

Minimum Bounded Degree Spanning Trees

Refuting a Conjecture of M. X. Goemans

Stephen Chestnut Martin Nägele Rico Zenklusen

Institute for Operations Research

March 1, 2016

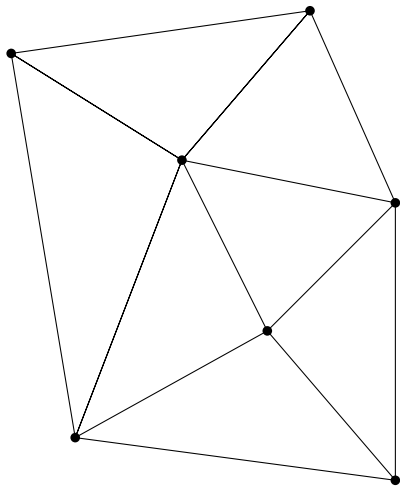
The Minimum Bounded Degree Spanning Tree Problem

(MBDST Problem)

The Minimum Bounded Degree Spanning Tree Problem (MBDST Problem)

Input:

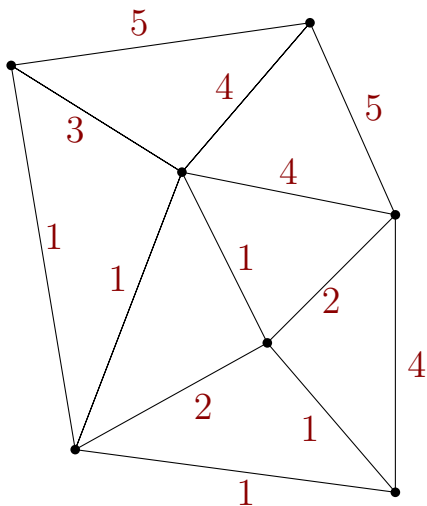
- ▶ graph $G = (V, E)$



The Minimum Bounded Degree Spanning Tree Problem (MBDST Problem)

Input:

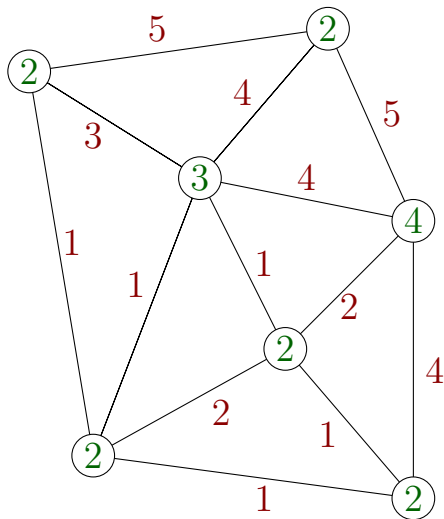
- ▶ graph $G = (V, E)$
- ▶ cost vector $c: E \rightarrow \mathbb{Q}_{>0}$



The Minimum Bounded Degree Spanning Tree Problem (MBDST Problem)

Input:

- ▶ graph $G = (V, E)$
- ▶ cost vector $c: E \rightarrow \mathbb{Q}_{>0}$
- ▶ degree bounds $d: V \rightarrow \mathbb{Z}_{>0}$



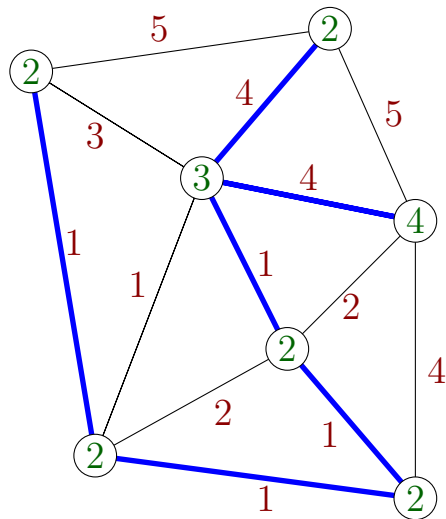
The Minimum Bounded Degree Spanning Tree Problem (MBDST Problem)

Input:

- ▶ graph $G = (V, E)$
- ▶ cost vector $c: E \rightarrow \mathbb{Q}_{>0}$
- ▶ degree bounds $d: V \rightarrow \mathbb{Z}_{>0}$

Output:

minimum cost **spanning tree**
satisfying the degree bounds



Goemans' Approximation Approach

Goemans' Approximation Approach

IP-formulation:

$$\min c^T x$$

$$\text{s.t. } x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset$$

$$x(E) = |V| - 1$$

$$x \geq 0$$

$$x(\delta(v)) \leq d(v) \quad \forall v \in V$$

$$x \in \{0, 1\}^E$$

Goemans' Approximation Approach

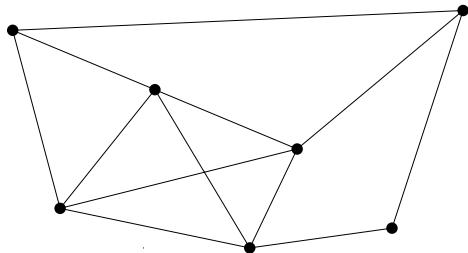
IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \begin{array}{l} \text{spanning tree constraints,} \\ \text{polytope } P_{ST}(G) \end{array}$$

Goemans' Approximation Approach

IP-formulation:

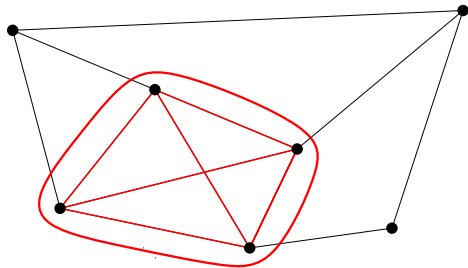
$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & \left. \begin{array}{l} x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ x(E) = |V| - 1 \\ x \geq 0 \end{array} \right\} \begin{array}{l} \text{spanning tree constraints,} \\ \text{polytope } P_{ST}(G) \end{array} \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array}$$



Goemans' Approximation Approach

IP-formulation:

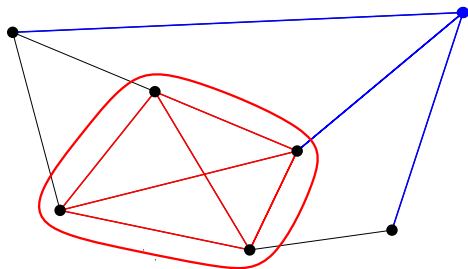
$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \begin{array}{l} \text{spanning tree constraints,} \\ \text{polytope } P_{ST}(G) \end{array}$$



Goemans' Approximation Approach

IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \begin{array}{l} \text{spanning tree constraints,} \\ \text{polytope } P_{ST}(G) \end{array}$$

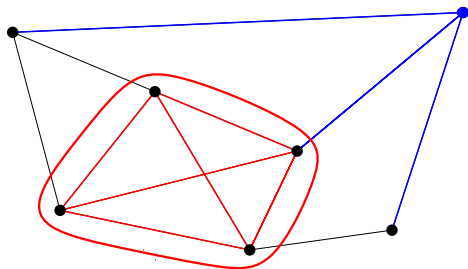


Goemans' Approximation Approach

IP-formulation:

$$\min c^T x$$

$$\left. \begin{array}{l} \text{s.t. } x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ x(E) = |V| - 1 \\ x \geq 0 \\ x(\delta(v)) \leq d(v) \quad \forall v \in V \\ x \in \{0, 1\}^E \end{array} \right\} \text{polytope } P_{BDST}(G, d)$$



Goemans' Approximation Approach

IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{polytope } P_{BDST}(G, d)$$

Main idea: use matroid intersection

Goemans' Approximation Approach

IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{polytope } P_{BDST}(G, d)$$

Main idea: use matroid intersection

- ▶ graphic matroid M_1

Goemans' Approximation Approach

IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{polytope } P_{BDST}(G, d)$$

Main idea: use matroid intersection

- ▶ graphic matroid M_1
- ▶ “degree constraint matroid” M_2 ?

