

## Simplex Algorithm: Geometric view

Linear optimization problem

$$\max\{c^T x \mid Ax \leq b, x \in \mathbb{R}^n\} \quad (\text{LP})$$

### Simplex-Algorithm (Dantzig 1947)

1. Find a vertex of  $P$ .
2. Proceed from vertex to vertex along edges of  $P$  such that the objective function  $z = c^T x$  increases.
3. Either a vertex will be reached that is optimal, or an edge will be chosen which goes off to infinity and along which  $z$  is unbounded.

## Basic solutions

- $Ax \leq b$ ,  $A \in \mathbb{R}^{m \times n}$ ,  $\text{rank}(A) = n$ .
- $M = \{1, \dots, m\}$  row indices,  $N = \{1, \dots, n\}$  column indices
- For  $I \subseteq M, J \subseteq N$  let  $A_{IJ}$  denote the submatrix of  $A$  defined by the rows in  $I$  and the columns in  $J$ .
- $I \subseteq M, |I| = n$  is called a *basis of  $A$*  iff  $A_{I*} = A_{IN}$  is non-singular.
- In this case,  $A_{I*}^{-1} b_I$ , where  $b_I$  is the subvector of  $b$  defined by the indices in  $I$ , is called a *basic solution*.
- If  $x = A_{I*}^{-1} b_I$  satisfies  $Ax \leq b$ , then  $x$  called a *basic feasible solution* and  $I$  is called a *feasible basis*.

## Algebraic characterization of vertices

### Theorem

For a non-empty polyhedron  $P = \{x \in \mathbb{R}^n \mid Ax \leq b\}$  the following holds:

1.  $P$  has at least one vertex if and only if  $\text{rank}(A) = n$ .
2. A vector  $v \in \mathbb{R}^n$  is a vertex of  $P$  if and only if it is a basic feasible solution of  $Ax \leq b$ , for some basis  $I$ .
3. If  $\text{rank}(A) = n$ , then for any  $c \in \mathbb{R}^n$ , either the maximum value of  $z = c^T x$  for  $x \in P$  is attained at a vertex of  $P$  or  $z$  is unbounded on  $P$ .

### Remark

It follows from (2) that a polyhedron has at most finitely many vertices.

In general, a vertex may be defined by several bases.

## Simplex Algorithm: Algebraic version

- Suppose  $\text{rank}(A) = n$  (otherwise apply Gaussian elimination).
- Suppose  $I$  is a feasible basis with corresponding vertex  $v = A_{I*}^{-1} b_I$ .
- Compute  $u^T \stackrel{\text{def}}{=} c^T A_{I*}^{-1}$  (vector of  $n$  components indexed by  $I$ ).
- If  $u \geq 0$ , then  $v$  is an optimal solution, because for each feasible solution  $x$

$$c^T x = u^T A_{I*} x \leq u^T b_I = u^T A_{I*} v = c^T v.$$

- If  $u \not\geq 0$ , choose  $i \in I$  such that  $u_i < 0$  and define the direction  $d \stackrel{\text{def}}{=} -A_{i*}^{-1} e_i$ , where  $e_i$  is the  $i$ -th unit basis vector in  $\mathbb{R}^I$ .
- Next increase the objective function value by going from  $v$  in direction  $d$ , while maintaining feasibility.

### Simplex Algorithm: Algebraic version <sup>(2)</sup>

1. If  $Ad \not\leq 0$ , the largest  $\lambda \geq 0$  for which  $v + \lambda d$  is still feasible is

$$\lambda^* = \min \left\{ \frac{b_p - A_{p*} v}{A_{p*} d} \mid p \in \{1, \dots, m\}, A_{p*} d > 0 \right\}. \quad (\text{PIV})$$

Let this minimum be attained at index  $k$ . Then  $k \notin I$  because  $A_{i*} d = -e_i \leq 0$ .

Define  $I' = (I \setminus \{i\}) \cup \{k\}$ , which corresponds to the vertex  $v + \lambda^* d$ .

Replace  $I$  by  $I'$  and repeat the iteration.

2. If  $Ad \leq 0$ , then  $v + \lambda d$  is feasible, for all  $\lambda \geq 0$ . Moreover,

$$c^T d = -c^T A_{i*}^{-1} e_i = -u^T e_i = -u_i > 0.$$

Thus the objective function can be increased along  $d$  to infinity and the problem is unbounded.

### Termination and complexity

- The method terminates if the indices  $i$  and  $k$  are chosen in the right way (such choices are called *pivoting rules*).
- Following the rule of Bland, one can choose the smallest  $i$  such that  $u_i < 0$  and the smallest  $k$  attaining the minimum in (PIV).
- For most known pivoting rules, sequences of examples have been constructed such that the number of iterations is exponential in  $m+n$  (e.g. Klee-Minty cubes).
- Although no pivoting rule is known to yield a polynomial time algorithm, the Simplex method turns out to work very well in practice.

### Simplex : Phase I

- In order to find an *initial feasible basis*, consider the auxiliary linear program

$$\max \{y \mid Ax - by \leq 0, -y \leq 0, y \leq 1\}, \quad (\text{Aux})$$

where  $y$  is a new variable.

- Given an arbitrary basis  $K$  of  $A$ , obtain a feasible basis  $I$  for (Aux) by choosing  $I = K \cup \{m+1\}$ . The corresponding basic feasible solution is 0.
- Apply the Simplex method to (Aux). If the optimum value is 0, then (LP) is infeasible. Otherwise, the optimum value has to be 1.
- If  $I'$  is the final feasible basis of (Aux), then  $K' = I' \setminus \{m+2\}$  can be used as an initial feasible basis for (LP).