Applied Algorithms CSCI-B505 / INFO-I500

Lecture 16.

Greedy Algorithms

- Greedy Algorithms
 - Interval Scheduling Problem
 - Interval Partitioning Problem
 - Bin-packing Heuristics

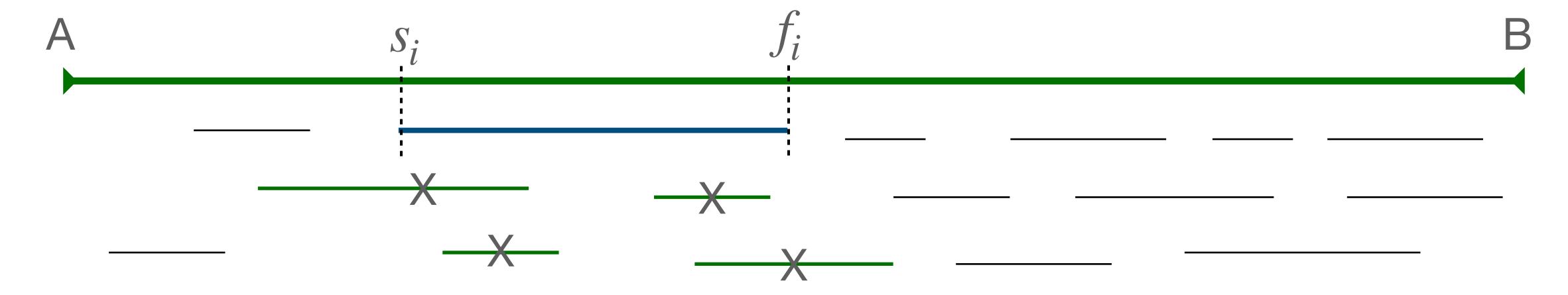
Greedy Algorithms

• While solving an optimization problem, we get through some steps, each of which requires a decision

 A greedy approach chooses the most appropriate decision at the moment without worrying about the future

• Greedy algorithms not always guarantee the optimum solution, but sometimes they do as well

• Given a list of n tasks, where the ith task starts at s_i and finishes at f_i , the aim is to select the largest subset of them without overlaps in time.



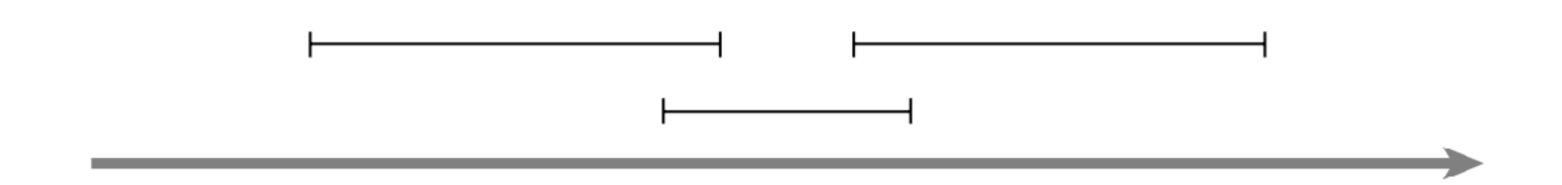
Select as many tasks as possible between A and B such that no two of them overlaps

• Let's sort them according to their starting times s_i , and select accordingly.



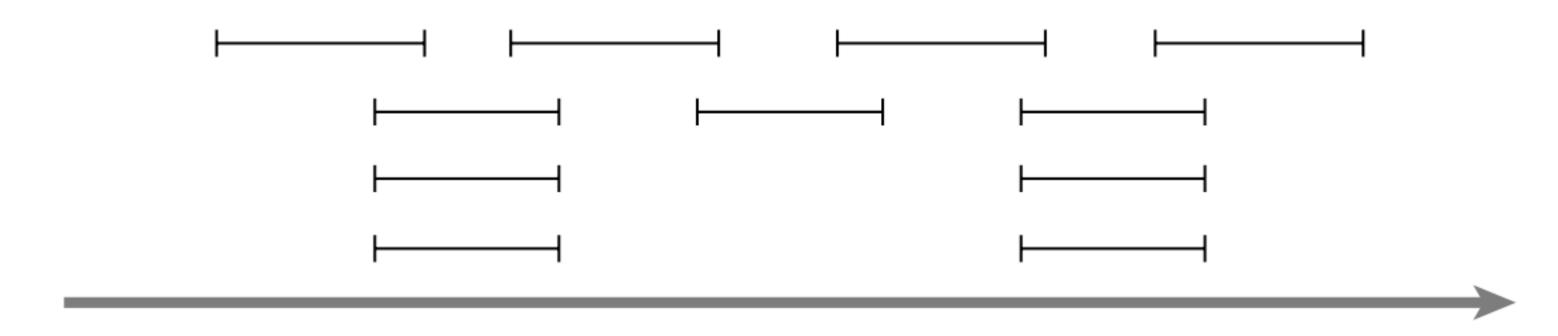
- Problem: Some long task that starts earlier may kill many smaller items that would otherwise increase the number of fulfilled was count.
- So, it worths to give credit to task duration?