Goemans' Approximation Approach

IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{polytope } P_{BDST}(G, d)$$

Main idea: use matroid intersection

- ▶ graphic matroid M_1
- ▶ “degree constraint matroid” M_2 ?

$$\mathcal{I}_2 = \{F \subseteq E \mid |F \cap \delta(v)| \leq d(v) \ \forall v\}$$

Goemans' Approximation Approach

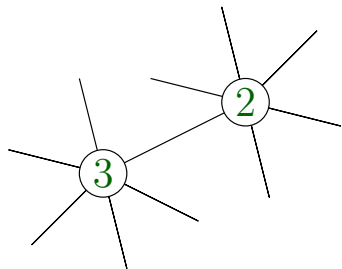
IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{polytope } P_{BDST}(G, d)$$

Main idea: use matroid intersection

- ▶ graphic matroid M_1
- ▶ “degree constraint matroid” M_2 ?

$$\mathcal{I}_2 = \{F \subseteq E \mid |F \cap \delta(v)| \leq d(v) \quad \forall v\}$$



Goemans' Approximation Approach

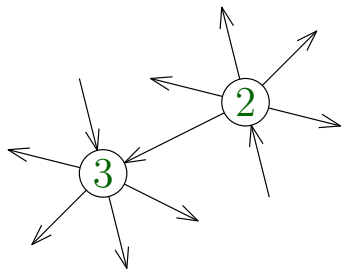
IP-formulation:

$$\begin{array}{ll} \min & c^T x \\ \text{s.t.} & x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ & x(E) = |V| - 1 \\ & x \geq 0 \\ & x(\delta(v)) \leq d(v) \quad \forall v \in V \\ & x \in \{0, 1\}^E \end{array} \left. \vphantom{\begin{array}{l} \min \\ \text{s.t.} \end{array}} \right\} \text{polytope } P_{BDST}(G, d)$$

Main idea: use matroid intersection

- ▶ graphic matroid M_1
- ▶ “degree constraint matroid” M_2 ?

$$\mathcal{I}_2 = \{F \subseteq E \mid |F \cap \delta(v)| \leq d(v) \quad \forall v\}$$



Goemans' Approximation Approach

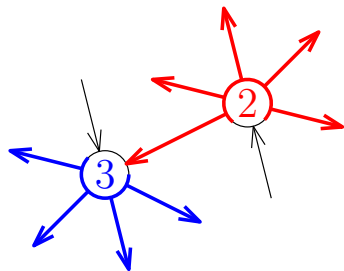
IP-formulation:

$$\begin{array}{l} \min c^T x \\ \text{s.t. } \left. \begin{array}{l} x(E[S]) \leq |S| - 1 \quad \forall S \subseteq V, S \neq \emptyset \\ x(E) = |V| - 1 \\ x \geq 0 \\ x(\delta(v)) \leq d(v) \quad \forall v \in V \\ x \in \{0, 1\}^E \end{array} \right\} \text{polytope } P_{BDST}(G, d) \end{array}$$

Main idea: use matroid intersection

- ▶ graphic matroid M_1
- ▶ “degree constraint matroid” M_2 ?

$$\mathcal{I}_2 = \{F \subseteq E \mid |F \cap \delta^+(v)| \leq d(v) \forall v\}$$



Goemans' Approximation Approach

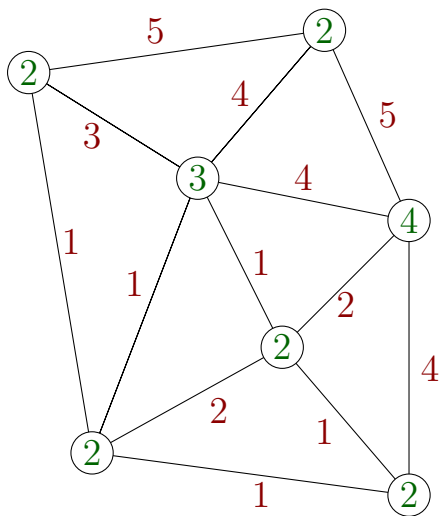
How to get a proper orientation?

Goemans' Approximation Approach

How to get a proper orientation?

► solve LP-relaxation

$$x^* = \arg \min_{x \in P_{BDST}(G,d)} c^\top x$$

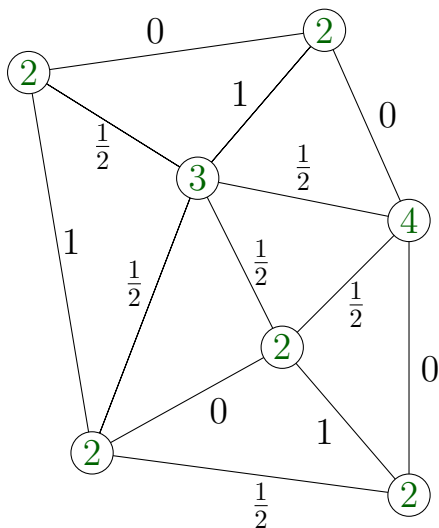


Goemans' Approximation Approach

How to get a proper orientation?

► solve LP-relaxation

$$x^* = \arg \min_{x \in P_{BDST}(G,d)} c^\top x$$



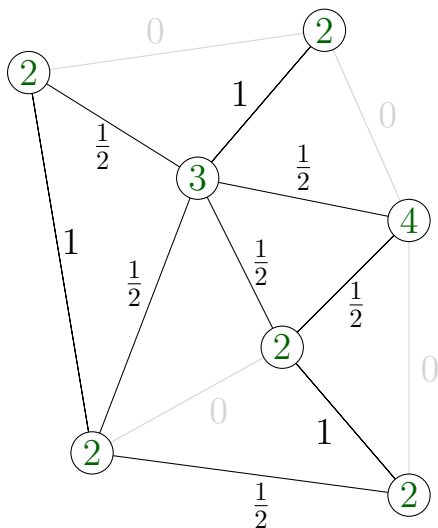
Goemans' Approximation Approach

How to get a proper orientation?

- ▶ solve LP-relaxation

$$x^* = \arg \min_{x \in P_{BDST}(G,d)} c^\top x$$

- ▶ restrict to $E^* := \text{supp } x^*$



Goemans' Approximation Approach

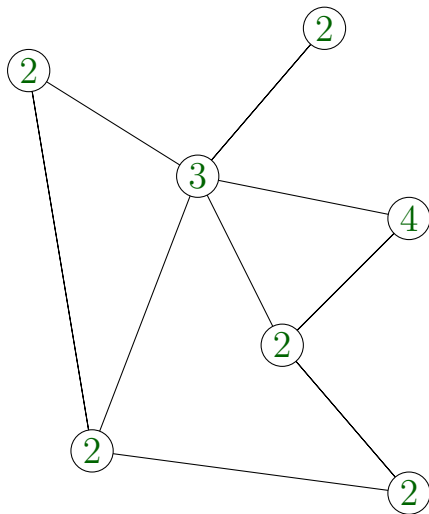
How to get a proper orientation?

- ▶ solve LP-relaxation

$$x^* = \arg \min_{x \in P_{BDST}(G,d)} c^\top x$$

- ▶ restrict to $E^* := \text{supp } x^*$
- ▶ sparsity:

$$|E^*[U]| \leq 2|U| - 3 \quad \forall U \subseteq V$$



Goemans' Approximation Approach

How to get a proper orientation?

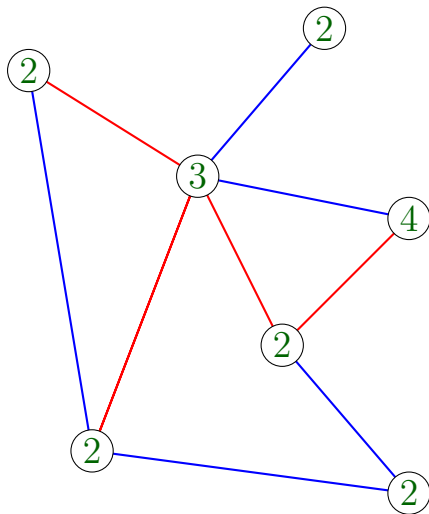
- ▶ solve LP-relaxation

$$x^* = \arg \min_{x \in P_{BDST}(G,d)} c^\top x$$

- ▶ restrict to $E^* := \text{supp } x^*$
- ▶ sparsity:

$$|E^*[U]| \leq 2|U| - 3 \quad \forall U \subseteq V$$

- ▶ partition into two forests



Goemans' Approximation Approach

How to get a proper orientation?

