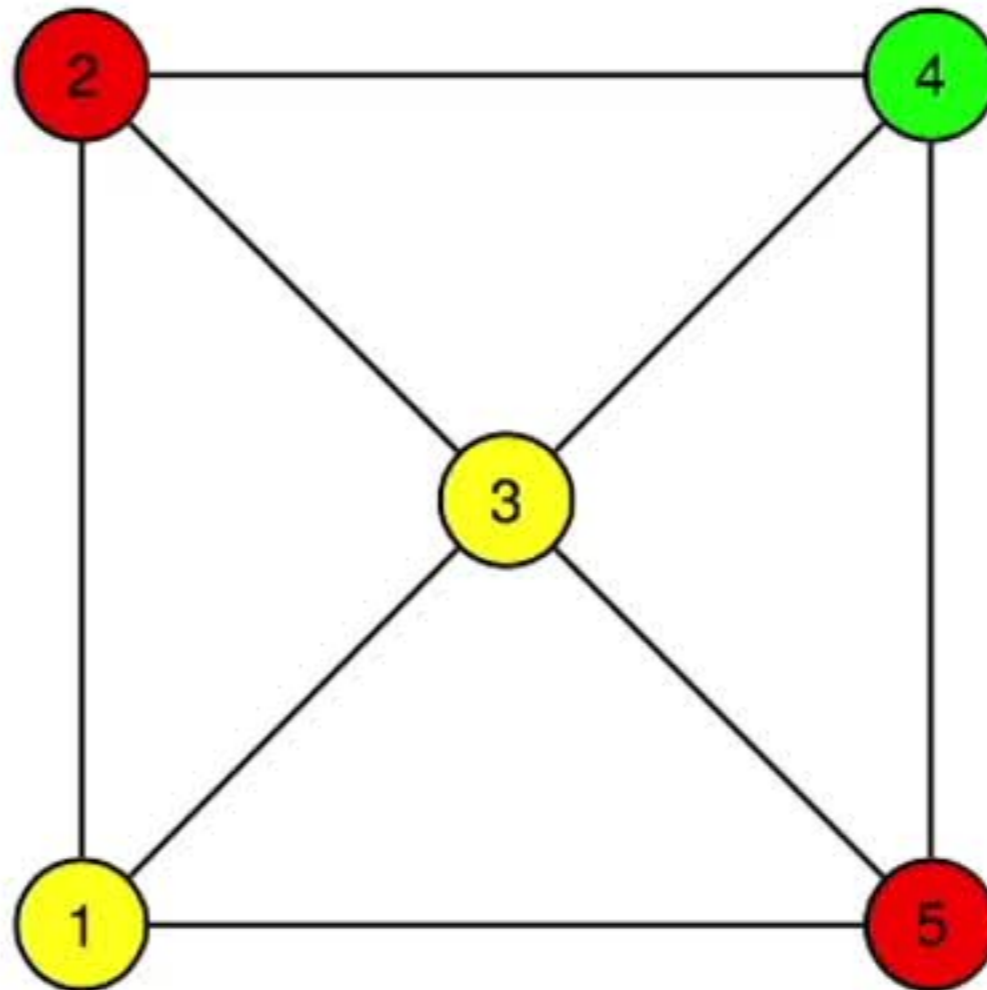
OR approach:

