

Theorem: (important for the exam)

A point x of a polyhedron P ($x \in P = \{x \geq 0 \mid Ax = b\} \neq \emptyset$) is a vertex of $P \iff x$ is a basic feasible solution (bfs) associated to the system $Ax = b$

\hookrightarrow There's a 1-1 correspondence between vertices and bfs (ignoring degenerate solutions)

Proof:

- " x is a bfs $\Rightarrow x$ is a vertex of P "

$$x = [x_1, \dots, x_m, 0, \dots, 0]^T$$

\Downarrow

$$x = [x_1, \dots, x_k, 0, \dots, 0]^T$$

$k \in [0, m]$ since x is a bfs and we also must consider degenerate cases: arrange the elements of x s.t. $x_1, \dots, x_k > 0$ and $x_{k+1}, \dots, x_n = 0$

Let's assume that x is not a vertex of P :

$$\exists y, z \in P, y \neq z, \exists \lambda \in]0, 1[\mid x = \lambda y + (1-\lambda)z$$

Since x is a convex combination of y and z and $y_i, z_i > 0 \forall i=1, \dots, n$, y and z must have the same form of x :

$$y = [y_1, \dots, y_n, 0, \dots, 0]^T$$

$$z = [z_1, \dots, z_n, 0, \dots, 0]^T$$

Since $y, z \in P$:

$\hookrightarrow A$'s columns are ordered following x 's elements re-arrangement

$$A_1 y_1 + \dots + A_k y_k + A_{k+1} 0 + \dots = b \quad (\text{I})$$

$$A_1 z_1 + \dots + A_k z_k + \dots = b \quad (\text{II})$$

$$\text{Let's now consider } (\text{I}) - (\text{II}) = A_1 (y_1 - z_1) + \dots + A_k (y_k - z_k) \underset{=0}{=} b - b$$

$$\text{let } \alpha_i = y_i - z_i \quad \forall i = 1, \dots, k$$

$$\text{Since } y \neq z, \exists i, i = 1, \dots, k \mid \alpha_i \neq 0 \Rightarrow (\text{I}) - (\text{II}) \neq 0$$

This means that A_1, \dots, A_k must be linear dependent, which is impossible since $A_1, \dots, A_k \in B$ (\Rightarrow they must be lin. independent)

Our only assumption was that x was not a vertex, hence this assumption must be wrong $\Rightarrow x$ is a vertex.

- "x is a vertex of $P \Rightarrow x$ is a bfs"

$$x = [x_1, x_2, \dots, x_n] \in P$$

\Downarrow

$$x = [x_1, \dots, x_k, 0, \dots] \in P$$

re-arrange the elements of x
s.t. $x_1, \dots, x_k > 0, x_{k+1}, \dots, x_n = 0$,
with $k \in [0, n]$

Since $x \in P \Rightarrow A_1 x_1 + \dots + A_k x_k = b$ (I) → A's columns are ordered following x's elements re-arrangement

We now have 2 possible cases:

- A_1, \dots, A_k are linear independent $\rightarrow k \leq m$

$$x = [\underbrace{x_1, \dots, x_k}_m, 0, \dots, 0, \dots]$$

$$A = [\underbrace{A_1 \dots A_k \dots A_m}_{B} \dots]$$

Arbitrarily pick the A_{k+1}, \dots, A_m columns st. A_1, \dots, A_m are lin. independent

$$Ax = b \rightarrow [B | \dots] x = b \rightarrow x = \begin{bmatrix} B^{-1}b \\ 0 \end{bmatrix} \text{ (bfs)}$$

- A_1, \dots, A_k are linear dependent (no limitations on k)

consider $\alpha_1, \dots, \alpha_k \in \mathbb{R}^m$ s.t. $A\alpha = 0$ (II)

given an $\varepsilon \geq 0$:

$$(I) + \varepsilon(II) = (x_1 + \varepsilon\alpha_1)A_1 + \dots + (x_k + \varepsilon\alpha_k)A_k = b$$

$$(I) - \varepsilon(II) = (x_1 - \varepsilon\alpha_1)A_1 + \dots + (x_k - \varepsilon\alpha_k)A_k = b$$

$$\text{let } x_i + \varepsilon\alpha_i = y_i, x_i - \varepsilon\alpha_i = z_i, \forall i = 1, \dots, k$$

With ε arbitrarily small, $y = [y_1, \dots, y_k, 0, \dots]^T$ and $z = [z_1, \dots, z_k, 0, \dots]^T$ have all positive values ($y_i, z_i > 0 \forall i = 1, \dots, k$)

We can rewrite the equations as follows

$$A_1 y = A_2 z = b$$

Since not all α_i are 0 and $\epsilon \neq 0$, $y \neq z$, also x can be written as the convex combination of y and z

$$\begin{cases} y = x + \epsilon \alpha \\ z = x - \epsilon \alpha \end{cases} \rightarrow x = \frac{1}{2} y + \frac{1}{2} z$$

This means x can't be a vertex, which is impossible
Hence A_1, \dots, A_k must be linear independent

As we've stated before, getting all the vertices would require computing $\binom{n}{m}$ bfs

Solution:

- 1) Choose a random vertex and check its optimality
- 2) If it's not an optimal solution go to a neighbor and check its optimality
- 3) Repeat step 2 till we find an optimal solution

Optimality test

Given an optimisation problem $\begin{cases} Ax = b \\ x \geq 0 \end{cases}$ and a basis B :

$$Ax = [B \ F] \begin{bmatrix} x_B \\ x_F \end{bmatrix} = Bx_B + Fx_F = b$$

$$x_B = B^{-1}b - B^{-1}Fx_F$$

the cost function can be written as follows:

$$\begin{aligned} c^T x &= [c_B^T \ c_F^T] \begin{bmatrix} x_B \\ x_F \end{bmatrix} \\ &= c_B^T x_B + c_F^T x_F \\ &= c_B^T (B^{-1}b - B^{-1}Fx_F) + c_F^T x_F \\ &= \underbrace{c_B^T B^{-1}b}_{c_0} + \underbrace{(c_F^T - c_B^T B^{-1}F)}_{\bar{c}_F^T} x_F \end{aligned}$$

$\bar{c}_F^T \rightarrow$ reduced cost vector

if we're on a vertex, $x_F = 0 \Rightarrow c^T x = c_0$

if $\bar{c}_F \geq 0$ then $c^T x \geq c_0$, then the vertex obtained by B is the optimal solution

$$\hookrightarrow \bar{c}_{F_i} \geq 0 \quad \forall i$$

\hookrightarrow no need to check more vertices