

# 6.215/6.255J/15.093J/IDS.200J Optimization Methods

## Lecture 2: Geometry of Linear Optimization

September 14, 2021

# Today's Lecture

## Outline

- Few remaining points from Lecture 1
- General Linear Optimization Problems
- Standard Form
- Preliminary Geometric Insights
- Geometric Concepts (Polyhedra, “Corners”)
- Equivalence of Algebraic and Geometric Concepts

# Another Formulation Example - Scheduling

Data & Constraints; Decision variables?

- Hospital wants to make weekly nightshift for its nurses
- $d_j$  demand for nurses,  $j = 1 \dots 7$
- Every nurse works 5 days in a row
- Goal: hire minimum number of nurses

# Another Formulation Example - Scheduling

Data & Constraints; Decision variables?

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- $d_j$  demand for nurses,  $j = 1 \dots 7$
- Every nurse works 5 days in a row
- Goal: hire minimum number of nurses

## Decision Variables

$x_j$ : # nurses starting their week on day  $j$

# Scheduling

## Formulation

$$\begin{array}{ll}
 \min & \sum_{j=1}^7 x_j \\
 \text{s.t.} & x_1 + \phantom{x_2} \phantom{x_3} \phantom{x_4} \phantom{x_5} \phantom{x_6} \phantom{x_7} \geq d_1 \\
 & x_1 + \phantom{x_2} \phantom{x_3} \phantom{x_4} \phantom{x_5} \phantom{x_6} \phantom{x_7} \geq d_2 \\
 & x_1 + \phantom{x_2} \phantom{x_3} \phantom{x_4} \phantom{x_5} \phantom{x_6} \phantom{x_7} \geq d_3 \\
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 & \phantom{x_1} \phantom{x_2} \phantom{x_3} \phantom{x_4} \phantom{x_5} \phantom{x_6} \phantom{x_7} \geq d_6 \\
 & \phantom{x_1} \phantom{x_2} \phantom{x_3} \phantom{x_4} \phantom{x_5} \phantom{x_6} \phantom{x_7} \geq d_7 \\
 & x_j \geq 0
 \end{array}$$

# Messages

## How to formulate?

- Define your decision variables clearly.
- Write constraints and objective function.
- No systematic method available.

### What is a good LO formulation?

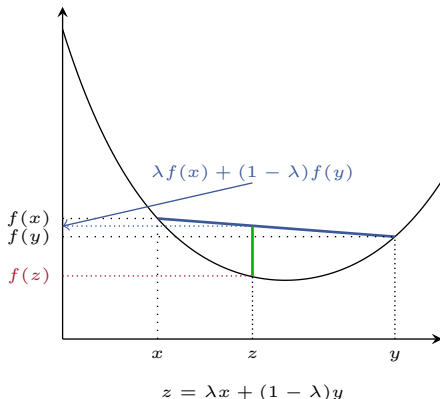
A formulation with a small number of variables and constraints, and the matrix  $\mathbf{A}$  is sparse.

# (Nonlinear) Optimization

The general problem

$$\begin{array}{ll}\min & f(x_1, \dots, x_n) \\ \text{s.t.} & g_1(x_1, \dots, x_n) \leq 0 \\ & \vdots \\ & g_m(x_1, \dots, x_n) \leq 0\end{array}$$

# From Linear to Non-Linear: The case of Convex Functions



- $f : S \rightarrow R$  is convex if for all  $x, y \in S$ , and every  $\lambda \in [0, 1]$

$$f(\lambda x + (1 - \lambda)y) \leq \lambda f(x) + (1 - \lambda)f(y)$$

- $f(\cdot)$  concave if  $-f(\cdot)$  convex.



