

# Minimum Sum Colorings of Chordal Graphs

Ian DeHaan and Zachary Friggstad



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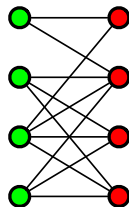
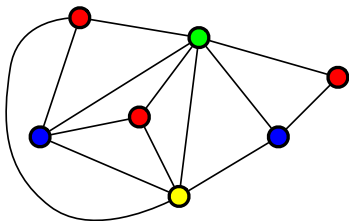


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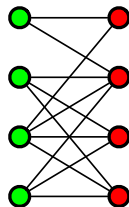
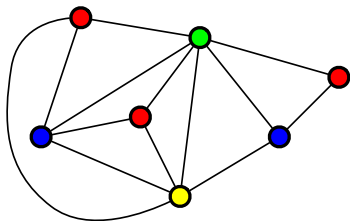
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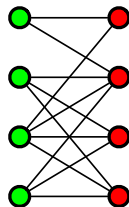
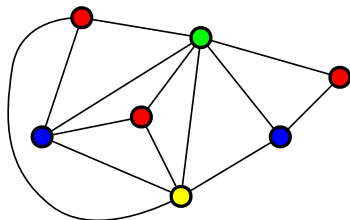
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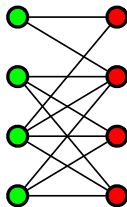
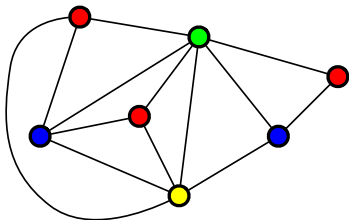
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## Scheduling Models

- ▶ Vertices: jobs to be processed
- ▶ Edges: resource conflicts
- ▶ Objective: minimize the final completion time

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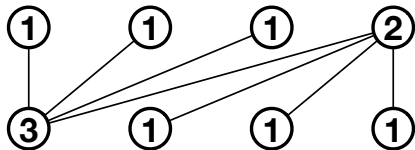
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- ▶ Vertices: jobs to be processed
- ▶ Edges: resource conflicts
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## Minimum-Sum Coloring Problem

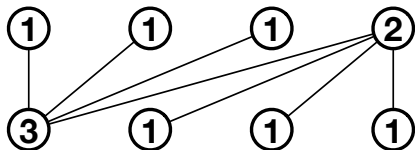
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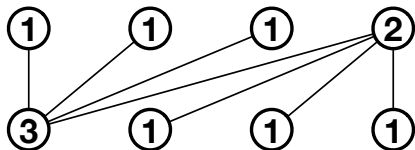


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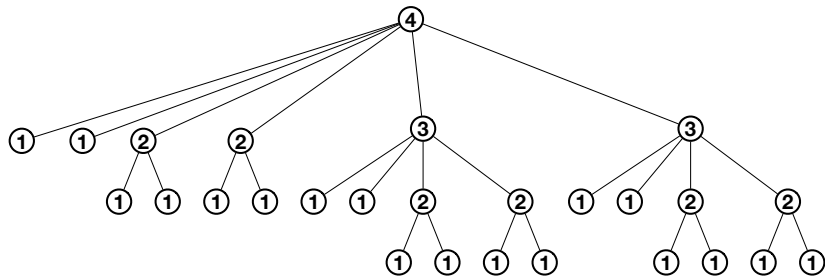
but allowing 3 colors gives a minimum sum of **11**.

There is no  $n^{0.999}$ -approximation for arbitrary graphs.

[Bar-Noy et al, 1998]



## Min-Sum Coloring on Trees



- ▶ The optimum coloring uses  $O(\log n)$  colors.  
[Kubicka and Schwenk, 1989]
- ▶ An optimal solution can be computed in  $O(n \cdot \log^2 n)$  time using dynamic programming.

## Min-Sum Coloring on Bipartite Graphs

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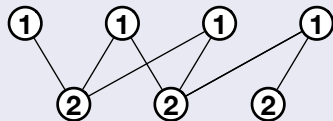
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## Example:

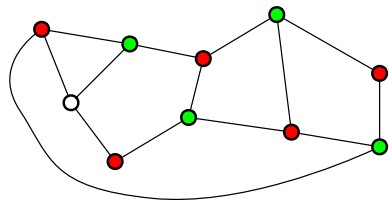
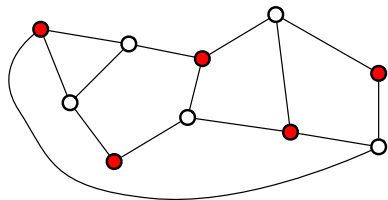
The natural 2-coloring is a 1.5-approximation.



