

① Finding a starting BFS

② Entering variable optimality condition
 non basic variable \rightarrow basic variable
 current 0

③ Leaving variable feasible condition
 basic variable \rightarrow non basic
 ≥ 0 $= a$

④ move from ~~one~~ BFS \rightarrow next BFS
 current
 (Gaussian Elimination)

Assume that x_j is entering.

x_{B_r} to be leaving which r.

$$\bar{Y} = B^{-1}A \Leftrightarrow A = B \bar{Y}$$

$m \times n$ $m \times m$ $m \times n$
 \uparrow

$B =$ basic matrix
 $= [b_1, \dots, b_m]$
 $= [\vec{a}_1, \dots, \vec{a}_m]$

feasibility test

$$[\bar{y}_1, \dots, \bar{y}_j, \dots, \bar{y}_n]$$

\uparrow
is the entering

$$\bar{y}_j = \begin{pmatrix} y_{1j} \\ \vdots \\ y_{mj} \end{pmatrix}$$

ensure that the next solution is feasible

current basic solution

$\min_{1 \leq i \leq m} \left\{ \frac{b_i}{y_{ij}} \right\} = y_{ij} > 0$

\downarrow r the

$y_{rj} > 0$ leaving variable

$$\max f(\vec{x}) = \vec{c}^T \vec{x} \leftarrow \text{linear}$$

$$\text{s.t. } \begin{cases} A\vec{x} = \vec{b} \\ \vec{x} \geq 0 \end{cases}$$

$$f(\vec{x}) = f(\vec{x}_0) + \nabla f(\vec{x}_0)(\vec{x} - \vec{x}_0) + \text{h.o.t.}$$

$$\Leftrightarrow f(\vec{x}) = f(\vec{x}_0) + \boxed{\nabla f(\vec{x}_0)(\vec{x} - \vec{x}_0)} \quad \# \quad (\because \text{linear})$$

$$f(\vec{x}) = f(\vec{x}_0) + \vec{c}^T (\vec{x} - \vec{x}_0)$$

$$\vec{c}^T \vec{x} = \vec{c}^T \vec{x}_0 + \underbrace{\vec{c}^T (\vec{x} - \vec{x}_0)}_{is > 0}$$



