

Linear Programming

Outline

- Linear Programming
 - Integer
 - Vertex Cover
 - Fractional
- How are fractional solutions useful?
 - Rounding
 - Approximate Vertex Cover

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Learning objectives:

You are able to

- describe the concept of a linear program
- explain the difference between an integer and a fractional linear program
- illustrate how to model the vertex cover as a linear program
- apply rounding to turn a fractional vertex cover into an integer vertex cover

Linear Programs

- Recall the optimization formulation

Minimize
 x

Objective function $f(x)$

Subject to

Constraints $g_1(x)$
 $g_2(x)$
 ...
 $g_m(x)$

Linear Programs

$$\mathbf{x} = x_1, x_2, \dots, x_n$$

Minimize
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Objective function $f(x)$

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Linear Programs

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$$x = x_1, x_2, \dots, x_n$$

Linear:

Each g_i is a linear
inequality on x_1, x_2, \dots, x_n
 f is a linear function

Linear Programs – Geometric Interpretation

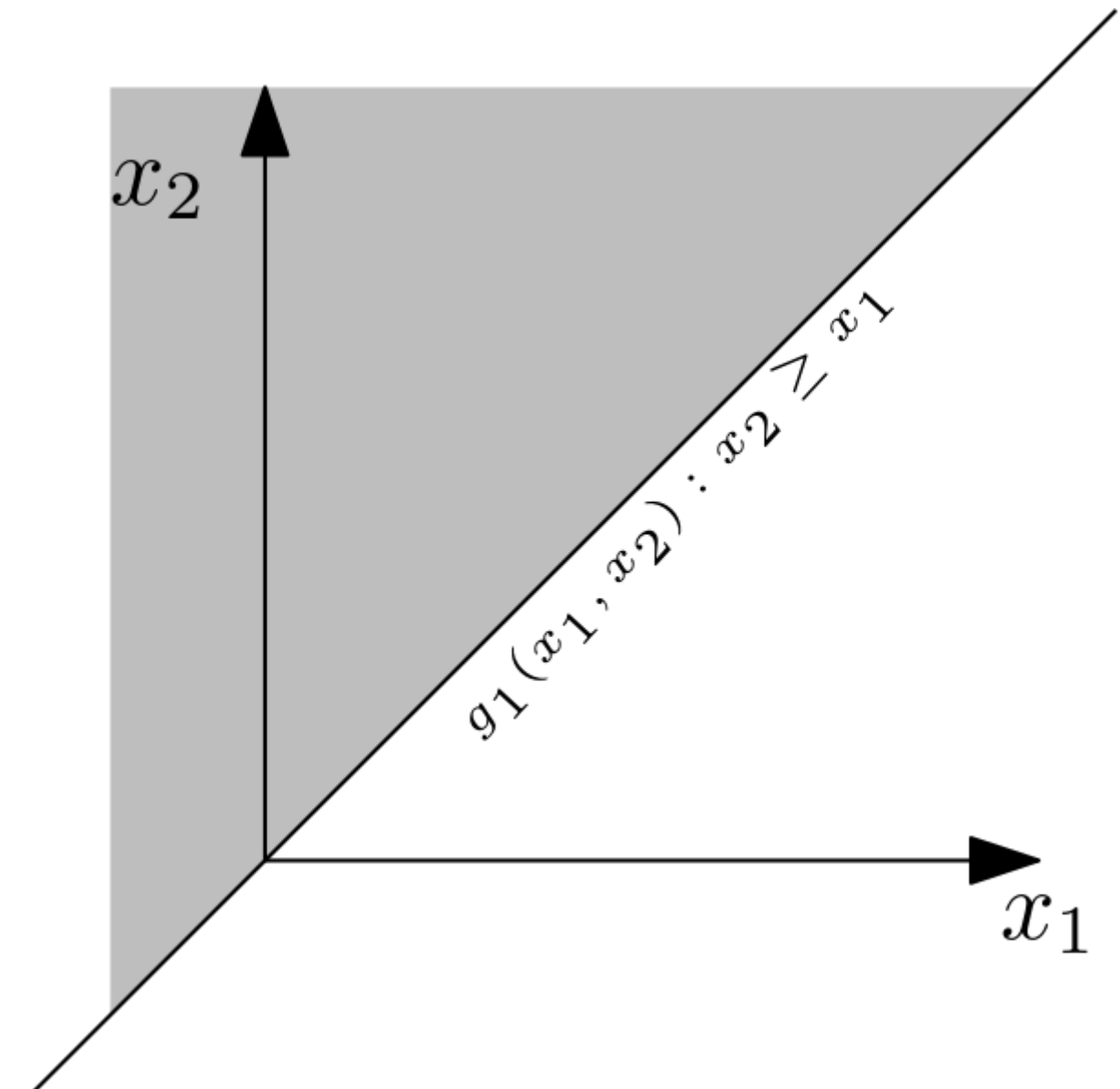
Maximize
 x

Objective function $f(x)$

Subject to

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 ...
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$x = x_1, x_2$



Linear Programs – Geometric Interpretation

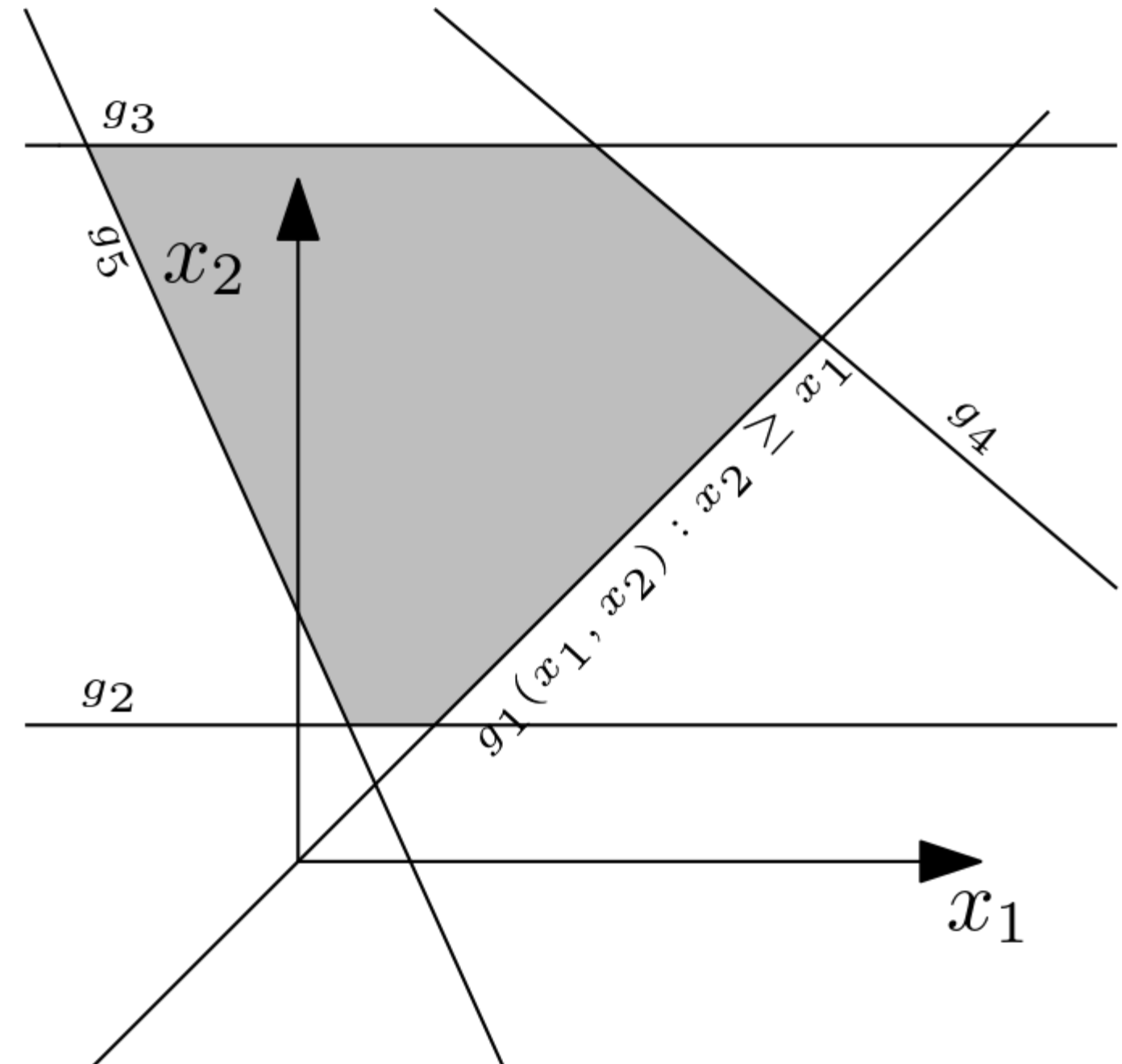
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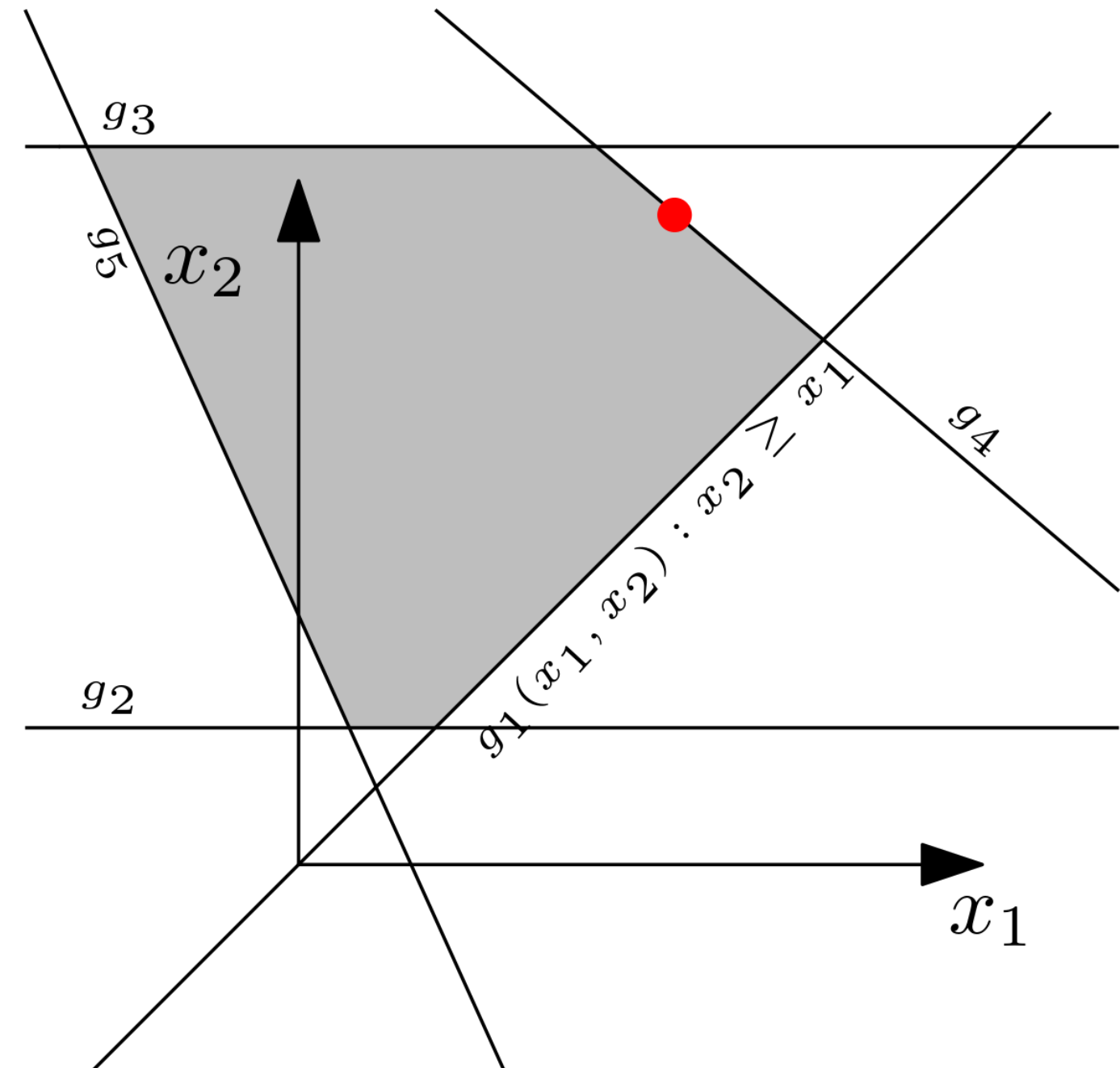
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Linear Programs – Geometric Interpretation