# General Linear Optimization Problems

All possible variants

$$\begin{array}{ll} \min / \max & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} = b_i \quad i \in M_1 \\ & \mathbf{a}_i^T \mathbf{x} \leq b_i \quad i \in M_2 \\ & \mathbf{a}_i^T \mathbf{x} \geq b_i \quad i \in M_3 \\ & x_j \geq 0 \quad j \in N_1 \\ & x_j \leq 0 \quad j \in N_2 \\ & x_j \text{ free} \quad j \in N_3 \end{array}$$

# Central Problem

Standard form

$$\begin{array}{ll}\min & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{Ax} = \mathbf{b} \\ & \mathbf{x} \geq 0\end{array}$$

## Characteristics:

- minimization problem
- equality constraints
- non-negative variables

**Claim:** specialization is “without loss of generality”.

# Equivalence Transformations I

Two problems are **equivalent** if

- Solution of either problem can be found “easily” once the other is solved
- A formal definition from complexity theory exists but is beyond our scope

**Example: From max to min**

$$\begin{array}{ll} \max & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} \leq b_i \\ & \mathbf{a}_i^T \mathbf{x} \geq b_i \\ & \mathbf{x} \text{ free} \end{array} \quad \Leftrightarrow \quad \begin{array}{ll} -\min & -\mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} \leq b_i \\ & \mathbf{a}_i^T \mathbf{x} \geq b_i \\ & \mathbf{x} \text{ free} \end{array}$$

# Equivalence Transformations II

Two problems are **equivalent** if

- Solution of either problem can be found “easily” once the other is solved
- A formal definition from complexity theory exists but is beyond our scope

**Example: Slack variables**

$$\begin{array}{ll} - \min & -\mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} \leq b_i \\ & \mathbf{a}_i^T \mathbf{x} \geq b_i \\ & \mathbf{x} \text{ free} \end{array} \quad \Leftrightarrow \quad \begin{array}{ll} - \min & -\mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} + s_i = b_i, \quad s_i \geq 0 \\ & \mathbf{a}_i^T \mathbf{x} - s_i = b_i, \quad s_i \geq 0 \\ & \mathbf{x} \text{ free} \end{array}$$

# Equivalence Transformations III

Two problems are **equivalent** if

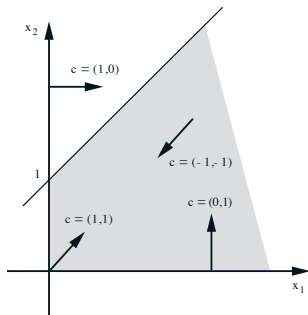
- Solution of either problem can be found “easily” once the other is solved
- A formal definition from complexity theory exists but is beyond our scope

**Example: Decomposition as difference**

$$\begin{array}{ll} - \min & -\mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{a}_i^T \mathbf{x} + s_i = b_i, \quad s_i \geq 0 \\ & \mathbf{a}_i^T \mathbf{x} - s_i = b_i, \quad s_i \geq 0 \\ & \mathbf{x} \text{ free} \end{array} \quad \Leftrightarrow \quad \begin{array}{ll} - \min & -\mathbf{c}^T (\mathbf{x}^+ - \mathbf{x}^-) \\ \text{s.t.} & \mathbf{a}_i^T (\mathbf{x}^+ - \mathbf{x}^-) + s_i = b_i, \quad s_i \geq 0 \\ & \mathbf{a}_i^T (\mathbf{x}^+ - \mathbf{x}^-) - s_i = b_i, \quad s_i \geq 0 \\ & \mathbf{x}^+ \geq 0, \quad \mathbf{x}^- \geq 0 \end{array}$$

# Preliminary Geometric Insights

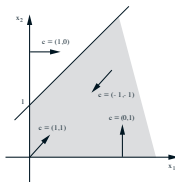
$$\begin{array}{ll}\text{minimize} & c_1 x_1 + c_2 x_2 \\ \text{subject to} & -x_1 + x_2 \leq 1 \\ & x_1 \geq 0 \\ & x_2 \geq 0\end{array}$$



# Preliminary Geometric Insights

## Wrap Up

- 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.

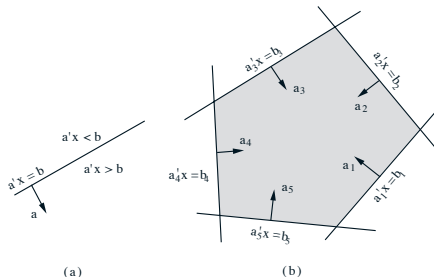


- The feasible set is empty.

