# **Multiple Machines**

- Model Multiple Available resources
  - people
  - $-\operatorname{time\ slots}$
  - queues
  - networks of computers
- Now concerned with both allocation to a machine and ordering on that machine.

# $P||C_{\max}|$

### NP-complete from partition.

### Example

- $j \quad p_j$
- 1 10
- 2 8
- 3 6
- 4 4
- **5** 2
- **6** 1
- What is the makespan on 2 machines?
- 3 machines ?
- 4 machines ?

# **Approxmiation Algorithms**

- Cannot come up with an optimal solution in polynomial time
- Will look at relative error :  $C_{\max}(\text{our algorithm})/C_{\max}(OPT)$
- Challenges:
  - Our algorithm's performance is different on different instances
  - We can't compute  $C_{\max}(OPT)$

## **Approxmiation Algorithms**

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- Challenges:
  - Our algorithm's performance is different on different instances
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#### Solution:

- We will use a worst case measure on performance
- We will use a lower bound on  $C_{\max}(OPT)$

### **Approximation Algorithms**

An algorithm A is a  $\rho$  approximation algorithm for a problem, if for all inputs

$$\frac{C_{\max}(A)}{C_{\max}(OPT)} \le \rho$$

In addition, A must run in polynomial time.

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We can't compute C_{\max}(OPT) .
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**Recipe:** 

 $\bullet$  Instead, we compute a lower bound  $\ LB(OPT)$  , such that

- -LB(OPT) is easy to compute
- $\, LB(OPT) \leq C_{\max}(OPT)$  .
- $\bullet$  We then show that  $\ C_{\max}(A) \leq \rho LB(OPT)$  .

Combining the previous two steps, we have:

 $C_{\max}(A) \le \rho LB(OPT) \le \rho C_{\max}(OPT)$ 

which can be rewritten as

 $\frac{C_{\max}(A)}{C_{\max}(OPT)} \le \rho$ 

Notes:

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- Must come up with a good lower bound
- Can replace  $C_{\text{max}}$  with any objective.

# **Lower Bounds for** $P||C_{\max}$

- Average load
- Longest job

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- Average load  $\lceil \sum p_j / m \rceil$
- Longest job  $p_{\max} = \max_{j} \{p_j\}$

# List Scheduling Algorithm

### A Greedy Algorithm

- 1. Make a list of the jobs (in any order)
- 2. When a machine becomes available, schedule the next job on the list.

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### Analysis

- Let t be the last time at which all machines are busy.
- $t \leq \sum_j p_j/m$
- $C_{\max} \le t + p_{\max} \le \sum_j p_j / m + p_{\max}$  .

Put this together with our lower bound:

$$C_{\max} \le t + p_{\max} \le \sum_{j} p_j / m + p_{\max} \le 2LB \le 2OPT$$

# **Improved Algorithm**

- Schedule length is average load plus last job.
- When last job is small, the schedule is shorter.
- Force last job to be small LPT (Longest Processing Time).

### LPT is a 4/3-approximation for $P||C_{max}$ .

#### **Proof Outline**

- If last job is small (  $\leq 1/3OPT$  ) then 4/3-approximation
- Otherwise, there are at most 2 jobs per machine and LPT is optimal.

Even better algorithms are possible: A polynomial-time approximation scheme (PTAS) is an algorithm that, given fixed  $\epsilon > 0$ , returns at  $(1 + \epsilon)$ -approximation in polynomial time. The running time can have a bad dependence on  $\epsilon$ , such as  $n^{O(1/\epsilon)}$ .

 $P||C_{\max}$  has a PTAS.

### **Precedence Constraints**

- $P\infty$  prec  $C_{max}$  is known as project scheduling.
- $P|\text{prec}|C_{\text{max}}$  has a 2-approximation.

What are good lower bounds for  $P|\text{prec}|C_{\text{max}}$ ?

### **Precedence Constraints**

- $P\infty$  prec  $C_{max}$  is known as project scheduling.
- $P|\text{prec}|C_{\text{max}}$  has a 2-approximation.

What are good lower bounds for  $P|\text{prec}|C_{\text{max}}$ ?

- Average load
- $p_{\max}$
- any path in the precedence graph
- the critical path is the longest path in the precedence graph.