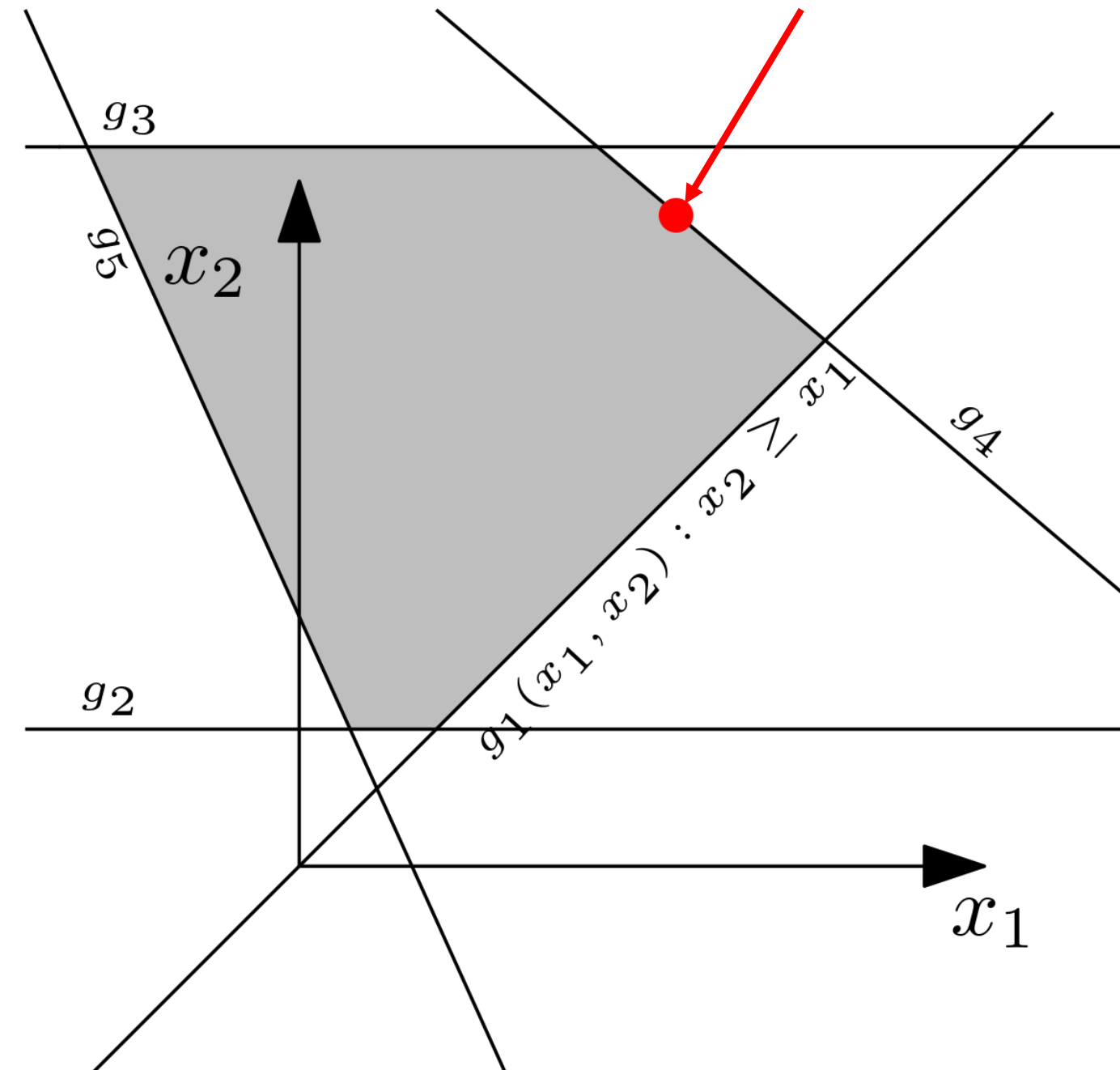
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Linear Programs – Geometric Interpretation