# Polyhedra

## Definitions

- The set  $\{\mathbf{x} \mid \mathbf{a}^T \mathbf{x} = b\}$  is called a **hyperplane**.
- The set  $\{\mathbf{x} \mid \mathbf{a}^T \mathbf{x} \geq b\}$  is called a **halfspace**.
- The (finite) intersection of many halfspaces is called a **polyhedron**.
- A polyhedron  $P = \{\mathbf{x} \mid \mathbf{A}\mathbf{x} \geq \mathbf{b}\}$  is a convex set, i.e.,  
if  $\mathbf{x}, \mathbf{y} \in P$ , then  $\lambda \mathbf{x} + (1 - \lambda)\mathbf{y} \in P$  for any  $0 \leq \lambda \leq 1$ .

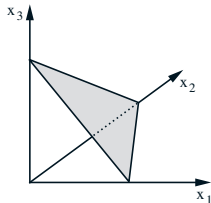




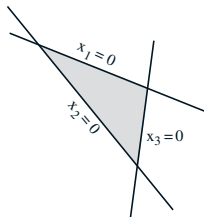
# Polyhedra - Standard Form

$$P = \{\mathbf{x} \in \mathbb{R}^n : \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq 0\}$$

- $\mathbf{A} \in \mathbb{R}^{m \times n}$  has full row rank  $m < n$
- $P$  lives in  $\mathbb{R}^{n-m}$  dimensional subspace



(a)



(b)

# Geometric vs Standard Representations

Geometric representation :  $\{\mathbf{x} : \mathbf{Ax} \geq \mathbf{b}\}$ .

- Easier to visualize
- Harder to manipulate algebraically

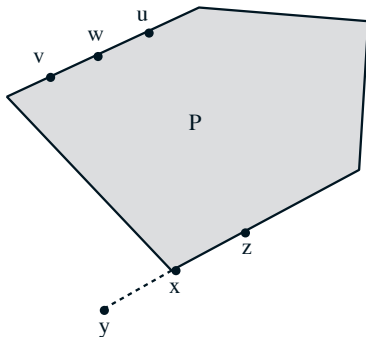
Standard Representation :  $\{\mathbf{x} : \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq 0\}$ .

- Harder to visualize
- Easier to manipulate algebraically

# Corners

## Extreme Points

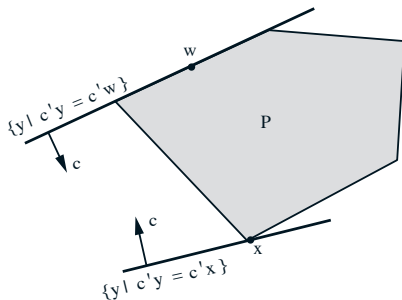
- Polyhedron  $P = \{x \mid Ax \geq b\}$
- $x \in P$  is **an extreme point** of  $P$  if there does not exist  $y, z \in P$ , both different from  $x$ , such that:  
$$x = \lambda y + (1 - \lambda)z, \quad 0 < \lambda < 1$$



# Corners

## Vertex

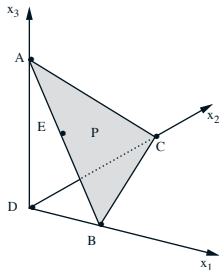
- Polyhedron  $P = \{x \mid Ax \geq b\}$
- $x \in P$  is a vertex of  $P$  if  $\exists c$  such that  $x$  is the unique optimum of
$$\begin{array}{ll}\text{minimize} & c^T y \\ \text{subject to} & y \in P\end{array}$$



# Corners

## Active Constraints

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

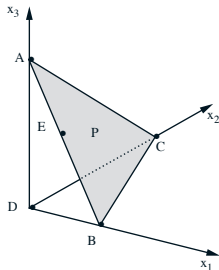


- Points A,B,C,D: 3 constraints active (active = satisfied as an equality)
- Point E: 2 constraints active

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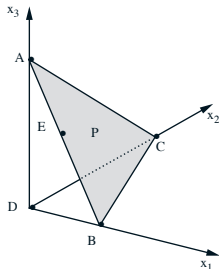


- Points A,B,C,D: 3 constraints active (active = satisfied as an equality)
- Point E: 2 constraints active
- Now, suppose we add  $2x_1 + 2x_2 + 2x_3 = 2 \Rightarrow$  3 hyperplanes are active at point E.

# Corners

## Active Constraints

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



- Points A,B,C,D: 3 constraints active (active = satisfied as an equality)
- Point E: 2 constraints active
- Now, suppose we add  $2x_1 + 2x_2 + 2x_3 = 2 \Rightarrow$  3 hyperplanes are active at point E. But these constraints are not linearly independent.

