

Integer Programming

ISE 418

Lecture 6

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Reading for This Lecture

- N&W Sections I.4.4 and I.4.6
- Wolsey Section 9.1
- CCZ Chapter 3

Describing Polyhedra

In Lecture 5, we derived the following fundamental results.

Theorem 1.

1. *Every full-dimensional polyhedron \mathcal{P} has a unique (up to scalar multiplication) representation that consists of one inequality representing each facet of \mathcal{P} .*
2. *If $\dim(\mathcal{P}) = n - k$ with $k > 0$, then \mathcal{P} is described by a maximal set of linearly independent rows of $(A^=, b^=)$, as well as one inequality representing each facet of \mathcal{P} .*

Theorem 2. *If a facet F of \mathcal{P} is represented by (π, π_0) , then the set of all representations of F is obtained by taking scalar multiples of (π, π_0) plus linear combinations of the equality set of \mathcal{P} .*

Extreme Points

Definition 1. x is an **extreme point** of \mathcal{P} if there do not exist $x^1, x^2 \in \mathcal{P}$ such that $x = \frac{1}{2}x^1 + \frac{1}{2}x^2$.

Proposition 1. x is an extreme point of \mathcal{P} if and only if x is a zero-dimensional face of \mathcal{P} .

Proposition 2. If $\mathcal{P} \neq \emptyset$ and $\text{rank}(A) = n - k$, then \mathcal{P} has a face of dimension k and no proper face of lower dimension.

- These three results together imply that \mathcal{P} has an extreme point if and only if $\text{rank}(A) = n$.
- This is the case for any polytope or any polyhedron lying in the non-negative orthant.
- Recall that in 406, we showed that a polyhedron has an extreme point if and only if it does not contain a line.
- **Don't confuse $\text{rank}(A) = n$ with \mathcal{P} being full-dimensional!**

Extreme Rays

Definition 2. The **recession cone** $\text{rec}(\mathcal{P})$ associated with \mathcal{P} is $\{r \in \mathbb{R}^n \mid Ar \leq 0\}$. Members of the recession cone are called **rays** of \mathcal{P} .

Definition 3. r is an **extreme ray** of \mathcal{P} if there do not exist rays r^1 and r^2 of \mathcal{P} such that $r = \frac{1}{2}r^1 + \frac{1}{2}r^2$.

Proposition 3. If $\mathcal{P} \neq \emptyset$, then r is an extreme ray of \mathcal{P} if and only if $\{\lambda r \mid \lambda \in \mathbb{R}_+\}$ is a one-dimensional face of the recession cone.

Definition 4. The **lineality space** of \mathcal{P} is $\text{lin}(\mathcal{P}) = \{r \in \mathbb{R}^n \mid Ar = 0\} = \text{rec}(\mathcal{P}) \cap (-\text{rec}(\mathcal{P}))$.

- We have that $\text{lin}(\mathcal{P}) = \{0\}$ if and only if $\text{rank}(A) = n$ and in that case, we call the polyhedron **pointed**.
- Note that if r is an extreme ray, then so is λr for $\lambda > 0$.
- We need only consider one “representative” of each one-dimensional face of the recession cone.
- We can do this by choosing extreme rays r with $\|r\| = 1$.
- The last two results together imply that a polyhedron has a finite number of extreme points and extreme rays.

Polarity

Definition 5. The **polar** of a set S is $S^* = \{y \in \mathbb{R}^n \mid yx \leq 1 \forall x \in S\}$.

Theorem 3. Given $a^1, \dots, a^m \in \mathcal{Q}^n$ and $0 \leq k \leq m$, let

$$\mathcal{Q}_1 = \{x \in \mathbb{R}^n \mid a^i x \leq 1, i = 1, \dots, k; a^i x \leq 0, i = k + 1, \dots, m\}$$

$$\mathcal{Q}_2 = \text{conv}(\{0, a^1, \dots, a^k\}) + \text{cone}(\{a^{k+1}, \dots, a^m\})$$

Then $\mathcal{Q}_1^* = \mathcal{Q}_2$ and $\mathcal{Q}_2^* = \mathcal{Q}_1$

- From this definition, we can see that if \mathcal{Q} is a polyhedron containing the origin, then have that
 1. \mathcal{Q}^* is also a polyhedron containing the origin;
 2. $\mathcal{Q}^{**} = \mathcal{Q}$;
 3. \mathcal{Q}^* is bounded if and only if \mathcal{Q} contains the origin in its interior;
 4. $\text{aff}(\mathcal{Q}^*)$ is the orthogonal complement of $\text{lin}(\mathcal{Q})$ and $\dim(\mathcal{Q}^*) + \dim(\text{lin}(\mathcal{Q})) = n$.
- \mathcal{Q}^* can be roughly interpreted as the set of all valid inequalities of \mathcal{Q} .
- When \mathcal{Q} is full-dimensional and has 0 in its interior, then the extreme of \mathcal{Q}^* are in one-to-one correspondence with the facets of \mathcal{Q} .

