

# 15.081J/6.251J Introduction to Mathematical Programming

Lecture 4: Geometry of Linear Optimization III

# 1 Outline

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1. Projections of Polyhedra
2. Fourier-Motzkin Elimination Algorithm
3. Optimality Conditions

# 2 Projections of polyhedra

SLIDE 2

- $\pi_k : \mathbb{R}^n \mapsto \mathbb{R}^k$  projects  $\mathbf{x}$  onto its first  $k$  coordinates:

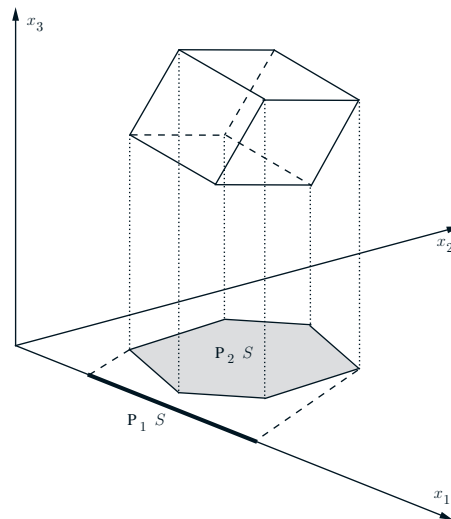
$$\pi_k(\mathbf{x}) = \pi_k(x_1, \dots, x_n) = (x_1, \dots, x_k).$$

- 

$$\Pi_k(S) = \{\pi_k(\mathbf{x}) \mid \mathbf{x} \in S\};$$

Equivalently

$$\Pi_k(S) = \left\{ (x_1, \dots, x_k) \mid \text{there exist } x_{k+1}, \dots, x_n \right. \\ \left. \text{s.t. } (x_1, \dots, x_n) \in S \right\}.$$



## 2.1 The Elimination Algorithm

### 2.1.1 By example

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- Consider the polyhedron

$$x_1 + x_2 \geq 1$$

$$\begin{aligned}
x_1 + x_2 + 2x_3 &\geq 2 \\
2x_1 + 3x_3 &\geq 3 \\
x_1 - 4x_3 &\geq 4 \\
-2x_1 + x_2 - x_3 &\geq 5.
\end{aligned}$$

- We rewrite these constraints

$$\begin{aligned}
0 &\geq 1 - x_1 - x_2 \\
x_3 &\geq 1 - (x_1/2) - (x_2/2) \\
x_3 &\geq 1 - (2x_1/3) \\
-1 + (x_1/4) &\geq x_3 \\
-5 - 2x_1 + x_2 &\geq x_3.
\end{aligned}$$

- Eliminate variable  $x_3$ , obtaining polyhedron  $Q$

$$\begin{aligned}
0 &\geq 1 - x_1 - x_2 \\
-1 + x_1/4 &\geq 1 - (x_1/2) - (x_2/2) \\
-1 + x_1/4 &\geq 1 - (2x_1/3) \\
-5 - 2x_1 + x_2 &\geq 1 - (x_1/2) - (x_2/2) \\
-5 - 2x_1 + x_2 &\geq 1 - (2x_1/3).
\end{aligned}$$

## 2.2 The Elimination Algorithm

SLIDE 4

1. Rewrite  $\sum_{j=1}^n a_{ij}x_j \geq b_i$  in the form

$$a_{in}x_n \geq -\sum_{j=1}^{n-1} a_{ij}x_j + b_i, \quad i = 1, \dots, m;$$

if  $a_{in} \neq 0$ , divide both sides by  $a_{in}$ . By letting  $\bar{x} = (x_1, \dots, x_{n-1})$  that  $P$  is represented by:

$$\begin{aligned}
x_n &\geq d_i + \mathbf{f}'_i \bar{x}, & \text{if } a_{in} > 0, \\
d_j + \mathbf{f}'_j \bar{x} &\geq x_n, & \text{if } a_{jn} < 0, \\
0 &\geq d_k + \mathbf{f}'_k \bar{x}, & \text{if } a_{kn} = 0.
\end{aligned}$$

2. Let  $Q$  be the polyhedron in  $\mathfrak{R}^{n-1}$  defined by:

$$\begin{aligned}
d_j + \mathbf{f}'_j \bar{x} &\geq d_i + \mathbf{f}'_i \bar{x}, & \text{if } a_{in} > 0 \text{ and } a_{jn} < 0, \\
0 &\geq d_k + \mathbf{f}'_k \bar{x}, & \text{if } a_{kn} = 0.
\end{aligned}$$

Theorem:

The polyhedron  $Q$  constructed by the elimination algorithm is equal to the projection  $\Pi_{n-1}(P)$  of  $P$ .

## 2.3 Implications

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- Let  $P \subset \mathbb{R}^{n+k}$  be a polyhedron. Then, the set

$$\{\mathbf{x} \in \mathbb{R}^n \mid \text{there exists } \mathbf{y} \in \mathbb{R}^k \text{ such that } (\mathbf{x}, \mathbf{y}) \in P\}$$

is also a polyhedron.

- Let  $P \subset \mathbb{R}^n$  be a polyhedron and let  $\mathbf{A}$  be an  $m \times n$  matrix. Then, the set  $Q = \{\mathbf{Ax} \mid \mathbf{x} \in P\}$  is also a polyhedron.
- The convex hull of a finite number of vectors is a polyhedron.