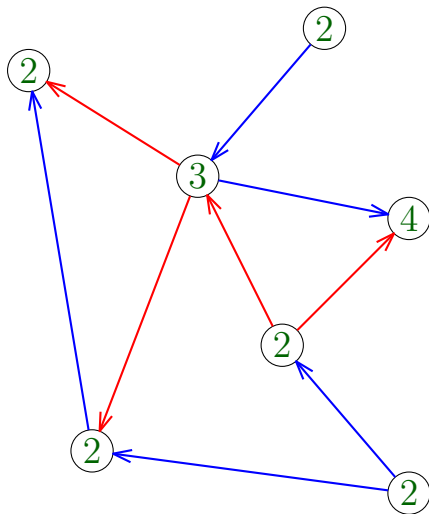
- ▶ solve LP-relaxation

$$x^* = \arg \min_{x \in P_{BDST}(G,d)} c^\top x$$

- ▶ restrict to $E^* := \text{supp } x^*$
- ▶ sparsity:

$$|E^*[U]| \leq 2|U| - 3 \quad \forall U \subseteq V$$

- ▶ partition into two forests
- ▶ orient towards leaves



Improving the Algorithm

Conjecture

For any extreme point x^ of $P_{BDST}(G, d)$ with support E^* , there exists an orientation A^* of E^* such that*

$$\sum_{e \in \delta_{A^*}^-(v)} (1 - x^*(e)) \leq 1 \quad \forall v \in V.$$

Improving the Algorithm

Conjecture

For any extreme point x^ of $P_{BDST}(G, d)$ with support E^* , there exists an orientation A^* of E^* such that*

$$\sum_{e \in \delta_{A^*}^-(v)} (1 - x^*(e)) \leq 1 \quad \forall v \in V.$$

- ▶ if true: get degree bound violation +1 using M_2 with

$$\mathcal{I}_2 = \{F \subseteq E^* \mid |F \cap \delta^+(v)| \leq \lceil x^*(\delta^+(v)) \rceil \quad \forall v \in V\}$$

Improving the Algorithm

Conjecture

For any extreme point x^ of $P_{BDST}(G, d)$ with support E^* , there exists an orientation A^* of E^* such that*

$$\sum_{e \in \delta_{A^*}^-(v)} (1 - x^*(e)) \leq 1 \quad \forall v \in V.$$

- ▶ if true: get degree bound violation +1 using M_2 with

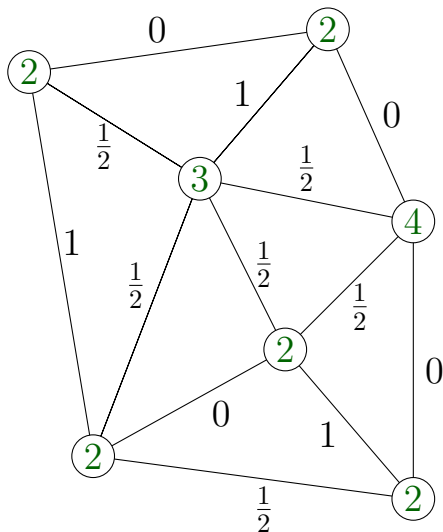
$$\mathcal{I}_2 = \{F \subseteq E^* \mid |F \cap \delta^+(v)| \leq \lceil x^*(\delta^+(v)) \rceil \quad \forall v \in V\}$$

- ▶ interpretation: orientation of spare $z^* := 1 - x^*$

Spare Orientation: An Example

- ▶ extreme point x^*
- ▶ restriction to E^*
- ▶ spare $z^* = 1 - x^*$
- ▶ orientation A^* such that

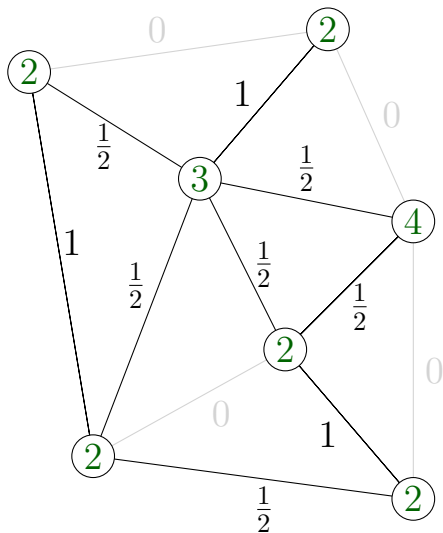
$$\sum_{e \in \delta_{A^*}^-(v)} z^*(e) \leq 1$$



Spare Orientation: An Example

- ▶ extreme point x^*
- ▶ restriction to E^*
- ▶ spare $z^* = 1 - x^*$
- ▶ orientation A^* such that

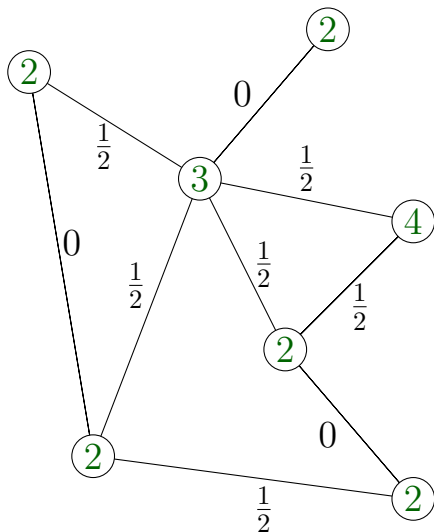
$$\sum_{e \in \delta_{A^*}^-(v)} z^*(e) \leq 1$$



Spare Orientation: An Example

- ▶ extreme point x^*
- ▶ restriction to E^*
- ▶ spare $z^* = 1 - x^*$
- ▶ orientation A^* such that

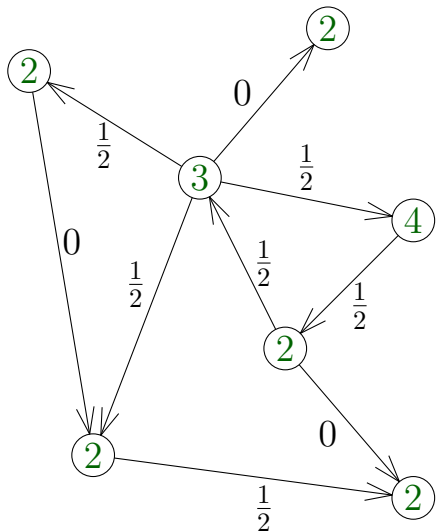
$$\sum_{e \in \delta_{A^*}^-(v)} z^*(e) \leq 1$$



Spare Orientation: An Example

- ▶ extreme point x^*
- ▶ restriction to E^*
- ▶ spare $z^* = 1 - x^*$
- ▶ orientation A^* such that

$$\sum_{e \in \delta_{A^*}^-(v)} z^*(e) \leq 1$$



Spare Orientation: How can it fail?

Spare Orientation: How can it fail?

- ▶ If there exists $U \subseteq V$ with

$$z^*(E^*[U]) > |U|, \text{ or equivalently, } x^*(E^*[U]) < |E^*[U]| - |U|,$$

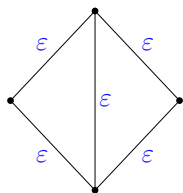
then it fails.

Spare Orientation: How can it fail?