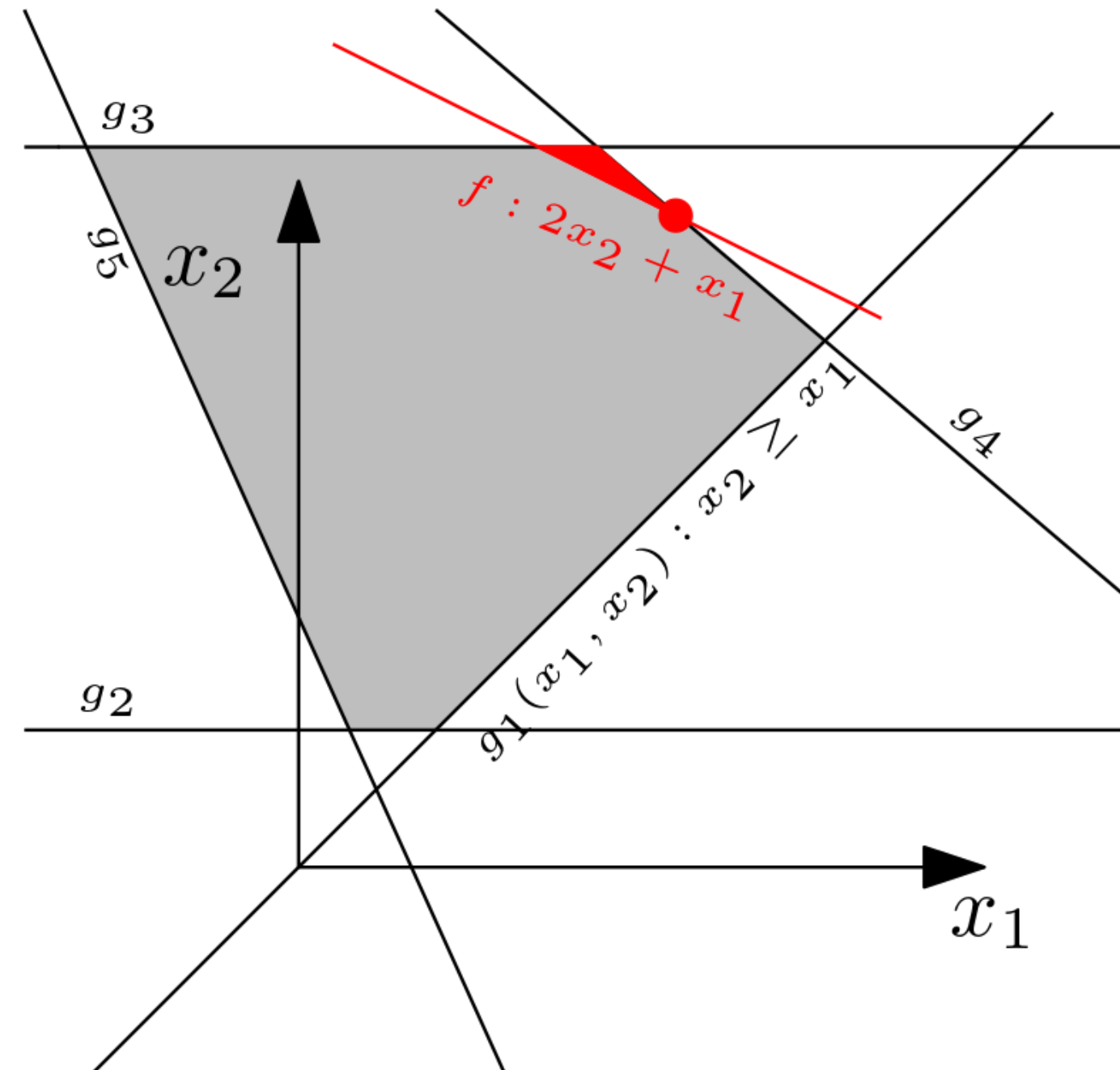
Maximize
 x

$$f(x) = 2x_2 + x_1$$

Subject to

Constraints $g_1(x)$
 $g_2(x)$
 ...
 $g_m(x)$

$$x = x_1, x_2$$



Linear Programs – Geometric Interpretation

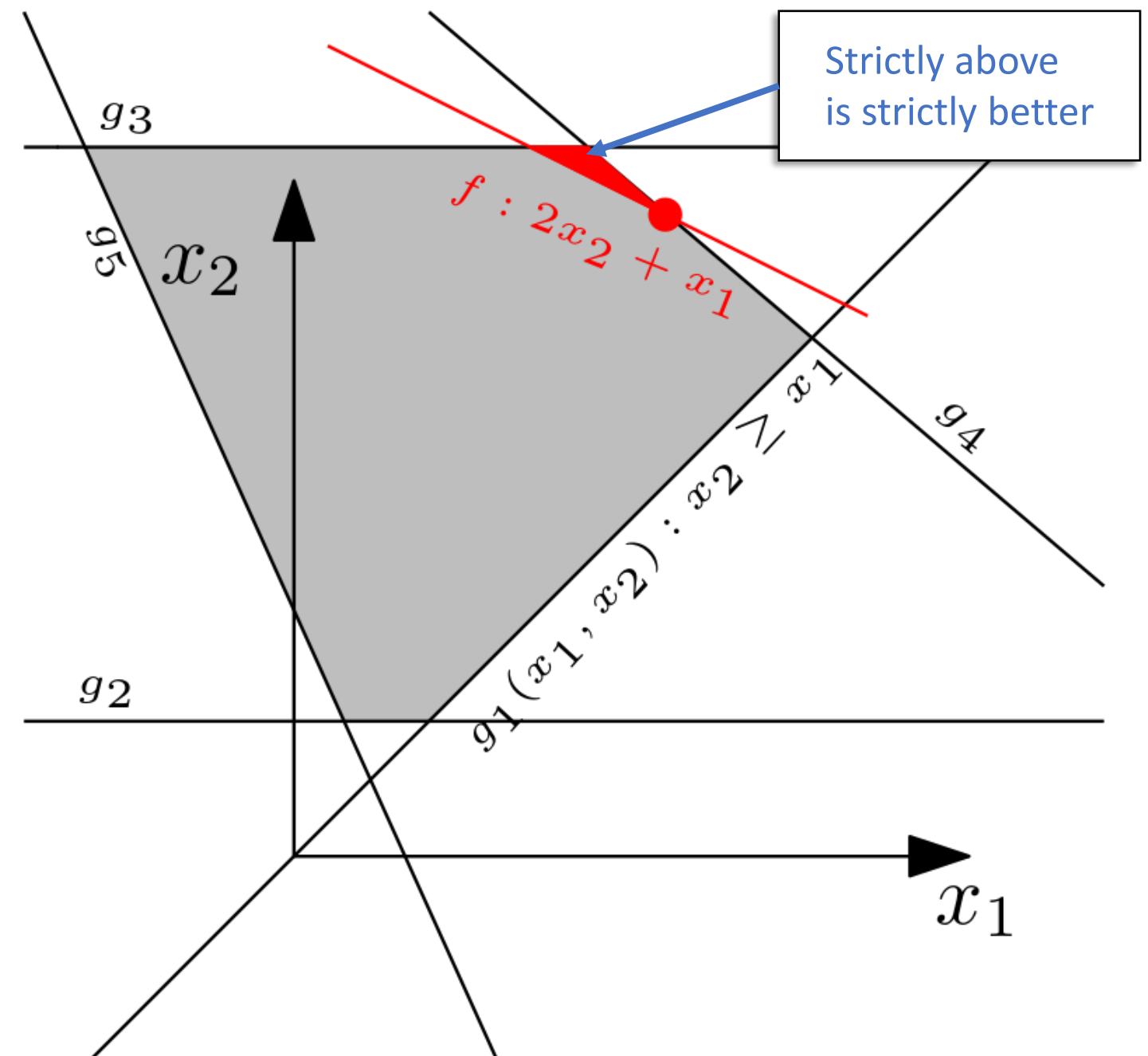
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Linear Programs – Geometric Interpretation

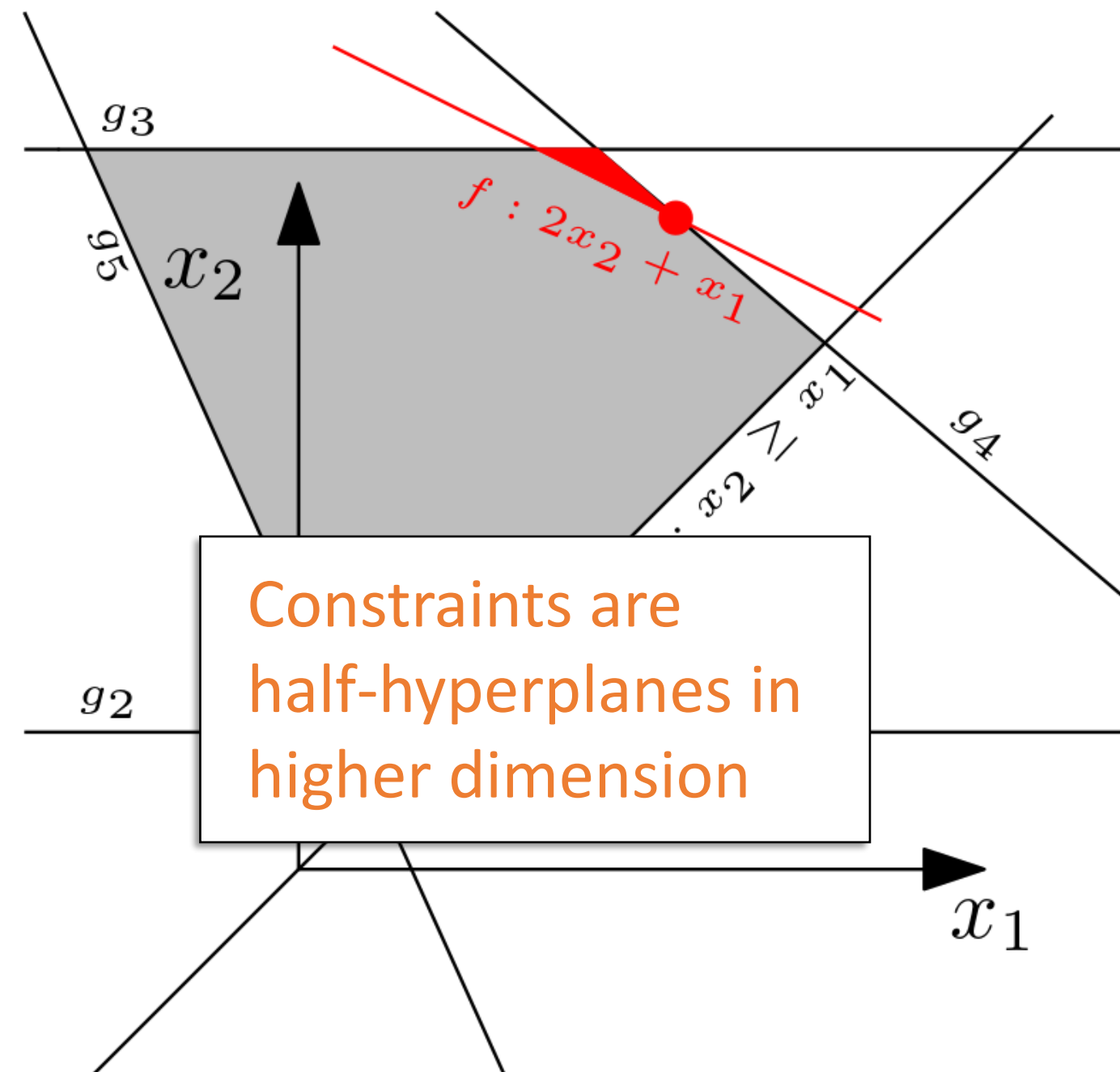
Maximize
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Subject to

