

Minimizing the Number of Tardy Jobs

$$1 \parallel \sum U_j$$

Example

j	p_j	d_j
1	10	10
2	2	11
3	7	13
4	4	15
5	8	20

Ideas:

- Need to choose a subset of jobs S that meet their deadlines.
- Schedule the jobs that meet their deadlines in EDD order (Why?)
- Schedule the remaining jobs in an arbitrary order.

Question: How do you choose the subset?

Algorithm for $1||\Sigma U_j$

- Give an incremental algorithm
- Consider jobs in deadline order
- **Invariant:** Maintain a maximum cardinality set of jobs that meet their deadlines, among such sets, choose the one with with the smallest total amount of processing time.

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Algorithm

- Sort jobs by deadlines; $S = \emptyset$
- For each job j in deadline order
 - $S = S \cup \{j\}$
 - if j doesn't meet its deadline when S is scheduled
 - * $S = S - \{ \text{job in } S \text{ with largest processing time} \}$

Analysis

- Run time

- Need to sort – $O(n \log n)$
- Need to maintain the schedule for S and delete the job with largest processing time. (Maintain a set of numbers doing insert, delete and delete max operations).

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- Need to maintain the schedule for S and delete the job with largest processing time. (Maintain a set of numbers doing insert, delete and delete max operations). – Use a priority queue, each operations is $O(\log n)$ time.

Analysis: Proof by Induction. After each step k , let S_k denote S .

- S_k schedules a maximum sized subset of $\{ 1, \dots, k \}$
- Among all such subsets S_k is the one with the minimum total processing time.

Another Example

j	p_j	d_j
1	3	5
2	4	7
3	2	8
4	6	10
5	6	11
6	1	14
7	5	15

Special Case of a common deadline

- $1 || \Sigma U_j$ is easy.
- What about $1 || \Sigma w_j U_j$

Example

j	p_j	w_j
1	10	10
2	20	50
3	30	20

D is 40.

- We are choosing a minimum weight subset of jobs that miss their deadline
- Equivalently: we are choosing a maximum weight subset of jobs that make their deadlines.
- Equivalently: Choosing a maximum weight set of jobs that fit in a “bin” of certain size.

Knapsack

$$\max \sum_j w_j x_j$$

$$\text{s.t. } \sum_j p_j x_j \leq D$$

A one constraint lp, a knapsack problem.

- If you can take objects fractionally, then the greedy algorithm (w_j/p_j) is optimal.
- What about the integral (non-preemptive case).

Example

j	p_j	w_j
1	11	12
2	9	9
3	90	89

D is 100.

Solving Knapack Via Dynamic Programming

1. Non-polynomial. We will explicitly solve the problem for **all** possible values of either time or weight (in this example time.)
2. Polynomial would be polynomial in $n, m, \log W, \log D$, where $W = \max_j w_j$.
3. Running time will be polynomial in n, m, W, D . Called **pseudopolynomial**.
4. Reasonable approach when W and/or D is not too large.

Main Ideas:

- Parameterize solution, and define optimal solutions of a certain size in terms of solutions with smaller parameter values.
- Build up a table of solutions, eventually obtaining the solution for the desired parameter value.

DP for Knapsack: maximum weight competing by deadline

- $f(j, t)$ will be the best way to schedule jobs $1, \dots, j$ with t or less total processing time.
- Best means maximum total weight.

- What is $f(n, D)$?
- Maximum weight way to schedule all the jobs using at most D total processing time.
- This is the problem we want to solve.

DP

To schedule jobs $1, \dots, j$ using t total processing time there are two cases:

- job j is not scheduled.
- job j is scheduled

DP

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-
- If j is not scheduled, then the optimal solution for $1, \dots, j$ is the same as the optimal solution for $1, \dots, j - 1$, hence $f(j, t) = f(j - 1, t)$
 - If j is scheduled, then we need to add j to the schedule, hence we have to look at the optimal schedule using $t - p_j$ units of processing, hence: $f(j, t) = f(j - 1, t - p_j) + w_j$.

We don't know which case happens, so we try all and take the maximum

$$f(j, t) = \max\{f(j - 1, t), f(j - 1, t - p_j) + w_j\}$$

We initialize with $f(0, \cdot) = 0, f(\cdot, 0) = 0$, and anything with a negative index has a value of $-\infty$.

Example

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