# Corners

## Basic Solution and Basic Feasible Solution (BFS)

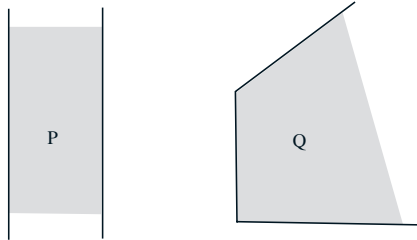
- (Classical Form) Polyhedron  $P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} \geq \mathbf{b}\}$
- **Intuition:** a basic or BFS is a point at which  $n$  inequalities are active and corresponding equations are linearly independent.
- Formally:
  - Let  $\mathbf{a}_1, \dots, \mathbf{a}_m$  the rows of  $\mathbf{A}$ ,  $\mathbf{x} \in \mathbb{R}^n$ , and  $I = \{i \mid \mathbf{a}_i^T \mathbf{x} = b_i\}$ .
    - $\mathbf{x}$  is a **basic solution** if subspace spanned by  $\{\mathbf{a}_i, i \in I\}$  is  $\mathbb{R}^n$ .
    - It is a **basic feasible solution (BFS)** if we also have  $\mathbf{x} \in P$ .



# Equivalence: vertex $\Leftrightarrow$ extreme point $\Leftrightarrow$ BFS

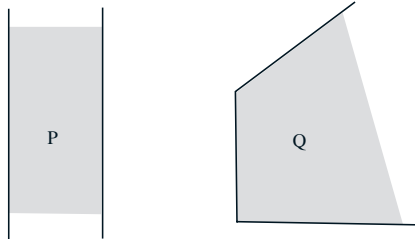
**Theorem:**  $P = \{\mathbf{x} \mid \mathbf{Ax} \geq \mathbf{b}\}$ . Let  $\mathbf{x} \in P$ . Then  $\mathbf{x}$  is a vertex  $\Leftrightarrow \mathbf{x}$  is an extreme point  $\Leftrightarrow \mathbf{x}$  is a BFS.

## Existence of extreme points



Note that  $P = \{(x_1, x_2) : 0 \leq x_1 \leq 1\}$  does not have an extreme point, while  $Q = \{(x_1, x_2) : x_1 + 5 \geq x_2, x_1 \geq 0, x_2 \geq 0\}$  has two. Why?

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### Definition

A polyhedron  $P \subset \mathbb{R}^n$  **contains a line** if there exists a vector  $\mathbf{x} \in P$  and a nonzero vector  $\mathbf{d} \in \mathbb{R}^n$  such that  $\mathbf{x} + \lambda \mathbf{d} \in P$  for all scalars  $\lambda$ .

# Existence of extreme points

## Theorem

Suppose that the polyhedron  $P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{a}_i^T \mathbf{x} \geq b_i, i = 1, \dots, m\}$  is nonempty. Then, the following are equivalent:

- (a) The polyhedron  $P$  has at least one extreme point.
- (b) The polyhedron  $P$  does not contain a line.
- (c) There exist  $n$  vectors out of the family  $\mathbf{a}_1, \dots, \mathbf{a}_m$ , which are linearly independent.

## Corollary

- (Nonempty) polyhedra in standard form contain an extreme point.
- (Nonempty) bounded polyhedra contain an extreme point.

# Optimality of extreme points

## Theorem

### Theorem

- Consider the LO

$$\begin{array}{ll}\min & \mathbf{c}^T \mathbf{x} \\ \text{s.t.} & \mathbf{x} \in P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} \geq \mathbf{b}\}.\end{array}$$

- Assume that  $P$  has no line and that the LO has an optimal solution.
- Then there exists an optimal solution which is an extreme point of  $P$ .

# BFS for standard form polyhedra

## Basic and BFS

- Standard Form Polyhedron  $P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq 0\}$
- Assume that the  $m \times n$  matrix  $\mathbf{A}$  has linearly independent rows
- $\mathbf{x} \in \mathbb{R}^n$  is a **basic solution** if and only if  $\mathbf{Ax} = \mathbf{b}$ , and there exist indices  $B(1), \dots, B(m)$  such that:
  - The columns  $\mathbf{A}_{B(1)}, \dots, \mathbf{A}_{B(m)}$  are linearly independent
  - If  $i \neq B(1), \dots, B(m)$ , then  $x_i = 0$
- Note: If we also have  $\mathbf{x} \geq 0$ , then  $\mathbf{x}$  is a **basic feasible solution**.