• Let's sort them according to their durations $(f_i - s_i)$, and select from shorter to longer ones.



- Problem: A short task may kill longer ones that would otherwise contribute to the number of fulfilled task count.
- So, starting earlier or being shorter seems problematic.

How about selecting the tasks that have minimum collisions with the others?



- Problem: The optimal solution above would contain four tasks. However, the proposed heuristics selects the middle in the middle row that makes the answer three.
- Although minimal collision seems a lot better, there is a better selection strategy that guarantees the optimality!

• Select the item that finishes earliest! In other words, sort the tasks according to f_i , and execute them accordingly.

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Initially let R be the set of all requests, and let A be empty While R is not yet empty

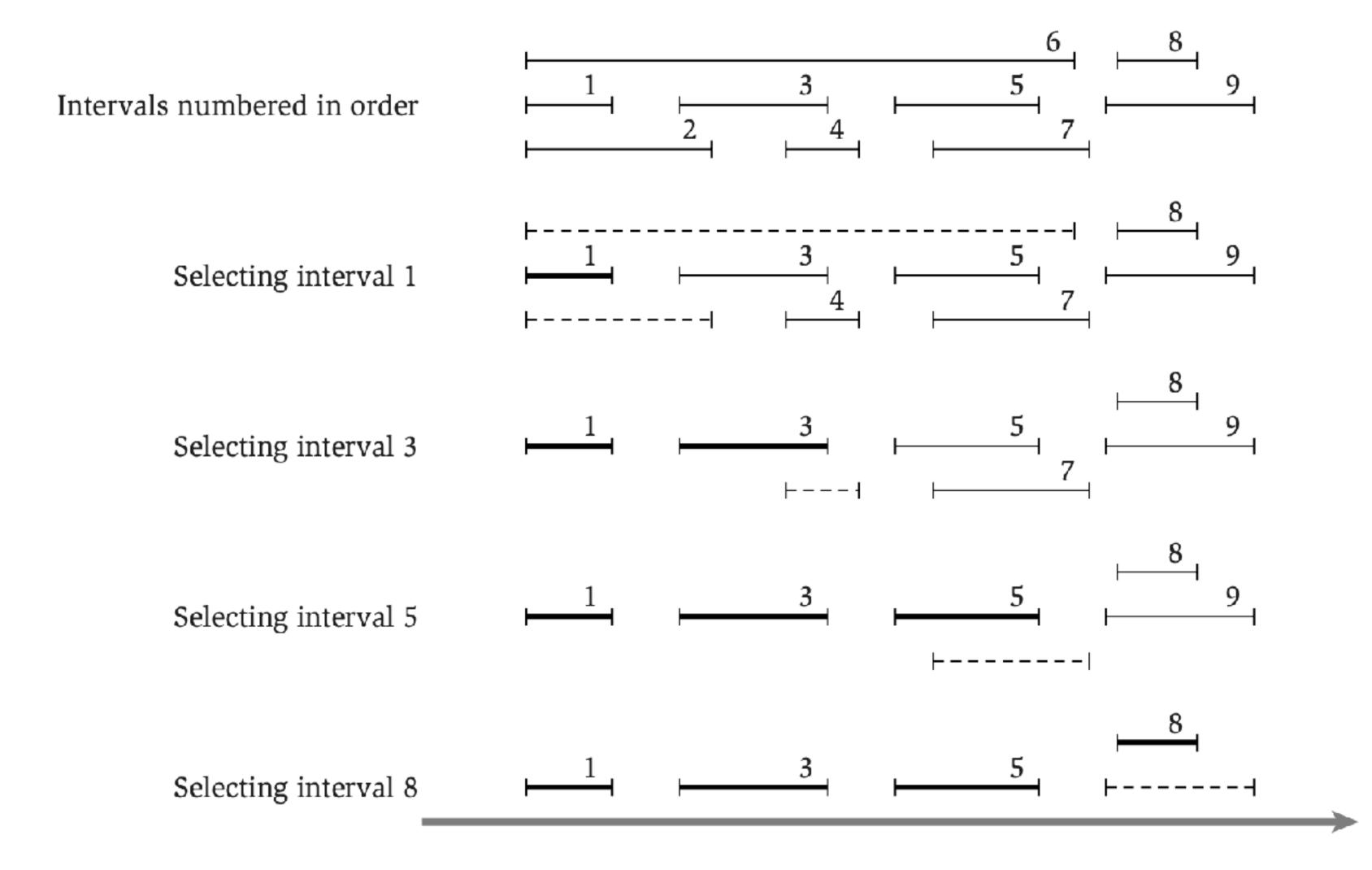
Choose a request i \in R that has the smallest finishing time Add request i to A