Overview

An integer program

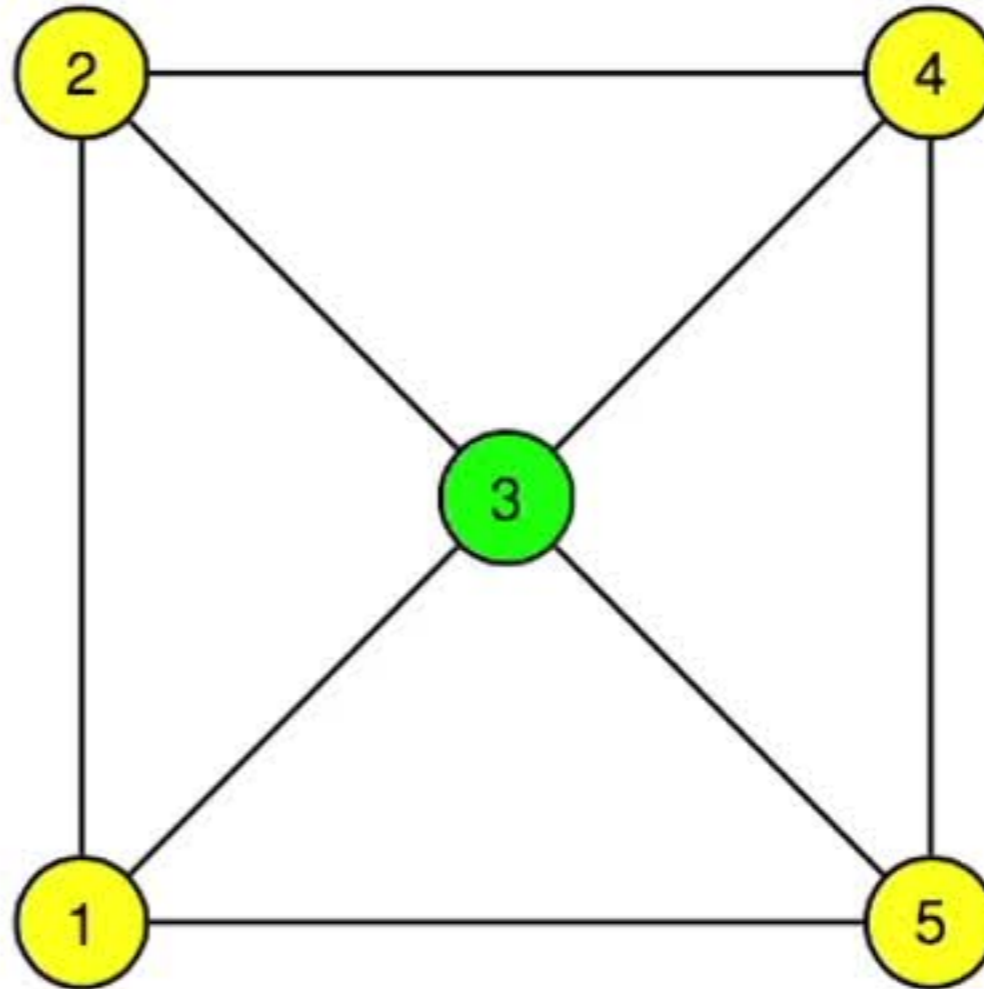
Future work

A *coloring* of G is a partition, \mathcal{C} , of the nodes in G .