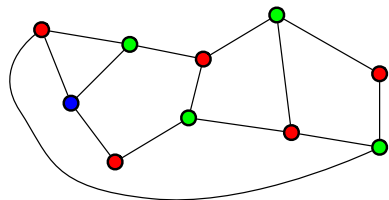
**Analysis:** The sum of colors is at most  $n + \frac{1}{2} \cdot n = 1.5 \cdot n \leq 1.5 \cdot OPT$ .

# A Greedy Algorithm

Greedly coloring maximum independent sets yields a 4-approximation. [Bar-Noy et al, 1998]



Red = 1  
Green = 2  
Blue = 3



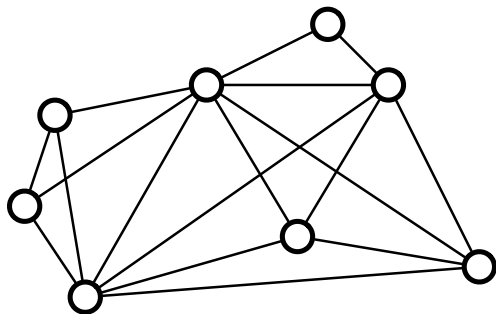
## Some Known Approximation Bounds for Min-Sum coloring

Graph Class	Upper Bound	Lower Bound
Perfect	3.591	APX-HARD
<b>Chordal</b>	<b>1.796 + <math>\epsilon</math></b>	APX-HARD
Interval	1.796	APX-HARD
Bipartite	27/26	APX-HARD
Planar	PTAS	NP-HARD
Line graphs	1.8298	APX-HARD

# Chordal Graphs

## Definition (Chordal Graphs)

No induced cycles of length  $\geq 4$ , a.k.a. triangulated graphs.







# Interval Graph Algorithm

Suppose  $G$  allows us to compute a maximum-size  $k$ -colorable subgraph (MKCS) in polynomial time for any  $k \geq 1$ .

Then, a 1.796-approximation is possible. [Halldórsson et al., 2008]

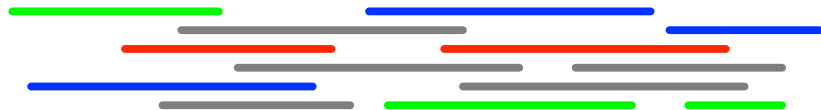


Figure: Interval graphs can be used to represent concurrent tasks.

# Interval Graph Algorithm

Set  $c := 3.591$  and pick some  $\delta \in [0, 1)$  uniformly randomly.

**for**  $k = c^\delta, c^{\delta+1}, c^{\delta+2}, c^{\delta+3}, \dots$

- ▶ Find a maximum  $k$ -colorable subgraph of  $G$ .
- ▶ Use the next  $k$  unused integers to color it.
- ▶ Remove these nodes from  $G$ .

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- ▶ Find a maximum  $k$ -colorable subgraph of  $G$ .
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- ▶ Remove these nodes from  $G$ .

This yields an approximation with guarantee

$$\frac{c+1}{2 \cdot \ln c} \approx 1.796$$

## MKCS in Chordal Graphs

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To get an improved approximation for min-sum coloring, we require an approach which can be used with MKCS approximations.

# MKCS in Chordal Graphs

## Theorem (D., Friggstad, 2023)

*Let  $\mathcal{G}$  be a graph class that is closed under taking induced subgraphs. If there is a PTAS for weighted MKCS in  $\mathcal{G}$ , then for any  $\epsilon > 0$ , there is a polynomial-time  $(1.796 + \epsilon)$ -approximation for minimum-sum coloring in  $\mathcal{G}$ .*

# MkCS in Chordal Graphs

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**General Statement:** Given a  $\rho$ -approximation for MkCS, our min-sum coloring approximation guarantee is:

$$\inf_{1 < c < \frac{1}{1-\rho}} \frac{\rho \cdot (c + 1)}{2 \cdot (1 - (1 - \rho) \cdot c) \cdot \ln c}$$