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$x = x_1, x_2, \dots, x_n$



Linear Programs – Geometric Interpretation

Maximize
 x

Objective function $f(x)$

Subject to

Constraints $g_1(x)$
 $g_2(x)$
 ...
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$x = x_1, x_2, \dots, x_n$

Integer linear program:

$\forall i, x_i = \{0,1\}$

Outline

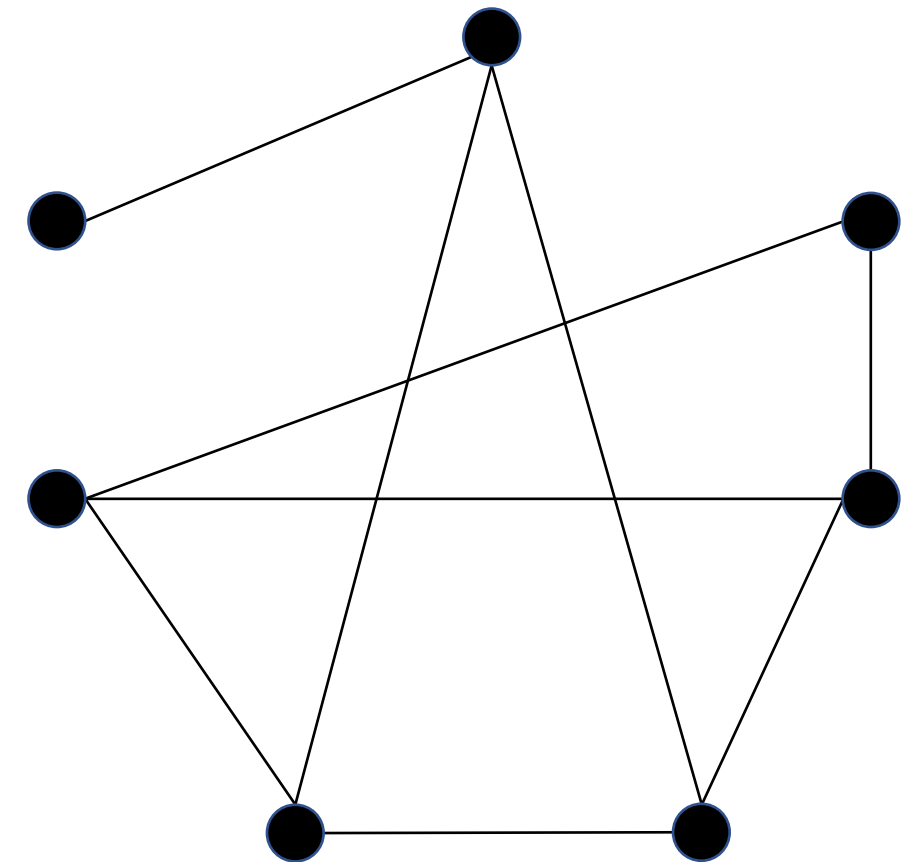
- Linear Programming
 - Integer
 - Vertex Cover
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- How are fractional solutions useful?
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Minimum Vertex Cover

An Informative Example Problem: Vertex Cover

Given a graph $G = (V, E)$

Find a set of nodes $C \subseteq V$ that
covers all edges.

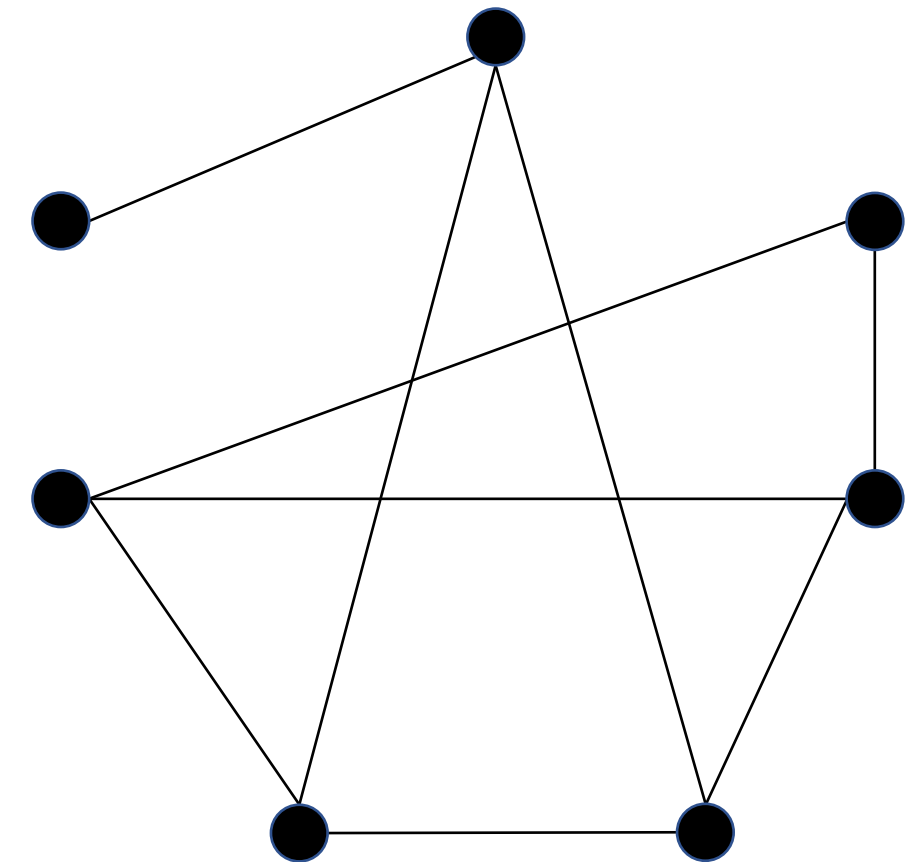


Minimum Vertex Cover

An Informative Example Problem: Vertex Cover

Given a graph $G = (V, E)$
Find a set of nodes $C \subseteq V$ that
covers all edges.

