

## Lecture 5-W4

2021年2月2日 星期二 09:35

### Pivot Rule:

$$\begin{aligned} \text{Max } & -2 + x_1 + 2x_2 - W_2 \\ \text{subject to: } & \begin{cases} W_1 = 1 - 3x_2 + W_2 \\ \lambda = 2 - x_1 - 2x_2 + W_2 \\ W_3 = 3 - x_1 - 3x_2 + W_2 \\ W_i \geq 0, x_i \geq 0, \lambda \geq 0 \end{cases} \end{aligned}$$

non-basic variables = independent variables,

basic variables = dependent variables.

While iterating the basic step of the simplex Method,

We operate two choices:

(1) A variable with positive coefficient from the objective function.

(2) a basic variable to substitute.

(a) a variable goes from non-basic to basic. - Entering variable

(b) a variable goes from basic to non-basic - Leaving variable

A computer needs a set of specific instructions. So we are required to choose a rule that specifies how to make this decision. Different rules lead to different algorithm.

A pivoting rule is a rule specifying how to choose the leaving and the entering variable at each iteration of the simplex method.

- Anstee's Rule
  - Blond's Rule
- } pivoting rule. only for positive-coefficient variables in the objective function.

**Rule:** A good pivoting rule is a rule for which the termination of the simplex method is guaranteed.

(1) Anstee's rule: (Not the best) rule of thumb / in exam. use this

- choose the entering variable with the largest possible coefficient.

If there is a tie, then choose the smallest subscript.

- (choose the dependent var to flip). If there is to choose the leaving variable, just choose the one with the smallest subscript.

(2) Blond's rule: good rule to get termination

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- everytime you have to choose a variable  $x_k$ , just choose the one with the smallest subscript  $k$ .

Example:

$$\max \quad Z = x_1 - 2x_2 - 2x_4$$

subject to

$$\begin{cases} w_1 = -0.5x_1 + 3.5x_2 + 2x_3 - 4x_4 \\ w_2 = -0.5x_1 + x_2 + 0.5x_3 - 0.5x_4 \\ w_3 = 1 - x_1 \\ x_i \geq 0, \quad w_i \geq 0 \end{cases}$$

( $w_1 \dots w_m, x_1 \dots x_n$ )

||

( $x_1 \dots x_m, x_{m+1} \dots x_{m+n}$ )

$$x_1 = x_2 = x_3 = x_4 = 0$$

$$w_1 = 0, \quad w_2 = 0, \quad w_3 = 1$$

← This is admissible so I don't need to run the auxiliary problem.

I increase the value of  $x_1$ , (choosing  $x_1$  as entering variable)

$$\begin{cases} w_1 = -0.5x_1 \geq 0 \\ w_2 = -0.5x_1 \geq 0 \\ w_3 = 1 - x_1 \geq 0 \end{cases} \quad \begin{cases} x_1 \geq 0 \\ x_1 \leq 0 \\ x_1 \leq 1 \end{cases} \Rightarrow \boxed{x_1 = 0}$$

no gain in the objective function but maybe I change dictionary and things can be improved in the next step of the algorithm.

For Anstee and Blond's rules, we choose  $w_1$  as leaving variable.

pivot,  $x_1$  enters and  $w_1$  leaves bringing us to:

$$\begin{aligned} Z &= -2w_1 + 5x_2 + 4x_3 - 10x_4 \\ x_1 &= -2w_1 + 7x_2 + 4x_3 - 8x_4 \\ w_2 &= w_1 - 2.5x_2 - 1.5x_3 + 3.5x_4 \\ w_3 &= 1 + 2w_1 - 7x_2 - 4x_3 + 8x_4 \end{aligned}$$

For the second iteration,  $x_2$  enters and  $w_2$  leaves bringing us to:  $w_2 \Rightarrow x_1$

$$\begin{aligned} Z &= -2w_2 + x_3 - 3x_4 \\ x_1 &= 0.8w_1 - 2.8w_2 - 0.2x_3 + 1.8x_4 \\ x_2 &= 0.4w_1 - 0.4w_2 - 0.6x_3 + 1.4x_4 \\ w_3 &= 1 - 0.8w_1 + 2.8w_2 + 0.2x_3 - 1.8x_4 \end{aligned}$$

