

15.081J/6.251J Introduction to Mathematical Programming

Lecture 2: Geometry of Linear Optimization I

1 Outline:

SLIDE 1

1. What is the central problem?
2. Standard Form.
3. Preliminary Geometric Insights.
4. Geometric Concepts (Polyhedra, "Corners").
5. Equivalence of algebraic and geometric concepts.

2 Central Problem

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$$\begin{array}{ll} \text{minimize} & \mathbf{c}'\mathbf{x} \\ \text{subject to} & \mathbf{a}_i'\mathbf{x} = b_i \quad i \in M_1 \\ & \mathbf{a}_i'\mathbf{x} \leq b_i \quad i \in M_2 \\ & \mathbf{a}_i'\mathbf{x} \geq b_i \quad i \in M_3 \\ & x_j \geq 0 \quad j \in N_1 \\ & x_j > < 0 \quad j \in N_2 \end{array}$$

2.1 Standard Form

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$$\begin{array}{ll} \text{minimize} & \mathbf{c}'\mathbf{x} \\ \text{subject to} & \mathbf{A}\mathbf{x} = \mathbf{b} \\ & \mathbf{x} \geq \mathbf{0} \end{array}$$

Characteristics

- Minimization problem
- Equality constraints
- Non-negative variables

2.2 Transformations

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$$\begin{array}{ll} \max \mathbf{c}'\mathbf{x} & - \min(-\mathbf{c}'\mathbf{x}) \\ \mathbf{a}_i'\mathbf{x} \leq b_i & \mathbf{a}_i'\mathbf{x} + s_i = b_i, \quad s_i \geq 0 \\ \Leftrightarrow & \\ \mathbf{a}_i'\mathbf{x} \geq b_i & \mathbf{a}_i'\mathbf{x} - s_i = b_i, \quad s_i \geq 0 \\ x_j > < 0 & x_j = x_j^+ - x_j^- \\ & x_j^+ \geq 0, \quad x_j^- \geq 0 \end{array}$$

2.3 Example

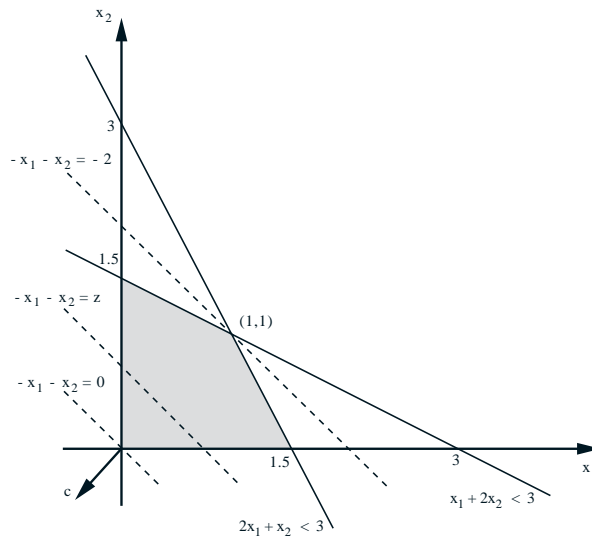
SLIDE 5

$$\begin{aligned}
 &\text{maximize} && x_1 - x_2 \\
 &\text{subject to} && x_1 + x_2 \leq 1 \\
 &&& x_1 + 2x_2 \geq 1 \\
 &&& x_1 \geq 0, x_2 \geq 0 \\
 &&& \Downarrow \\
 &-\text{minimize} && -x_1^+ + x_1^- + x_2 \\
 &\text{subject to} && x_1^+ - x_1^- + x_2 + s_1 = 1 \\
 &&& x_1^+ - x_1^- + 2x_2 - s_2 = 1 \\
 &&& x_1^+, x_1^-, x_2, s_1, s_2 \geq 0
 \end{aligned}$$

3 Preliminary Insights

SLIDE 6

$$\begin{aligned}
 &\text{minimize} && -x_1 - x_2 \\
 &\text{subject to} && x_1 + 2x_2 \leq 3 \\
 &&& 2x_1 + x_2 \leq 3 \\
 &&& x_1, x_2 \geq 0
 \end{aligned}$$

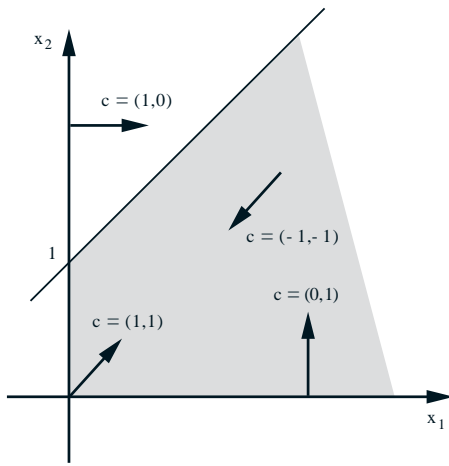


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$$\begin{aligned}
 -x_1 + x_2 &\leq 1 \\
 x_1 &\geq 0 \\
 x_2 &\geq 0
 \end{aligned}$$

SLIDE 8

- There exists a unique optimal solution.
- There exist multiple optimal solutions; in this case, the set of optimal solutions can be either bounded or unbounded.