Intuition

- Theorem 3 may seem mysterious at first, but it's actually quite intuitive.
- There are two perspectives we can take to understand it.
- First recall that inequalities valid for non-empty polyhedron $\mathcal{P} = \{x \in \mathbb{R}^m \mid Ax \leq b\}$ can be generated by solving the system

$$uA = \pi$$

$$ub \leq \pi_0$$

$$u \geq 0$$

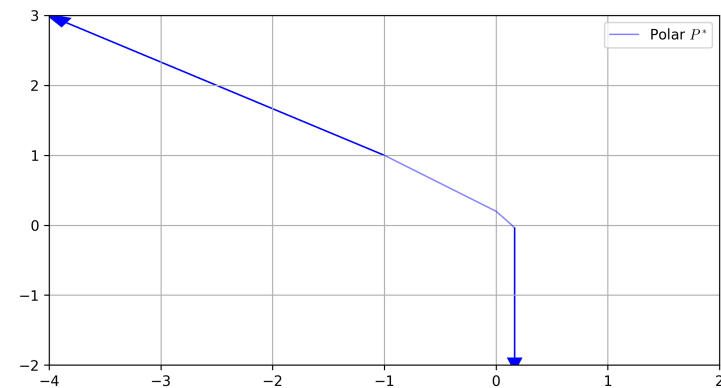
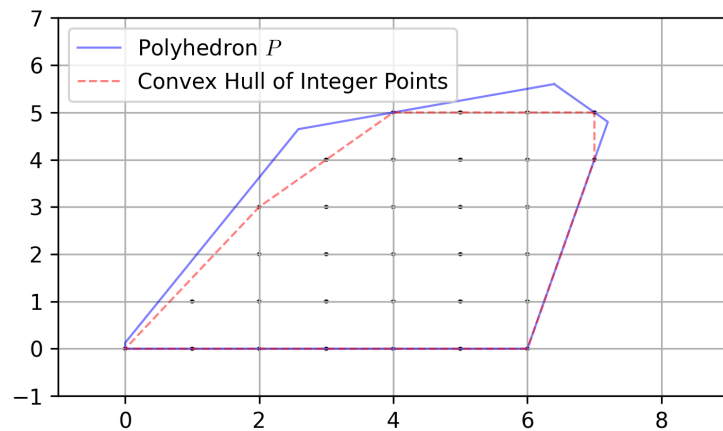
- Alternatively, the conic hull of the rows of A includes all non-dominated valid inequalities.
- Assuming A has full rank, then the rows of A are the extreme rays of this conic hull.
- To get to the polar, we normalize, which truncates the cone so that the extreme rays of the conic hull become extreme points of the polar.
- Inequalities through the origin have right-hand side zero and can't be normalized, so if the origin is not in the interior, the polar is unbounded.

Example

Let $\mathcal{P} = \{x \in \mathbb{R}_+ \mid Ax \leq b\}$, where

$$A = \begin{bmatrix} -1 & 4 \\ -14 & 8 \\ 4 & -1 \\ 1 & 1 \end{bmatrix}, \quad b = \begin{bmatrix} 16 \\ 1 \\ 24 \\ 12 \end{bmatrix}$$

and let $\mathcal{S} = \mathcal{P} \cap \mathbb{Z}^2$. Then according to Theorem 3, the polyhedron and its polar are



Some Results from Linear Optimization

Theorem 4. *If $\mathcal{P} \neq \emptyset$, $\text{rank}(A) = n$, and $\max\{cx \mid x \in \mathcal{P}\}$ is finite, then there is an optimal solution that is an extreme point.*

Theorem 5. *For a given extreme point x^* , there exists a $c \in \mathbb{Z}^n$ such that x^* is the optimal solution to $\max\{cx \mid x \in \mathcal{P}\}$*

Theorem 6. *If $\mathcal{P} \neq \emptyset$, $\text{rank}(A) = n$, and $\max\{cx \mid x \in \mathcal{P}\}$ is unbounded, then there is an extreme ray r^* with $cr^* > 0$.*

- Note again that the set of all optimal solutions to a linear optimization problem is a face of the associated polyhedron.
- We call this the *optimal face*.
- Combining these results, we get [Minkowski's Theorem](#).

