

July 3rd, 2018

Recall:

(Algorithm 1) (\*)

For  $t = 0, 1, 2, \dots, t_0$

let  $A_t$  = matrix of remaining processing times

let  $\Delta_t$  = max {row or column sum of  $A_t$ }

let  $I_t = \{i \in I : \text{sum of row } i \text{ in } A_t = \Delta_t\}$

let  $J_t = \{j \in J : \text{sum of col } j \text{ in } A_t = \Delta_t\}$

let  $G_t$  = bipartite graph on  $I_t \cup J_t$  s.t.  $ij \in E(G_t)$  iff  $[A_t]_{ij} > 0$

Compute  $M_t$  : a  $(I_t \cup J_t)$ -perfect matching in  $G_t$

Schedule jobs according to  $M_t$  from time  $t$  to  $t+1$  (\*\*)

Recall:

We were finishing Algorithm 1.

→ We can reduce  $R(\text{optimal Cmax}) \leq p \cdot \text{optimal Cmax}$

Claim (from last class)

Algorithm 1 computes a schedule of makespan  $\Delta_0$  [Proof]

We showed that since  $\Delta_{t+1} = \Delta_t - 1 \Rightarrow \Delta_t = \Delta_0 - t$   
 $\Rightarrow$  @ time  $\Delta_0$ , all jobs are completed

$$C_{\max} \geq \sum_{i \in I} P_{ij} \quad \forall j \in J$$

$$C_{\max} \geq \sum_{j \in J} P_{ij} \quad \forall i \in I$$

$$C(LB) = \max \left\{ \max_{j \in J} \sum_{i \in I} P_{ij}, \max_{i \in I} \sum_{j \in J} P_{ij} \right\}$$

Hilroy

DP



## Issues to resolve!

- (1) @ time  $t$ , why does a  $J_t \cup I_t$ -perfect matching exist?
- (2) When can we reuse  $M_t$  as  $M_{t+\delta}$ ?

Goal: Bound # <sup>of</sup> matchings we use  $\leq mn + m + n$ .

Idea from last class:

What prevents  $M_t$  from being a  $(J_{t+\delta} \cup I_{t+\delta})$ -perfect matching in  $G_{t+\delta}$ ?

- $\leq m n_1$ ) Complete a job if  $p_{ij} = 1$ .
- $\leq m n_2$ )  $J_{t+\delta} \neq J_t$ , there's a new tight machine
- $\leq n n_3$ )  $I_{t+\delta} \neq I_t$ , there's a new tight job

How long to use  $M_t$ ?

"Run" the schedule given by  $M_t$  for a time interval of length  $\delta$  UNTIL

- one of the  $p_{ij}$ 's goes to 0,
- $\Delta_t - \delta$  becomes equal to row  $i$ 's sum for  $i$  not matched by  $M_t$
- $\Delta_t - \delta$  becomes equal to column  $j$ 's sum for  $j$  not matched by  $M_t$

Then we can replace  $C$  in  $A$  by  $A$

→ (\*\*) w: let  $t=0$ , while  $t < \Delta_0$

→ (\*\*\*) w:

Compute:

$$\delta = \min \begin{cases} \min_{i \notin M_t} p_{ij} \\ \min_{i \text{ not matched by } M_t} (\Delta_t - \text{row sum}_t(i)) \\ \min_{j \text{ not matched by } M_t} (\Delta_t - \text{col sum}_t(j)) \end{cases}$$

Schedule according to  $M_t$  from time  $t$  to  $t + \delta$ . Advance  $t \leftarrow t + \delta$

Claim:

We can always find a  $(I_t \cup J_t)$ -perfect matching in  $G_t$  at time  $t$

[Proof]

If  $G_t$  is bipartite, recall seen fractional  $S$ -perfect matching, can find integral  $S$ -perfect matching

Consider:

$$x_{ij} = \frac{p_{ij}}{\Delta_t} \geq 0$$

For  $i \in I$

$\rightarrow$  and  $= 1$  iff  $i \in I_t$

$$\sum_{j \in J} x_{ij} = \frac{1}{\Delta_t} \underbrace{\sum_{j \in J} p_{ij}}_{\leq \Delta_t} \leq 1$$

For  $j \in J$ :