Delete all requests from R that are not compatible with request i

EndWhile

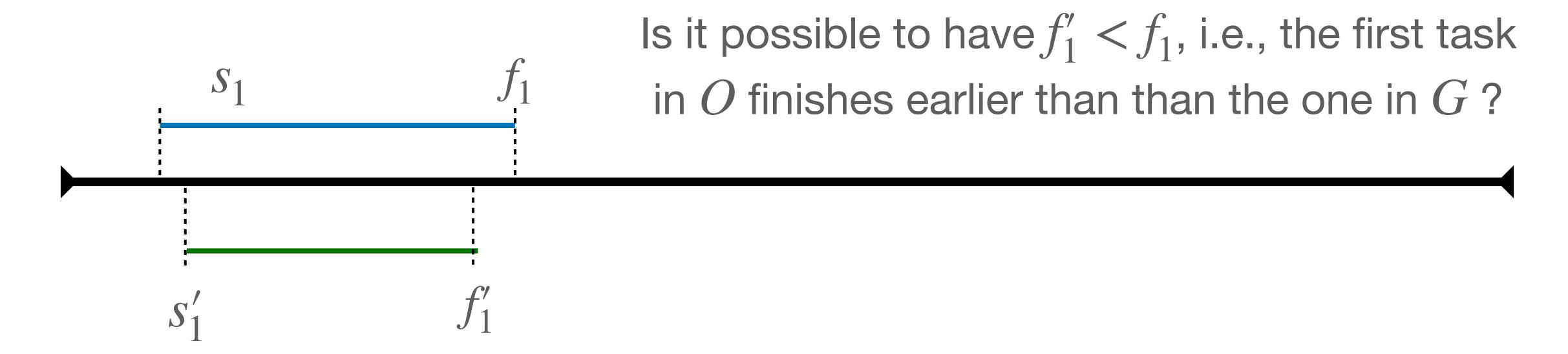
Return the set A as the set of accepted requests
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Will that generate the optimal solution?



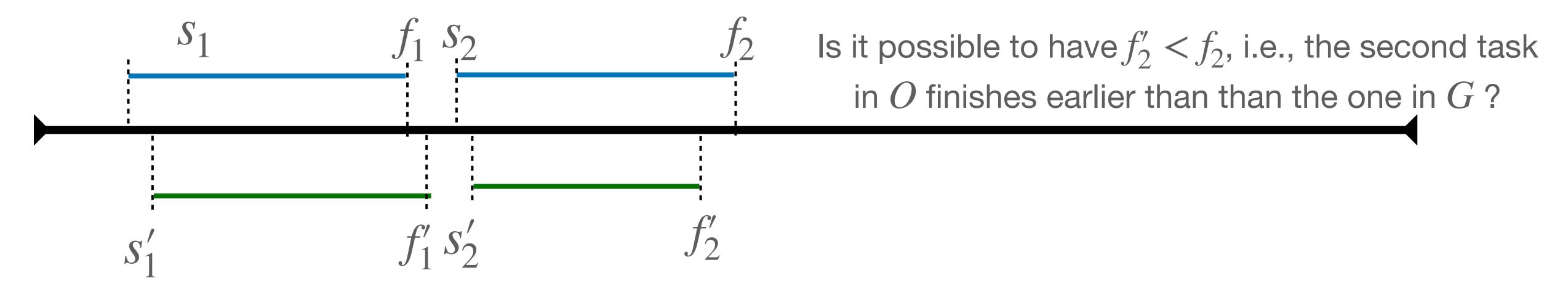
Will that generate the optimal solution?

