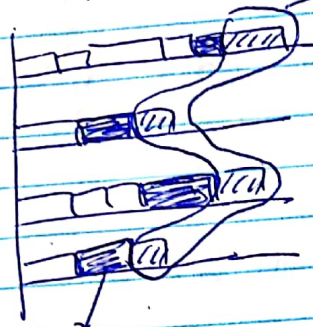


454-16

Pr 2.1:



$\square = \text{"slot 2"} , \text{ and etc.}$

A job  $j$  is assigned to slot  $k$  on m/c  $i$  means that there are  $k-1$  jobs on m/c  $i$  after job  $j$ .

Why is this useful?

Every job  $j$  assigned to slot 1 on m/c  $i$  contributes  $p_j$  to an objective. And, in general, if  $j$  is assigned to slot  $k$  on m/c  $i$ , it contributes  $k \cdot p_j$  to  $\sum C_j$ . (Since ~~there~~ it contributes  $p_j$  to its completion time, and  $(k-1)p_j$  to the jobs that come after it on <sup>its</sup> m/c).

Goal: Assign jobs to slots so as to minimize

$$\sum_{j \rightarrow \text{slot } k} k p_j.$$

"Assign largest  $m$  jobs to  $m$  slot 1's  
Next  $m$  largest jobs to  $m$  slot 2's,  
and so on."

i.e. Sort jobs so that  $p_1 \geq \dots \geq p_n$ . Let  $J = \{1, \dots, n\}$ .  
 $k=1, 1, 1, 1, 1$

Cor

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Repeat until  $J = \emptyset$ :

- Schedule  $\min_{\text{first}} C_m, |J|$  jobs from  $J$  in

slot  $k$  of the  $m$  m/c's (arbitrarily)

-  $J \leftarrow J \setminus \{ \text{first } \min(C_m, |J|) \text{ jobs of } J \}$

-  $k \leftarrow k+1$ .

This is clearly polynomial in time.

Theorem 1: The above algorithm produces an optimal schedule for  $R||\sum C_j$ .

[Proof] Exercise

Exercise: Show that running LIST-SCHEDULING with ~~the~~ jobs sorted in increasing  $P_j$  order mimics the above algorithm and produces an optimal schedule.

$R||\sum C_j$ :

(Recall: job  $j$  takes time  $P_{ij}$  to be processed on m/c  $i$ )

IF  $j \mapsto \text{slot}(i, k)$ , then there are  $(k-1)$  jobs following  $j$  on m/c  $i$ , and  $j$  contributes  $k \cdot P_{ij}$  to  $\sum C_j$ .

Then, our task can be restated as:

$$\min \sum C_j \equiv \text{minimize}$$

Job  $\mapsto$  slot assignment problem

Assign jobs to  $(i, k)$  slot to

$$\min \sum_{j \mapsto (i, k)} k P_{ij}.$$



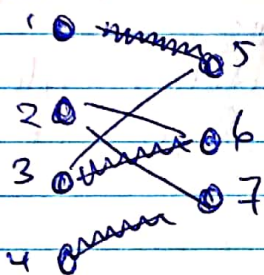
## Excursion into matchings in bipartite graphs:

Let  $G = (V := A \cup B, E)$  be a bipartite graph with bipartition  $A \cup B$ .

Defn: (Matching, S-perfect)

- A matching  $M$  is a set of edges s.t. no 2 edges of  $M$  share an endpoint
- Given  $S \subseteq V$ , a matching  $M$  is said to be S-perfect, if  $\forall v \in S, \exists e \in M$  incident to  $v$ .

Ex:



--- matching  $M$ .

For  $S = \{1, 5, 6, 7\}$ ,  $M$  is S-perfect,  
~~but not S-perfect~~  
 For  $S = \{2\}$ ,  $M$  is not S-perfect.

Defn (Fractional Matching, Fractional S-perfect)

A fractional matching  $x$  is an assignment  $\{x_e \geq 0 \mid e \in E\}$  of ~~numbers~~ to edges  $s-t$ .

$$\sum_{e \in \delta(v)} x_e \leq 1 \quad \forall v \in V. \quad (\delta(v): \text{set of edges incident to } v).$$

Further, a fractional matching  $x$  is S-perfect (for  $S \subseteq V$ ) if

$$\sum_{e \in \delta(v)} x_e = 1 \quad \forall v \in S.$$

Note: If  $x \in \{0, 1\}^E$ , then fractional matching corresponds to matchings and similarly for S-perfect matchings.