$\rightarrow$  and  $= 1$  iff  $j \in J_t$

$$\sum_{i \in I} x_{ij} = \frac{1}{\Delta_t} \underbrace{\sum_{i \in I} p_{ij}}_{\leq \Delta_t} \leq 1$$

□

← (This is a fractional  $S$ -perfect matching, so there is an integral  $S$ -perfect matching)

Hilary



Ex:

Iteration 1 ( $t=0$ )

$$P = \begin{bmatrix} 2 & 1 & 0 & 2 \\ 0 & 4 & 1 & 0 \\ 2 & 0 & 0 & 2 \end{bmatrix} = A_0$$

1.  $\Delta_0 = \max \text{ row sum or col sum}$

$$\begin{bmatrix} 2 & 1 & 0 & 2 \\ 0 & 4 & 1 & 0 \\ 2 & 0 & 0 & 2 \end{bmatrix} \begin{matrix} = 5 \\ = 5 \\ = 4 \end{matrix}$$

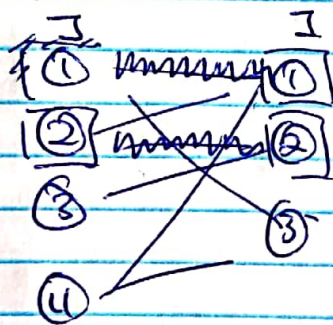
4   5   1   4

So,  $\Delta_0 = 5$

2.  $I_0 = \{1, 2\}$

3.  $J_0 = \{2\}$

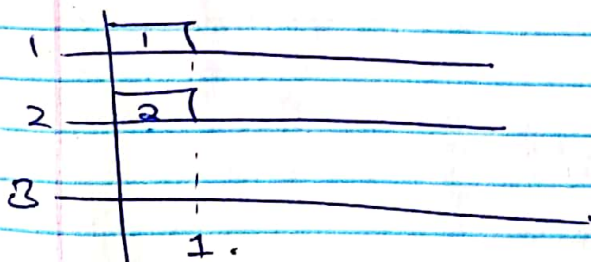
4.  $M_0$



$M_0 = \{(1,1), (2,2)\}$

$$5. \delta = \min \begin{cases} \min \{2, 4\} \\ \min \{5-4\} \\ \min \{5-1, 5-4\} \end{cases} = 1$$

$\nearrow PC(1,1)$   
 $\nwarrow PC(2,2)$



← cost

Iteration 2 (t=1)

$$A_1 = \begin{bmatrix} \textcircled{1} & 1 & 0 & 2 \\ 0 & \textcircled{3} & 1 & 0 \\ 2 & 0 & 0 & \textcircled{2} \end{bmatrix} \begin{matrix} = 4 \\ = 4 \\ = 4 \end{matrix}$$

" " " "

3 4 1 4

$$\Delta_1 = 4$$

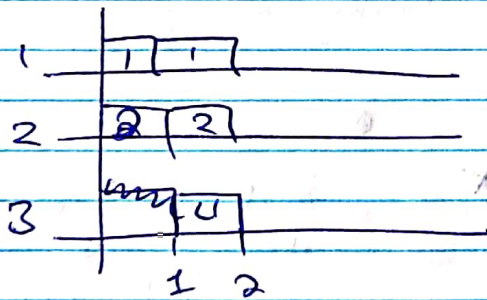
$$I_1 = \{1, 2, 3\}$$

$$J_1 = \{2, 4\}$$

$$M_1 = (\text{See circles}) \quad (J, I)$$

→ So, we have:  $\{(1, 1), (2, 2), (3, 4)\}$

$$\delta = \min \begin{cases} \min \{1, 3, 2\} \\ \min \{ \infty \} \\ \min \{4-1\} \end{cases} = 1$$