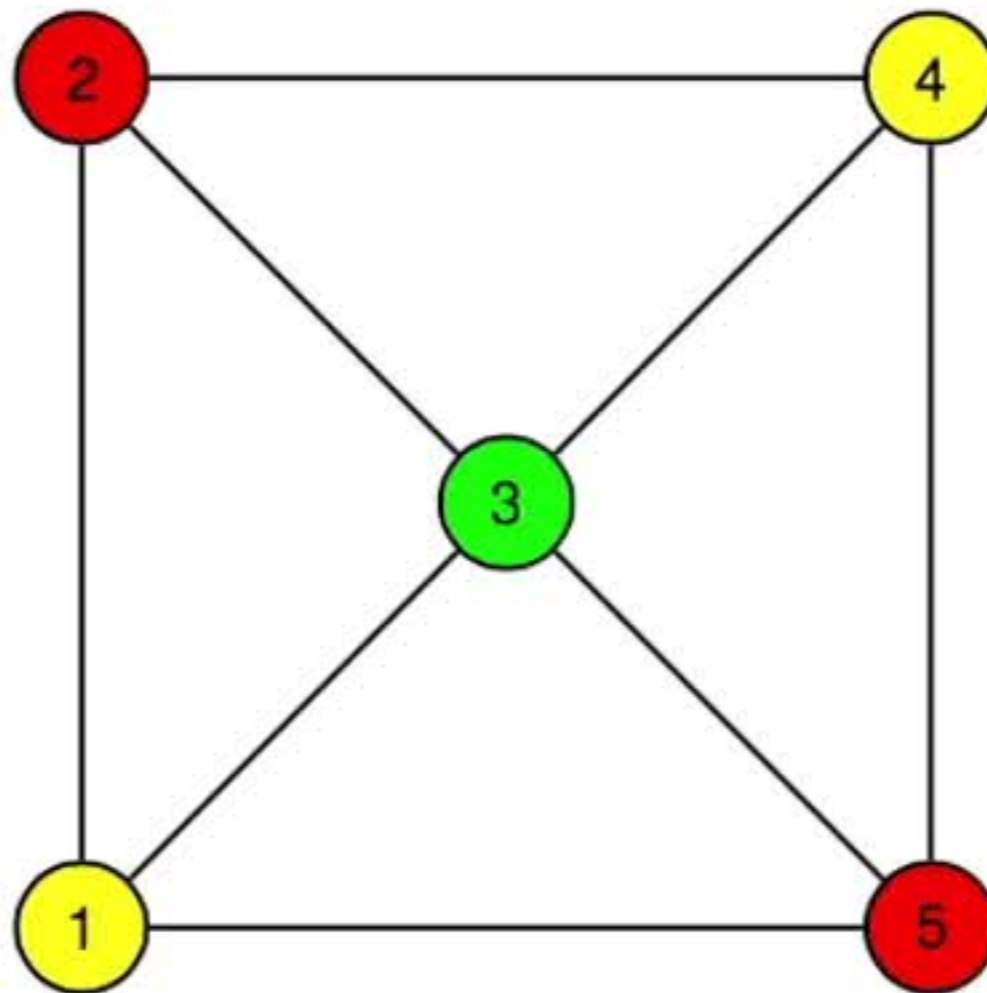
$$\mathcal{C} = \{\{1, 3\}, \{2, 5\}, \{4\}\}$$

A *balanced coloring*: for every $S, T \in \mathcal{C}$ and $u, v \in S$, the number of edges from u and v to nodes in T is equal.



Balanced

A *minimal balanced coloring* is the balanced coloring using as few colors as possible.