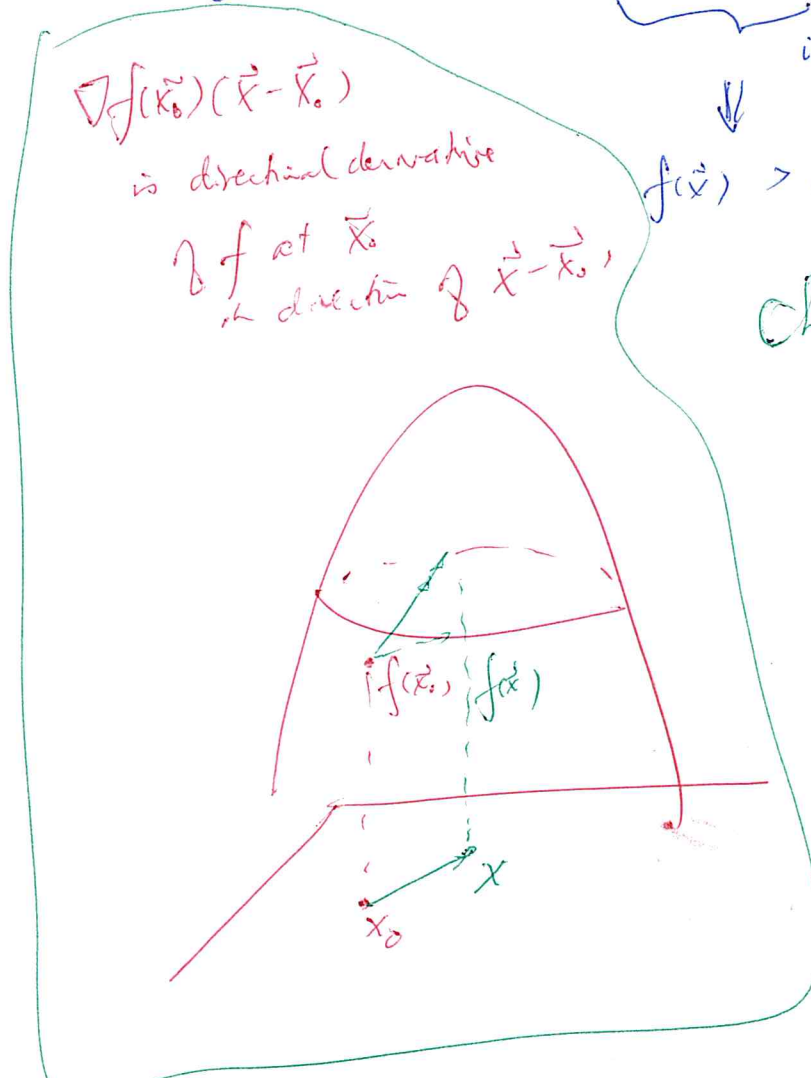
$$f(\vec{x}) > f(\vec{x}_0)$$

Choose \vec{x} st.

$$\nabla f(\vec{x}_0)(\vec{x} - \vec{x}_0) > 0$$

$$\Leftrightarrow \vec{c}^T (\vec{x} - \vec{x}_0) > 0$$

next BFS current BFS



$$\nabla f(\vec{x}_0)(\vec{x} - \vec{x}_0)$$

is directional derivative

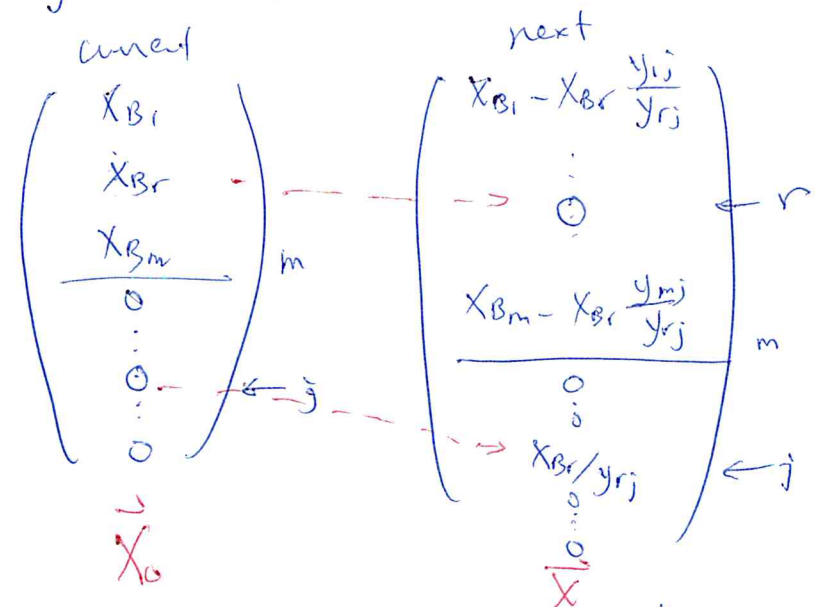
∂f at \vec{x}_0

in direction of $\vec{x} - \vec{x}_0$

current

$$A = [B | R] \quad \vec{x} = \begin{pmatrix} \vec{x}_B \\ \vec{0} \end{pmatrix}$$

x_j is entering & x_{Br} is leaving



durchzeit
durch

$$\vec{c}^T (\vec{x} - \vec{x}_0) = \vec{c}^T \begin{pmatrix} -x_{Br} \frac{y_{1j}}{y_{rj}} \\ \vdots \\ -x_{Br} \\ -x_{Br} \frac{y_{mj}}{y_{rj}} \\ \vdots \\ x_{Br} \frac{1}{y_{rj}} \end{pmatrix} = (c_{B_1}, \dots, c_{B_m}, \dots, -c_j, \dots, 0)$$

$$= \sum_{\substack{i=1 \\ i \neq r}}^m c_{B_i} \left(-x_{Br} \frac{y_{ij}}{y_{rj}} \right) - c_{B_r} \cdot x_{Br} \left(\frac{y_{rj}}{y_{rj}} \right) + c_j \cdot \frac{x_{Br}}{y_{rj}}$$

$$= - \sum_{i=1}^m c_{B_i} x_{Br} \left(\frac{y_{ij}}{y_{rj}} \right) + c_j \left(\frac{x_{Br}}{y_{rj}} \right)$$

$$= \frac{x_{Br}}{y_{rj}} \left\{ c_j - \sum_{i=1}^m c_{B_i} \frac{y_{ij}}{y_{rj}} \right\} = \frac{x_{Br}}{y_{rj}} [c_j - z_j]$$

$$f(\vec{x}) = f(\vec{x}_0) + \text{d.d on } x_j \text{ direction}$$

$$= f(\vec{x}_0) + \frac{x_j - x_{0j}}{y_{rj}} \{c_j - z_j\}$$

↑
Current
Objective
value

0 (feasibility condition)
in the next
step.