### **Unit Processing Times**

 $P|p_j = 1, \text{prec}|C_{\text{max}}$  is **NP-hard.** 

### Heuristics

- Critical Path (CP) rule
  - The job at the head of the longest string of jobs in the constraint graph has the highest priority
  - $-P|p_j = 1, tree|C_{\max}$  is solved by CP.
- Largest Number of Successors First (LNS)
  - The job with the largest total number of successors in the constraint graph has highest priority.
  - For in-trees and chains, LNS is identical to CP
  - LNS is also optimal for  $P|p_j = 1, outtree|C_{max}$
- Generalization to arbitrary processing times is possible

#### **Fixed Number of Processors**

- $P2|p_j = 1$ , prec $|C_{\text{max}}|$  is solvable in polynomial time
- $P3|p_j = 1$ , prec $|C_{\text{max}}|$  is a big open question.

# **Preemptions:** $P|\text{pmtn}|C_{\text{max}}$

• McNaughton's wrap-around rule is optimal.

### Example

- *j p<sub>j</sub>* **A** 7 **B** 10 **C** 1 **D** 4
- E 9

# **LP for** $P|\text{pmtn}|C_{\text{max}}$

 $x_{ij}$  is the time that job j runs on machine i .  $C_{\max}$  is also Variables: a variable.

#### Constraints

- Each job runs for  $p_j$  units of time
- Each machine runs for at most  $C_{\text{max}}$  time.
- $C_{\text{max}}$  is more than any processing time.

$$\min C_{\max} \tag{1}$$

$$s.t.$$

$$\sum_{i=1}^{m} x_{ij} = p_j \qquad j = 1 \dots n \tag{3}$$

$$\sum_{j=1}^{n} x_{ij} \le C_{\max} \quad i = 1 \dots m \tag{4}$$

$$\sum_{i=1}^{m} x_{ij} \le C_{\max} \quad j = 1 \dots n \tag{5}$$

(6)

Note that LP only assigns pieces of jobs to machines. Need to also assign jobs to times.

### Machines with speeds $-Q|pmtn|C_{max}$

- Machines  $M_1, \ldots, M_m$  with speeds  $v_1, \ldots, v_m$  .
- Assume wlog that  $v_1 \ge v_2 \ge v_m$
- Assume wlog that  $p_1 \ge p_2 \ge p_n$
- If a job runs for one unit of time on machine  $M_i$ , it uses up  $v_i$  units of processing.
- If job j runs on machine  $M_i$ , then it takes  $p_j/v_i$  time units to complete.

#### Example

- $\begin{array}{ccc} j & p_j \\ \mathbf{A} & \mathbf{20} \end{array}$
- B 16
- **C** 2
- D 1

What are the lower bounds

# Lower bounds for $Q|pmtn|C_{max}$

- What is the analog of  $p_{\text{max}}$  ?
- What is the analog of average load ?
- Are there others ?

## Lower bounds for $Q | pmtn | C_{max}$

- What is the analog of  $p_{\text{max}}$  ?  $p_1/v_1$
- What is the analog of average load ?  $-\sum p_j/\sum v_i$
- Are there others ? Yes

**General Lower Bound** 

$$C_{\max} \ge \max\left(\frac{p_1}{v_1}, \frac{p_1 + p_2}{v_1 + v_2}, \dots, \frac{\sum_{j=1}^{m-1} p_j}{\sum_{i=1}^{m-1} v_i}, \frac{\sum_{j=1}^n p_j}{\sum_{i=1}^m v_i}\right)$$

### Lower Bound

$$C_{\max} \ge \max\left(\frac{p_1}{v_1}, \frac{p_1 + p_2}{v_1 + v_2}, \dots, \frac{\sum_{j=1}^{m-1} p_j}{\sum_{i=1}^{m-1} v_i}, \frac{\sum_{j=1}^n p_j}{\sum_{i=1}^m v_i}\right)$$

What is the lower bound for our example?

Can we achieve this lower bound?

### $\underline{\mathbf{LRPT}}$

Longest Remaining Processing Time on Fastest Machines

#### Example 1

- j  $p_j$
- A 20
- B 16
- **C** 2
- D 1

v = (4, 2, 1)

#### Example 2

- j  $p_j$
- A 20
- B 16
- C 12
- D 1

### Notes:

- LRPT-FM is optimal in continuous time
- LRPT-FM is near otimal in discrete time, for small time steps.