

15.083J/6.859J Integer Optimization

Lecture 10: Solving Relaxations

1 Outline

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- The key geometric result behind the ellipsoid method
- The ellipsoid method for the feasibility problem
- The ellipsoid method for optimization
- Problems with exponentially many constraints

2 The Ellipsoid method

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- D is an $n \times n$ positive definite symmetric matrix
- A set E of vectors in \mathfrak{R}^n of the form

$$E = E(\mathbf{z}, \mathbf{D}) = \{\mathbf{x} \in \mathfrak{R}^n \mid (\mathbf{x} - \mathbf{z})' \mathbf{D}^{-1} (\mathbf{x} - \mathbf{z}) \leq 1\}$$

is called an **ellipsoid** with center $\mathbf{z} \in \mathfrak{R}^n$

2.1 The algorithm intuitively

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- Problem: Decide whether a given polyhedron

$$P = \{\mathbf{x} \in \mathfrak{R}^n \mid \mathbf{A}\mathbf{x} \geq \mathbf{b}\}$$

is nonempty

- Key property: We can find a new ellipsoid E_{t+1} that covers the half-ellipsoid and whose volume is only a fraction of the volume of the previous ellipsoid E_t

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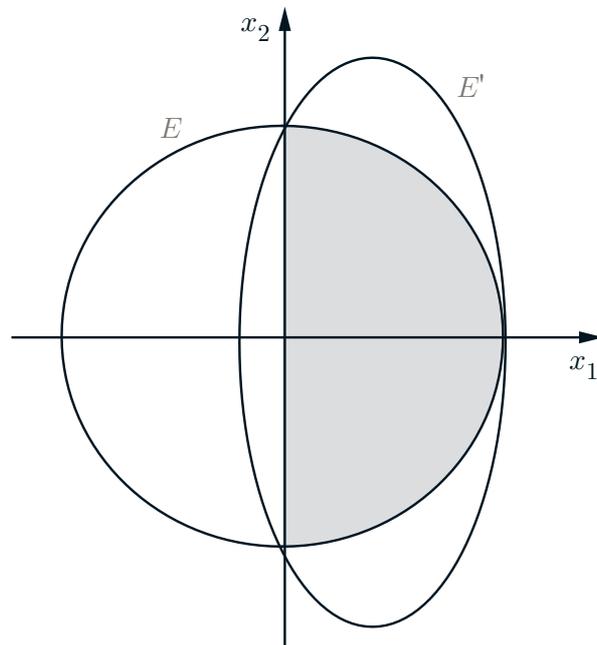
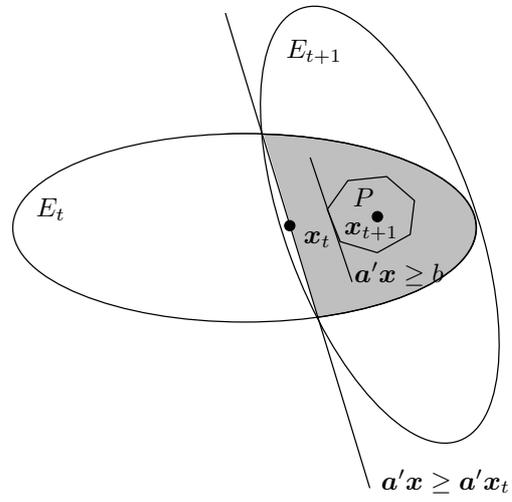
2.2 Key Theorem