Optimality Condition

Choose j (x_j) st $c_j - z_j > 0$

$$z_j = \sum_{i=1}^m c_{B_i} y_{ij}$$

$$\Leftrightarrow (z_1, \dots, z_n) = (c_{B_1}, \dots, c_{B_m}) [Y]$$

$(\vec{c}^T - \vec{z}^T)$ choose j st the entry is > 0 .

$$\vec{z}^T = \vec{c}_B^T Y$$

Since \mathbf{x} is a feasible solution, $A\mathbf{x} = \mathbf{b}$. Thus by (2.24),

$$\mathbf{b} = \sum_{j=1}^n x_j \mathbf{a}_j = \sum_{j=1}^n x_j (B\mathbf{y}_j) = \sum_{j=1}^n \left(\sum_{i=1}^m y_{ij} \mathbf{b}_i \right) x_j = \sum_{i=1}^m \left(\sum_{j=1}^n y_{ij} x_j \right) \mathbf{b}_i = \sum_{i=1}^m \tilde{x}_i \mathbf{b}_i = B\tilde{\mathbf{x}}.$$

Since B is non-singular and we already have $B\mathbf{x}_B = \mathbf{b}$, it follows that $\tilde{\mathbf{x}} = \mathbf{x}_B$. Thus by (2.27),

$$z \leq \sum_{i=1}^m x_{B_i} c_{B_i} = z_0$$

for all \mathbf{x} in the feasible region. □

Example 2.4. Let us consider the LPP with

$$A = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 0 & 1 \end{bmatrix}, \quad \mathbf{c}^T = [2, 5, 6, 8] \quad \text{and} \quad \mathbf{b} = \begin{bmatrix} 5 \\ 2 \end{bmatrix}.$$

Let us choose our starting B as

$$B = [\mathbf{a}_1, \mathbf{a}_4] = [\mathbf{b}_1, \mathbf{b}_2] = \begin{bmatrix} 1 & 4 \\ 1 & 1 \end{bmatrix}.$$

Then it is easily checked that the corresponding basic solution is $\mathbf{x}_B = [1, 1]^T$, which is clearly feasible with objective value

$$z = \mathbf{c}_B^T \mathbf{x}_B = [2, 8] \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 10.$$

Since

$$B^{-1} = \frac{1}{3} \begin{bmatrix} -1 & 4 \\ 1 & -1 \end{bmatrix},$$

by (2.24), the y_{ij} are given by

$$y_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad y_2 = \begin{bmatrix} -\frac{2}{3} \\ \frac{2}{3} \end{bmatrix} \quad y_3 = \begin{bmatrix} -1 \\ 1 \end{bmatrix} \quad y_4 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}.$$

Hence

$$z_2 = [2, 8] \begin{bmatrix} -\frac{2}{3} \\ \frac{2}{3} \end{bmatrix} = 4,$$

and

$$z_3 = [2, 8] \begin{bmatrix} -1 \\ 1 \end{bmatrix} = 6.$$

Since $z_2 - c_2 = -1$ and $z_3 - c_3 = 0$, we see that \mathbf{a}_2 is the *entering* column. As remarked above,

$$z_1 - c_1 = z_4 - c_4 = 0,$$

because x_1 and x_4 are basic variables. Looking at the column entries of \mathbf{y}_2 , we find that y_{22} is the only positive entry. Hence $\mathbf{b}_2 = \mathbf{a}_4$ is the *leaving* column. Thus

$$\hat{B} = [\mathbf{a}_1, \mathbf{a}_2] = \begin{bmatrix} 1 & 2 \\ 1 & 0 \end{bmatrix},$$

and the corresponding basic solution is found to be $\hat{\mathbf{x}}_B = [2, \frac{3}{2}]^T$, which is clearly feasible as expected. The new objective value is given by

$$\hat{z} = [2, 5] \begin{bmatrix} 2 \\ \frac{3}{2} \end{bmatrix} = 11.5 > z.$$

$A = B\mathbf{Y}$
 $\mathbf{Y} = B^{-1}A$
 $\hat{\mathbf{z}}^T = \mathbf{c}_B^T \mathbf{Y}$