- Optimal solution $O = \langle (s'_1, f'_1), (s'_2, f'_2), ...(s'_m, f'_m) \rangle$
- Greedy solution $G = \langle (s_1, f_1), (s_2, f_2), \ldots, (s_k, f_k) \rangle$
- If m=k, then G is optimal. Notice there may be multiple optimal solutions.



No! Because the greedy algorithm chooses the one with the earliest finish time among the list always.

- Optimal solution $O = \langle (s'_1, f'_1), (s'_2, f'_2), ...(s'_m, f'_m) \rangle$
- Greedy solution $G = \langle (s_1, f_1), (s_2, f_2), ..., (s_k, f_k) \rangle$
- If m=k, then G is optimal. Notice there may be multiple optimal solutions.



- Since $s_2' > f_1' > f_1$, the task (s_2', f_2') cannot conflict with the first task (s_1, f_1) .
- Therefore, if $f_2' < f_2$, then it should have been selected by the G.
- So, $f_2' < f_2$ is also impossible.
- And this is the case for all i = 1, 2, 3, ..., k (induction).

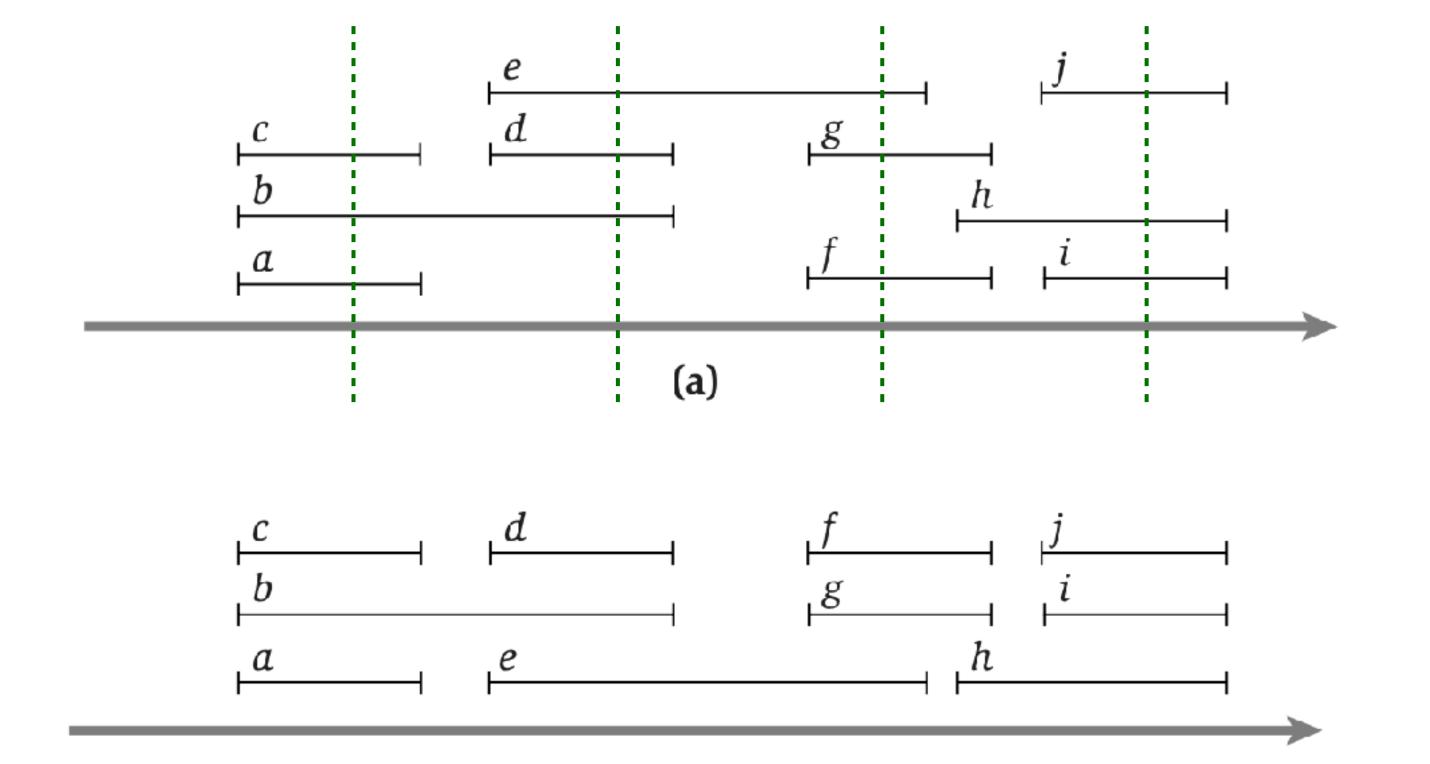
- Optimal solution $O = \langle (s'_1, f'_1), (s'_2, f'_2), ...(s'_m, f'_m) \rangle$
- Greedy solution $G = \langle (s_1, f_1), (s_2, f_2), \ldots, (s_k, f_k) \rangle$
- If m = k, then G is optimal. Notice there may be multiple optimal solutions.