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- $E = E(\mathbf{z}, \mathbf{D})$ be an ellipsoid in \mathfrak{R}^n ; \mathbf{a} nonzero n -vector.
- $H = \{\mathbf{x} \in \mathfrak{R}^n \mid \mathbf{a}'\mathbf{x} \geq \mathbf{a}'\mathbf{z}\}$

$$\begin{aligned}\bar{\mathbf{z}} &= \mathbf{z} + \frac{1}{n+1} \frac{\mathbf{D}\mathbf{a}}{\sqrt{\mathbf{a}'\mathbf{D}\mathbf{a}}}, \\ \bar{\mathbf{D}} &= \frac{n^2}{n^2-1} \left(\mathbf{D} - \frac{2}{n+1} \frac{\mathbf{D}\mathbf{a}\mathbf{a}'\mathbf{D}}{\mathbf{a}'\mathbf{D}\mathbf{a}} \right).\end{aligned}$$

- The matrix $\bar{\mathbf{D}}$ is symmetric and positive definite and thus $E' = E(\bar{\mathbf{z}}, \bar{\mathbf{D}})$ is an ellipsoid
- $E \cap H \subset E'$
- $\text{Vol}(E') < e^{-1/(2(n+1))} \text{Vol}(E)$



2.3 Illustration

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2.4 Assumptions

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- A polyhedron P is **full-dimensional** if it has positive volume
- The polyhedron P is bounded: there exists a ball $E_0 = E(\mathbf{x}_0, r^2\mathbf{I})$, with volume V , that contains P
- Either P is empty, or P has positive volume, i.e., $\text{Vol}(P) > v$ for some $v > 0$
- E_0, v, V , are a priori known
- We can make our calculations in infinite precision; square roots can be computed exactly in unit time

2.5 Input-Output

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Input:

- A matrix \mathbf{A} and a vector \mathbf{b} that define the polyhedron $P = \{\mathbf{x} \in \Re^n \mid \mathbf{a}'_i \mathbf{x} \geq b_i, i = 1, \dots, m\}$
- A number v , such that either P is empty or $\text{Vol}(P) > v$
- A ball $E_0 = E(\mathbf{x}_0, r^2\mathbf{I})$ with volume at most V , such that $P \subset E_0$

Output: A feasible point $\mathbf{x}^* \in P$ if P is nonempty, or a statement that P is empty

2.6 The algorithm

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1. (Initialization)
Let $t^* = \lceil 2(n+1) \log(V/v) \rceil$; $E_0 = E(\mathbf{x}_0, r^2\mathbf{I})$; $\mathbf{D}_0 = r^2\mathbf{I}$; $t = 0$.
2. (Main iteration)
 - If $t = t^*$ stop; P is empty.
 - If $\mathbf{x}_t \in P$ stop; P is nonempty.
 - If $\mathbf{x}_t \notin P$ find a violated constraint, that is, find an i such that $\mathbf{a}'_i \mathbf{x}_t < b_i$.
 - Let $H_t = \{\mathbf{x} \in \Re^n \mid \mathbf{a}'_i \mathbf{x} \geq \mathbf{a}'_i \mathbf{x}_t\}$. Find an ellipsoid E_{t+1} containing $E_t \cap H_t$: $E_{t+1} = E(\mathbf{x}_{t+1}, \mathbf{D}_{t+1})$ with

$$\mathbf{x}_{t+1} = \mathbf{x}_t + \frac{1}{n+1} \frac{\mathbf{D}_t \mathbf{a}_i}{\sqrt{\mathbf{a}'_i \mathbf{D}_t \mathbf{a}_i}},$$
$$\mathbf{D}_{t+1} = \frac{n^2}{n^2 - 1} \left(\mathbf{D}_t - \frac{2}{n+1} \frac{\mathbf{D}_t \mathbf{a}_i \mathbf{a}'_i \mathbf{D}_t}{\mathbf{a}'_i \mathbf{D}_t \mathbf{a}_i} \right).$$

- $t := t + 1$.

