

$$P = \{x \in \mathbb{R}^n \mid A \cdot x = \delta, x \geq 0\} \quad A \in \mathbb{R}^{m \times n}, \text{rank}(A) = m$$

basic solution can be specified by choosing m linearly indep. columns of A : $(A_{\beta(1)}, \dots, A_{\beta(m)}) =: B$
 $x_B = \begin{pmatrix} x_{\beta(1)} \\ \vdots \\ x_{\beta(m)} \end{pmatrix} = B^{-1} \cdot \delta$, $x_j = 0$ for $j \notin \beta(1), \dots, \beta(m)$

$$A \cdot x = B \cdot x_B + \sum_{j \notin \beta(1), \dots, \beta(m)} A_j \cdot x_j = B \cdot B^{-1} \cdot \delta + 0 = \delta$$

$$\sum_{j=1}^n A_j \cdot x_j$$

2.5 Existence of extreme points

Def: A polyhedron P contains a line if there is a $x \in P$ and some direction $d \neq 0$ such that $x + \lambda \cdot d \in P \forall \lambda \in \mathbb{R}$.

Theorem: Let $P = \{x \in \mathbb{R}^n \mid Ax \geq \delta\} \neq \emptyset$. The following are equivalent:

- (i) There is an extreme point $x \in P$
- (ii) P does not contain a line
- (iii) matrix A contains n linearly independent rows.

Proof: (i) \Rightarrow (iii): Clear by def. of basic feasible solutions.

(iii) \Rightarrow (ii): Assume by contradiction that $x \in P$ and $x + \lambda \cdot d \in P \forall \lambda \in \mathbb{R}$ for some d

$$\Rightarrow A \cdot (x + \lambda \cdot d) = A \cdot x + \lambda \cdot A \cdot d \geq \delta \quad \forall \lambda \in \mathbb{R}$$

$$\Rightarrow A \cdot d = 0$$

Since $\text{rank}(A) = n \Rightarrow d = 0$

(ii) \Rightarrow (i): Let $x \in P$ such that $|\{i \mid a_i^T \cdot x = \delta_i\}|$ is maximum. Assume by contradiction that $\{a_i \mid a_i^T x = \delta_i\}$ does not contain n linearly indep. vectors.

$$\Rightarrow \exists d \in \mathbb{R}^n \setminus \{0\} \text{ such that } a_i^T \cdot d = 0 \quad \forall i \in I$$

$$\Rightarrow a_i^T \cdot (x + \lambda \cdot d) = \delta_i \quad \forall \lambda \in \mathbb{R}$$

Since $x + \lambda \cdot d$ is not contained in P for all $\lambda \in \mathbb{R}$, there is a constraint $a_j^T x \geq \delta_j$ that is eventually violated for some $\lambda \in \mathbb{R}$. $\rightarrow \exists \lambda_0 \in \mathbb{R}$ with

$$a_j^T (x + \lambda_0 \cdot d) = \delta_j \text{ and } x + \lambda_0 \cdot d \in P.$$

This contradicts the choice of x . \square

Corollary: Every non-empty bounded polyhedron and every non-empty polyhedron in standard form contain an extreme points.

$$\begin{aligned} Ax = \delta \\ I \cdot x \geq 0 \end{aligned} \iff \begin{pmatrix} A \\ -A \\ I \end{pmatrix} \cdot x \geq \begin{pmatrix} \delta \\ -\delta \\ 0 \end{pmatrix}$$

Example:

$$\left. \begin{aligned} x_1 + x_2 &\geq 1 \\ x_1 + 2x_2 &\geq 0 \end{aligned} \right\} P$$

x_1, x_2, x_3
contains a line
because

$$\begin{pmatrix} 1 \\ 0 \end{pmatrix} + \lambda \cdot \begin{pmatrix} 0 \\ 1 \end{pmatrix} \in P \quad \forall \lambda \in \mathbb{R}$$

2.6 Optimality of extreme points

Theorem: Let $P \subseteq \mathbb{R}^n$ polyhedron and $c \in \mathbb{R}^n$. If P has an extreme point and $c^T x$ is bounded from below for $x \in P$, then there exists an extreme point that is optimal.

Proof: We prove that for any $x \in P$ there exists an extreme point $y \in P$ with $c^T y \leq c^T x$.

Let $x \in P$ and $I := \{i \mid a_i^T x = \delta_i\}$. If $\{a_i \mid i \in I\}$ contains n linearly indep. vectors we are done. Otherwise $\exists d \in \mathbb{R}^n$ $d \neq 0$ with $a_i^T \cdot d = 0 \quad \forall i \in I$. Assume that $c^T \cdot d \leq 0$ (otherwise replace d by $-d$).

1. Case: $c^T \cdot d < 0 \Rightarrow c^T \cdot (x + \lambda \cdot d) \rightarrow -\infty$ for $\lambda \rightarrow \infty$

Thus $\exists j : a_j^T \cdot d < 0$. Let $\lambda_0 := \min_{j \notin I : a_j^T d < 0} \frac{b_j - a_j^T x}{a_j^T d}$

Remark: $a_j^T (x + \lambda_0 \cdot d) = a_j^T x + \lambda_0 \cdot a_j^T d \geq \delta_j$

Then $x + \lambda_0 \cdot d \in P$ and $\{i \mid a_i^T(x + \lambda_0 \cdot d) = b_i\} \supseteq \{i \mid a_i^T x = b_i\}$ contains more linearly independent vectors than $\{i \mid a_i^T x = b_i\}$.

Notice that $c^T x \geq c^T(x + \lambda_0 \cdot d)$. Repeat argument with $x + \lambda_0 \cdot d$.

2. Case: $c^T \cdot d = 0 \Rightarrow c^T(x + \lambda \cdot d) = c^T x \quad \forall \lambda \in \mathbb{R}$

Since P does not contain a line, we find an additional constraint in direction d or $-d$.

See above ...

□

Corollary: For any linear programming problem over a nonempty polyhedron, either the optimal cost is $-\infty$ or there exists an optimal solution.

Proof: Any linear program is equivalent to a linear program in standard form.

Example: $\min \frac{1}{x}$ does not have opt. sol.
s.t. $x > 0$

Chapter 3: The Simplex Method

Let $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $c \in \mathbb{R}^n$, $\text{rank}(A) = m$

Let $P = \{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$ and consider the LP

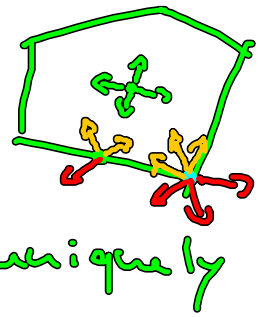
$$\begin{aligned} \min \quad & c^T x \\ \text{s.t.} \quad & Ax = b \\ & x \geq 0 \end{aligned}$$

3.1 Optimality conditions

Def: For $x \in P \subseteq \mathbb{R}^n$ (polyhedron), the vector $d \in \mathbb{R}^n \setminus \{0\}$ is a feasible direction at x if there exists $\theta > 0$

with $x \in \mathbb{R}^n$.

Observation: Let $B = (A_{B(1)}, \dots, A_{B(m)})$ be a basis matrix. Then the basic variables $x_{B(1)}, \dots, x_{B(m)}$ in the system $Ax = b$ are uniquely determined by the non-basic variables.



$$Ax = b \iff B \cdot x_B + \sum_{j \in B^c} A_j \cdot x_j = b$$

$$\iff x_B = B^{-1} \cdot b - \sum_{j \in B^c} B^{-1} \cdot A_j \cdot x_j$$