

## Section 3.2: Makespan Scheduling - A PTAS

### Sketch of PTAS:

1. Schedule **long jobs** ( $> \epsilon \cdot \text{OPT}$ ) using **rounding and dyn. prg.**  
 $\Rightarrow C_{\max} \leq (1+\epsilon) \text{OPT}$
2. Add **short jobs** ( $\leq \epsilon \cdot \text{OPT}$ ) to the schedule using **LPT**.  
 $\Rightarrow C_{\max} \leq (1+\epsilon) \text{OPT}$

How to identify long/short jobs when we don't know  $\text{OPT}$ ?

- We **need the short jobs to be  $\leq \epsilon \cdot \text{OPT}$**  to ensure the approx. factor. For this purpose, we could use any lower bound on  $\text{OPT}$ , like  $P/m$ .
- But we also **need the long jobs to be  $\geq \epsilon \cdot \text{OPT}$**  to ensure the running time.

We will develop a family of algorithms with an algorithm  $B_k$  for each  $k \in \mathbb{Z}^+$ . ( $\varepsilon = \frac{1}{k}$ )

Scheduling the long jobs:

- (1) „Guess“ an optimal makespan  $T$ .  
The long jobs are those longer than  $T/k$ .
- (2) Round down each job size to the nearest multiple of  $T/k^2$ .
- (3) Use dyn. prg. to check whether optimal makespan  $\leq T$  for rounded long jobs.

Do binary search for  $T$  on the interval  $[L, U]$ , where

$$L = \max \left\{ \left\lceil \frac{P}{m} \right\rceil, p_{\max} \right\} \text{ and}$$

$$U = \left\lfloor \frac{P}{m} + \left(1 - \frac{1}{m}\right) p_{\max} \right\rfloor,$$

where  $P$  is the total size of long jobs.

$\beta_k(I)$

$$L \leftarrow \max \left\{ \left\lceil \frac{P}{m} \right\rceil, p_{\max} \right\}; \quad U \leftarrow \left\lceil \frac{P}{m} + \left(1 - \frac{1}{m}\right) p_{\max} \right\rceil$$

While  $L \neq U$

$$T \leftarrow \frac{1}{2} \lceil L + U \rceil$$



$I_x \leftarrow \{ \text{job } j \in I \mid p_j > T/k \}$  // Update set of long jobs

$I'_x \leftarrow I_x$  with each job size rounded down to nearest multiple of  $T/k^2$

Use **dyn. prg.** to pack  $I'_x$  in **bins of size T**

If #bins  $\leq m$

$$U \leftarrow T$$



else

$$L \leftarrow T + 1$$



$S'_x \leftarrow$  schedule of  $I'_x$  corresponding to the packing found by dyn. prg.

$S_x \leftarrow$  schedule of  $I_x$  corresponding to  $S'_x$

$S \leftarrow$  schedule of  $I$  obtained by adding **short jobs** to  $S_x$  using **LPT**

Binary search for T

Dyn. prog. as for bin packing:

$S'_x$  places  $\leq k$  jobs on each machine:

Each long job has size  $\geq T/k$

Since  $T/k$  is a multiple of  $T/k^2$ , each job in  $I'_x$  also has size  $\geq T/k$ .

There are  $\leq k^2$  different job sizes in  $I'_x$ , since no job is longer than  $T$ .

Hence, the configuration of a machine can be represented by a vector  $(s_1, s_2, \dots, s_{k^2})$ , where  $0 \leq s_i \leq k$ .

Thus,  $|\mathcal{B}| \leq (k+1)^{k^2}$ .