2.7 Correctness

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Theorem: Let P be a bounded polyhedron that is either empty or full-dimensional and for which the prior information \mathbf{x}_0, r, v, V is available. Then, the ellipsoid method decides correctly whether P is nonempty or not, i.e., if $\mathbf{x}_{t^*-1} \notin P$, then P is empty

2.8 Proof

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- If $\mathbf{x}_t \in P$ for $t < t^*$, then the algorithm correctly decides that P is nonempty
- Suppose $\mathbf{x}_0, \dots, \mathbf{x}_{t^*-1} \notin P$. We will show that P is empty.
- We prove by induction on k that $P \subset E_k$ for $k = 0, 1, \dots, t^*$. Note that $P \subset E_0$, by the assumptions of the algorithm, and this starts the induction.

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- Suppose $P \subset E_k$ for some $k < t^*$. Since $\mathbf{x}_k \notin P$, there exists a violated inequality: $\mathbf{a}'_{i(k)}\mathbf{x} \geq \mathbf{b}_{i(k)}$ be a violated inequality, i.e., $\mathbf{a}'_{i(k)}\mathbf{x}_k < \mathbf{b}_{i(k)}$, where \mathbf{x}_k is the center of the ellipsoid E_k
- For any $\mathbf{x} \in P$, we have

$$\mathbf{a}'_{i(k)}\mathbf{x} \geq \mathbf{b}_{i(k)} > \mathbf{a}'_{i(k)}\mathbf{x}_k$$

- Hence, $P \subset H_k = \{\mathbf{x} \in \mathfrak{R}^n \mid \mathbf{a}'_{i(k)}\mathbf{x} \geq \mathbf{a}'_{i(k)}\mathbf{x}_k\}$
- Therefore, $P \subset E_k \cap H_k$

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By key geometric property, $E_k \cap H_k \subset E_{k+1}$; hence $P \subset E_{k+1}$ and the induction is complete

$$\frac{\text{Vol}(E_{t+1})}{\text{Vol}(E_t)} < e^{-1/(2(n+1))}$$

$$\frac{\text{Vol}(E_{t^*})}{\text{Vol}(E_0)} < e^{-t^*/(2(n+1))}$$

$$\text{Vol}(E_{t^*}) < V e^{-\lceil 2(n+1) \log \frac{V}{v} \rceil / (2(n+1))} \leq V e^{-\log \frac{V}{v}} = v$$

If the ellipsoid method has not terminated after t^* iterations, then $\text{Vol}(P) \leq \text{Vol}(E_{t^*}) \leq v$. This implies that P is empty

2.9 Binary Search

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- $P = \{x \in \mathfrak{R} \mid x \geq 0, x \geq 1, x \leq 2, x \leq 3\}$
- $E_0 = [0, 5]$, centered at $x_0 = 2.5$

- Since $x_0 \notin P$, the algorithm chooses the violated inequality $x \leq 2$ and constructs E_1 that contains the interval $E_0 \cap \{x \mid x \leq 2.5\} = [0, 2.5]$
- The ellipsoid E_1 is the interval $[0, 2.5]$ itself
- Its center $x_1 = 1.25$ belongs to P
- This is binary search

2.10 Boundedness of P

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Let \mathbf{A} be an $m \times n$ integer matrix and let \mathbf{b} a vector in \Re^n . Let U be the largest absolute value of the entries in \mathbf{A} and \mathbf{b} .

Every extreme point of the polyhedron $P = \{\mathbf{x} \in \Re^n \mid \mathbf{A}\mathbf{x} \geq \mathbf{b}\}$ satisfies

$$-(nU)^n \leq x_j \leq (nU)^n, \quad j = 1, \dots, n$$

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- All extreme points of P are contained in

$$P_B = \{\mathbf{x} \in P \mid |x_j| \leq (nU)^n, j = 1, \dots, n\}$$

- Since $P_B \subseteq E(\mathbf{0}, n(nU)^{2n}\mathbf{I})$, we can start the ellipsoid method with $E_0 = E(\mathbf{0}, n(nU)^{2n}\mathbf{I})$