Assume m > k, which means there is s'_{k+1}, f'_{k+1}

- Since $s'_{k+1} > f'_k > f_k$, the task (s'_{k+1}, f'_{k+1}) has no conflict with the task (s_k, f_k) .
- Then, the task (s'_{k+1}, f'_{k+1}) should not have been eliminated from the list
- This means the greedy algorithm cannot stop at such a k value.
- This proves $(k \ge m)$, and since m is optimal, k = m!
- Greedy approach is optimal...

Interval PARTITIONING Problem

- Now, assume multiple processors are available.
- We would like to finish all tasks with MINIMUM number of processors.
- Given a list of n tasks, where the ith task starts at s_i and finishes at f_i , the aim is to select the largest subset of them without overlaps in time.

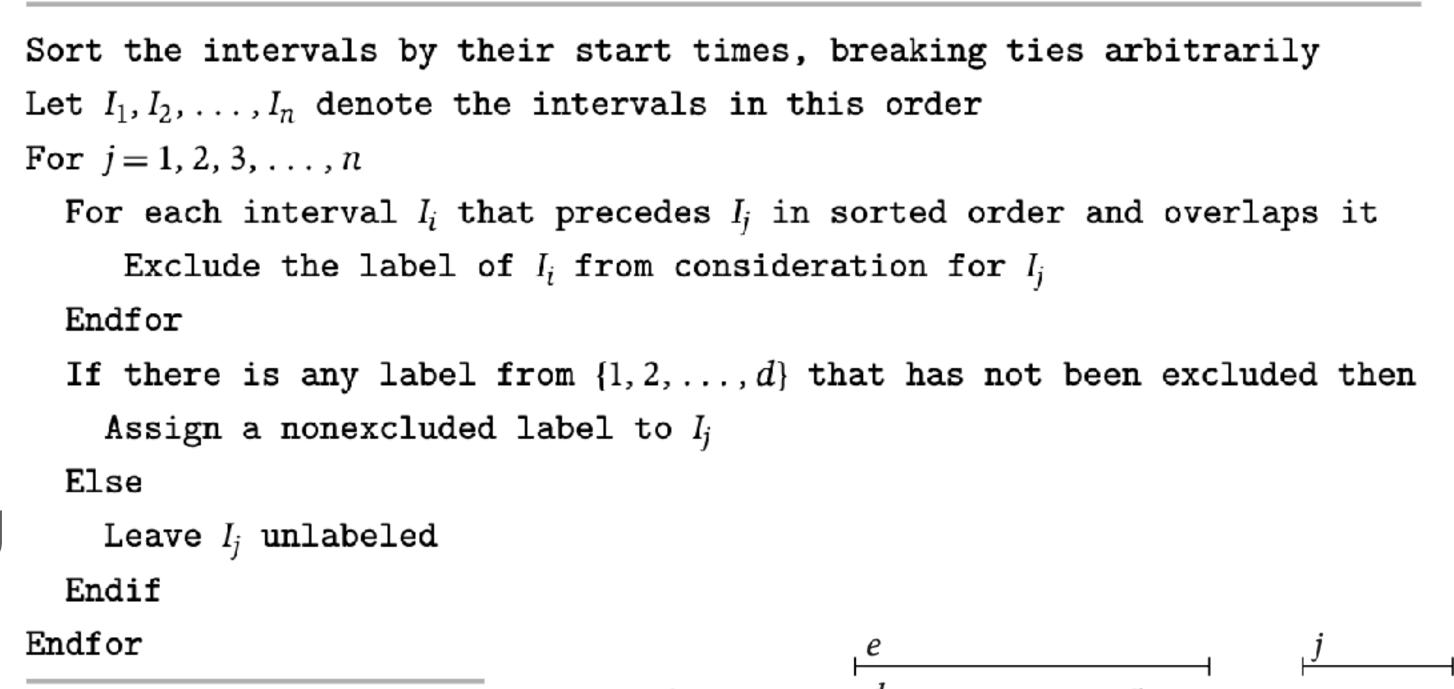


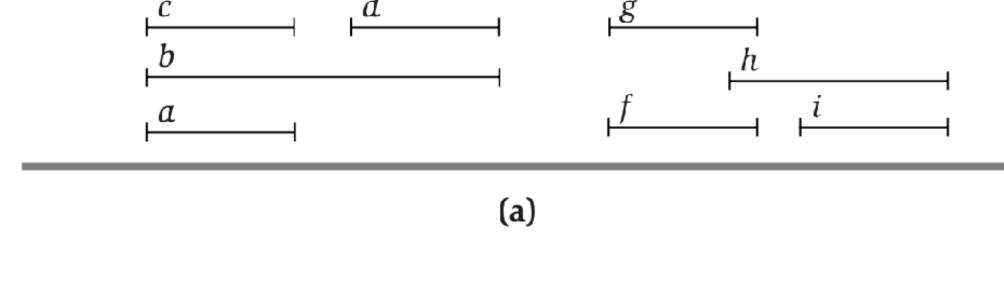
(b)

- What is the least number of processors we will need?