Iteration 3:

$$A_2 = \begin{bmatrix} 0 & 1 & 0 & \textcircled{2} \\ 0 & \textcircled{2} & 1 & 0 \\ \textcircled{2} & 0 & 0 & 1 \\ 2 & 3 & 1 & 3 \end{bmatrix} \begin{matrix} 3 \\ 3 \\ 3 \\ \end{matrix}$$

$$\Delta_2 = \{2\}, I_2 = \{1, 2, 3\}, J_2 = \{2, 4\}$$

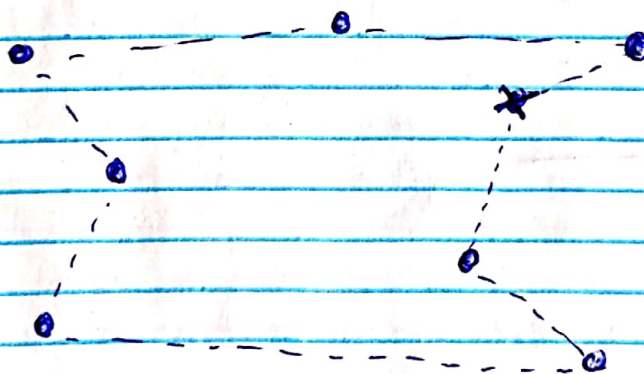
(continuing on  $\beta = 2$ )

And, iteration 4 looks like:

$$A_3 = \begin{bmatrix} 0 & \boxed{1} & 0 & 0 \\ 0 & 0 & \boxed{1} & 0 \\ 0 & 0 & 0 & \boxed{1} \\ \end{bmatrix}, \text{ etc.}$$

Travelling Salesman Problem:

→ We can show that this is equivalent to "jobshop w/ setup times".



→ Want to travel through all nodes and returning to starting node  $\Rightarrow$  w shortest distance/cost.

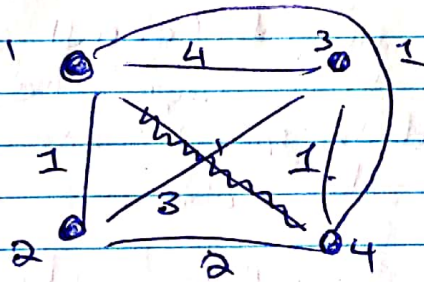
Def'n (Travelling Salesman Problem - TSP)

Given complete graph  $G = (V, E)$  with edge costs  $c_e \geq 0$  for all  $e \in E$ . We want to find a (simple) cycle of min. cost that visits all the nodes (Euler tour in this context)  
 No repeated vertices.

Ex:

$G = (\{1, 2, 3, 4\}, \{12, 13, 14, 23, 24, 34\})$   
 $c_e =$ 

1	4	1	3	2	1
↓	↓	↓	↓	↓	↓



Q: Is this problem NP-hard?  
A: YES.

Q: Can we find a  $k$ -approx algorithm?  
A: No, not for arbitrary edge weights.

But, we can do this if the edge weights we assume satisfy the triangle inequality

Def'n

Edge costs  $c_e$  for all  $e \in E$  satisfy the triangle inequality if:

$$c_{uv} \leq c_{uw} + c_{wv} \text{ for } u, v, w \in V.$$

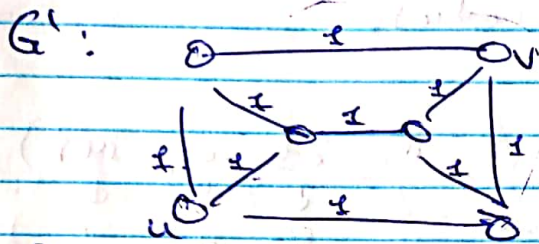
Hilroy



Def'n

We call TSP instance metric if the edge costs satisfy the triangle inequality

Ex: Metric TSP instance



If  $e \in G'$ , then  $c_e$  = length of the shortest

$u-v$  path ( $e = uv$ )  
 ↳ i.e.  $uv$  is not in this graph, the cost of  $uv$  would be ~~at least~~ 2. ~~at least~~ in order to preserve the metric.

Claim  
Lemma: If edge costs satisfy the triangle inequality, then  $\forall u, v \in V$ ,  $c_{uv} \leq c(P)$  for every  $u, v$ -path  $P$ .

(If  $P = u = w_0, w_1, \dots, w_k = v$  then  
 $c(P) = \sum_{i=0}^{k-1} c_{w_i w_{i+1}} = \sum_{e \in P} c_e$ .)

[Proof]

$$\begin{aligned} c(P) &= c_{w_0 w_1} + c_{w_1 w_2} + \dots + c_{w_{k-1} w_k} \\ &\geq c_{w_0 w_2} \\ &\geq c_{w_0 w_3} \end{aligned}$$

→ Do induction ~~at~~ on the length of the path.  $\square$