For each $\{u, v\} \in E$,
either $u \in C$ or $v \in C$

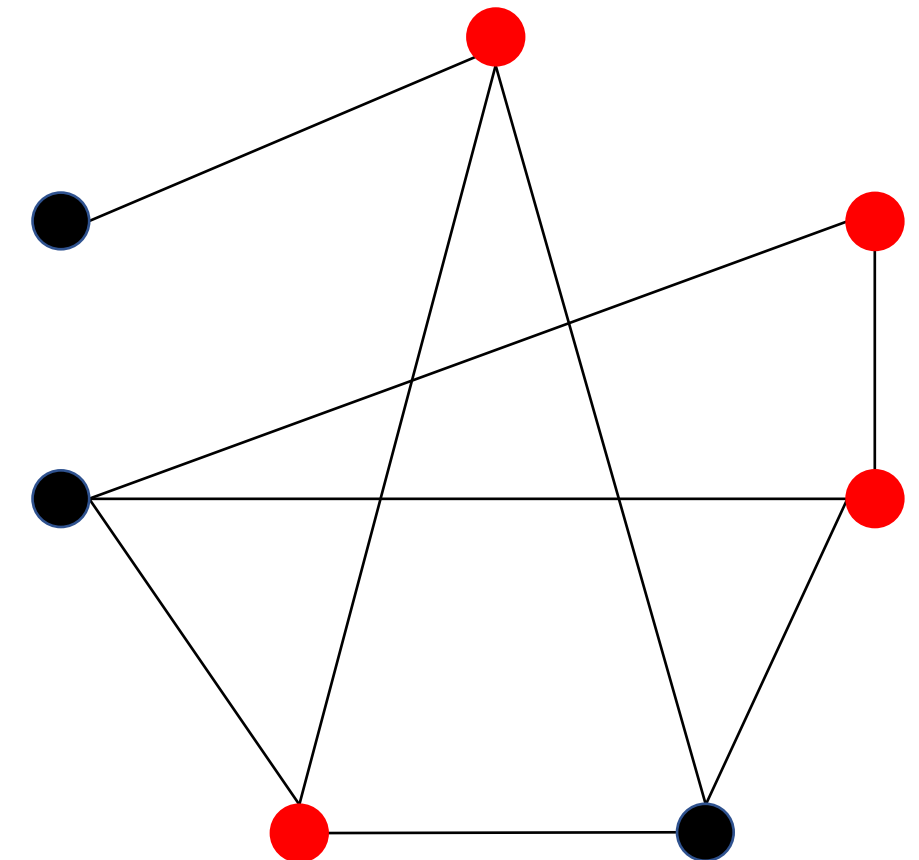


Minimum Vertex Cover

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Find a set of nodes $C \subseteq V$ that
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For each $\{u, v\} \in E$,
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Minimum Vertex Cover

Integer program:

$$x_i = \begin{cases} 1 & \text{if node } i \text{ in the cover} \\ 0 & \text{otherwise} \end{cases}$$

Nodes numbered
from 1 to n

Minimum Vertex Cover

Minimize
 x

$$\sum_{i=1}^n x_i$$

Integer program:

$$x_i = \begin{cases} 1 & \text{if node } i \text{ in the cover} \\ 0 & \text{otherwise} \end{cases}$$

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Minimum Vertex Cover

Minimize
 x

$$\sum_{i=1}^n x_i$$

Subject to

$$x_i + x_j \geq 1 \quad \text{for all } \{i, j\} \in E$$

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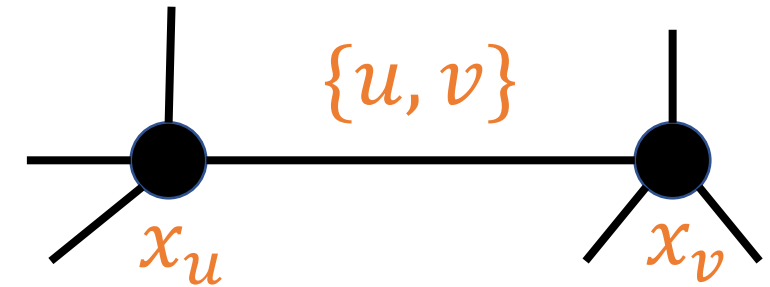
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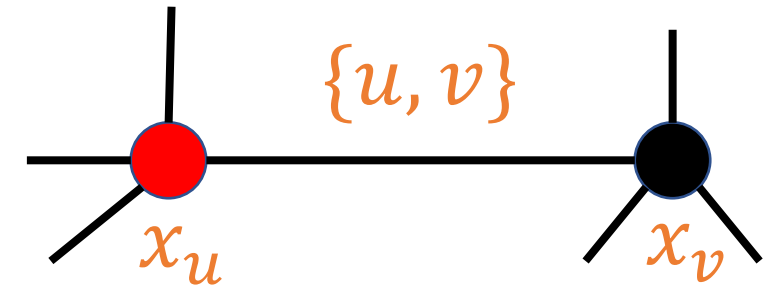
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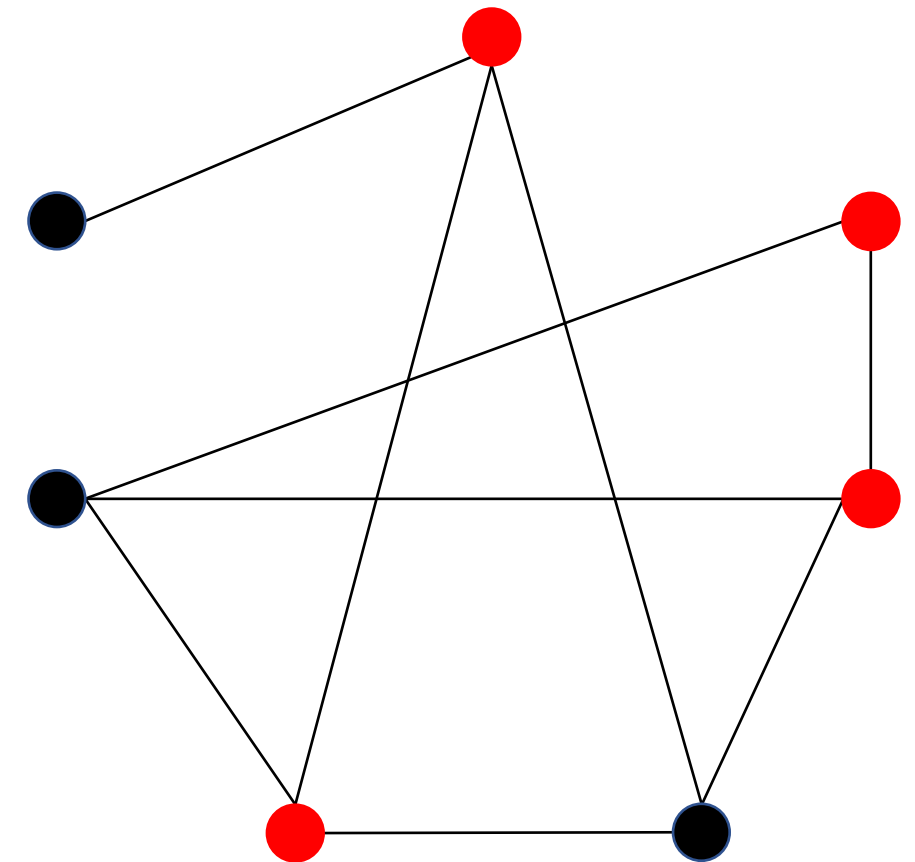
Minimum Vertex Cover

Minimize
 x

$$\sum_{i=1}^n x_i$$

Subject to

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \in \{0, 1\} & \text{for all } i \in V \end{array}$$



Minimum Vertex Cover

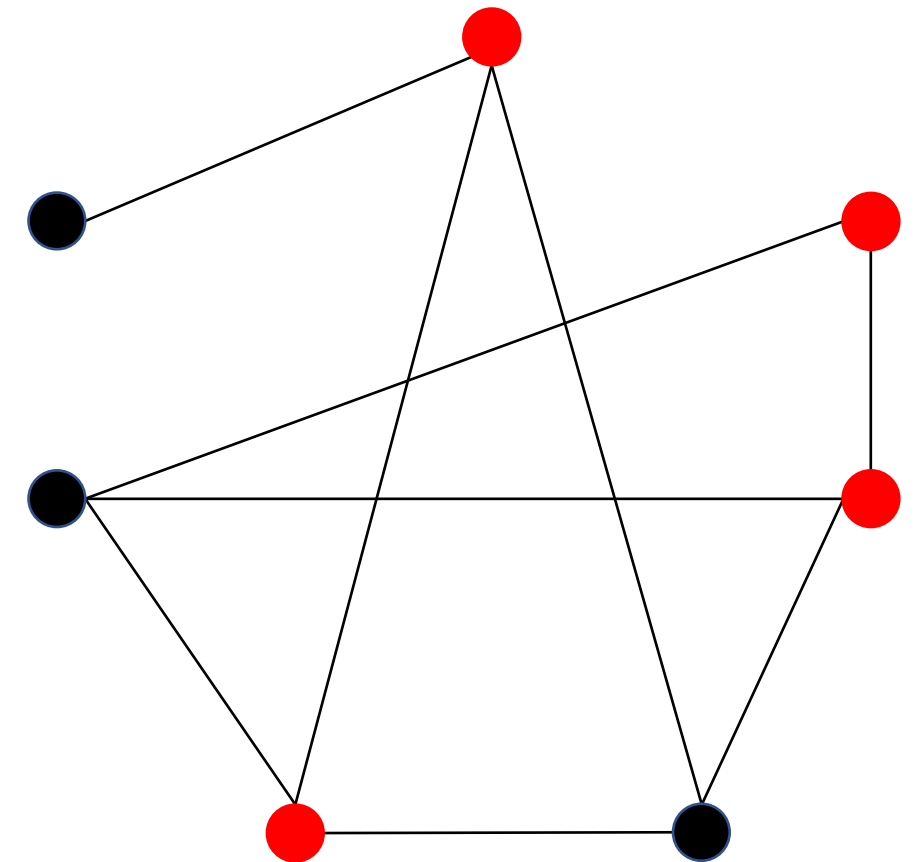
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Integer program IP