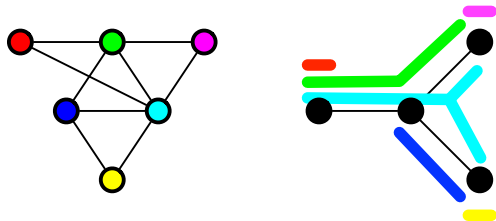
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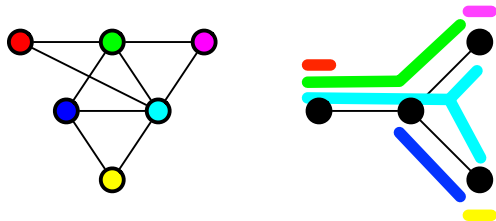
Any chordal graph is the intersection graph of subtrees of a tree  $\mathcal{T}$ .



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### Dynamic Program:

For  $v \in \mathcal{T}$  and any  $\leq k$  subtrees  $S$  spanning  $v$ :

$f(v, S)$  = best solution in the subtree under  $v$  that uses  $S$  at  $v$

## Approximations for Large $k$

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To get the  $(1 - \epsilon)$ -approximation for any constant  $\epsilon$ :

- ▶ If  $k \leq 8/\epsilon^3$ , use the exact algorithm (running time  $n^{O(1/\epsilon^3)}$ ).
- ▶ Otherwise, use our new  $(1 - 2/k^{1/3}) \geq (1 - \epsilon)$  approximation.

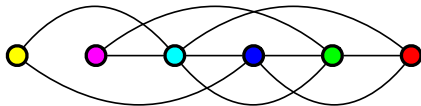
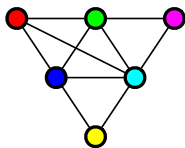
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An ordering  $v_1, \dots, v_n$  of the nodes such that for each  $v_i$ , its right neighbors  $N^r(v_i)$  form a clique.

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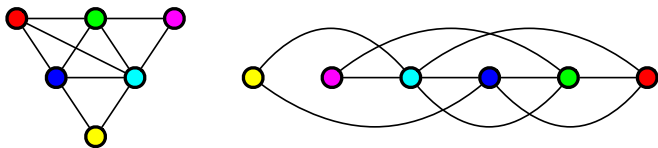
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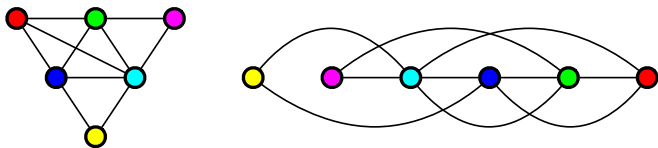
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### Remark

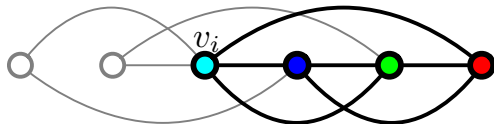
$S \subseteq V$  induces a  $k$ -colorable subgraph iff for each  $v_i \in S$ ,

$$|N^r(v_i) \cap S| \leq k - 1.$$

# Linear Programming Formulation

Let  $x_i$  indicate if we select  $v_i$  in our  $k$ -colorable subgraph.

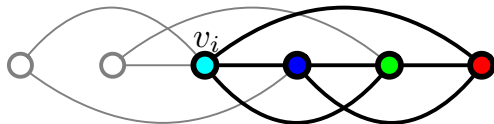
$$\begin{aligned} \text{maximize :} & \quad \sum_i w_i \cdot x_i \\ \text{subject to :} & \quad x_i + \sum_{v_j \in N^r(v_i)} x_j \leq k \quad \forall i \\ & \quad x_i \in [0, 1] \quad \forall i \end{aligned}$$



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## Rounding Algorithm

- ▶  $S \leftarrow \emptyset$
- ▶ **for**  $i$  from  $n$  down to 1:
  - ▶ Flip a coin  $c_i$  with bias  $(1 - 1/k^{1/3}) \cdot x_i$  towards heads.
  - ▶ If  $S \cup \{v_i\}$  is feasible and  $c_i = \text{heads}$ , add  $v_i$  to  $S$ .