# BFS for standard form polyhedra

## Algebraic construction of BFS

### Procedure for constructing basic (feasible) solutions

- 1 Choose  $m$  linearly independent columns  $\mathbf{A}_{B(1)}, \dots, \mathbf{A}_{B(m)}$
- 2 Let  $x_i = 0$  for all  $i \neq B(1), \dots, B(m)$
- 3 Solve  $\mathbf{Ax} = \mathbf{b}$  for  $x_{B(1)}, \dots, x_{B(m)}$

$$\mathbf{Ax} = \mathbf{b} \rightarrow \mathbf{Bx}_B + \mathbf{Nx}_N = \mathbf{b}$$

$$\mathbf{x}_N = 0, \quad \mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b}$$

- 4 If we have  $\mathbf{x}_B \geq 0$ , then  $\mathbf{x}$  is a BFS.

# BFS for standard form polyhedra

## Example 1

$$\begin{bmatrix} 1 & 1 & 2 & 1 & 0 & 0 & 0 \\ 0 & 1 & 6 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \mathbf{x} = \begin{bmatrix} 8 \\ 12 \\ 4 \\ 6 \end{bmatrix}$$

- $\mathbf{A}_4, \mathbf{A}_5, \mathbf{A}_6, \mathbf{A}_7$  basic columns
- Solution:  $\mathbf{x}^T = (0, 0, 0, 8, 12, 4, 6)$ , a BFS
- Another basis:  $\mathbf{A}_3, \mathbf{A}_5, \mathbf{A}_6, \mathbf{A}_7$  basic columns.
- Solution:  $\mathbf{x}^T = (0, 0, 4, 0, -12, 4, 6)$ , not a BFS

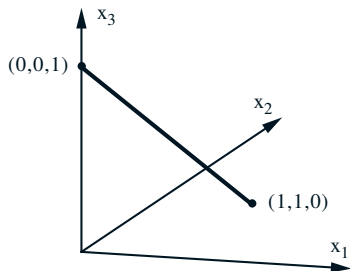


# Basic Solution - Degeneracy

With various representations of polyhedra

- Classical: Polyhedron  $P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} \geq \mathbf{b}\}$ ,  $\mathbf{A} \in \mathbb{R}^{m \times n}$ .
  - degenerate if more than  $n$  of the constraints are active at  $\mathbf{x}$
- Standard:  $P = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{Ax} = \mathbf{b}, \mathbf{x} \geq 0\}$ ,  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(\mathbf{A}) = m$ .
  - degenerate if it contains more than  $n - m$  zeros
- General:  $P$  defined by both equality and inequality constraints.
  - degenerate if more than  $n$  of the constraints are active at  $\mathbf{x}$

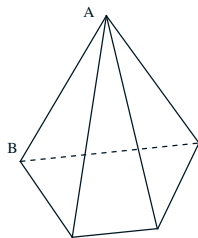
## Degeneracy is representation dependent



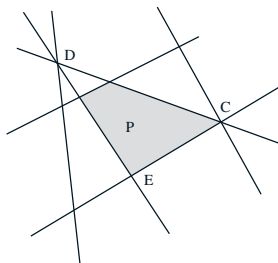
- $P = \left\{ (x_1, x_2, x_3) \mid x_1 - x_2 = 0, x_1 + x_2 + 2x_3 = 2, x_1, x_2, x_3 \geq 0 \right\}$
- $n = 3, m = 2$  and  $n - m = 1$   
 $\Rightarrow (1, 1, 0)$  is nondegenerate, while  $(0, 0, 1)$  is degenerate.
- $P = \left\{ (x_1, x_2, x_3) \mid x_1 - x_2 = 0, x_1 + x_2 + 2x_3 = 2, x_1 \geq 0, x_3 \geq 0 \right\}$   
 $\Rightarrow (0, 0, 1)$  is now nondegenerate !

# Checking ...

Are the points  $A, B, C, D, E$  basic solutions?, BFS?, Denegerate?



(a)



(b)