Still balanced, not minimal

Premise

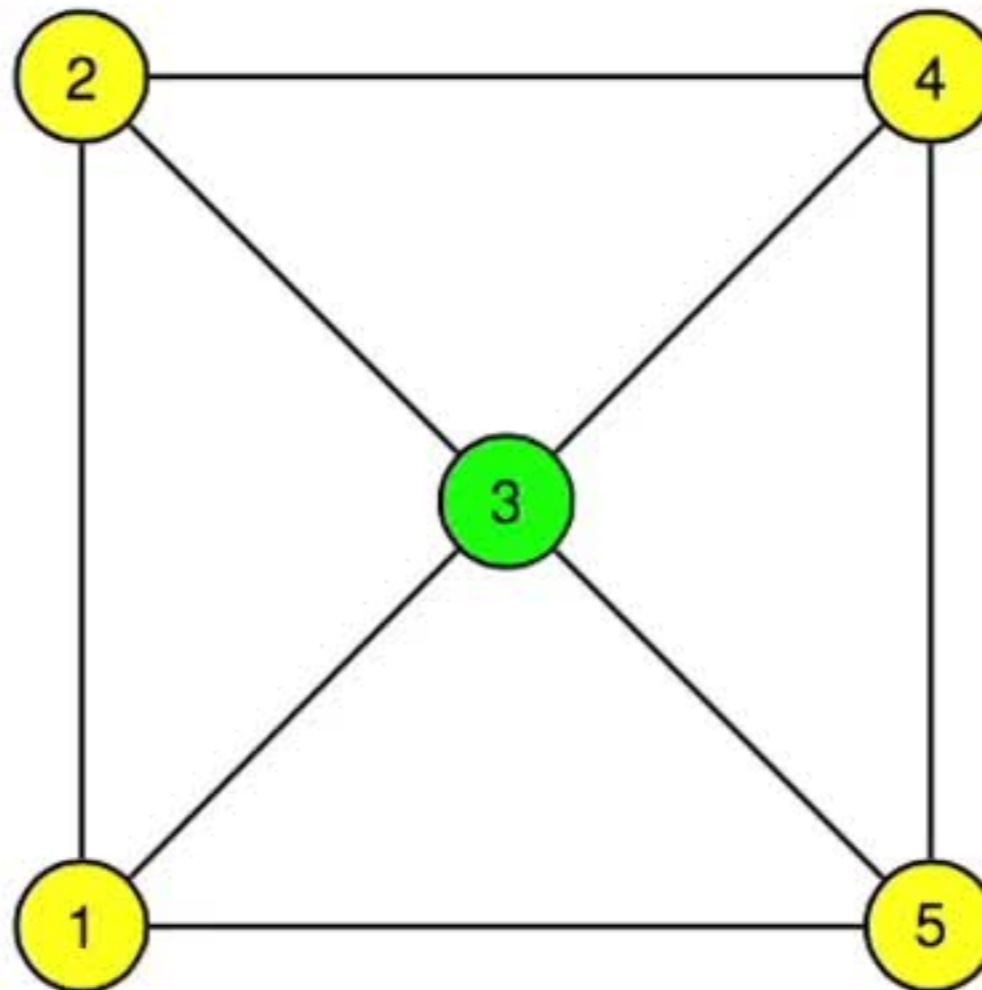
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Minimal balanced coloring finds the largest synchronous clusters

- A greedy algorithm by Belykh and Hasler (2011) finds the minimal balanced coloring in polynomial time.
- What about other generalizations?
 - Approximate edge weights, isolating nodes, directed graphs, etc.

Mathematical programming:

$$\begin{array}{ll} \max & f(x) \\ \text{subject to} & x \in S \end{array}$$

$$f : S \subset \mathbf{R}^n \rightarrow \mathbf{R}$$

Linear programming: f is linear, S is a polyhedron, i.e.,

$$f(x) = \sum_i c_i x_i,$$

$$S = \{x \in \mathbf{R}^n : Ax \leq b\}, A \in \mathbf{R}^{m \times n}, b \in \mathbf{R}^m.$$

x are decision variables

Goal: find the optimal x

(Linear) Integer programming: f linear, S is a polyhedron intersected with \mathbf{Z}^n .

Implementations exist to solve linear and integer programs.

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$G = (V, E)$, $C = \text{set of colors}$

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$G = (V, E)$, $C =$ set of colors

Decision variables

t = number of colors used

z_c = indicator of color c used, $c \in C$

y_{ic} = indicator of i colored c , $i \in V$, $c \in C$

x_{ijcd} = indicator of i colored c , j colored d , $ij \in E$, $c, d \in C$

$1 + |C| + |V||C| + |E||C|^2 = O(|E||C|^2)$ variables 😞

Idea: Use linear equations and inequalities to create the set of all feasible balanced colorings.

