

Recall:

primal LP

$$\min c^T x$$

$$\text{s.t. } Ax = b$$

$$x \geq 0$$

dual LP

$$\max b^T y$$

$$\text{s.t. } A^T y \leq c$$

optimality cito:

$$Ax = b$$

$$A^T y + z = c$$

$$z \geq 0, y \geq 0$$

$$x^T z = 0$$

$$F = \{x \in \mathbb{R}^n \mid Ax = b, x \geq 0\}$$

= polyhedron in std form

x is a vertex if:

$$x = \lambda x_1 + (1-\lambda) x_2, \quad \lambda \in (0,1), \quad x_1, x_2 \in F$$

$$\Rightarrow x_1 = x_2 = x$$

Theorem Let F be a polyhedron in std form then

$x \in F$ is a vertex iff columns A_i of A for which $x_i > 0$ are lin indep.

proof (here only \Leftarrow part)

Let $x \in F$, $I = \{i \mid x_i > 0\}$, $A_i, i \in I$ lin indep

Let $x^1, x^2 \in F$ and $\lambda \in (0,1)$ st.

$$x = \lambda x^1 + (1-\lambda) x^2$$

Since $x^1, x^2 \geq 0$ and $x_i = 0$ for $i \notin I \Rightarrow$ here:

$$x_i = x_i^1 = 0 \quad \text{for } i \notin I \quad \text{lin indep}$$

Thus:

$$0 = b - b = Ax^1 - Ax^2 = \sum_{i \in I} A_i (x_i^1 - x_i^2) \Rightarrow x_i^1 = x_i^2, i \in I$$

$$\Rightarrow x^1 = x^2.$$

Def Let F be a polyhedron in std form. We say $x \in F$ is a basic feasible point if there is some index set $B \subset \{1 \dots n\}$ with $|B| = m$ and $x_j = 0$ for $j \notin B$ and $A_i : i \in B$ are lin indep

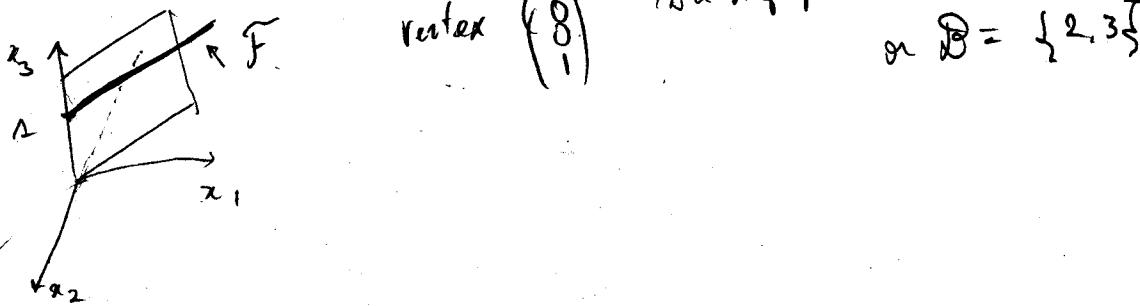
Theorem: Let F be a polyhedron in std form s.t.

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

x is a vertex of $F \Leftrightarrow x$ is a b.f.p.

Note: representations of a vertex as a b.f.p. do not unique. We can have degenerate b.f.p.:

$$A = \begin{pmatrix} 1 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad b = \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \quad x_1, x_2, x_3 \geq 0.$$



Theorem: F is a polyhedron in std form. $A \in \mathbb{R}^{m \times n}, \text{rank}(A) = m$

then:

- $F \neq \emptyset \Rightarrow F$ contains a b.f.p.
- F has finitely many b.f.p.
- If $\min_{x \geq 0} c^T x$ has a solution, then one of the sol is b.f.p.

Simplex Method:

Let x be a b.f.p. and let $B \subset \{1..n\}$ be the concp. index set.

$$\text{let } I = B^C = \{1..n\} \setminus B$$

Since $x_I = 0$

$$Ax = A_B x_B + A_I x_I = A_B x_B = b \Rightarrow x_B = A_B^{-1}b$$

How do we check x is a minimum? Use optimality conditions:

$$\begin{pmatrix} A_B^T \\ A_I^T \end{pmatrix} y + \begin{pmatrix} \bar{z}_B \\ \bar{z}_I \end{pmatrix} = \begin{pmatrix} c_B \\ c_I \end{pmatrix} \quad (\forall L = 0)$$

$$x^T \bar{z} = x_B^T \bar{z}_B = 0 \quad (\text{Compl})$$

Setting $\bar{z}_B = 0$ to satisfy compl. cdto we get:

$$\begin{cases} y = A_B^{-T} c_B \\ \bar{z}_I = c_I - A_I^T y \end{cases}$$

If $\bar{z}_I \geq 0$, then x_I is optimal (since all optimality cond hold)

What to do if current point is not a min?

then $\exists q \in I$ s.t. $\bar{z}_q < 0$

Let Δ_B be s.t.

$$A_B \Delta_B = -A_q$$

and Δ_I be s.t.

$$\delta_i = \begin{cases} 1 & \text{if } i = q \\ 0 & \text{if } i \in I \setminus \{q\} \end{cases}$$

then:

$$A \begin{pmatrix} \Delta_B \\ \Delta_I \end{pmatrix} = A_B \Delta_B + A_I \Delta_I = 0$$

\overbrace{s}^j step.