$$\max f = 2x_1 + 5x_2 + 6x_3 + 8x_4$$

$$s.t. \begin{cases} \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 0 & 1 \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 5 \\ 2 \end{pmatrix} \\ x_1, \dots, x_4 \geq 0 \end{cases}$$

$$B = \begin{bmatrix} 1 & 4 \\ 1 & 1 \end{bmatrix}$$

② $(\vec{a}_1, \vec{a}_2, \vec{a}_3, \vec{a}_4) A = B \bar{I}$
 $= [\vec{a}_1, \vec{a}_4] \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}$

$$\bar{I} = B^{-1} A$$

$$= \begin{bmatrix} 1 & -2/3 & -1 & 0 \\ 0 & 2/3 & 1 & 1 \end{bmatrix}$$

$$\vec{C}_B = (C_1, C_4) = (2, 8)$$

① BFS. $f = 2 \cdot 1 + 8 \cdot 1 = 10$

$$\begin{pmatrix} 1 & 4 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_4 \end{pmatrix} = \begin{pmatrix} 5 \\ 2 \end{pmatrix}$$

$$x_1, x_4 \geq 0$$

$$\Leftrightarrow \begin{pmatrix} x_1 \\ x_4 \end{pmatrix} = \begin{pmatrix} 1 & 4 \\ 1 & 1 \end{pmatrix}^{-1} \begin{pmatrix} 5 \\ 2 \end{pmatrix}$$

$$= \begin{pmatrix} 1 \\ 1 \end{pmatrix} \text{ starting BFS.}$$

Corresponding to the basic variables \bar{I} 's columns = I .

$$\vec{z} = (2, 8) \cdot \bar{I} = (2, 4, 6, 8)$$

$$\vec{C} - \vec{z}^T = (2, 5, 6, 8) - (2, 4, 6, 8)$$

$$= (0, 1, 0, 0) \quad \text{if } z_2 - z_2 > 0$$

x_2 is the entering variable

Optimality condition

③

$$\bar{I} = \begin{pmatrix} 1 & -2/3 & -1 & 0 \\ 0 & 2/3 & 1 & 1 \end{pmatrix}$$

leaving variable

entering x_2

$$\min \left\{ \frac{x_{Br}}{y_{r2}} : y_{r2} > 0 \right\} \left\{ \frac{x_1}{y_{12}}, \frac{x_4}{y_{22}}, y_{kr} > 0 \right\}$$

$$\min \left\{ \frac{1}{-2/3}, \frac{1}{2/3} \right\} \rightarrow \text{leaving variable } x_1$$

leaving variable x_2 is entering

④

$$\text{BFS} \begin{pmatrix} x_1 = 1 \\ x_2 = 0 \\ x_3 = 0 \\ x_4 = 1 \end{pmatrix} \rightarrow \begin{pmatrix} x_1 > 0 \\ x_2 > 0 \\ x_3 = 0 \\ x_4 = 0 \end{pmatrix}$$

Since

$$\hat{B}^{-1} = \frac{1}{2} \begin{bmatrix} 0 & 2 \\ 1 & -1 \end{bmatrix},$$

by (2.24), the \hat{y}_{ij} are given by

$$\hat{y}_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \hat{y}_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \quad \hat{y}_3 = \begin{bmatrix} 0 \\ \frac{3}{2} \end{bmatrix} \quad \hat{y}_4 = \begin{bmatrix} 1 \\ \frac{3}{2} \end{bmatrix}.$$

$$A = B, \bar{Y}_1$$

$$\bar{z}_1^T = \bar{c}_B^T \bar{Y}_1$$

Hence

$$z_3 - c_3 = [2, 5] \begin{bmatrix} 0 \\ \frac{3}{2} \end{bmatrix} - 6 = 1.5,$$

and

$$z_4 - c_4 = [2, 5] \begin{bmatrix} 1 \\ \frac{3}{2} \end{bmatrix} - 8 = 1.5.$$

Since all $z_j - c_j \geq 0$, $1 \leq j \leq 4$, we see that the point $[2, \frac{3}{2}]$ is an optimal solution.