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Minimum Vertex Cover

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$$\sum_{i=1}^n x_i$$

Subject to

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \geq 0 & \text{for all } i \in V \end{array}$$

Relaxation:

Allow *fractional*
values

Integer program:

$$x_i = \begin{cases} 1 & \text{if node } i \text{ in the cover} \\ 0 & \text{otherwise} \end{cases}$$

Nodes numbered
from 1 to n

Minimum Vertex Cover

Minimize
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$$\sum_{i=1}^n x_i$$

Subject to

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \geq 0 & \text{for all } i \in V \end{array}$$



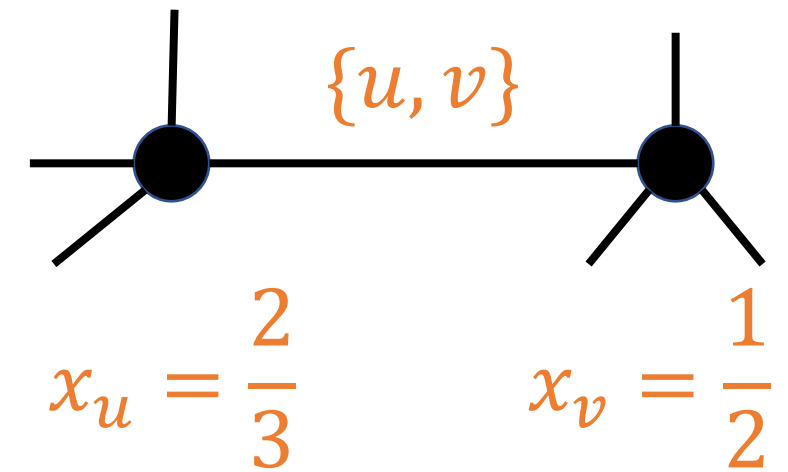
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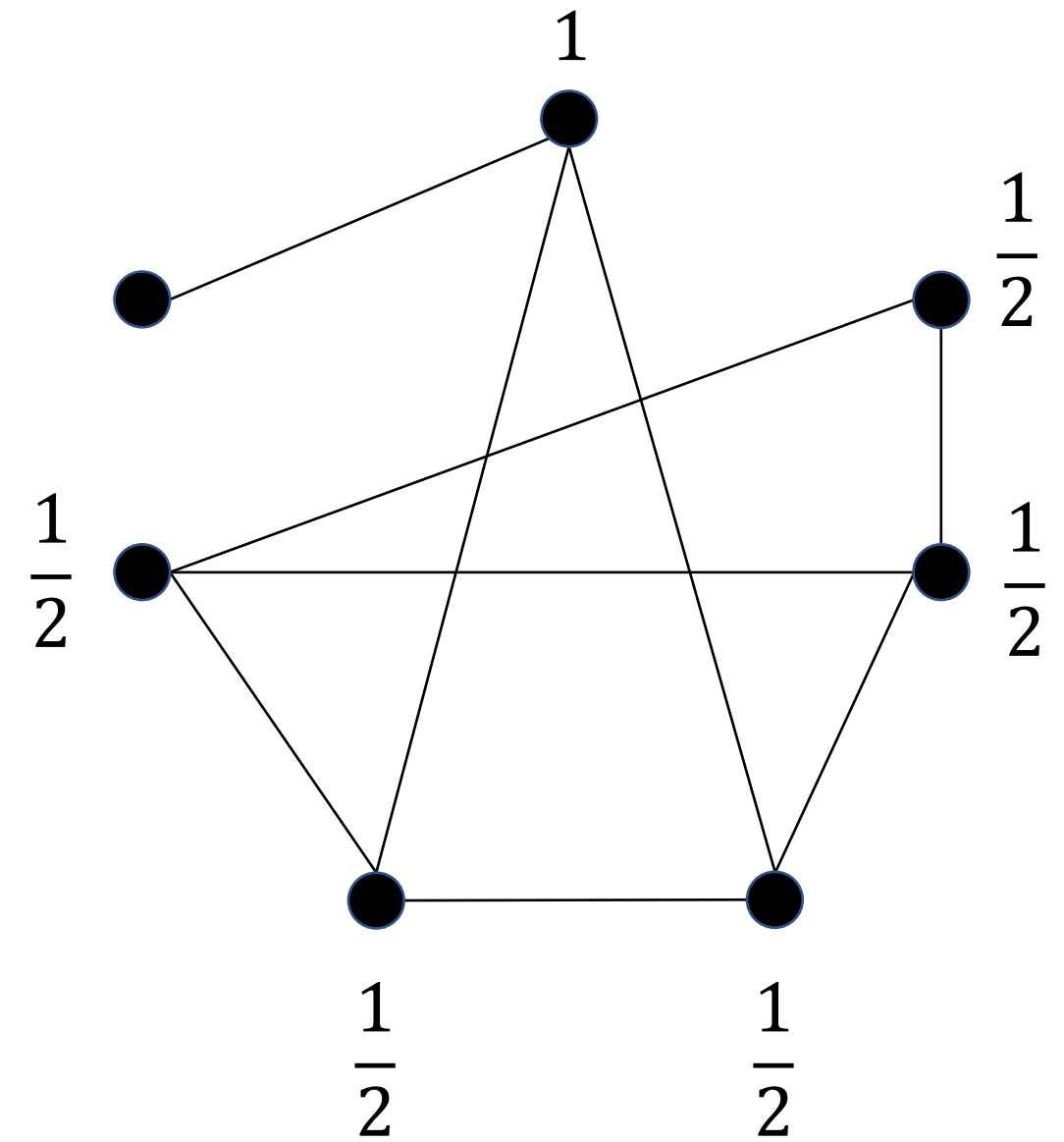
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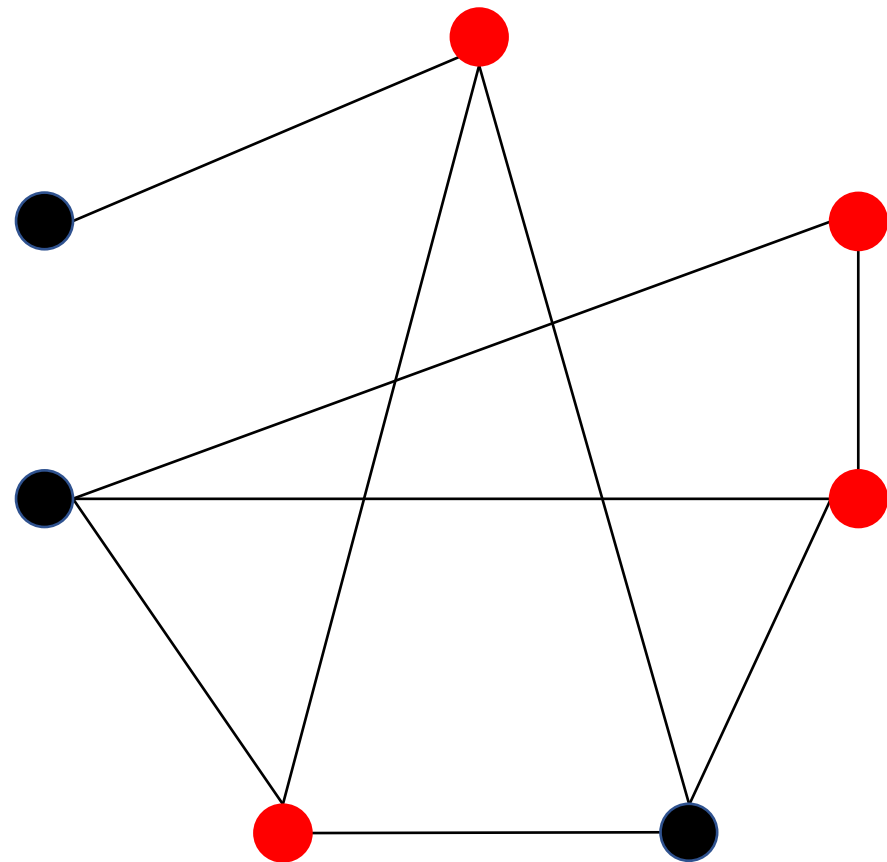