Minkowski's Theorem

Theorem 7. *If $\mathcal{P} \neq \emptyset$ and $\text{rank}(A) = n$, then*

$$\mathcal{P} = \left\{ \sum_{k \in \mathcal{E}} \lambda_k x^k + \sum_{j \in \mathcal{R}} \mu_j r^j \mid \lambda_k \geq 0 \text{ for } k \in \mathcal{E}, \mu_j \geq 0 \text{ for } j \in \mathcal{R}, \sum_{k \in \mathcal{E}} \lambda_k = 1 \right\},$$

where $\{x^k\}_{k \in \mathcal{E}}$ are the extreme points and $\{r^j\}_{j \in \mathcal{R}}$ are the (representative) extreme rays.

Corollary 1. *A nonempty polyhedron is bounded if and only if it has no extreme rays.*

Corollary 2. *A polytope is the convex hull of its extreme points.*

- A set of the form given above is called *finitely generated* when \mathcal{R} and \mathcal{E} are finite sets.
- If \mathcal{R} or \mathcal{E} were not finite, then the feasible region would be that of a *semi-infinite optimization problem*.
- This result is often stated as “every polyhedron is finitely generated.”

More Results from Linear Optimization

Define the following:

- $\mathcal{P}^+ = \{x \in \mathbb{R}_+^n \mid Ax \leq b\}$, $z = \max\{cx \mid x \in \mathcal{P}^+\}$
- $\mathcal{Q}^+ = \{u \in \mathbb{R}_+^m \mid uA \geq c\}$, $w = \min\{ub \mid u \in \mathcal{Q}^+\}$
- $\{x^k\}_{k \in \mathcal{E}}$, $\{u^i\}_{i \in \mathcal{I}}$ are the extreme points of \mathcal{P}^+ and \mathcal{Q}^+ respectively.
- $\{r^j\}_{j \in \mathcal{R}}$, $\{v^t\}_{t \in \mathcal{T}}$ are the extreme rays of \mathcal{P}^+ and \mathcal{Q}^+ respectively.

Theorem 8 (Farkas). $\mathcal{P}^+ \neq \emptyset \Leftrightarrow v^t b \geq 0 \forall t \in \mathcal{T}$

Theorem 9. *The following are equivalent when $\mathcal{P}^+ \neq \emptyset$:*

1. z is unbounded from above;
2. there exists $j \in \mathcal{R}$ with $cr^j > 0$; and
3. $\mathcal{Q}^+ = \emptyset$.

Theorem 10. *If $\mathcal{P}^+ \neq \emptyset$ and z is bounded, then*

$$z = \max_{k \in \mathcal{E}} cx^k = w = \min_{i \in \mathcal{I}} u^i b$$

The Projection of a Polyhedron

- Let $\mathcal{Q} = \{(x, y) \in \mathbb{R}^r \times \mathbb{R}^{n-r} \mid Gx + Hy \leq b\}$
- The projection of \mathcal{P} into the space of just the x variables is

$$\begin{aligned} \text{proj}_x(\mathcal{Q}) &= \{x \in \mathbb{R}^n \mid \exists y \in \mathbb{R}^p, (x, y) \in \mathcal{P}\} \\ &= \{x \in \mathbb{R}^n \mid v^t(b - Ax) \geq 0 \forall t \in \mathcal{T}\} \end{aligned}$$

where $\{v^t\}_{t \in \mathcal{T}}$ are the extreme rays of $Q = \{v \in \mathbb{R}_+^m \mid vH = 0\}$.

- This immediately implies that **the projection of a polyhedron is a polyhedron**.
- Note that this notion of projection does not technically coincide with the one from Lecture 4, as projection usually takes place in the same space.
- In other words, the ordinary notion would be to project \mathcal{Q} into the subspace $\{(x, y) \in \mathbb{R}^r \times \mathbb{R}^{n-r} \mid y = 0\}$.
- The projection of a point (x, y) into this subspace is the point $(x, 0)$, where $x \in \text{proj}_x(\mathcal{Q})$.