For the third iteration,  $x_3$  enters and  $x_1$  leaves:

$$\begin{aligned} Z &= 4w_1 - 16w_2 - 5x_1 + 6x_4 \\ x_3 &= 4w_1 - 14w_2 - 5x_1 + 9x_4 \\ x_2 &= -2w_1 + 8w_2 + 3x_1 - 4x_4 \\ w_3 &= 1 - x_1 \end{aligned}$$

For the fourth iteration,  $x_4$  enters and  $x_2$  leaves:

only substitute with

the sharpest equation!

In this case,  $w_3$  is excluded.

$$\begin{aligned}
 x_2 &= -2w_1 + 8w_2 + 3x_1 - 4x_4 \\
 w_3 &= 1 - x_1.
 \end{aligned}$$

For the fourth iteration,  $x_4$  enters and  $x_2$  leaves:

$$\begin{aligned}
 \zeta &= w_1 - 4w_2 - 0.5x_1 - 1.5x_2 \\
 x_3 &= -0.5w_1 + 4w_2 + 1.75x_1 - 2.25x_2 \\
 x_4 &= -0.5w_1 + 2w_2 + 0.75x_1 - 0.25x_2 \\
 w_3 &= 1 - x_1.
 \end{aligned}$$

In the fifth iteration,  $w_1$  enters and  $x_3$  leaves:

$$\begin{aligned}
 \zeta &= -2x_3 + 4w_2 + 3x_1 - 6x_2 \\
 w_1 &= -2x_3 + 8w_2 + 3.5x_1 - 4.5x_2 \\
 x_4 &= x_3 - 2w_2 - x_1 + 2x_2 \\
 w_3 &= 1 - x_1.
 \end{aligned}$$

Lastly, for the sixth iteration,  $w_2$  enters and  $x_4$  leaves:

$$\begin{aligned}
 \zeta &= -2x_4 + x_1 - 2x_2 \\
 w_1 &= 2x_3 - 4x_4 - 0.5x_1 + 3.5x_2 \\
 w_2 &= 0.5x_3 - 0.5x_4 - 0.5x_1 + x_2 \\
 w_3 &= 1 - x_1.
 \end{aligned}$$

→ The same objective function as before.

**Rmk**: if the algorithm does not terminate then we enter an infinite cycle.

$x_1 \dots x_n$  independent.

$w_1 \dots w_m$  dependent.

$n = \#$  of variables

$m = \#$  of constraints.

Every configuration of the simplex algorithm is usually determined by the choice of the  $n$  independent variables.

$$\underbrace{w_1 \dots w_m \ x_1 \dots x_n}_{\text{pick } n \text{ of them}} \Rightarrow \binom{n+m}{n} = \frac{(n+m)!}{n! \cdot m!}$$

number of all possible configurations we can hit with the simplex method.

This is a large number but finite: If I keep iterating forever I must come back on some of the configurations.

Long story short:

**THEOREM**: if we make decisions with Bland's rule, cycling cannot happen.



## Matrix Notation

$$\max f = \overset{\text{scalar product}}{C_0} + \underbrace{C_1 x_1 + \dots + C_n x_n}_{\text{linear}}$$

$$\vec{v} = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \quad \vec{w} = \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$$

$$\vec{v} \cdot \vec{w} = v_1 \cdot w_1 + v_2 \cdot w_2 + \dots + v_n \cdot w_n$$

$$\vec{x} = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \quad \vec{c} = \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}$$

$$\Rightarrow f = C_0 + \vec{x} \cdot \vec{c}$$

$$\vec{v} = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix} \quad \vec{w} = \begin{bmatrix} w_1 \\ \vdots \\ w_n \end{bmatrix}$$