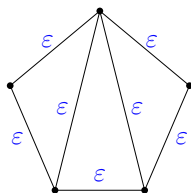
- ▶ If there exists $U \subseteq V$ with

$$z^*(E^*[U]) > |U|, \text{ or equivalently, } x^*(E^*[U]) < |E^*[U]| - |U|,$$

then it fails.



$$\epsilon < \frac{1}{5}$$



$$\epsilon < \frac{2}{7}$$

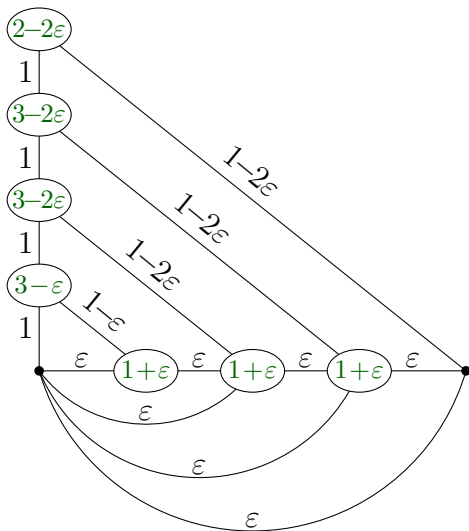
Constructing a Counterexample

Constructing a Counterexample

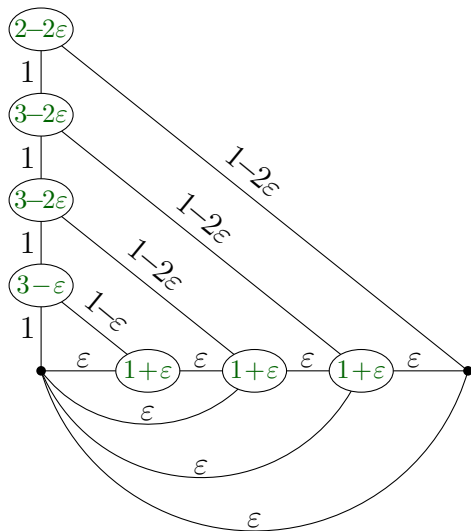
- ▶ non-integral degree bounds

Constructing a Counterexample

- ▶ non-integral degree bounds

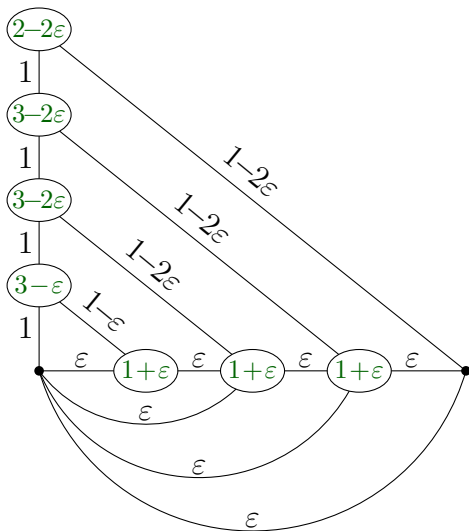


Checking Feasibility in $P_{BDST}(G, d)$



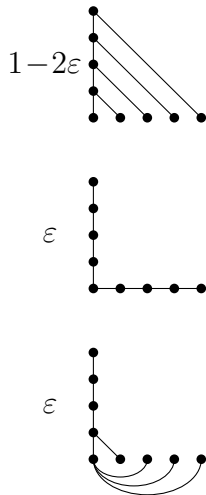
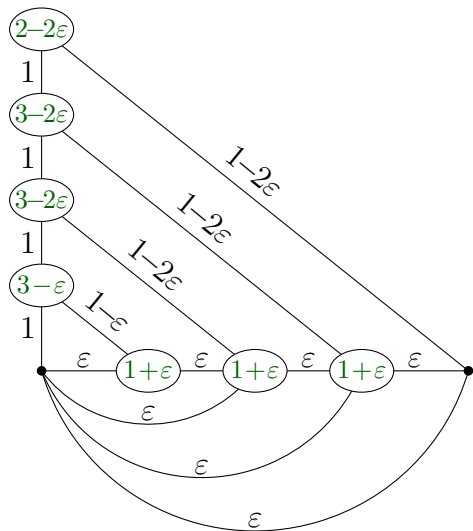
Checking Feasibility in $P_{BDST}(G, d)$

- decomposition as convex combination of spanning trees

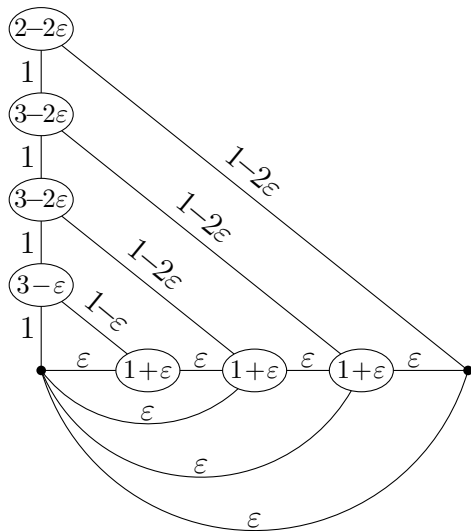


Checking Feasibility in $P_{BDST}(G, d)$

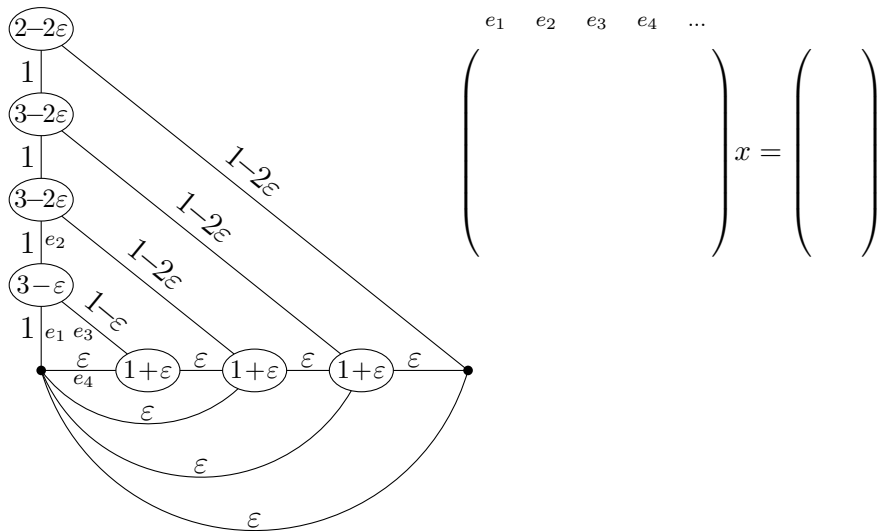
- decomposition as convex combination of spanning trees



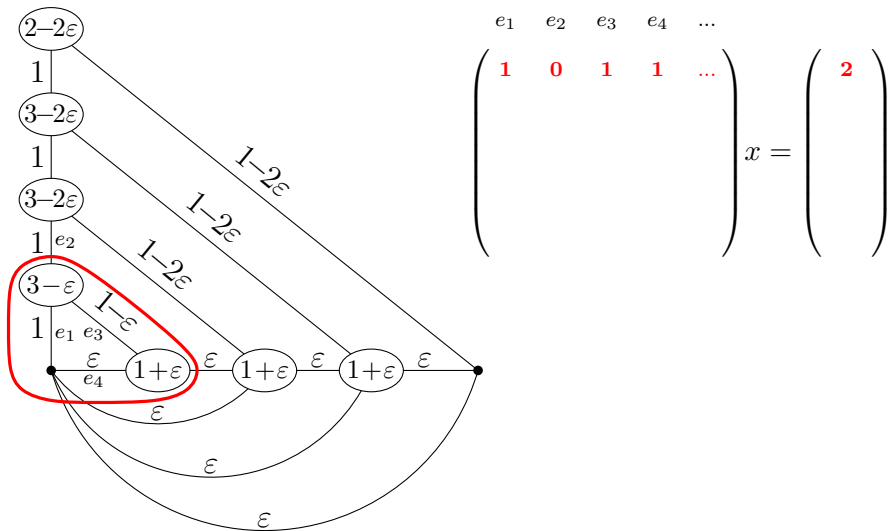
Tight Constraints form a Full-Rank System



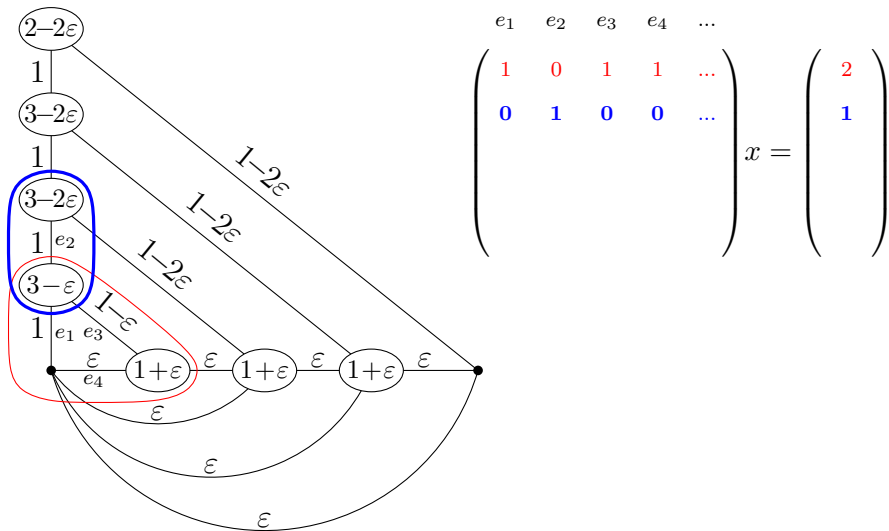
Tight Constraints form a Full-Rank System