Weyl's Theorem

Theorem 11. *If*

$$Q = \left\{ \sum_{k \in \mathcal{E}} \lambda_k x^k + \sum_{j \in \mathcal{R}} \mu_j r^j \mid \lambda_k \geq 0 \text{ for } k \in \mathcal{E}, \mu_j \geq 0 \text{ for } j \in \mathcal{R}, \sum_{k \in \mathcal{E}} \lambda_k = 1 \right\},$$

where $\{x^k\}_{k \in \mathcal{E}}$ and $\{r^j\}_{j \in \mathcal{R}}$ are given sets of rational vectors, then Q is a rational polyhedron.

- This is the converse of Minkowski's Theorem.
- This says roughly “every finitely generated set is a polyhedron” (remember the rationality assumption).
- The proof is easy using projection.

The Fundamental Theorem

- We have already discussed informally the fact that an integer optimization problem can, in theory, be reduced to a linear optimization problem.
- We now make these ideas more formal.
- To do so, we would now like to show the following:

Theorem 12. (*The Fundamental Theorem of Integer Optimization*)
If $\mathcal{P} = \{x \in \mathbb{R}^n \mid Ax \leq b\}$, where $A \in \mathbb{Q}^{m \times n}$, $b \in \mathbb{Q}^m$, and $\mathcal{S} = \mathcal{P} \cap (\mathbb{Z}^p \times \mathbb{R}^{n-p})$, then $\text{conv}(\mathcal{S})$ is a rational polyhedron with the same recession cone as \mathcal{P} .

Proving \mathcal{S} Is Finitely Generated

- This result is easily proven if \mathcal{S} is bounded (how?).
- If \mathcal{S} is not bounded, then it is not so obvious.
- Our approach will be to show that \mathcal{S} itself can be finitely generated.
- It then follows that $\text{conv}(\mathcal{S})$ is finitely generated.

Proving \mathcal{S} Is Finitely Generated (cont.)

- Consider \mathcal{P} and \mathcal{S} from Theorem 12.
- By Minkowski's Theorem, we can write

$$\mathcal{P} = \left\{ \sum_{k \in \mathcal{E}} \lambda_k x^k + \sum_{j \in \mathcal{R}} \mu_j r^j \mid \lambda_k \geq 0 \text{ for } k \in \mathcal{E}, \mu_j \geq 0 \text{ for } j \in \mathcal{R}, \sum_{k \in \mathcal{E}} \lambda_k = 1 \right\},$$

with $\{x^k\}_{k \in \mathcal{E}}$ the extreme points and $\{r^j\}_{j \in \mathcal{R}}$ the extreme rays.

- We can assume **wlog** that the extreme rays are integral.
- Then \mathcal{S} is finitely generated by $\mathcal{Q} \cap \mathcal{S}$ and the extreme rays of \mathcal{P} , where

$$\mathcal{Q} = \left\{ \sum_{k \in \mathcal{E}} \lambda_k x^k + \sum_{j \in \mathcal{R}} \mu_j r^j \mid \lambda_k \geq 0 \text{ for } k \in \mathcal{E}, 0 \leq \mu_j < 1 \text{ for } j \in \mathcal{R}, \sum_{k \in \mathcal{E}} \lambda_k = 1 \right\},$$

Example

- Let's find a finite set of generators for the set $\mathcal{S} = \mathcal{P} \cap \mathbb{Z}^2$, where

$$\mathcal{P} = \{x \in \mathbb{R}_+^2 \mid 5x_1 + 3x_2 \geq 10, 5x_1 - 5x_2 \geq -1, -x_1 + 2x_2 \geq -2\}$$

- The generators for \mathcal{S} are the set of integer points inside the set \mathcal{Q} defined previously.
- Set \mathcal{P} and its generator are shown in Figure 1 on the next slide.
- The set \mathcal{Q} is defined as

$$\mathcal{Q} = \{\lambda_1 e_1 + \lambda_2 e_2 + \mu_1 r_1 + \mu_2 r_2 \mid \lambda_1, \lambda_2 \in \mathbb{R}_+, \lambda_1 + \lambda_2 = 1, \mu_1, \mu_2 \in [0, 1)\}$$

- The generators for \mathcal{S} itself are then the points

$$\{(2, 0), (2, 1), (2, 2), (3, 1), (3, 2), \text{ and } (4, 1)\},$$

along with the extreme rays $(1, 1)$ and $(2, 1)$ of the recession cone.

- In this case, just the points $(2, 0), (2, 1), (2, 2)$ are a minimal set of generators, since the other points above can be generated by those.