## 2.4 Algorithm for LO

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- Consider  $\min \mathbf{c}'\mathbf{x}$  subject to  $\mathbf{x} \in P$ .
- Define a new variable  $x_0$  and introduce the constraint  $x_0 = \mathbf{c}'\mathbf{x}$ .
- Apply the elimination algorithm  $n$  times to eliminate the variables  $x_1, \dots, x_n$ .
- We are left with the set

$$Q = \{x_0 \mid \text{there exists } \mathbf{x} \in P \text{ such that } x_0 = \mathbf{c}'\mathbf{x}\},$$

and the optimal cost is equal to the smallest element of  $Q$ .

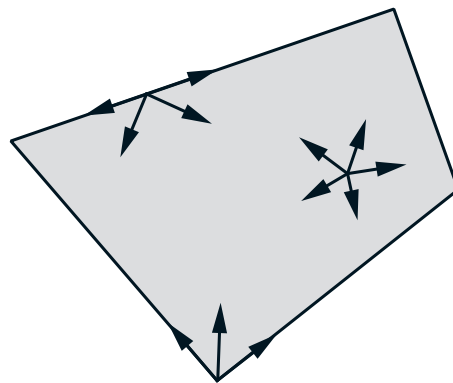
## 3 Optimality Conditions

### 3.1 Feasible directions

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- We are at  $\mathbf{x} \in P$  and we contemplate moving away from  $\mathbf{x}$ , in the direction of a vector  $\mathbf{d} \in \mathbb{R}^n$ .
- We need to consider those choices of  $\mathbf{d}$  that do not immediately take us outside the feasible set.
- A vector  $\mathbf{d} \in \mathbb{R}^n$  is said to be a **feasible direction** at  $\mathbf{x}$ , if there exists a positive scalar  $\theta$  for which  $\mathbf{x} + \theta\mathbf{d} \in P$ .

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- $\mathbf{x}$  be a BFS to the standard form problem corresponding to a basis  $\mathbf{B}$ .
- $x_i = 0, i \in N, \mathbf{x}_B = \mathbf{B}^{-1}\mathbf{b}$ .
- We consider moving away from  $\mathbf{x}$ , to a new vector  $\mathbf{x} + \theta\mathbf{d}$ , by selecting a nonbasic variable  $x_j$  and increasing it to a positive value  $\theta$ , while keeping the remaining nonbasic variables at zero.
- Algebraically,  $d_j = 1$ , and  $d_i = 0$  for every nonbasic index  $i$  other than  $j$ .
- The vector  $\mathbf{x}_B$  of basic variables changes to  $\mathbf{x}_B + \theta\mathbf{d}_B$ .
- Feasibility:  $\mathbf{A}(\mathbf{x} + \theta\mathbf{d}) = \mathbf{b} \Rightarrow \mathbf{A}\mathbf{d} = \mathbf{0}$ .
- $\mathbf{0} = \mathbf{A}\mathbf{d} = \sum_{i=1}^n \mathbf{A}_i d_i = \sum_{i=1}^m \mathbf{A}_{B(i)} d_{B(i)} + \mathbf{A}_j = \mathbf{B}\mathbf{d}_B + \mathbf{A}_j \Rightarrow \mathbf{d}_B = -\mathbf{B}^{-1}\mathbf{A}_j$ .
- Nonnegativity constraints?
  - If  $\mathbf{x}$  nondegenerate,  $\mathbf{x}_B > \mathbf{0}$ ; thus  $\mathbf{x}_B + \theta\mathbf{d}_B \geq \mathbf{0}$  for  $\theta$  is sufficiently small.
  - If  $\mathbf{x}$  degenerate, then  $\mathbf{d}$  is not always a feasible direction. Why?
- Effects in cost?
 

Cost change:  $\mathbf{c}'\mathbf{d} = c_j - \mathbf{c}'_B \mathbf{B}^{-1}\mathbf{A}_j$  This quantity is called **reduced cost**  $\bar{c}_j$  of the variable  $x_j$ .

### 3.2 Theorem

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- $\mathbf{x}$  BFS associated with basis  $\mathbf{B}$
- $\bar{\mathbf{c}}$  reduced costs  
Then
- If  $\bar{\mathbf{c}} \geq \mathbf{0} \Rightarrow \mathbf{x}$  optimal
- $\mathbf{x}$  optimal and non-degenerate  $\Rightarrow \bar{\mathbf{c}} \geq \mathbf{0}$

### 3.3 Proof

- $\mathbf{y}$  arbitrary feasible solution
- $\mathbf{d} = \mathbf{y} - \mathbf{x} \Rightarrow \mathbf{A}\mathbf{x} = \mathbf{A}\mathbf{y} = \mathbf{b} \Rightarrow \mathbf{A}\mathbf{d} = \mathbf{0}$

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$$\Rightarrow \mathbf{B}\mathbf{d}_B + \sum_{i \in N} \mathbf{A}_i d_i = \mathbf{0}$$

$$\Rightarrow \mathbf{d}_B = - \sum_{i \in N} \mathbf{B}^{-1} \mathbf{A}_i d_i$$

$$\begin{aligned} \Rightarrow \mathbf{c}'\mathbf{d} &= \mathbf{c}'_B \mathbf{d}_B + \sum_{i \in N} c_i d_i \\ &= \sum_{i \in N} (c_i - \mathbf{c}'_B \mathbf{B}^{-1} \mathbf{A}_i) d_i = \sum_{i \in N} \bar{c}_i d_i \end{aligned}$$

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- Since  $y_i \geq 0$  and  $x_i = 0, i \in N$ , then  $d_i = y_i - x_i \geq 0, i \in N$
  - $\mathbf{c}'\mathbf{d} = \mathbf{c}'(\mathbf{y} - \mathbf{x}) \geq 0 \Rightarrow \mathbf{c}'\mathbf{y} \geq \mathbf{c}'\mathbf{x}$   
 $\Rightarrow \mathbf{x}$  optimal
- (b) Your turn

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