$$\vec{v} = \vec{w} \text{ means } \begin{matrix} v_1 = w_1 \\ v_2 = w_2 \\ \vdots \end{matrix}$$

$$\text{assume } \vec{v} \leq \vec{w} \text{ if } \begin{matrix} v_1 \leq w_1 \\ v_2 \leq w_2 \\ \vdots \end{matrix}$$

$$\begin{aligned} \Rightarrow \begin{cases} a_{11}x_1 + \dots + a_{1n}x_n \leq b_1 \\ a_{21}x_1 + \dots + a_{2n}x_n \leq b_2 \\ \vdots \\ a_{m1}x_1 + \dots + a_{mn}x_n \leq b_m \\ x_1, \dots, x_n \geq 0 \end{cases} & \Rightarrow \vec{A} \cdot \vec{x} \leq \vec{b} \\ & \downarrow \\ & \begin{bmatrix} a_{11} & \dots & a_{1n} \\ \vdots & & \vdots \\ a_{m1} & \dots & a_{mn} \end{bmatrix} \cdot \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} \\ & \quad \vec{A} \quad \vec{x} \end{aligned}$$

$$\begin{aligned} \Rightarrow \text{O.F.} = \max f &= C_0 + \vec{c} \cdot \vec{x} \\ \text{subject to } &\begin{cases} \vec{A} \cdot \vec{x} \leq \vec{b} \\ \vec{x} \geq 0 \end{cases} \end{aligned}$$

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Step 0: introduce additional variables  $w_1, \dots, w_m$ .

$$\begin{cases} w_1 = b_1 - a_{11}x_1 - \dots - a_{1n}x_n \\ w_2 = b_2 - a_{21}x_1 - \dots - a_{2n}x_n \\ \vdots \\ w_m = b_m - a_{m1}x_1 - \dots - a_{mn}x_n \\ x_1, \dots, x_n \geq 0 \end{cases}$$

$$\vec{w} = \begin{bmatrix} w_1 \\ \vdots \\ w_m \end{bmatrix} \quad \begin{aligned} \text{Max } f &= C_0 + \vec{c} \cdot \vec{x} \\ \begin{cases} \vec{w} &= \vec{b} - \vec{A} \cdot \vec{x} \\ \vec{w} \geq 0, \vec{x} \geq 0 \end{cases} \end{aligned}$$

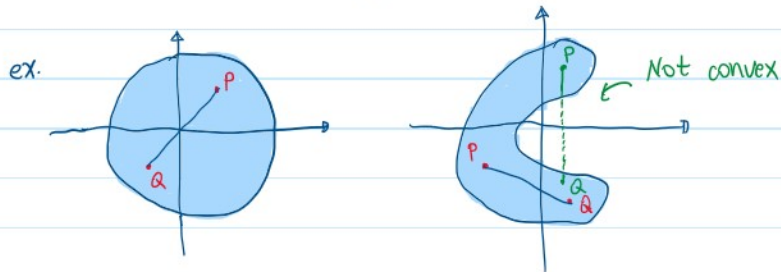
**Rmk:** since there are  $\binom{n+m}{n}$  dictionaries for a problem in  $n$  variables on  $m$  constraints the simplex method needs at most  $\binom{n+m}{n}$  steps to end.

There are examples where exactly  $\binom{n+m}{n}$  steps are required. (worst scenario).

### Convexity

← vector

$S \subseteq \mathbb{R}^n$  is convex if whenever I choose two points  $P, Q \in S$  the line segment & joining  $P$  and  $Q$  is totally contained in  $S$ , i.e.  $\ell \subseteq S$ .



**Rmk:** the points in the line segment joining  $P$  and  $Q$  can be written as  $tP + (1-t)Q$ ,  $0 \leq t \leq 1$

$\forall =$  for all

$\exists =$  there exist,

$S \subseteq \mathbb{R}^n$  is convex if  $\forall P, Q \in S$  we have  $tP + (1-t)Q \in S \quad \forall 0 \leq t \leq 1$