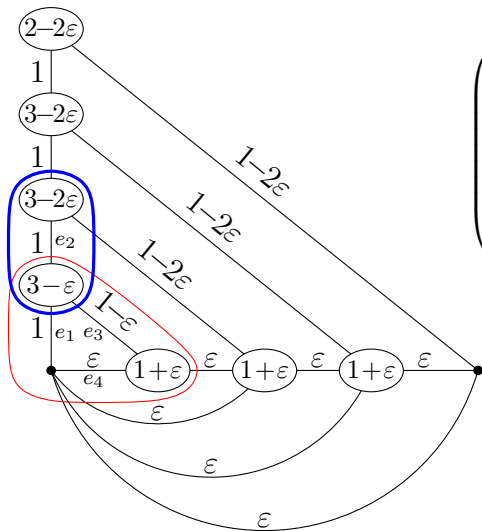
Tight Constraints form a Full-Rank System



Tight Constraints form a Full-Rank System

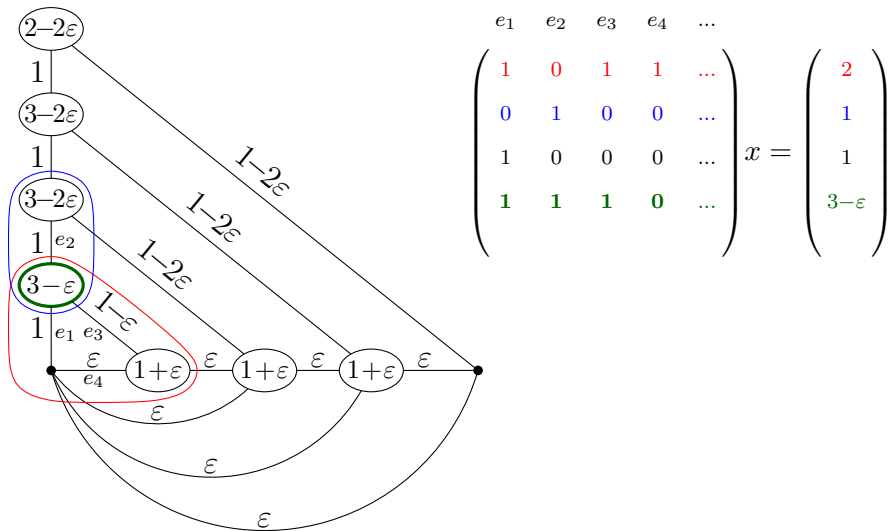


Tight Constraints form a Full-Rank System

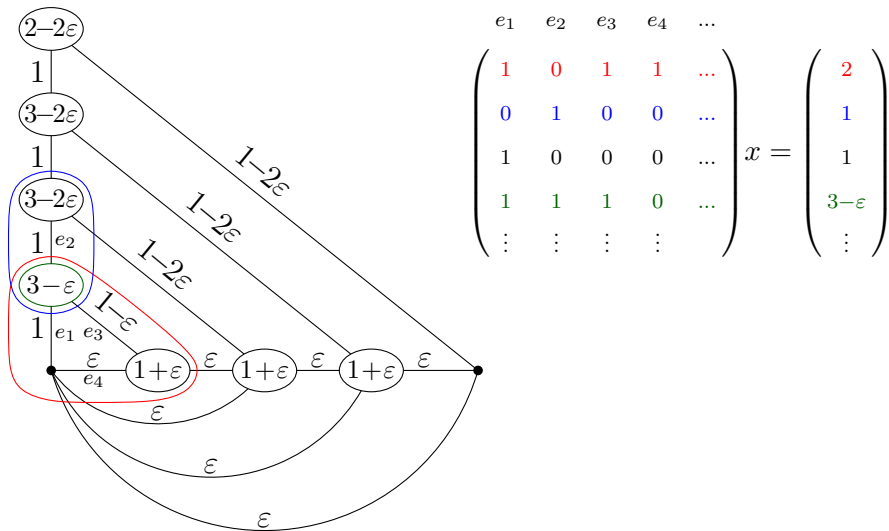


$$\begin{matrix}
 & e_1 & e_2 & e_3 & e_4 & \dots \\
 \begin{pmatrix}
 1 & 0 & 1 & 1 & \dots \\
 0 & 1 & 0 & 0 & \dots \\
 1 & 0 & 0 & 0 & \dots
 \end{pmatrix}
 & x = &
 \begin{pmatrix}
 2 \\
 1 \\
 1
 \end{pmatrix}
 \end{matrix}$$

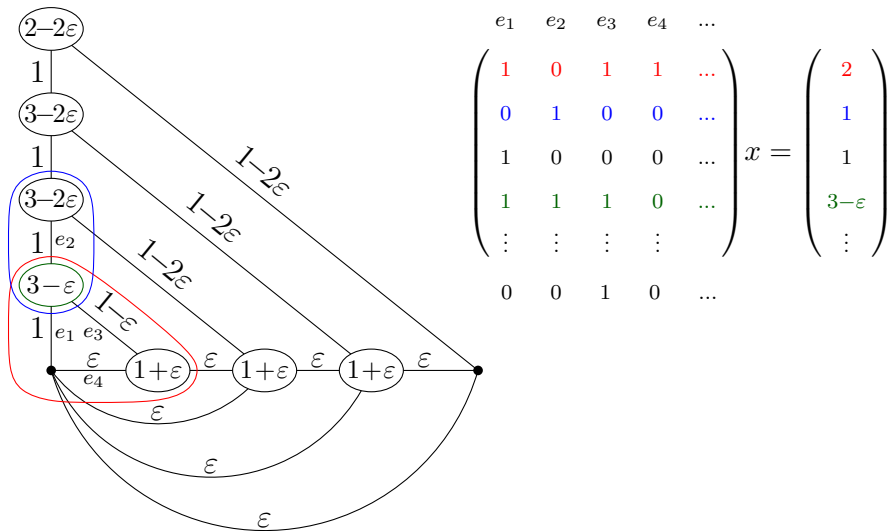
Tight Constraints form a Full-Rank System



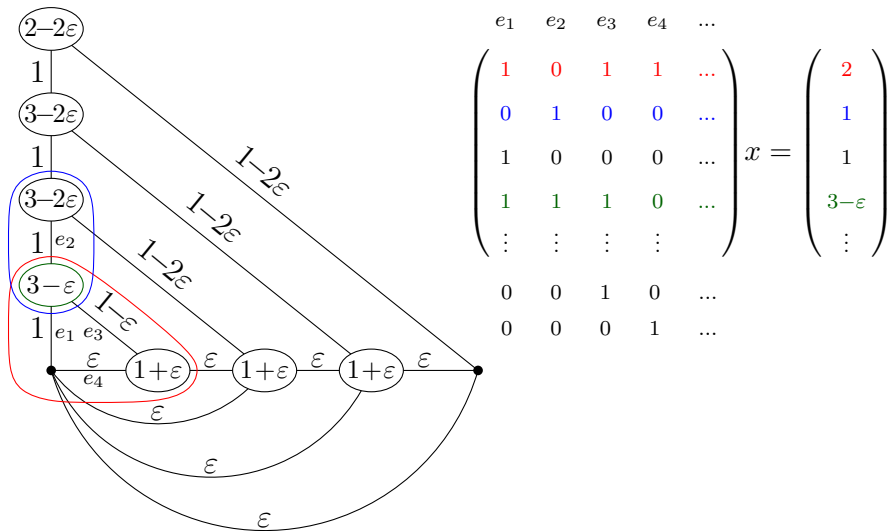
Tight Constraints form a Full-Rank System