Now let's see what happens with function values

$$\begin{aligned} c^T(x + ts) &= c_B^T x_B + t c_B^T s_B + t c_q \\ &= c_B^T x_B - t c_B^T A_B^{-1} A_q + t c_q \end{aligned}$$

Recall $\gamma_x = c_I - A_I^T A_B^{-1} c_B$

$$\gamma_q = c_q - A_q^T A_B^{-1} c_B < 0 \quad (\text{assumed})$$

$$\Rightarrow c^T(x + ts) = c_B^T x_B + t \underline{\gamma_q} \leq c_B^T x_B = c^T x$$

So this step makes function values decrease.

We need to ensure $x + ts \geq 0$, take largest t for which

$$x_B + ts_B \geq 0$$

- if $s_B \geq 0$, then LP is unbounded and does not have a sol since:
 $x_B + ts_B \geq 0 \forall t$, and thus $t \gamma_q \rightarrow -\infty$.

- otherwise there must be some $s_i < 0 : i \in \mathbb{Q}$:

$$t = \min_{\substack{i \in \mathbb{Q} \\ s_i < 0}} -\frac{x_i}{s_i} \quad (*)$$

New iterate:

$$x_{\text{new}} = x + t s$$

Let $r \in \mathbb{B}$ be index for which min in (*) is taken:

$$-\frac{x_r}{s_r} = \min_{\substack{i \in \mathbb{Q} \\ s_i < 0}} -\frac{x_i}{s_i}$$

$$B_{\text{new}} = B \cup \{q\} \setminus \{r\}$$

was zero, becomes non zero

was non zero, becomes zero

Claim: x_{new} is a b.f.p. with corresp. idx set \mathcal{B}_{new} . (73)

first: $(x_{\text{new}})_i = 0 \quad i \notin \mathcal{B}_{\text{new}} \quad (\text{by construction})$

$$0 = \sum_{i \in \mathcal{B}_{\text{new}}} A_i x_i$$

$$= \sum_{\substack{i \in \mathcal{B} \\ i \neq r}} A_i x_i + \cancel{x_q A_q} = -A_B \delta_B$$

$$= \sum_{\substack{i \in \mathcal{B} \\ i \neq r}} A_i (x_i - \cancel{x_q}) - \cancel{x_q} \Delta^r A_r$$

Using that $A_i, i \in \mathcal{B}$ are lin indep: $\begin{cases} x_i - \cancel{x_q} = 0, & i \in \mathcal{B}, i \neq r \\ \cancel{x_q} \Delta^r = 0 \end{cases}$

$$\Rightarrow \begin{cases} x_i = 0, & i \in \mathcal{B}, i \neq r \\ x_q = 0 & \text{(since index } r \text{ is s.t. } \Delta^r < 0) \end{cases}$$

$\Rightarrow x_i = 0, i \in \mathcal{B}_{\text{new}}. \quad \text{QED.}$

Simplex ALGO: Given bfp $x, \mathcal{B} \subset \{1..n\}$

1. $y = A_B^{-T} C_B$

$$z_I = c_I - A_I^T \underbrace{A_B^{-T} C_B}_y$$

2. If $z_I \geq 0$, stop. optimal sol. x found.

3. Select $q \in I$ s.t. $z_q < 0$

4. Compute descent dir: $\Delta_B = -A_B^{-1} A_q$

$$\rightarrow I: \Delta_i = \begin{cases} 1 & \text{if } i = q \\ 0 & \text{if } i \in I \setminus \{q\} \end{cases}$$

5. If $\Delta_i \geq 0, i \in \mathcal{B}$. STOP LP has unbd opt fun (no sol)

6. Compute $r \in \mathcal{B}$ s.t.

$$-\frac{x_r}{\Delta_r} = \min_{i \in \mathcal{B}} -\frac{x_i}{\Delta_i}$$

7. $x \leftarrow x - \frac{x_r}{\Delta_r} \Delta; \mathcal{B} \leftarrow (\mathcal{B} \cup \{q\}) \setminus \{r\}$