Table (B):

$\leq k^2$  dimensions (one for each size in  $I'_x$ )

$n_i + 1$  rows in dim.  $i$  ( $n_i = \#$  items of size  $i \cdot T/k^2$  in  $I'_x$ )

$$B(n_1, \dots, n_{k^2}) = 1 + \min_{S \in \mathcal{B}} \{ B(n_1 - s_1, \dots, n_{k^2} - s_{k^2}) \}$$

Running time:

# table entries:  $O(n^{k^2})$

Time per entry:  $|\mathcal{B}| \leq (k+1)^{k^2}$

# iterations of while loop:  $\log(U-L) \leq \log(p_{\max})$

Total time:  $O(n^{k^2} (k+1)^{k^2} \log(p_{\max}))$

## Approximation ratio:

When  $B_k$  terminates the while loop,  
 $\text{makespan}(S'_k) = T = \text{OPT}(I'_k)$

Since each of the  $\leq k$  jobs on a machine loses  $< T/k^2$  in the rounding,

$$\begin{aligned}\text{makespan}(S_k) &< \text{makespan}(S'_k) + k \cdot \frac{T}{k^2} \\ &= T + \frac{T}{k} \\ &= \left(1 + \frac{1}{k}\right) \text{OPT}(I'_k) \\ &\leq \left(1 + \frac{1}{k}\right) \text{OPT}(I), \text{ since } I_k \subseteq I, \text{ and} \\ &\quad \text{the job sizes are rounded down} \\ &\quad \text{to obtain } I'_k.\end{aligned}$$

Thus, if the last job to finish is a long job,  
 $B_k(I) < \left(1 + \frac{1}{k}\right) \text{OPT}(I)$ .

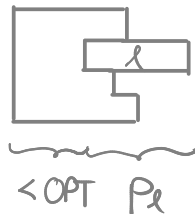
Otherwise, the last job to finish has

$$p_\ell \leq \frac{T}{k} \leq \frac{\text{OPT}(I'_k)}{k} \leq \frac{\text{OPT}(I)}{k}$$

Hence,

$$B_k(I) < \text{OPT}(I) + p_\ell \leq \left(1 + \frac{1}{k}\right) \text{OPT}$$

By the same argument  
as in the analysis of LS:



Thus, in both cases,  $B_k(I) < \left(1 + \frac{1}{k}\right) \text{OPT}$ .

Theorem 3.7 :  $\{B_k\}$  is a PTAS

Proof:

Let  $k = \lceil \frac{1}{\epsilon} \rceil$ . Then,

$B_k$  achieves an approx. factor of  $1+\epsilon$  with running time

$$O\left(\left(\left(\frac{1}{\epsilon}+1\right)n\right)^{\left(\frac{1}{\epsilon}\right)^2} \cdot \log(p_{\max})\right).$$

If  $\epsilon \in O(1)$ , this is poly. in the input size, since it takes  $\geq \log(p_{\max})$  bits to represent the job sizes.  $\square$

$\{B_k\}$  is not a FPTAS, since the running time is exponential in  $\frac{1}{\epsilon}$ .

Note that we did not expect a FPTAS, since the problem is strongly NP-complete...

The problem is **strongly NP-complete**, meaning that even the special case where  $\exists$  polynomial  $q$  s.t.  $p_{\max} \leq q(n)$ , for all input instances, is NP-complete.

(This means that, in contrast to Knapsack,  $\nexists$  pseudopoly. alg., unless  $P=NP$ .)

This implies that  **$\nexists$  FPTAS, unless  $P=NP$** :

Assume to the contrary that  $\exists$  FPTAS for the problem, i.e.,  $\exists$  family of algorithms  $\{A_\epsilon\}$ ,  $\epsilon > 0$ , with approx. factor  $1+\epsilon$  and running time poly. in  $n$  and  $\frac{1}{\epsilon}$ .

Consider the special case of the problem where  $\exists$  polynomial  $q$  s.t.  $p_{\max} \leq q(n)$ , for all instances. In this case,  $P \leq n \cdot q(n) \equiv p(n)$ .

For  $\epsilon = \frac{1}{p(n)}$ ,

- $\frac{1}{\epsilon}$  is poly. in  $n$ , so the running time of  $A_\epsilon$  is poly. in  $n$ .

- $A_\epsilon(I) \leq (1 + \frac{1}{p(n)}) \cdot \text{OPT}(I)$ , for any input  $I$   
 $< \text{OPT}(I) + 1$ , since  $\text{OPT}(I) < P \leq p(n)$

Thus, since  $A_\epsilon(I)$  is integer,  $A_\epsilon(I) = \text{OPT}(I)$ .

If  $P \neq NP$ , this **contradicts** the fact that the problem is **strongly NP-complete**.