## Analysis of the Rounding Algorithm

For each  $v_i$ , the **expected** number of heads flipped from  $N^r(v_i)$  is at most  $k \cdot (1 - k^{-1/3}) = k - k^{2/3}$ .

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Thus,  $v_i$  will be added with probability at least:

$$\Pr[|\#\text{heads from } N^r(v_i)| < k \wedge c_i = \text{heads}] \geq (1 - k^{-1/3}) \cdot (1 - k^{-1/3}) \cdot x_i.$$

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$\therefore$  The expected size of  $S$  is at least  $(1 - 2/k^{1/3}) \cdot OPT_{LP}$ .

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This analysis approach was presented because there are generic ways to efficiently de-randomize such algorithms if the analysis only uses second moments.

With more hands-on work, we can de-randomize tighter analysis (using Chernoff bounds) to get a deterministic algorithm with guarantee

$$1 - \frac{O(1)}{k \log k}.$$

## $k$ -Colorable Subgraphs to Min-Sum Coloring

### Theorem (D., Friggstad 2023)

*Let  $\mathcal{G}$  be a graph class that is closed under taking induced subgraphs. If there is a PTAS for weighted MKCS in  $\mathcal{G}$ , then for any  $\epsilon > 0$ , there is a polynomial-time  $(1.796 + \epsilon)$ -approximation for minimum-sum coloring in  $\mathcal{G}$ .*

# Linear Program

- ▶  $x_{v,k}$  - indicates that  $v$  has color  $k$ .
- ▶  $z_{C,k}$  - indicates that  $C$  is the set of nodes that are  $\leq k$ -colored.

$$\text{minimize: } \sum_{v \in V} \sum_{k=1}^n w_v \cdot k \cdot x_{v,k} \quad (1)$$

$$\text{subject to: } \sum_{k=1}^n x_{v,k} = 1 \quad \forall v \in V \quad (2)$$

$$\sum_{C \in \mathcal{C}_k} z_{C,k} \leq 1 \quad \forall 1 \leq k \leq n \quad (3)$$

$$\sum_{C \in \mathcal{C}_k: v \in C} z_{C,k} \geq \sum_{k' \leq k} x_{v,k'} \quad \forall v \in V, 1 \leq k \leq n \quad (4)$$

$$x, z \geq 0$$

**Last constraint:** (partial) agreement between  $z$  and  $x$  on the statement  $v$  is colored by color at most  $k$ .

## Taking the Dual

$$\text{maximize: } \sum_{v \in V} \alpha_v - \sum_{k=1}^n \beta_k$$

$$\text{subject to: } \alpha_v \leq w_v \cdot k + \sum_{\hat{k}=k}^n \theta_{v,\hat{k}} \quad \forall v \in V, 1 \leq k \leq n \quad (5)$$

$$\sum_{v \in C} \theta_{v,k} \leq \beta_k \quad \forall 1 \leq k \leq n, C \in \mathcal{C}_k \quad (6)$$

$$\beta, \theta \geq 0 \quad (7)$$

## Solving and Rounding

Via the ellipsoid method for solving LPs, this yields a solution  $(x, z)$  with value  $\leq OPT$  for the following slightly modified LP.

$$\begin{aligned}\sum_t x_{v,t} &= 1 && \forall v \\ \sum_S z_{S,t} &= 1/\rho && \forall t \\ \sum_{S \ni v} z_{S,t} &\geq \sum_{t' \leq t} x_{t',v} && \forall v, t \\ x, z &\geq 0\end{aligned}$$

### Rounding Algorithm

$c \leftarrow 3.591$

$\delta \sim [0, 1)$  uniformly at random

For  $k := c^\delta, c^{\delta+1}, c^{\delta+2}, \dots$

- ▶ Sample a  $\lfloor k \rfloor$ -colorable subset  $S$  from the distribution  $\rho \cdot z_S$ .
- ▶ Randomly permute the coloring.
- ▶ Concatenate this coloring to the coloring of  $G$  so far.

## Next Steps: Min-Sum Coloring on Perfect Graphs

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The “latency constant”  $c \approx 3.591$  shows up in the current best approximation ratio for interval, chordal, and perfect graphs.

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### Question:

Is  $c$  a fundamental lower bound, or can these approximations be improved?

Thank you!