The last example illustrates how one can find the optimal solution by searching through the basic feasible solutions. That is exactly what *simplex method* does. However, the simplex method uses tableaus to minimize the book-keeping work that we encountered in the last example.

next time x_1, x_2 are basic

$$\hat{B} = \begin{pmatrix} 1 & 2 \\ 1 & 0 \end{pmatrix}$$

$$\Rightarrow \hat{B}^{-1} = \frac{1}{2} \begin{pmatrix} 0 & 2 \\ 1 & -1 \end{pmatrix}$$

Solution $\begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix} = \begin{pmatrix} 2 \\ 3/2 \\ 0 \\ 0 \end{pmatrix}$

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \hat{B}^{-1} \vec{b} = \begin{pmatrix} 2 \\ 3/2 \end{pmatrix} \Rightarrow$$

$$f = 2 \cdot 2 + 5 \cdot \frac{3}{2} = 4 + \frac{15}{2} = 11.5$$

$$A = \hat{B} \hat{Y}$$

$$\hat{Y} = \hat{B}^{-1} A = \begin{bmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 3/2 & 3/2 \end{bmatrix}$$

$x_4 = 1 = 10 + (1.5)$
 $1.5 \neq \frac{x_B}{y_{rj}} (c_j - z_j)$
 $\frac{1}{3}$
 $\frac{1}{1}$
 $\frac{1}{3}$

$$\hat{z} = \hat{c}_B^T \hat{Y} = (2 \quad 5) \hat{Y} = \left(\begin{array}{cccc} 2 & 5 & \frac{15}{2} & \frac{19}{2} \end{array} \right) \frac{1}{\frac{2}{3}} \cdot 1 = 3$$

new
"current"
 $c_1 \quad c_2$
 current
 basis

$$C^T - \hat{z}^T = (2 \quad 5 \quad 0 \quad 8) - (2 \quad 5 \quad \frac{15}{2} \quad \frac{19}{2})$$

$$= (0 \quad 0 \quad -1.5 \quad -1.5) \Rightarrow \text{"optimal" of LPP}$$

None are positive $\Rightarrow x_3 \downarrow$ check $x_4 \downarrow$ check

"objective is linear"

Thm 2.7

$c_j - z_j \leq 0 \quad \forall j \Rightarrow$ then we are
at the optimum

$[B | R] \begin{pmatrix} \vec{x}_B \\ 0 \end{pmatrix} = \vec{b}$ $\Leftrightarrow B \vec{x}_B = \vec{b} \Leftrightarrow \vec{x}_B = B^{-1} \vec{b}$

pf: ① $A \begin{pmatrix} \vec{x}_B \\ 0 \end{pmatrix} = \vec{b} \quad , \quad \vec{x}_B \geq 0$ ③

$\vec{c}_B^T \vec{x}_B = \vec{c}^T \begin{pmatrix} \vec{x}_B \\ 0 \end{pmatrix} =$ current objective value

② $\forall \vec{u} \in FR \quad \left(\begin{array}{l} A \vec{u} = \vec{b} \\ \vec{u} \geq \vec{0} \end{array} \right) \quad \vec{u} = \begin{pmatrix} u_1 \\ \vdots \\ u_n \end{pmatrix}$

$\Rightarrow \boxed{\vec{c}^T \begin{pmatrix} \vec{x}_B \\ 0 \end{pmatrix} \geq \vec{c}^T \vec{u}}$

$\forall \vec{u} \in FR: \vec{c}^T \vec{u} = \sum_{i=1}^n c_i u_i \leq \sum_{i=1}^n z_j u_j$ (by assumption)

$= \vec{z}^T \vec{u}$

$= \cancel{\vec{z}^T A} \quad (\vec{z}^T = \vec{c}_B^T \Sigma)$

$= \vec{c}_B^T \Sigma \vec{u} \quad (A = B \Sigma)$

$= \vec{c}_B^T B^{-1} A \vec{u} \quad (A \vec{u} = \vec{b})$

$= \vec{c}_B^T B^{-1} \vec{b} \quad (B^{-1} \vec{b} = \vec{x}_B)$

$= \vec{c}_B^T \vec{x}_B \quad \neq$

Ex. 3.4

$$\text{Max } X_0 = 3X_1 + 2X_2 + 3X_3 + 0X_4 + 0X_5 + 0X_6$$

$$\left\{ \begin{array}{ccc|ccc} 2 & 1 & 1 & 1 & 0 & 0 \\ 1 & 2 & 3 & 0 & 1 & 0 \\ 2 & 2 & 1 & 0 & 0 & 1 \end{array} \right\} \begin{array}{l} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \\ X_6 \end{array} = \begin{array}{l} 2 \\ 5 \\ 6 \end{array}$$

$X_1, \dots, X_6 \geq 0$

$$\begin{pmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \\ X_5 \\ X_6 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 2 \\ 5 \\ 6 \end{pmatrix} \geq 0$$

non-basic variable
basic variable

