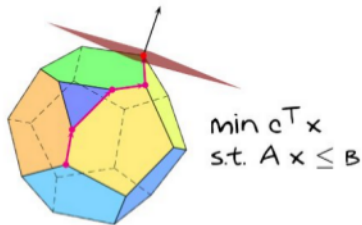


## The simplex method

- ▶ Bases and degeneracy
- ▶ Moving to a better neighbor



# Bases

A subset  $B \subseteq \{1, \dots, m\}$  of the row-indices with  $|B| = n$  and  $A_B$  non-singular is called *basis* of the LP.

If in addition  $A_B^{-1}b_B$  is feasible, then  $B$  is called *feasible basis*.

$x^* \in P = \{x \in \mathbb{R}^n : Ax \leq b\}$  is vertex  $\Leftrightarrow \exists \underline{B} \subseteq \{1, \dots, m\}$

s.t.  $|B| = n$ ,  $A_B$  non-singular

and  $x^* = A_B^{-1} \cdot b_B$

$$\boxed{\begin{array}{l} \text{MAX } C^T \cdot x \\ x \in P \end{array}}$$

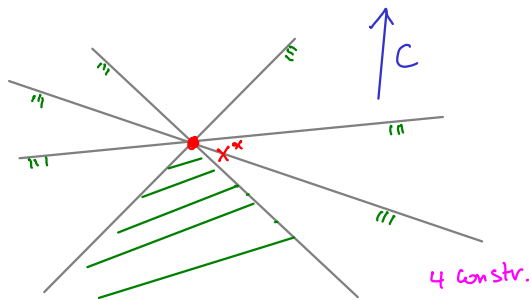
↑  
LP

# Vertices and bases

A vertex  $x^* \in P$  is represented by a basis  $B$ .

$$x^* = A_B^{-1} \cdot b_B$$

A vertex  $x^*$  can be represented by several bases.



$$\begin{aligned} \max & c^T \cdot x \\ & Ax \leq b \\ & x \in \mathbb{R}^2 \end{aligned}$$

4 constr.

$$\binom{4}{2} = 6$$

Quiz:

How many bases represent  $x^*$ ?

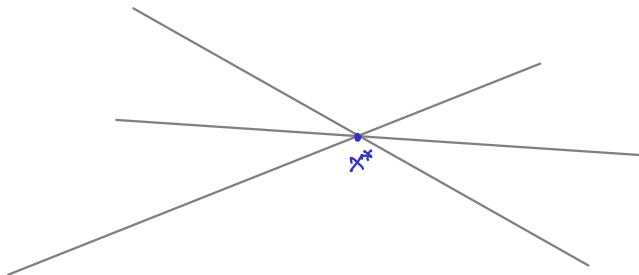
☐ 2

☐ 4

☒ 6

# Degeneracy

A linear program  $\max\{c^T x : x \in \mathbb{R}^n, Ax \leq b\}$  is *degenerate* if there exists an  $x^* \in \mathbb{R}^n$  such that there are more than  $n$  constraints of  $Ax \leq b$  that are active at  $x^*$ .



$$n = 2$$

3 constraints  
active at  $x^*$

# Simplex algorithm

George Dantzig (1914 - 2005)

*Basic idea:*

Start with vertex  $x^*$

while  $x^*$  is not optimal

Find vertex  $x'$  adjacent to  $x^*$  with  $c^T x' > c^T x^*$

update  $x^* := x'$

Or assert that LP is unbounded.

$$\begin{aligned} \max \quad & c^T x \\ \text{s.t.} \quad & Ax \leq b \\ & x \in \mathbb{R}^n \end{aligned}$$

$$P = \{x \in \mathbb{R}^n : Ax \leq b\}$$

# Optimal bases

A basis  $B$  is called *optimal* if it is feasible and the unique  $\hat{J} \in \mathbb{R}^m$  with

$$\hat{J}^T A = c^T \quad \text{and} \quad \hat{J}_i = 0, i \notin B$$

$$\lambda_B^T \underline{A_B} = c^T$$

$$\lambda_B^T = c^T \cdot A_B^{-1}$$

satisfies  $\hat{J} \geq 0$ .