Minimum Vertex Cover

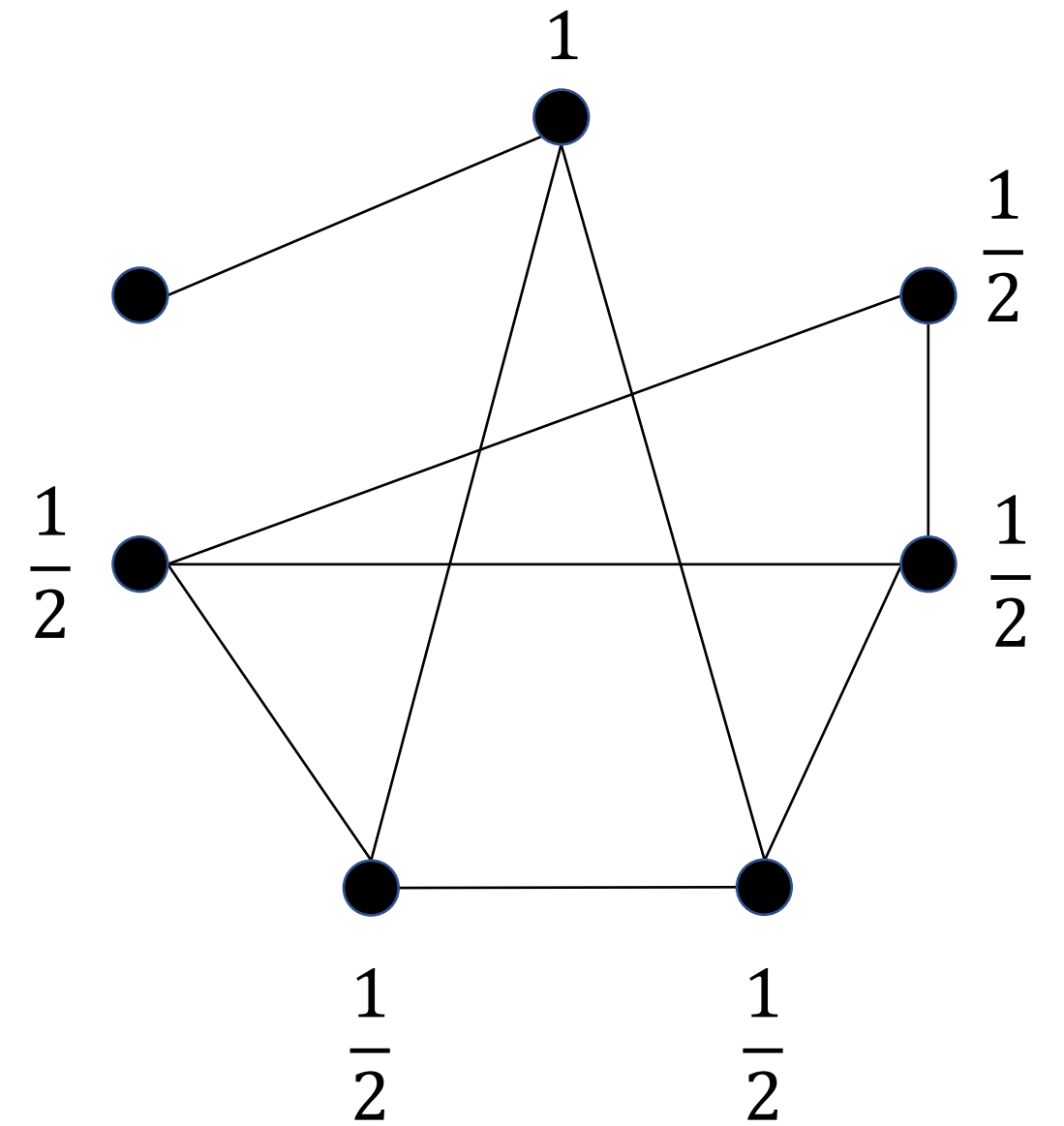
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Minimum Vertex Cover



$$f(x) = 4$$



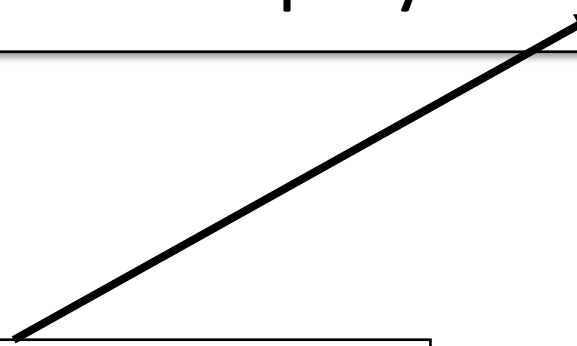
$$f(x) = 3 + \frac{1}{2} < 4$$

Solving Linear Programs

Theorem:

Linear program relaxations can be solved in polynomial time

Usually the definition of “efficient”

A black arrow points from the top-right corner of the bottom box to the bottom-left corner of the top box.

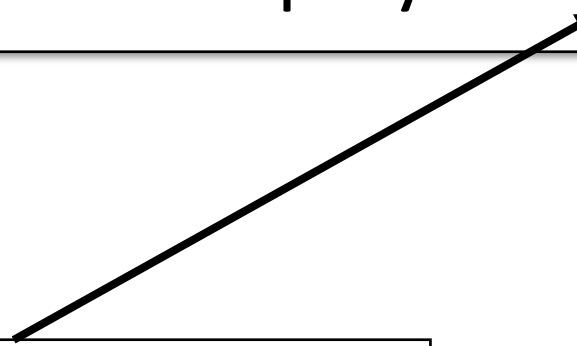
Solving Linear Programs

In this course, we use this theorem as a tool

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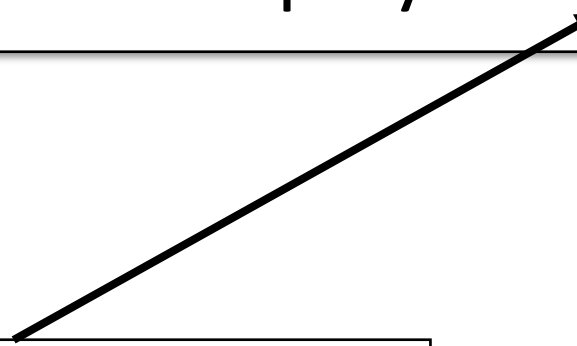


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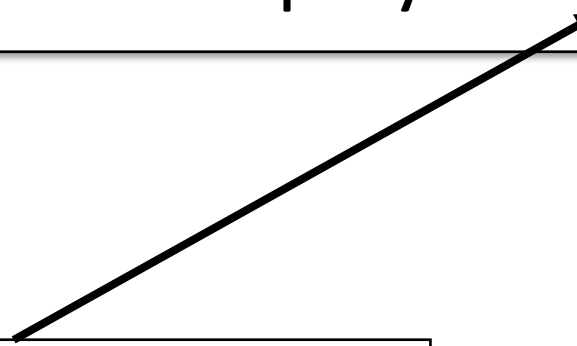
Simplex is a classic text-book algorithm

Solving Linear Programs

Theorem:

Linear program relaxations can be solved in polynomial time

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In this course, we use this theorem as a tool

Simplex is a classic text-book algorithm

The ellipsoid method achieves polynomial time

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Optimum of Relaxation is Better

Observation:

A solution to the integer program
is a solution to the relaxation.

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$$\sum_{i=1}^n x_i$$

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ \textcolor{red}{x_i} \geq 0 & \text{for all } i \in V \end{array}$$

$$\sum_{i=1}^n x_i$$

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ \textcolor{red}{x_i} \in \{0, 1\} & \text{for all } i \in V \end{array}$$

Optimum of Relaxation is Better

Observation:

A solution to the integer program is a solution to the relaxation.

Therefore, the optimum x_{IP} to the integer program is at most as good as the solution to the relaxation x_{LP} .

$$\sum_{i=1}^n x_i$$

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \geq 0 & \text{for all } i \in V \end{array}$$

$$\sum_{i=1}^n x_i$$

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \in \{0, 1\} & \text{for all } i \in V \end{array}$$

Rounding

Optimum integer x_{IP}

Optimum fractional x_{LP}

$$f(x_{LP}) \leq f(x_{IP})$$

Rounding

Optimum integer x_{IP}
Optimum fractional x_{LP}

$$f(x_{LP}) \leq f(x_{IP})$$

Idea:

Find an optimum x_{LP} .

Turn x_{LP} into an integer solution
with almost the same cost.