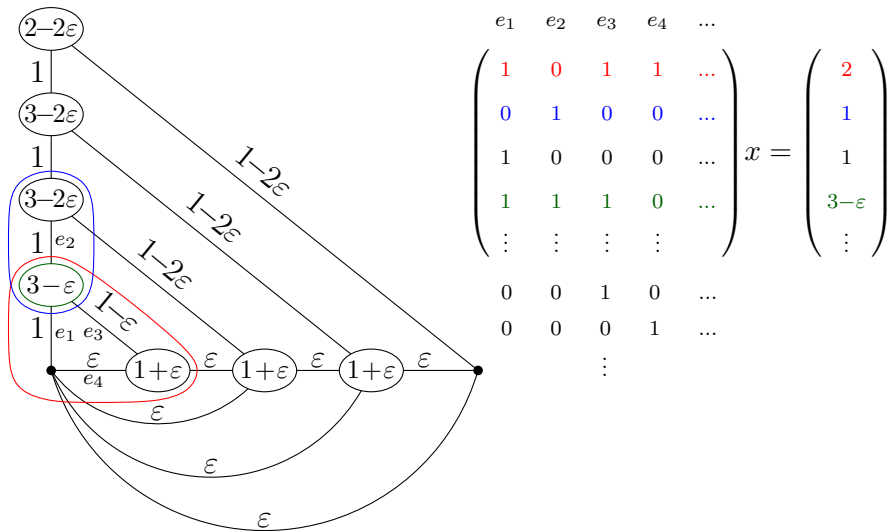
Tight Constraints form a Full-Rank System



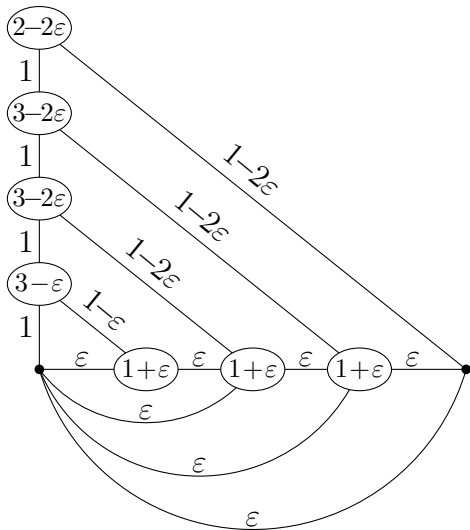
Tight Constraints form a Full-Rank System



Tight Constraints form a Full-Rank System



How to Impose Fractional Degree Bounds?



Feasible Degree Bound Slacks

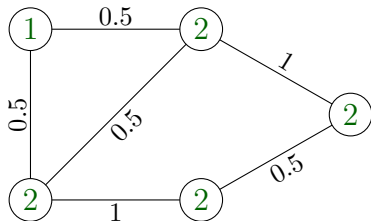
Definition

A number $r \in \mathbb{Q}_{\geq 0}$ is a *feasible degree bound slack* if there exists an integral instance (G, d) , an extreme point $x \in P_{BDST}(G, d)$ and a vertex v such that $d(v) - x(\delta(v)) = r$.

Feasible Degree Bound Slacks

Definition

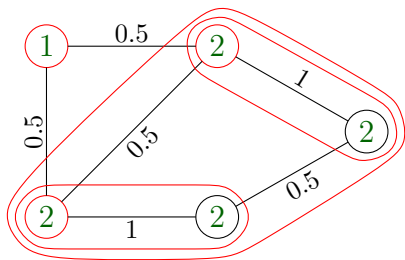
A number $r \in \mathbb{Q}_{\geq 0}$ is a *feasible degree bound slack* if there exists an integral instance (G, d) , an extreme point $x \in P_{BDST}(G, d)$ and a vertex v such that $d(v) - x(\delta(v)) = r$.



Feasible Degree Bound Slacks

Definition

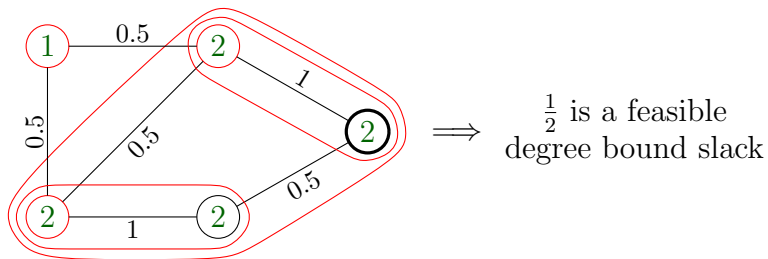
A number $r \in \mathbb{Q}_{\geq 0}$ is a *feasible degree bound slack* if there exists an integral instance (G, d) , an extreme point $x \in P_{BDST}(G, d)$ and a vertex v such that $d(v) - x(\delta(v)) = r$.



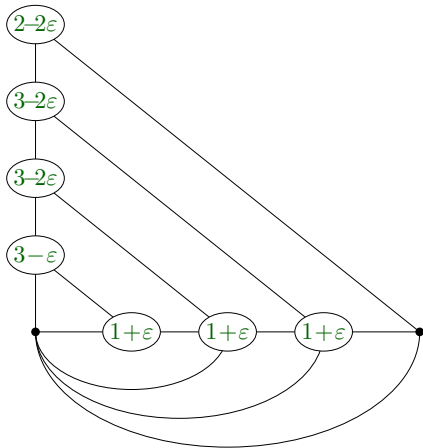
Feasible Degree Bound Slacks

Definition

A number $r \in \mathbb{Q}_{\geq 0}$ is a *feasible degree bound slack* if there exists an integral instance (G, d) , an extreme point $x \in P_{BDST}(G, d)$ and a vertex v such that $d(v) - x(\delta(v)) = r$.

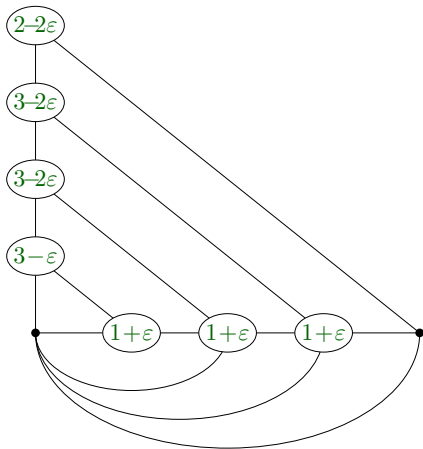


Modelling Fractional Degree Bounds

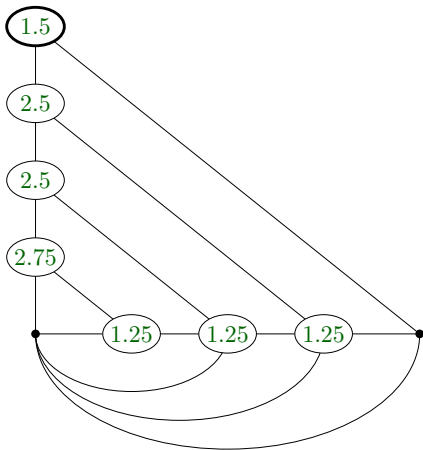


Modelling Fractional Degree Bounds

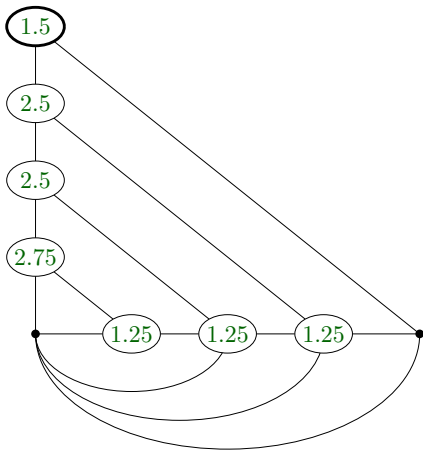
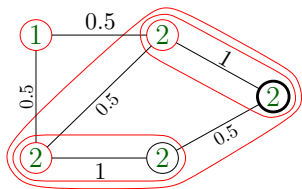
$$\varepsilon = \frac{1}{4}$$



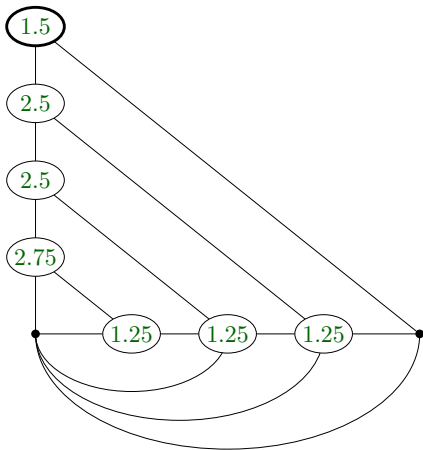
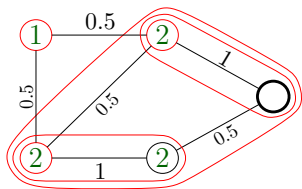
Modelling Fractional Degree Bounds