## Theorem

If  $B$  is optimal basis, then  $x^* = A_B^{-1} b_B$  is optimal solution of LP.

$$\left. \begin{array}{l} a_1^T x \leq b_1 \quad (*\lambda_1) \\ \vdots \\ a_m^T x \leq b_m \quad (*\lambda_m) \end{array} \right\} \begin{array}{l} \text{valid ineq.} \\ \Rightarrow \end{array}$$

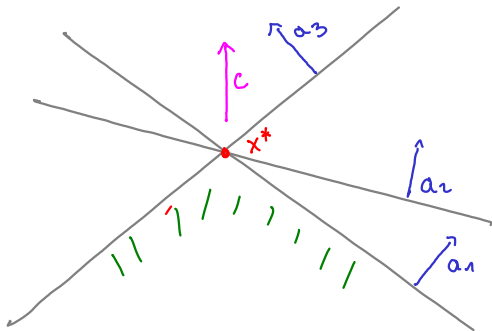
$$\begin{aligned} \lambda^T \cdot A \cdot x &\leq \lambda^T \cdot b = \lambda_B^T \cdot \underbrace{b_B}_{= A_B \cdot x^*} = \underbrace{\lambda_B^T \cdot A_B}_{= C^T} \cdot x^* \\ &= \underbrace{\lambda_B^T A_B}_{C^T} \cdot x^* \end{aligned}$$

$$C^T \cdot x \leq C^T \cdot x^* \text{ valid ineq.}$$

$\Rightarrow x^*$  optimal.

# Quiz

Which bases are optimal?



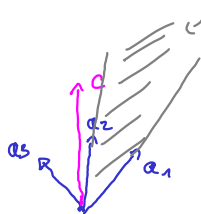
☐  $\{1,2,3\}$

☒  $\{1,3\}$

☒  $\{2,3\}$

non-neg. linear  
comb. of  $a_2$  and  $a_1$

$\nabla c \neq$



$$\lambda_{\{1,2\}}^T \cdot A_{\{1,2\}} = c^T$$

$$\lambda_{\{1,2\}} \neq 0$$

# The non-degenerate case

$$\max c^T x, \quad Ax \leq b \quad \begin{cases} A_B x \leq b_B & \text{active at } x^* \\ A_{\bar{B}} x \leq b_{\bar{B}} & \text{inactive.} \end{cases}$$

## Theorem

Suppose the LP is non-degenerate and  $B$  is a feasible but not optimal basis, then  $x^* = A_B^{-1} b_B$  is not an optimal solution.

$$\lambda^T A = c^T, \quad \lambda_j = 0 \quad \forall j \notin B, \quad \lambda_i < 0 \text{ for some } i \in B.$$

$$\text{Compute } d \in \mathbb{R}^n \text{ s.t. } A_B \lambda_i a_i \cdot d = 0, \quad a_i^T \cdot d = -1 \quad (A_B \text{ non-sing})$$

$$c^T \cdot d = \lambda_B^T \cdot A_B \cdot d = \underbrace{\lambda_i}_{< 0} \cdot \underbrace{a_i^T \cdot d}_{=-1}$$

$$\begin{matrix} > 0 \\ < 0 \\ \leq 0 \end{matrix}$$



$$\text{For } \underline{\varepsilon} > 0 \quad \begin{matrix} A_B(x^* + \varepsilon \cdot d) \\ b_B + \varepsilon \cdot \begin{pmatrix} 0 \\ \vdots \\ -1 \\ \vdots \\ 0 \end{pmatrix} \leftarrow i \end{matrix} \begin{matrix} \leq \\ = \\ \neq \end{matrix} b_B$$

$$\exists \varepsilon^* > 0 \text{ s.t.}$$

$$x^* + \varepsilon^* \cdot d \text{ is feasible.}$$

$$c^T(x^* + \varepsilon^* \cdot d) = c^T x^* + \varepsilon^* \cdot \underbrace{c^T \cdot d}_{> 0}$$

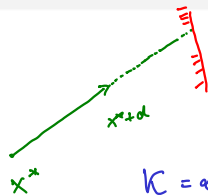




## Moving to a better neighbor

LP NON-DEGR.

- ▶  $B$  not an optimal basis
- ▶  $x^* = A_B^{-1} b_B$  corresponding basic feasible solution
- ▶  $\bar{a}_i < 0$  for some  $i \in B$
- ▶  $a_j^T d = 0, j \in B \setminus \{i\}$   
 $a_i^T d = -1$
- ▶  $c^T d > 0$
- ▶ there exists  $\varepsilon > 0$  such that  $x^* + \varepsilon d$  feasible



$$\begin{aligned} a_1^T x &\leq b_1 \\ &\vdots \\ a_m^T x &\leq b_m \end{aligned}$$

$$K = \{d : 1 \leq k \leq m, a_k^T d > 0\}$$

CASE 1:  $K = \emptyset$

LP UNBOUNDED

Question: How large can  $\varepsilon$  be?