Rounding

Optimum integer x_{IP}
Optimum fractional x_{LP}

$$f(x_{LP}) \leq f(x_{IP})$$

Idea:

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Turn x_{LP} into an integer solution
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Using a
black box

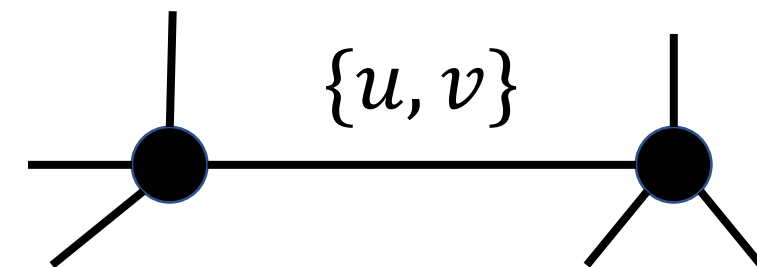


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Rounding - Minimum Vertex Cover

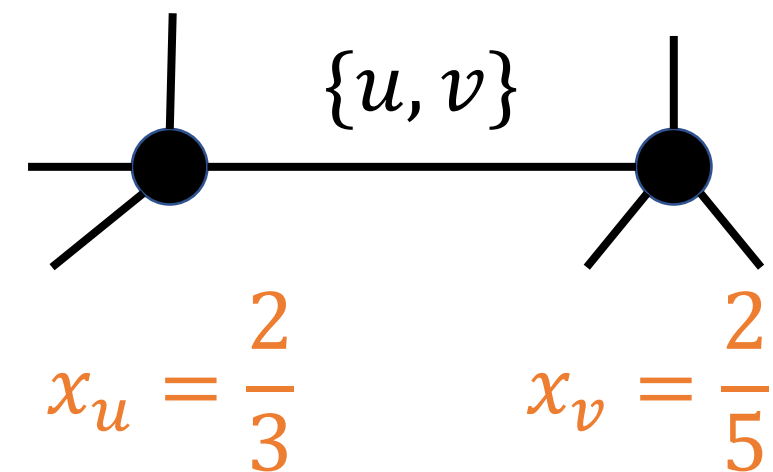
$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \geq 0 & \text{for all } i \in V \end{array}$$



Rounding - Minimum Vertex Cover

$$\begin{array}{ll} x_i + x_j \geq 1 & \text{for all } (i, j) \in E \\ x_i \geq 0 & \text{for all } i \in V \end{array}$$

Get fractional values from a black box.

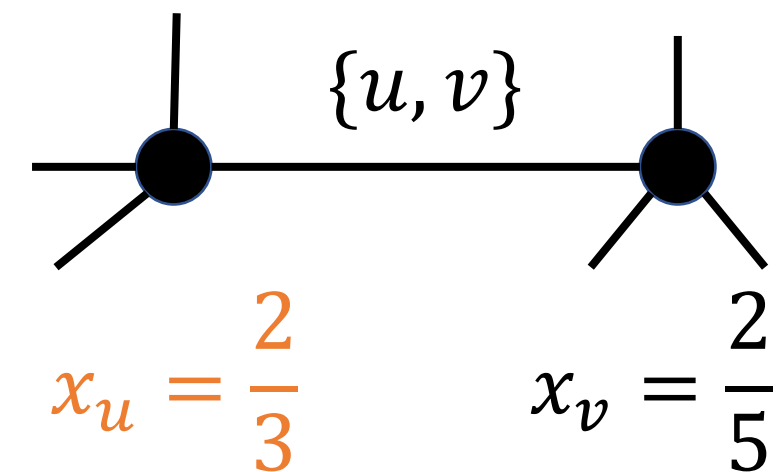


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Get fractional values from a black box.

At least one endpoint has
 $x_u \geq 1/2$

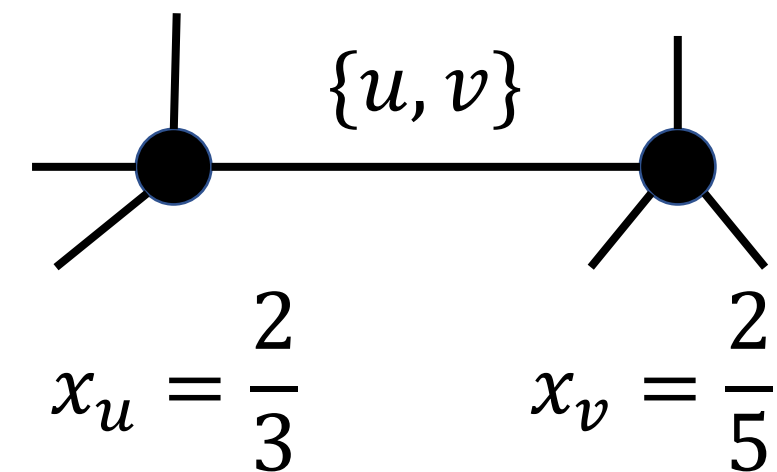


Rounding - Minimum Vertex Cover

Algorithm:

- 1) Find an optimum fractional solution
- 2) Set all $x_i \geq 1/2$ to 1
- 3) Set all $x_i < 1/2$ to 0

At least one endpoint has
 $x_u \geq 1/2$



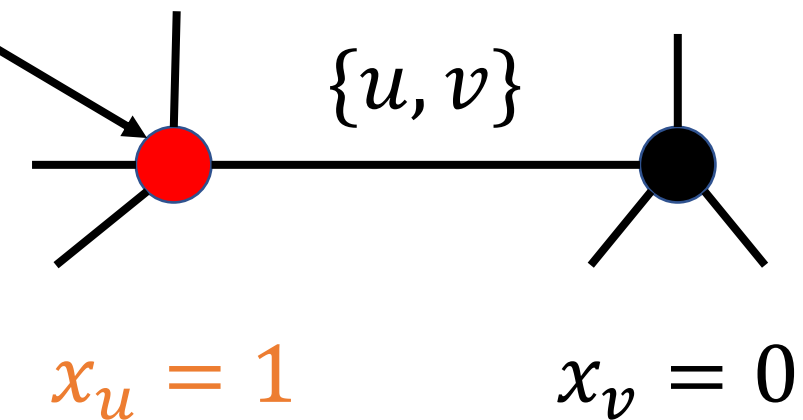
Rounding - Minimum Vertex Cover

Algorithm:

- 1) Find an optimum fractional solution
- 2) Set all $x_i \geq 1/2$ to 1
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At least one endpoint has
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Rounding:
Add to the cover

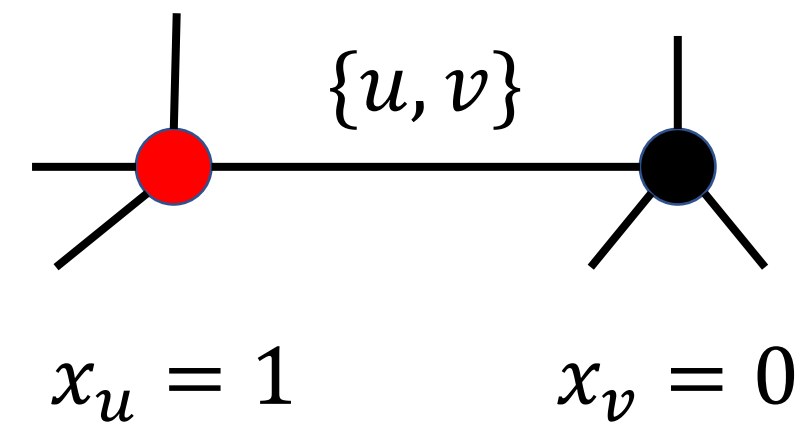


Rounding - Minimum Vertex Cover

Algorithm:

- 1) Find an optimum fractional solution
- 2) Set all $x_i \geq 1/2$ to 1
- 3) Set all $x_i < 1/2$ to 0

At least one endpoint has
 $x_u \geq 1/2$



Analysis:

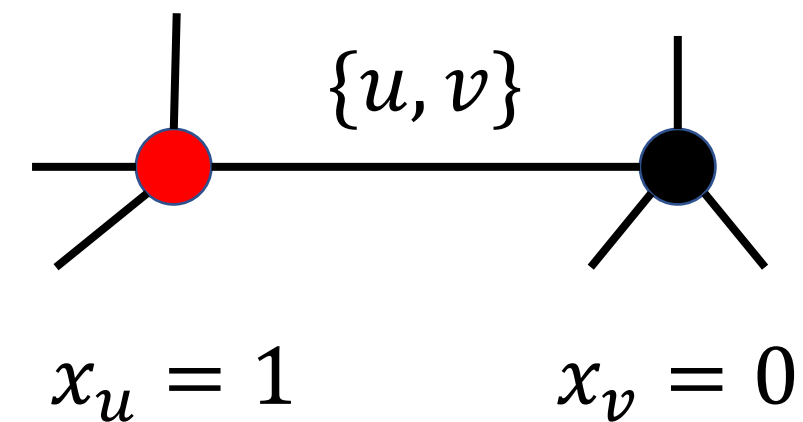
- i) Cost increases by at most a factor of 2

Rounding - Minimum Vertex Cover

Algorithm:

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- 2) Set all $x_i \geq 1/2$ to 1
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At least one endpoint has
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Analysis:

- i) Cost increases by at most a factor of 2
- ii) At least one endpoint is in the cover

Rounding - Minimum Vertex Cover

Algorithm:

- 1) Find an optimum fractional solution
- 2) Set all $x_i \geq 1/2$ to 1
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At least one endpoint has
 $x_u \geq 1/2$

Optimum integer x_{IP}
Optimum fractional x_{LP}

$$f(x_{LP}) \leq f(x_{IP})$$

2-approximation!

Analysis:

- i) Cost increases by at most a factor of 2.
- ii) At least one endpoint is in the cover

Wrap-up

Linear program:

A linear objective function and a set of linear constraints.

Has a nice geometric interpretation, especially in the case of two variables.

Rounding:

We know how to solve fractional LPs efficiently.

Sometimes, fractional solutions can be easily turned into integer solutions.