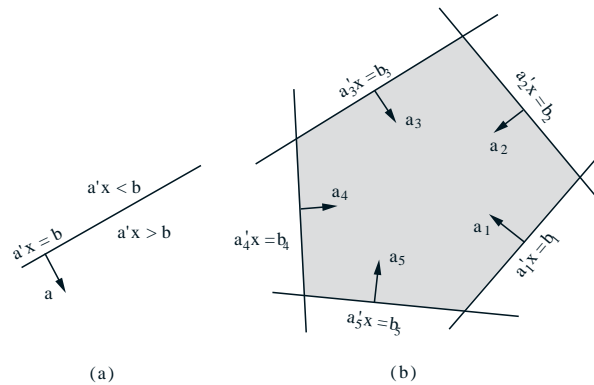
- The optimal cost is $-\infty$, and no feasible solution is optimal.
- The feasible set is empty.

4 Polyhedra

4.1 Definitions

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- The set $\{x \mid a'x = b\}$ is called a **hyperplane**.
- The set $\{x \mid a'x \geq b\}$ is called a **halfspace**.
- The intersection of many halfspaces is called a **polyhedron**.
- A polyhedron P is a convex set, i.e., if $x, y \in P$, then $\lambda x + (1 - \lambda)y \in P$.

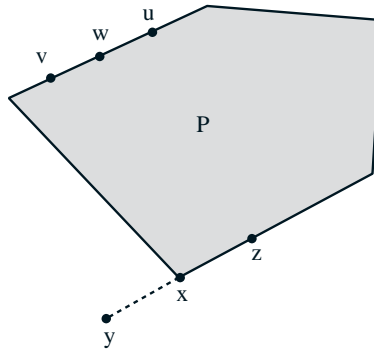


5 Corners

5.1 Extreme Points

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- Polyhedron $P = \{x \mid Ax \geq b\}$
- $x \in P$ is an extreme point of P
if $\nexists y, z \in P (y \neq x, z \neq x)$:
 $x = \lambda y + (1 - \lambda)z, 0 < \lambda < 1$



5.2 Vertex

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- $x \in P$ is a vertex of P if $\exists c$:
 x is the unique optimum

$$\begin{array}{ll} \text{minimize} & c'y \\ \text{subject to} & y \in P \end{array}$$

5.3 Basic Feasible Solution

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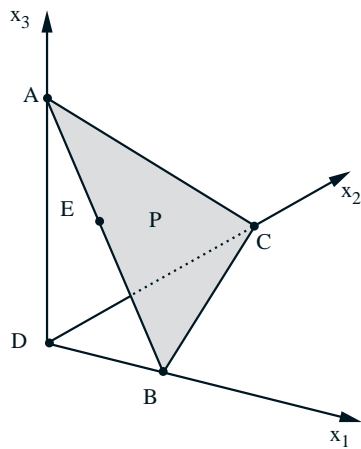
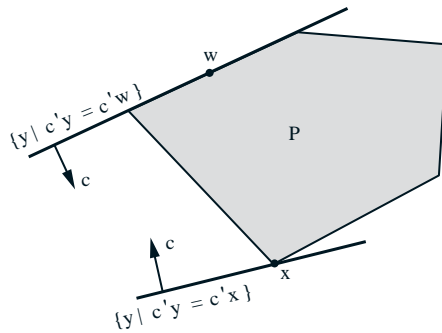
$$P = \{(x_1, x_2, x_3) \mid x_1 + x_2 + x_3 = 1, x_1, x_2, x_3 \geq 0\}$$

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Points A,B,C : 3 constraints active

Point E: 2 constraints active

suppose we add $2x_1 + 2x_2 + 2x_3 = 2$.



Then 3 hyperplanes are tight, but constraints are not linearly independent.

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Intuition: a point at which n inequalities are tight and corresponding equations are linearly independent.

$$P = \{x \in \mathbb{R}^n \mid Ax \leq b\}$$

- a_1, \dots, a_m rows of A
- $x \in P$
- $I = \{i \mid a_i'x = b_i\}$

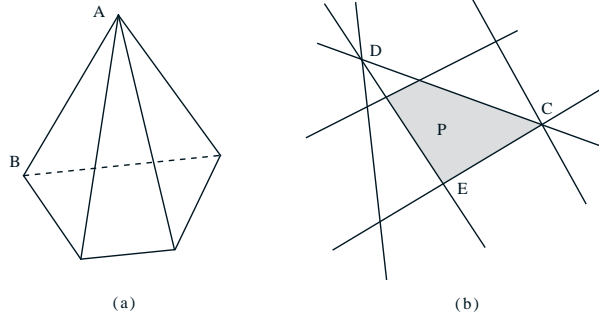
Definition x is a basic feasible solution if subspace spanned by $\{a_i, i \in I\}$ is \mathbb{R}^n .