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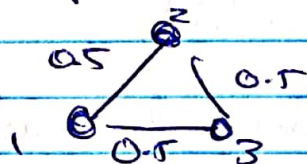


Fact 1: Given edge costs  $\{c_e\}_{e \in E}$  and any  $S \subseteq V$ , we can decide if  $\exists S$ -perfect matching and if so compute a min cost  $S$ -perfect matching in polytime.  
 Cost of matching  $M = \sum_{e \in M} c_e$

Fact 2: In a bipartite graph, given edge costs  $\{c_e\}_{e \in E}$  and a fractional  $S$ -perfect matching  $x$ , we can compute (in polytime) an  $S$ -perfect matching  $M$  of cost  $\sum_{e \in M} c_e$ .  
~~On a bipartite graph, given edge costs  $\{c_e\}_{e \in E}$  and a fractional  $S$ -perfect matching  $x$ , we can compute (in polytime) an  $S$ -perfect matching  $M$  of cost  $\sum_{e \in M} c_e$ .~~

(In particular, an  $S$ -perfect matching always exists if there is a fractional  $S$ -perfect matching)

(Ex. Counter example)



$x$ : Fractional  $\{1, 2, 3\}$ -perfect matching.

But, there is no way to obtain a  $\{1, 2, 3\}$ -perfect matching.

i.e. The assumption of bipartite graphs is important.



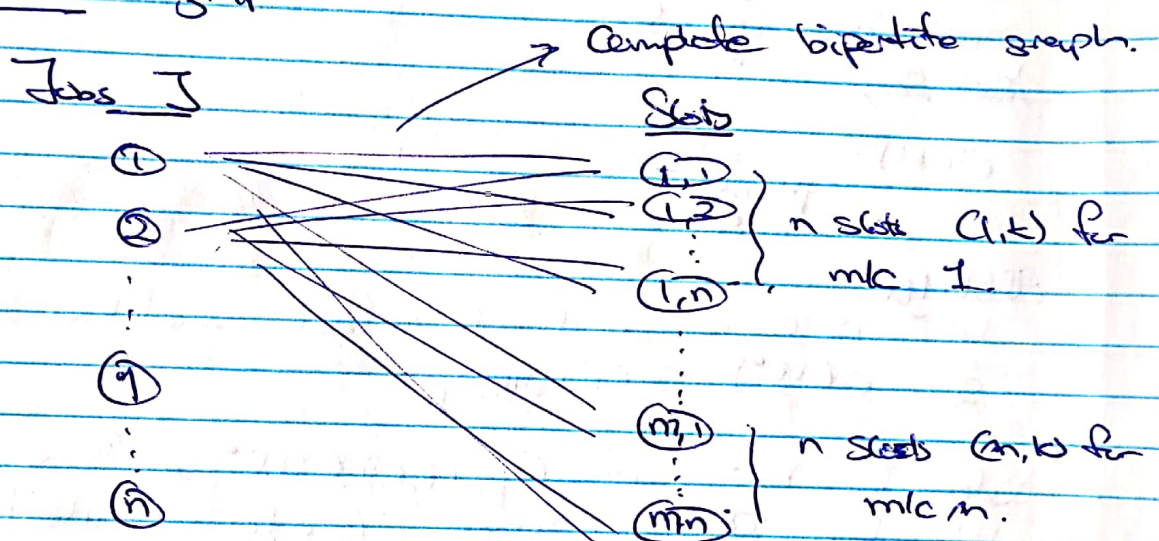
Back to  $R(\sum C_j)$

Recall we want to sched.

assign jobs to slots to minimize  $\sum_{j \rightarrow (i,k)} k p_{ij}$

Strategy: We will view this as a min-cost  $S$ -matching problem in a bipartite graph.

Bipartite graph:



Create node-jobs  $j$  for all jobs, and node  $(i,k)$  for each slot  $\forall i=1, \dots, m, k=1, \dots, n$ .

The cost on edge  $(j, (i,k))$   $\forall$  jobs  $j$ , slot  $(i,k)$  is  $k \cdot p_{ij}$ .

Theorem:

A min cost  $J$ -perfect matching  $M^*$  yields an optimal schedule for  $R(\sum C_j)$  where  $j \rightarrow (i,k)$  iff  $(j, (i,k)) \in M^*$ .

$\hookrightarrow$

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[Proof]

Suppose we have a schedule  $S$ . Then, consider the edge set  $M = \{C_j, (i, k) : j \text{ is scheduled in slot } (i, k) \text{ in } S\}$

Then  $M$  is a  $J$ -perfect matching since every job is assigned to exactly 1 slot and every slot is assigned at most one job.  $\square$

And so,

$$C(M) = \sum_{\substack{j \rightarrow (i, k) \\ \text{in } S}} k p_{ij} = \sum C_j S$$

Also,

$M^*$  is a min-cost  $J$ -perfect matching, so  $C(M^*) \leq \text{OPT}_{R||C_j}$

(Note: In  $M^*$  if  $C_j, (i, k) \in M^*$ , then  $\forall k' < k$ , can assume  $\nexists$  edge incident to  $(i, k')$  in  $M^*$ , since otherwise, we can replace  $C_j, (i, k)$  in  $M^*$  by  $C_j, (i, k')$  to get lower cost  $J$ -perfect matching.)

So, schedule constructed by  $j \rightarrow (i, k)$  iff  $C_j, (i, k) \in M^*$  is a valid schedule.

i.e.  $j \rightarrow (i, k)$  means  $k-1$  jobs follow  $j$  on machine  $i$  and  $\sum C_j$  of schedule =  $C(M^*) \leq \text{OPT}_{R||C_j}$

$\square$