-

$$\text{Vol}(E_0) \leq V = (2n(nU)^n)^n = (2n)^n (nU)^{n^2}$$

2.11 Full-dimensionality

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Let $P = \{\mathbf{x} \in \Re^n \mid \mathbf{A}\mathbf{x} \geq \mathbf{b}\}$. We assume that \mathbf{A} and \mathbf{b} have integer entries, which are bounded in absolute value by U . Let

$$\epsilon = \frac{1}{2(n+1)}((n+1)U)^{-(n+1)}.$$

Let

$$P_\epsilon = \{\mathbf{x} \in \Re^n \mid \mathbf{A}\mathbf{x} \geq \mathbf{b} - \epsilon\mathbf{e}\},$$

where $\mathbf{e} = (1, 1, \dots, 1)$.

(a) If P is empty, then P_ϵ is empty.

(b) If P is nonempty, then P_ϵ is full-dimensional.

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Let $P = \{\mathbf{x} \in \Re^n \mid \mathbf{A}\mathbf{x} \geq \mathbf{b}\}$ be a full-dimensional bounded polyhedron, where the entries of \mathbf{A} and \mathbf{b} are integer and have absolute value bounded by U . Then,

$$\text{Vol}(P) > v = n^{-n} (nU)^{-n^2(n+1)}$$

2.12 Complexity

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- $P = \{\mathbf{x} \in \Re^n \mid \mathbf{Ax} \geq \mathbf{b}\}$, where \mathbf{A} , \mathbf{b} have integer entries with magnitude bounded by some U and has full rank. If P is bounded and either empty or full-dimensional, the ellipsoid method decides if P is empty in $O(n \log(V/v))$ iterations
- $v = n^{-n}(nU)^{-n^2(n+1)}$, $V = (2n)^n(nU)^{n^2}$
- Number of iterations $O(n^4 \log(nU))$

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- If P is arbitrary, we first form P_B , then perturb P_B to form $P_{B,\epsilon}$ and apply the ellipsoid method to $P_{B,\epsilon}$
- Number of iterations is $O(n^6 \log(nU))$.
- It has been shown that only $O(n^3 \log U)$ binary digits of precision are needed, and the numbers computed during the algorithm have polynomially bounded size
- The linear programming feasibility problem with integer data can be solved in polynomial time

3 The ellipsoid method for optimization

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$$\begin{array}{ll} \min & \mathbf{c}'\mathbf{x} \\ \text{s.t.} & \mathbf{Ax} \geq \mathbf{b}, \end{array} \qquad \begin{array}{ll} \max & \mathbf{b}'\boldsymbol{\pi} \\ \text{s.t.} & \mathbf{A}'\boldsymbol{\pi} = \mathbf{c} \\ & \boldsymbol{\pi} \geq \mathbf{0}. \end{array}$$

By strong duality, both problems have optimal solutions if and only if the following system of linear inequalities is feasible:

$$\mathbf{b}'\mathbf{p} = \mathbf{c}'\mathbf{x}, \quad \mathbf{Ax} \geq \mathbf{b}, \quad \mathbf{A}'\mathbf{p} = \mathbf{c}, \quad \mathbf{p} \geq \mathbf{0}.$$

LO with integer data can be solved in polynomial time.

3.1 Sliding objective

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- We first run the ellipsoid method to find a feasible solution $\mathbf{x}_0 \in P = \{\mathbf{x} \in \Re^n \mid \mathbf{Ax} \geq \mathbf{b}\}$.
- We apply the ellipsoid method to decide whether the set

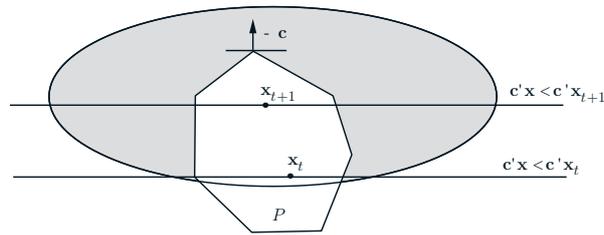
$$P \cap \{\mathbf{x} \in \Re^n \mid \mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{x}_0\}$$

is empty.