5.3.1 Degeneracy

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- If $|I| = n$, then $a_i, i \in I$ are linearly independent; x nondegenerate.

- If $|I| > n$, then there exist n linearly independent $\{\mathbf{a}_i, i \in I\}$; \mathbf{x} degenerate.



5.3.2 Example

$$\begin{array}{llll}
 \min & x_1 + & 5x_2 & -2x_3 \\
 \text{s.t.} & x_1 + & x_2 + & x_3 \leq 4 \\
 & x_1 & & \leq 2 \\
 & & & x_3 \leq 3 \\
 & & 3x_2 + & x_3 \leq 6 \\
 & x_1, & x_2, & x_3 \geq 0
 \end{array}$$

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6 Equivalence of definitions

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Theorem: $P = \{\mathbf{x} \mid \mathbf{Ax} \leq \mathbf{b}\}$. Let $\mathbf{x} \in P$.

\mathbf{x} is a vertex $\Leftrightarrow \mathbf{x}$ is an extreme point $\Leftrightarrow \mathbf{x}$ is a BFS.

6.1 Proof

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1. Vertex \Rightarrow extreme point

$$\exists \mathbf{c} : \mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{y} \quad \forall \mathbf{y} \in P$$

If \mathbf{x} is not an extreme point $\exists \mathbf{y}, \mathbf{z} \neq \mathbf{x}$:

$$\mathbf{x} = \lambda \mathbf{y} + (1 - \lambda) \mathbf{z}. \text{ But } \mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{y}, \quad \mathbf{c}'\mathbf{x} < \mathbf{c}'\mathbf{z}$$

$$\Rightarrow \mathbf{c}'\mathbf{x} = \lambda \mathbf{c}'\mathbf{y} + (1 - \lambda) \mathbf{c}'\mathbf{z} < \mathbf{c}'\mathbf{x} \text{ contradiction}$$

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2. Extreme point \Rightarrow BFS

Suppose \mathbf{x} is not a BFS.

Let $I = \{i : \mathbf{a}'_i \mathbf{x} = b_i\}$. But \mathbf{a}_i do not span all of $\mathfrak{R}^n \Rightarrow \exists \mathbf{z} \in \mathfrak{R}^n : \mathbf{a}'_i \mathbf{z} = 0, i \in I$

Let

$$\begin{aligned} \mathbf{x}_1 &= \mathbf{x} + \epsilon \mathbf{z} \\ \mathbf{x}_2 &= \mathbf{x} - \epsilon \mathbf{z} \\ \left. \begin{aligned} \mathbf{a}'_i \mathbf{x}_1 &= b_i \\ \mathbf{a}'_i \mathbf{x}_2 &= b_i \end{aligned} \right\} i \in I \end{aligned}$$

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$i \notin I : \mathbf{a}'_i \mathbf{x} < b_i \Rightarrow \mathbf{a}'_i (\mathbf{x} + \epsilon \mathbf{z}) < b_i, \mathbf{a}'_i (\mathbf{x} - \epsilon \mathbf{z}) < b_i$
for ϵ small enough.

$\Rightarrow \mathbf{x}_1, \mathbf{x}_2 \in P$: yet $\mathbf{x} = \frac{\mathbf{x}_1 + \mathbf{x}_2}{2} \Rightarrow$

\mathbf{x} not an extreme point: **contradiction**

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3. BFS \Rightarrow vertex

\mathbf{x}^* BFS

$$I = \{i : \mathbf{a}'_i \mathbf{x}^* = b_i\}$$

$$\text{Let } d_i = \begin{cases} 1 & i \in I \\ 0 & i \notin I. \end{cases}$$

$$\mathbf{c}' = -\mathbf{d}' \mathbf{A}$$

$$\text{Then } \mathbf{c}' \mathbf{x}^* = -\mathbf{d}' \mathbf{A} \mathbf{x}^* = -\sum_{i=1}^m d_i \mathbf{a}'_i \mathbf{x}^* = -\sum_{i \in I} \mathbf{a}'_i \mathbf{x}^* = -\sum_{i \in I} b_i.$$

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But $\forall \mathbf{x} \in P : \mathbf{a}'_i \mathbf{x} \leq b_i \Rightarrow$

$$\mathbf{c}' \mathbf{x} = -\sum_{i \in I} \mathbf{a}'_i \mathbf{x} \geq -\sum_{i \in I} b_i = \mathbf{c}' \mathbf{x}^* \quad \begin{array}{l} \mathbf{x}^* \text{ optimum} \\ \min \mathbf{c}' \mathbf{x} \\ \mathbf{x} \in P. \end{array}$$

Why unique?

Equality holds if $\mathbf{a}'_i \mathbf{x} = b_i, i \in I$; since \mathbf{a}_i spans \mathfrak{R}^n , $\mathbf{a}'_i \mathbf{x} = b_i$ has a unique solution $\mathbf{x} = \mathbf{x}^*$.

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