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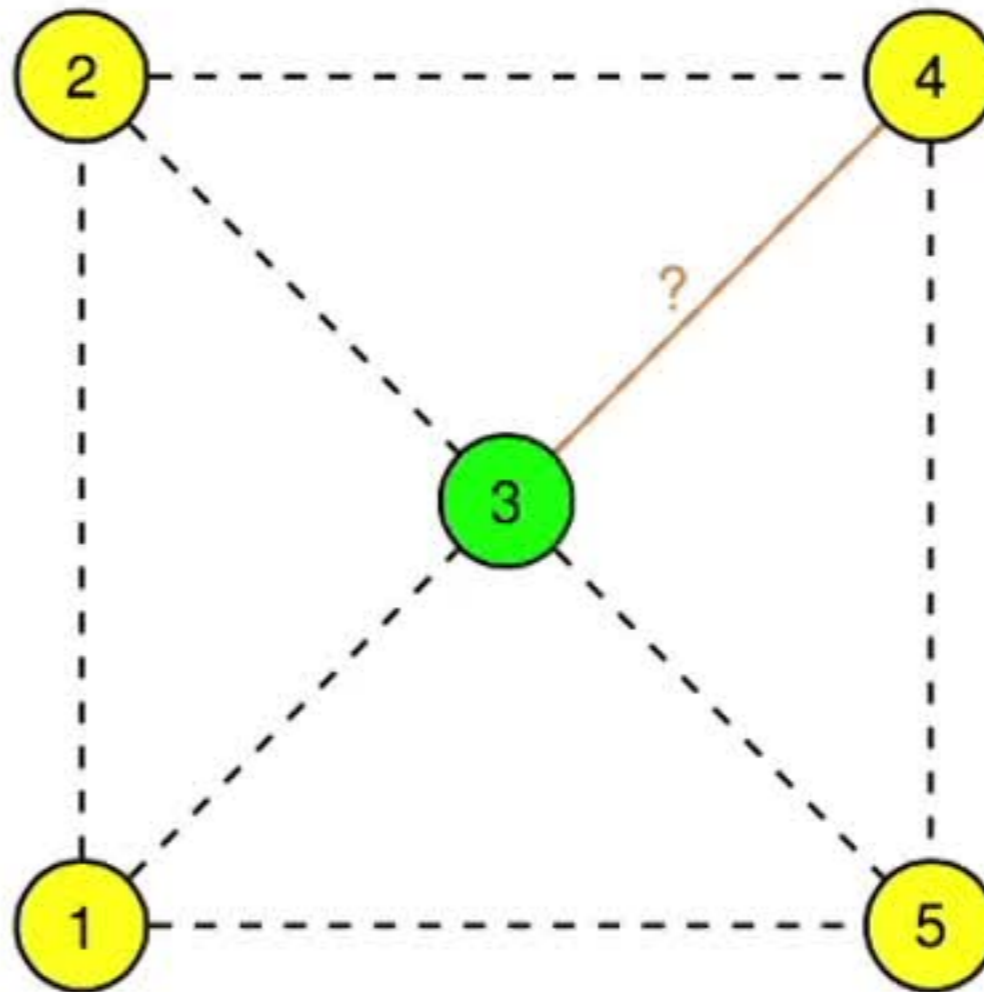
An Integer program

Future work

$$\begin{array}{llll}
 \min & t & & \\
 \text{s.t.} & \sum_{c \in C} y_{ic} = 1 & i \in V & \text{(a)} \\
 & \sum_{c \in C, d \in C} x_{ijcd} = 1 & ij \in E & \text{(b)} \\
 & x_{ijcd} \leq y_{ic} & ij \in E, c, d \in C & \text{(c1)} \\
 & x_{ijcd} \leq y_{jd} & (i, j) \in E, c, d \in C & \text{(c2)} \\
 & \sum_{ij \in E} x_{ijcd} \leq \sum_{pq \in E} x_{pqcd} + M(1 - y_{ik}) + M(1 - y_{pk}) & i, p \in V, c, d \in C & \text{(d1)} \\
 & \sum_{ji \in E} x_{jicd} \leq \sum_{qp \in E} x_{qpdc} + M(1 - y_{ik}) + M(1 - y_{pk}) & i, p \in V, c, d \in C & \text{(d2)} \\
 & y_{ic} \leq z_c & i \in V, c \in C & \text{(e)} \\
 & t = \sum_{c \in C} z_c & & \text{(f)} \\
 & x_{ijcd}, y_{ic}, z_c \in \{0, 1\} & \forall i, j, c, d &
 \end{array}$$

$$\begin{aligned}
 & |V| + |E| + 2|E||C|^2 + 2|V|^2|C|^2 + |V||C| + 1 \\
 & = O(|V|^2|C|^2) \text{ constraints.}
 \end{aligned}$$