# Simplex algorithm

George Dantzig (1914 - 2005)

*Basic idea:*

Start with vertex  $x^*$

while  $x^*$  is not optimal

Find vertex  $x'$  adjacent to  $x^*$  with  $c^T x' > c^T x^*$

update  $x^* := x'$

Or assert that LP is unbounded.

$$\begin{aligned} \max \quad & c^T \cdot x \\ \text{s.t.} \quad & Ax \leq b \\ & x \in \mathbb{R}^n \end{aligned}$$

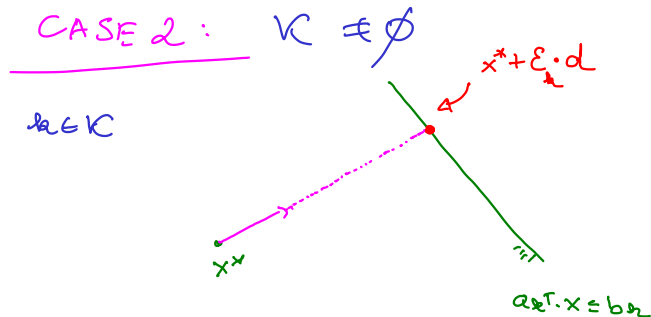
$$P = \{x \in \mathbb{R}^n : Ax \leq b\}$$

$$K = \emptyset$$

## Moving to a better neighbor

- ▶  $B$  not an optimal basis
- ▶  $x^* = A_B^{-1} b_B$  corresponding basic feasible solution
- ▶  $\bar{a}_i < 0$  for some  $i \in B$
- ▶  $a_j^T d = 0, j \in B \setminus \{i\}$   
 $a_i^T d = -1$
- ▶  $c^T d > 0$
- ▶ there exists  $\varepsilon > 0$  such that  $x^* + \varepsilon d$  feasible

Question: How large can  $\varepsilon$  be?



$$a_k^T(x^* + \varepsilon_k \cdot d) = b_k$$
$$\Leftrightarrow \varepsilon_k = \frac{b_k - a_k^T \cdot x^*}{\underbrace{a_k^T \cdot d}_{>0}}$$
$$\varepsilon^* = \min_{k \in K} \varepsilon_k$$

$k^* \in K$  Index where min is attained.

$x' = x^* + \varepsilon^* \cdot d$ ,  $B' = B \setminus \{i\} \cup \{k^*\}$  is BASIS

## Moving to a better neighbor

- ▶  $B$  not an optimal basis
- ▶  $x^* = A_B^{-1} b_B$  corresponding basic feasible solution
- ▶  $\bar{a}_i < 0$  for some  $i \in B$
- ▶  $a_j^T d = 0, j \in B \setminus \{i\}$   
 $a_i^T d = -1$
- ▶  $c^T d > 0$
- ▶ there exists  $\varepsilon > 0$  such that  $x^* + \varepsilon d$  feasible

**Question:** How large can  $\varepsilon$  be?

$$B' = B \setminus \{i\} \cup \{k^*\} \text{ BASIS.}$$

$$d \perp a_j, j \in B \setminus \{i\}$$

$$d \not\perp a_{k^*}, d^T a_{k^*} > 0$$

$\Rightarrow Q_{k^*}$  IS NOT A LINEAR COMB. OF THE  $a_j, j \in B \setminus \{i\}$ .



The ineq.  $a_j^T \cdot x \leq b_j, j \in B'$  are active at  $x'$ .

$\Rightarrow x'$  IS A VERTEX, ADJACENT TO  $x^*$

# Simplex algorithm

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while  $x^*$  is not optimal

Find vertex  $x'$  adjacent to  $x^*$  with  $c^T x' > c^T x^*$

update  $x^* := x'$

Or assert that LP is unbounded.

$$\max c^T x$$

$$Ax \leq b$$

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

$$P = \{x \in \mathbb{R}^n : Ax \leq b\}$$

$$K \neq \emptyset$$

$$K = \emptyset$$