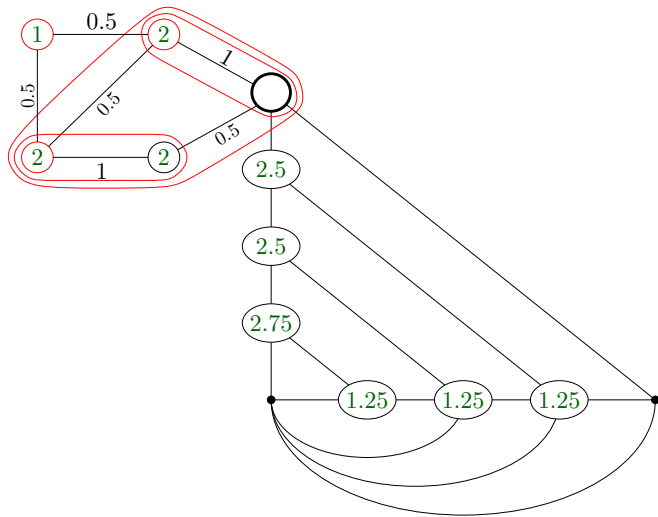
Modelling Fractional Degree Bounds



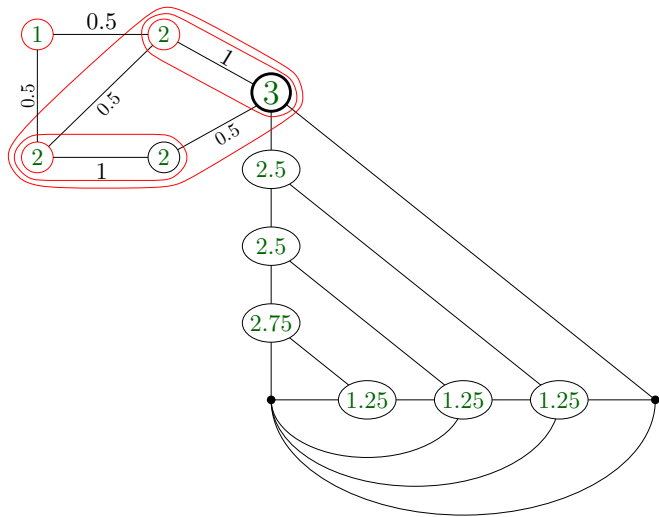
Modelling Fractional Degree Bounds



Modelling Fractional Degree Bounds



Modelling Fractional Degree Bounds

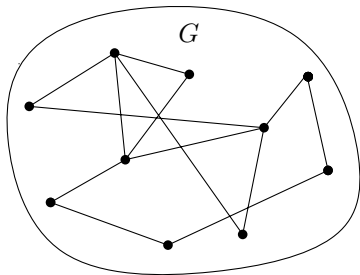


Modelling Fractional Degree Bounds

- ▶ works if $d(v) - \lfloor d(v) \rfloor$ is a feasible degree bound slack

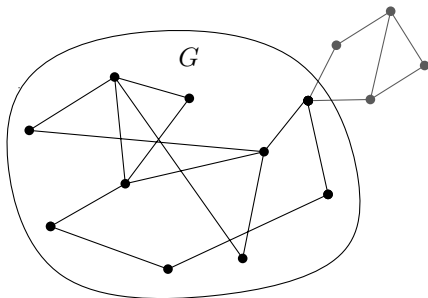
Modelling Fractional Degree Bounds

- ▶ works if $d(v) - \lfloor d(v) \rfloor$ is a feasible degree bound slack



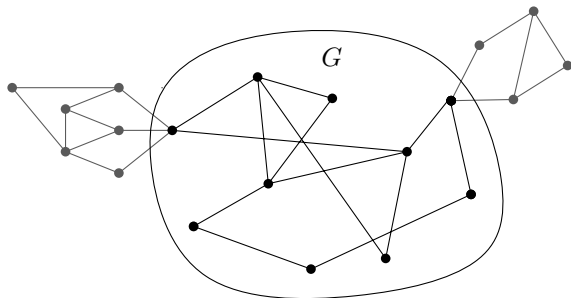
Modelling Fractional Degree Bounds

- ▶ works if $d(v) - \lfloor d(v) \rfloor$ is a feasible degree bound slack



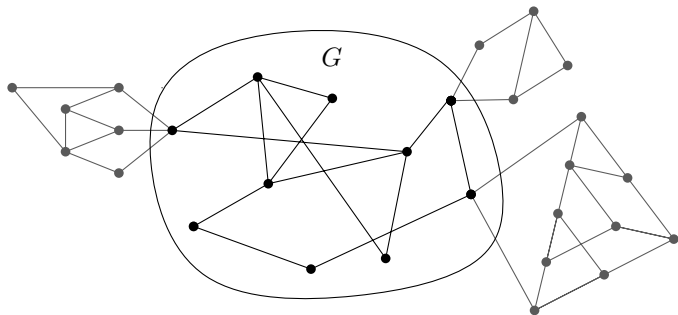
Modelling Fractional Degree Bounds

- ▶ works if $d(v) - \lfloor d(v) \rfloor$ is a feasible degree bound slack



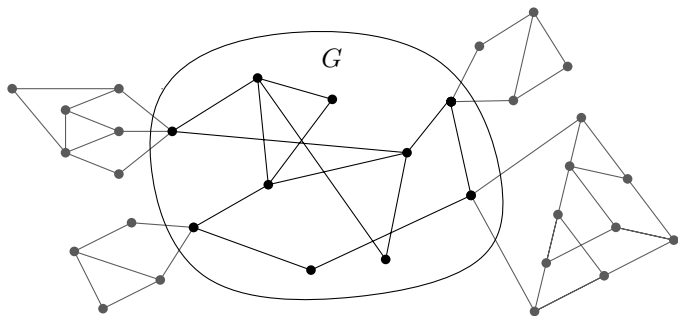
Modelling Fractional Degree Bounds

- ▶ works if $d(v) - \lfloor d(v) \rfloor$ is a feasible degree bound slack



Modelling Fractional Degree Bounds

- ▶ works if $d(v) - \lfloor d(v) \rfloor$ is a feasible degree bound slack



Feasible Degree Bound Slacks

Theorem

Every $r \in \mathbb{Q}_{\geq 0}$ is a feasible degree bound slack.

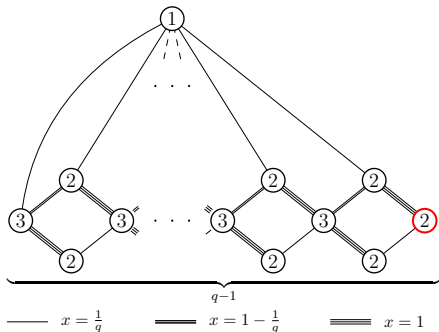
Feasible Degree Bound Slacks

Theorem

Every $r \in \mathbb{Q}_{\geq 0}$ is a feasible degree bound slack.

Lemma 1

For $q \geq 2$, $\frac{q-1}{q}$ is feasible.



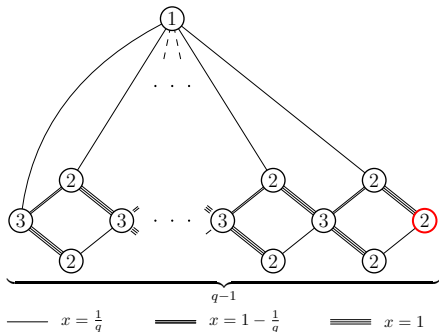
Feasible Degree Bound Slacks

Theorem

Every $r \in \mathbb{Q}_{\geq 0}$ is a feasible degree bound slack.

