

# One iteration of the simplex algorithm

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Start with feasible basis  $B$

while  $B$  is not optimal

Let  $i \in B$  be index with  $\lambda_i < 0$

Compute  $d \in \mathbb{R}^n$  with  $a_j^T d = 0, j \in B \setminus \{i\}$  and  $a_i^T d = -1$

Determine  $K = \{k: 1 \leq k \leq m, a_k^T d > 0\}$

if  $K = \emptyset$

assert LP unbounded

else

Let  $k \in K$  index where  $\min_{k \in K} (b_k - a_k^T x^*) / a_k^T d$  is attained

update  $B := B \setminus \{i\} \cup \{k\}$



# One iteration of the simplex algorithm

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## Theorem

*One iteration of the simplex algorithm requires a total number of  $O(m \cdot n)$  operations on rational numbers whose size is polynomial in the input size.*



# Smale's 18 problems for the next century

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Problem 18: Can linear programming be solved in strongly polynomial time?

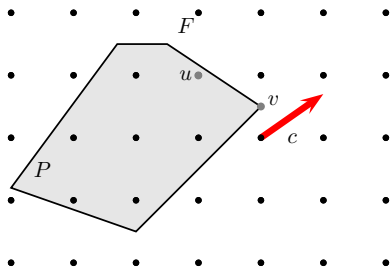


# Integer Programming

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$$\begin{aligned} \max & c^T x \\ & Ax \leq b \\ & x \in \mathbb{Z}^n \end{aligned}$$

with  $A \in \mathbb{Z}^{m \times n}$ ,  $b \in \mathbb{Z}^m$   
and  $c \in \mathbb{Z}^n$ .



# Complexity of integer programming

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## Theorem

*The integer feasibility problem is NP-complete.*





# The LP-relaxation

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## Theorem

If  $x^*$  is an optimal solution of *linear programming relaxation*  
 $\max\{c^T x : Ax \leq b, x \in \mathbb{R}^n\}$ , then  $c^T x^* \geq c^T x_I$ , for each integer feasible solution  $x_I$ .

# The LP-relaxation

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## Theorem

*If  $x^*$  is integral optimal solution of the linear programming relaxation  $\max\{c^T x : Ax \leq b, x \in \mathbb{R}^n\}$ , then  $x^*$  is also an optimal solution of the integer programming problem  $\max\{c^T x : Ax \leq b, x \in \mathbb{Z}^n\}$*

# Undirected graphs – Matchings

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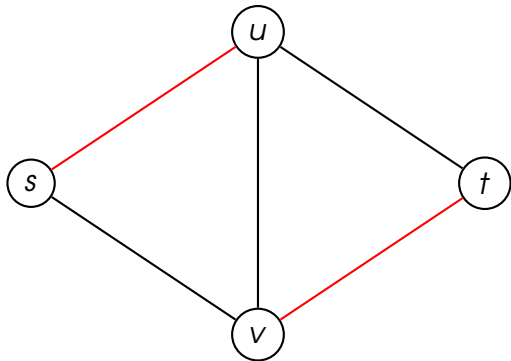
## Definition

An **undirected graph** is a tuple  $G = (V, E)$

$V$  is finite set of **vertices**

$E \subseteq \binom{V}{2}$  is set of **edges** of  $G$ .

**Matching:**  $M \subseteq E$  such that for all  $e_1 \neq e_2 \in M$  one has  $e_1 \cap e_2 = \emptyset$ .



# $\delta(v)$

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For a vertex  $v \in V$ , the set  $\delta(v) = \{e \in E : v \in e\}$  denotes the **incident** edges to  $v$ .

# Max-weight matching as IP

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$$\begin{aligned} & \max \sum_{e \in E} w(e)x(e) \\ & v \in V : \sum_{e \in \delta(v)} x(e) \leq 1 \\ & e \in E : x(e) \geq 0 \\ & x \in \mathbb{Z}^{|E|}. \end{aligned}$$

# Example

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# Integral polyhedra

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## Definition

A rational polyhedron is called **integral** if each nonempty face of  $P$  contains an integer vector.



# Simplex on polyhedra whose vertices are integral

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## Lemma

*Let  $P = \{x \in \mathbb{R}^n : Ax \leq b\}$  be an integral polyhedron with  $A \in \mathbb{R}^{m \times n}$  full-column rank. If the linear program*

$$\max\{c^T x : x \in \mathbb{R}^n, Ax \leq b\} \tag{7}$$

*is feasible and bounded, then the simplex method computes an optimal integral solution to the linear program.*

# The study of integral polyhedra

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## Lemma

*Let  $A \in \mathbb{Z}^{n \times n}$  be an integral and invertible matrix. One has  $A^{-1}b \in \mathbb{Z}^n$  for each  $b \in \mathbb{Z}^n$  if and only if  $\det(A) = \pm 1$ .*

# Total unimodularity

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## Definition

An integral matrix  $A \in \{0, \pm 1\}^{m \times n}$  is called **totally unimodular** if each of its square sub-matrices has determinant  $0, \pm 1$ .

## Example: Node-edge incidence matrix of bipartite graph

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# Hoffman-Kruskal Theorem

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## Theorem

*Let  $A \in \mathbb{Z}^{m \times n}$  be an integral matrix. The polyhedron  $P = \{x \in \mathbb{R}^n \mid Ax \leq b, x \geq 0\}$  is integral for each integral  $b \in \mathbb{Z}^m$  if and only if  $A$  is totally unimodular.*

