Example

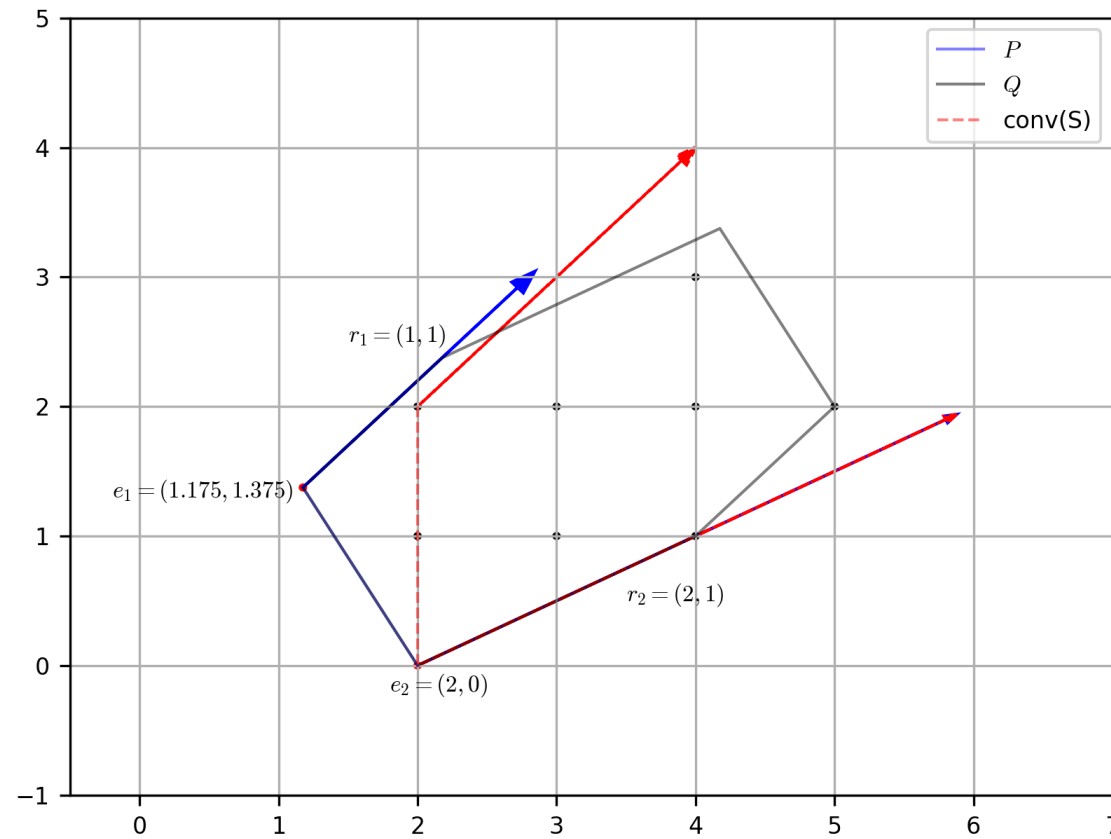


Figure 1: Generators for \mathcal{P} , the convex hull of \mathcal{S} , and \mathcal{Q} .

Consequences

- Once we have that S is finitely generated then we can easily show that $\text{conv}(S)$ is a rational polyhedron.
- Note that this result extends easily to the mixed case with rational data.
- Note also that if $\mathcal{P} \cap S \neq \emptyset$, then the extreme rays of \mathcal{P} and $\text{conv}(S)$ coincide.
- This also shows that solving the IP $\max\{cx \mid x \in S\}$ is essentially equivalent to solving the LP $\max\{cx \mid x \in \text{conv}(S)\}$.
 - The objective function of the IP is unbounded if and only if the objective function of the LP is unbounded.
 - If the LP has a bounded optimal value, then it has an optimal solution that is an optimal solution to the IP (an extreme point of $\text{conv}(S)$).
 - if \hat{x} is an optimal solution to IP, then it is an optimal solution to the LP.
- We can also show that an IP is either infeasible, unbounded, or has an optimal solution.

Implicitly Described Polyhedra

- $\text{conv}(S)$ is an “implicitly defined” polyhedron in the sense that we do not generally have a description of it in terms of half-spaces or generators.
- Knowing that $\text{conv}(S)$ is a polyhedron does not help much in obtaining an explicit description of it.
- It will, however, help in proving convergence of solution methods and in other important ways.
- In some case, we will try to generate parts of the description of this polyhedron.
- Not all the inequalities appearing in the formulation will be facet-defining for it.
- Using the properties of polyhedra that we know, we will try to determine which inequalities from the formulation are the facet-defining ones.
- We will also try to generate new valid inequalities that are facet-defining.
- Adding these to the formulation will necessarily increase its “strength.”