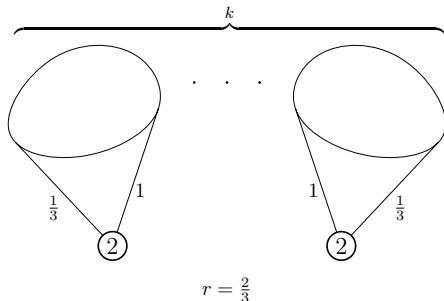
Lemma 1

For $q \geq 2$, $\frac{q-1}{q}$ is feasible.



Lemma 2

If r is feasible, then so is kr for all $k \in \mathbb{Z}$.



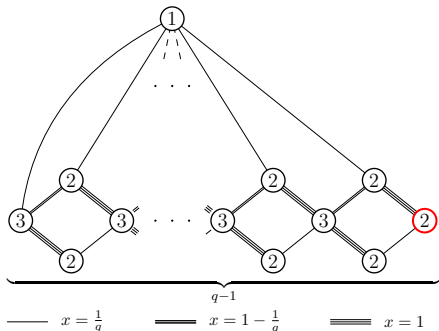
Feasible Degree Bound Slacks

Theorem

Every $r \in \mathbb{Q}_{\geq 0}$ is a feasible degree bound slack.

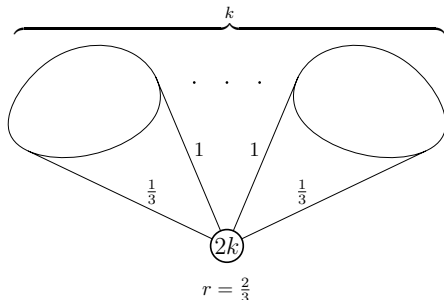
Lemma 1

For $q \geq 2$, $\frac{q-1}{q}$ is feasible.

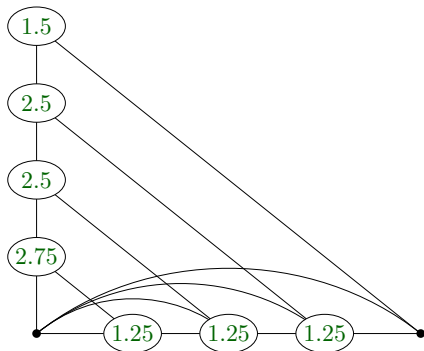


Lemma 2

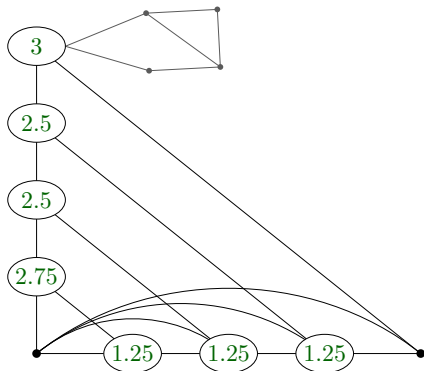
If r is feasible, then so is kr for all $k \in \mathbb{Z}$.



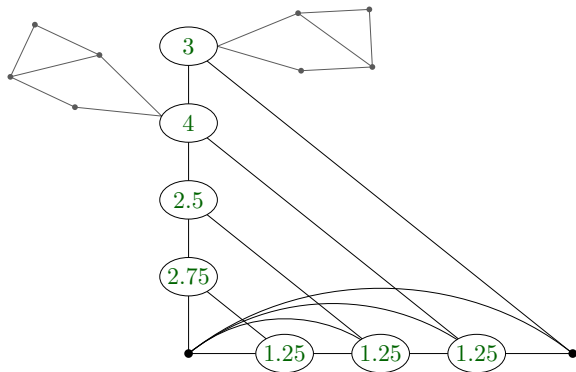
Conclusion



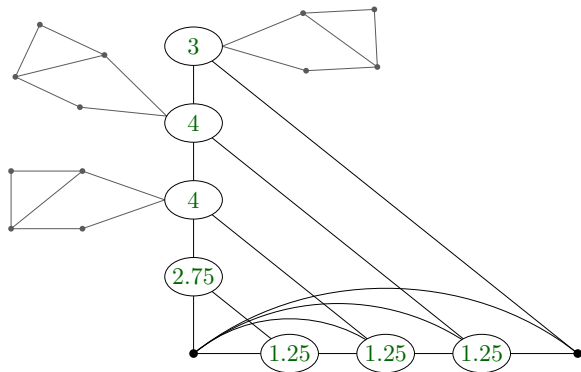
Conclusion



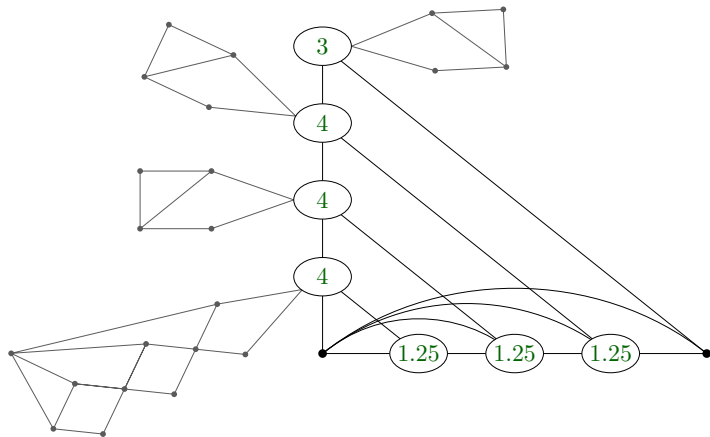
Conclusion



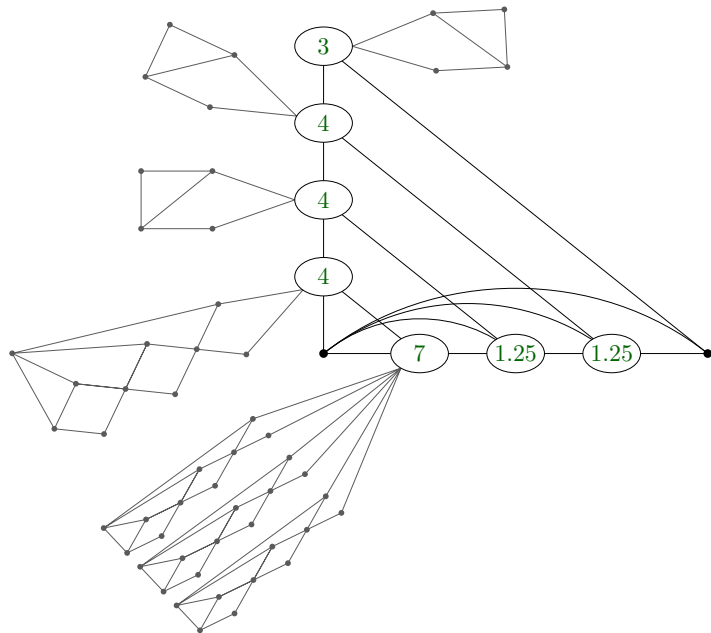
Conclusion



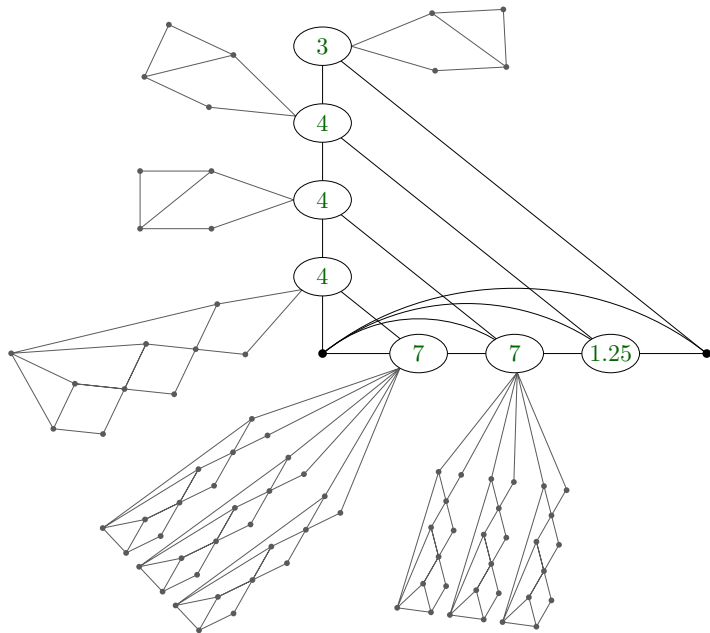
Conclusion



Conclusion



Conclusion



Conclusion