- If it is empty, then \mathbf{x}_0 is optimal. If it is nonempty, we find a new solution \mathbf{x}_1 in P with objective function value strictly smaller than $\mathbf{c}'\mathbf{x}_0$.

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- More generally, every time a better feasible solution \mathbf{x}_t is found, we take $P \cap \{\mathbf{x} \in \Re^n \mid \mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{x}_t\}$ as the new set of inequalities and reapply the ellipsoid method.



3.2 Performance in practice

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- Very slow convergence, close to the worst case
- Contrast with simplex method
- The ellipsoid method is a tool for classifying the complexity of linear programming problems

4 Problems

4.1 Example

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$$\min \sum_i c_i x_i$$

$$\sum_{i \in S} a_i x_i \geq |S|, \quad \text{for all subsets } S \text{ of } \{1, \dots, n\}$$

- There are 2^n constraints, but are described concisely in terms of the n scalar parameters a_1, \dots, a_n
- Question: Suppose we apply the ellipsoid algorithm. Is it polynomial?
- In what?

4.2 The input

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- Consider $\min \mathbf{c}'\mathbf{x}$ s.t. $\mathbf{x} \in P$
- P belongs to a family of polyhedra of special structure
- A typical polyhedron is described by specifying the dimension n and an integer vector \mathbf{h} of *primary data*, of dimension $O(n^k)$, where $k \geq 1$ is some constant.
- In example, $\mathbf{h} = (a_1, \dots, a_n)$ and $k = 1$
- U_0 be the largest entry of \mathbf{h}

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- Given n and \mathbf{h} , P is described as $\mathbf{Ax} \geq \mathbf{b}$
- \mathbf{A} has an arbitrary number of rows
- U largest entry in \mathbf{A} and \mathbf{b} . We assume

$$\log U \leq Cn^\ell \log^\ell U_0$$

5 The separation problem

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Given a polyhedron $P \subset \mathbb{R}^n$ and a vector $\mathbf{x} \in \mathbb{R}^n$, the **separation problem** is to:

- Either decide that $\mathbf{x} \in P$, or
- Find a vector \mathbf{d} such that $\mathbf{d}'\mathbf{x} < \mathbf{d}'\mathbf{y}$ for all $\mathbf{y} \in P$

What is the separation problem for

$$\sum_{i \in S} a_i x_i \geq |S|, \quad \text{for all subsets } S \text{ of } \{1, \dots, n\}?$$

6 Polynomial solvability

6.1 Theorem

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If we can solve the separation problem (for a family of polyhedra) in time polynomial in n and $\log U$, then we can also solve linear optimization problems in time polynomial in n and $\log U$. If $\log U \leq Cn^\ell \log^\ell U_0$, then it is also polynomial in $\log U_0$

- Proof ?
- Converse is also true
- Separation and optimization are polynomially equivalent

6.2 MST

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$$\begin{aligned} IZ_{MST} = \min & \sum_{e \in E} c_e x_e \\ \text{s.t.} & \sum_{e \in \delta(S)} x_e \geq 1 \quad \forall S \subseteq V, S \neq \emptyset, V \\ & \sum_{e \in E} x_e = n - 1 \\ & x_e \in \{0, 1\}. \end{aligned}$$

How can you solve the LP relaxation?

6.3 TSP

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$$x_e = \begin{cases} 1, & \text{if edge } e \text{ is included in the tour.} \\ 0, & \text{otherwise.} \end{cases}$$

$$\begin{aligned} \min \quad & \sum_{e \in E} c_e x_e \\ \text{s.t.} \quad & \sum_{e \in \delta(S)} x_e \geq 2, \quad S \subseteq E \\ & \sum_{e \in \delta(i)} x_e = 2, \quad i \in V \\ & x_e \in \{0, 1\} \end{aligned}$$

How can you solve the LP relaxation?