Constraint examples:

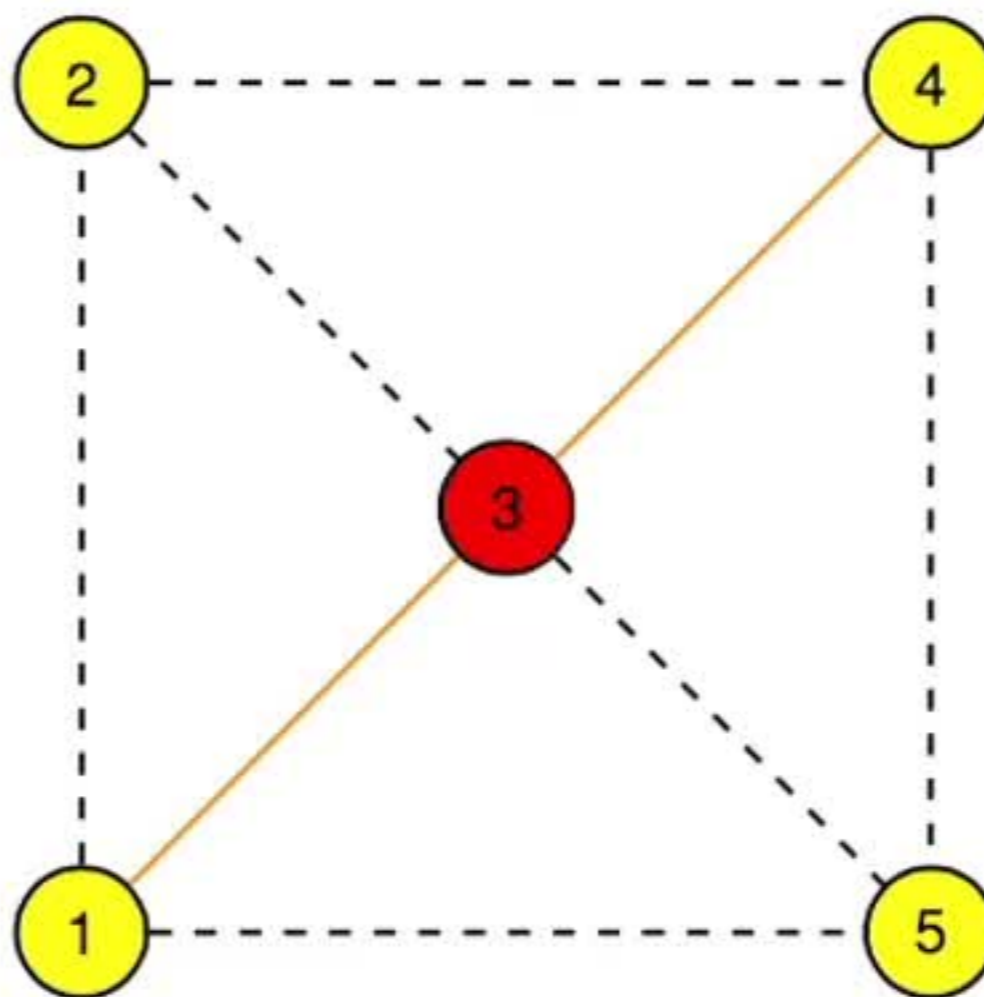


$$x_{34GR} \leq y_{3G}$$

$$x_{34GR} \not\leq y_{4R}$$

Forces $x_{34GR} = 0$ with this color setting

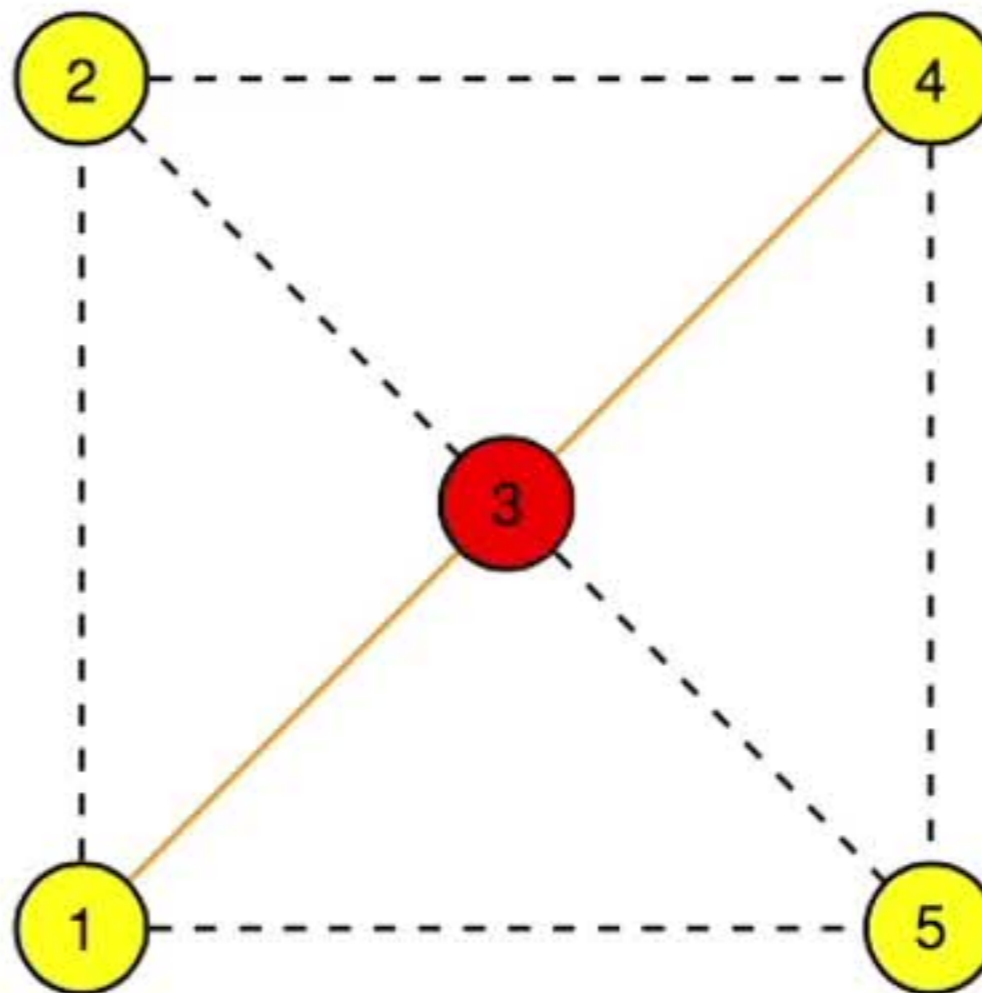
Constraint examples:



$$\cancel{X_{12}YR} + X_{13}YR + \cancel{X_{15}YR} \leq \cancel{X_{42}YR} + X_{43}YR + \cancel{X_{45}YR} + \cancel{M(1 - y_{1Y})} + \cancel{M(1 - y_{4Y})}$$

Symmetric constraint results in $X_{13}YR = X_{43}YR$ for this color setting

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Ongoing work

- Improve the runtime for solving the integer program
 - Can solve up to 50 node graphs tractably 😞
 - Adding families of valid inequalities
 - $O(|E||C|^2)$: Reducing the size of the problem, specifically bounding the number of colors to consider
 - More sophisticated techniques
- Extensions to model
 - Fixing the color number to find additional colorings
 - Approximate coloring: weighting the edges
 - Solving the Laplacian case