6.4 Probability Theory

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- Events A_1, A_2
- $P(A_1) = 0.5, P(A_2) = 0.7, P(A_1 \cap A_2) \leq 0.1$
- Are these beliefs consistent?
- General problem: Given n events $A_i, i \in N = \{1, \dots, n\}$, beliefs

$$P(A_i) \leq p_i, \quad i \in N,$$

$$P(A_i \cap A_j) \geq p_{ij}, \quad i, j \in N, i < j.$$

- Given the numbers p_i and p_{ij} , which are between 0 and 1, are these beliefs consistent?

6.4.1 Formulation

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$$\begin{aligned} x(S) &= P\left(\left(\bigcap_{i \in S} A_i\right) \cap \left(\bigcap_{i \notin S} \bar{A}_i\right)\right), \\ \sum_{\{S | i \in S\}} x(S) &\leq p_i, \quad i \in N, \\ \sum_{\{S | i, j \in S\}} x(S) &\geq p_{ij}, \quad i, j \in N, i < j, \\ \sum_S x(S) &= 1, \\ x(S) &\geq 0, \quad \forall S. \end{aligned}$$

The previous LP is feasible if and only if there does not exist a vector $(\mathbf{u}, \mathbf{y}, z)$ such that

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$$\begin{aligned} \sum_{i, j \in S, i < j} y_{ij} + \sum_{i \in S} u_i + z &\geq 0, \quad \forall S, \\ \sum_{i, j \in N, i < j} p_{ij} y_{ij} + \sum_{i \in N} p_i u_i + z &\leq -1, \\ y_{ij} \leq 0, u_i &\geq 0, \quad i, j \in N, i < j. \end{aligned}$$

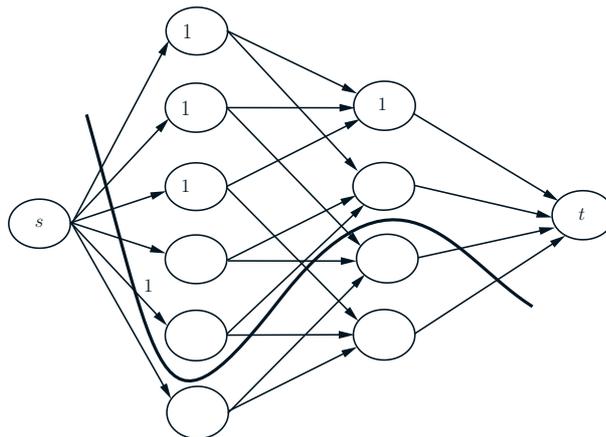
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Separation problem:

$$z^* + \min_S f(S) = \sum_{i,j \in S, i < j} y_{ij}^* + \sum_{i \in S} u_i^* \geq 0?$$

Example: $y_{12}^* = -2, y_{13}^* = -4, y_{14}^* = -4, y_{23}^* = -4, y_{24}^* = -1, y_{34}^* = -7,$
 $u_1^* = 9, u_2^* = 6, u_3^* = 4, u_4^* = 2,$ and $z^* = 2$

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- The minimum cut corresponds to $S_0 = \{3, 4\}$ with value $c(S_0) = 21$.
- $f(S_0) = \sum_{i,j \in S_0, i < j} y_{ij}^* + \sum_{i \in S_0} u_i^* = -7 + 4 + 2 = -1$
- $f(S) + z^* \geq f(S_0) + z^* = -1 + 2 = 1 > 0, \quad \forall S$
- Given solution $(\mathbf{y}^*, \mathbf{u}^*, z^*)$ is feasible

7 Conclusions

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- Ellipsoid algorithm can characterize the complexity of solving LOPs with an exponential number of constraints
- For practical purposes use dual simplex
- Ellipsoid method is an important theoretical development, not a practical one

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