- The maximum number of conflicting tasks at a moment during the whole interval!
- Call this number depth.

Interval PARTITIONING Problem

- Sort all tasks according to their start times.
- For each task, assign the first label that has not been assigned to previous conflicting tasks.
 - Do we ever need labels more than the depth number?
 - No! Since depth is the maximum conflicting items at a time.
 - Thus, the greedy approach is optimal!





Greedy Algorithms

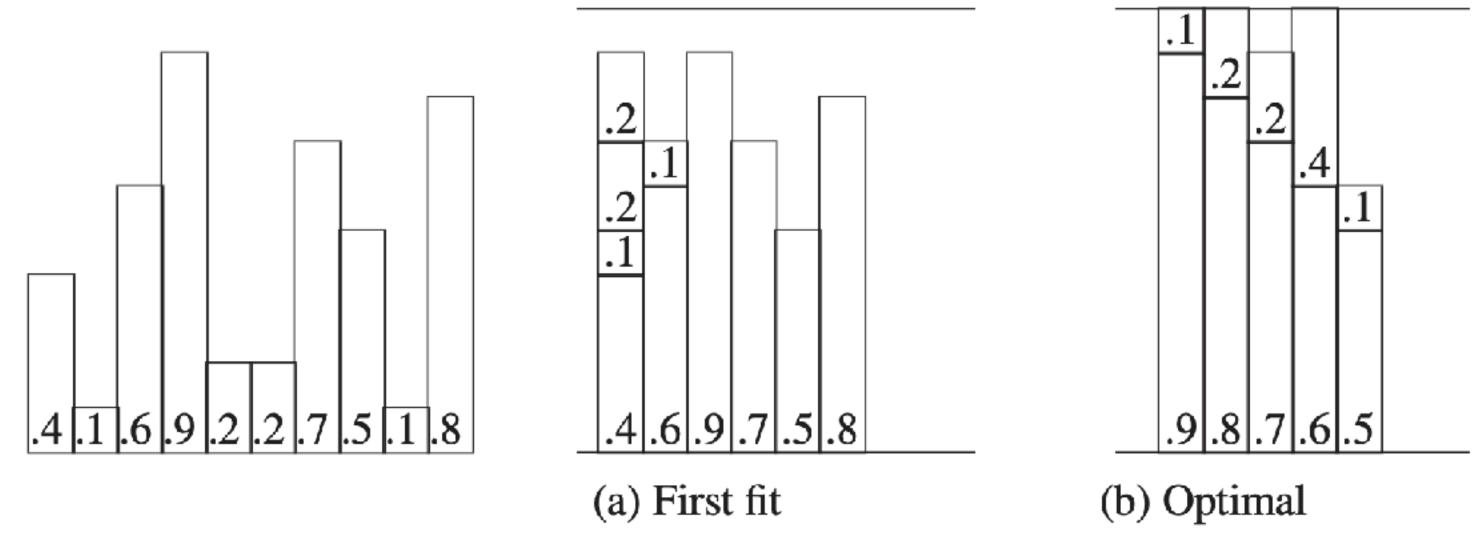


Figure 3.11. Bin packing. A list of 10 weights in the range [0,1]. In (a) the weights are packed by the first fit algorithm, which uses 6 bins. In (b) the weights are packed optimally in 5 bins.

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- Greedy algorithms cannot always produce the optimal!
- But, maybe preferred for good heuristics

Greedy Algorithms

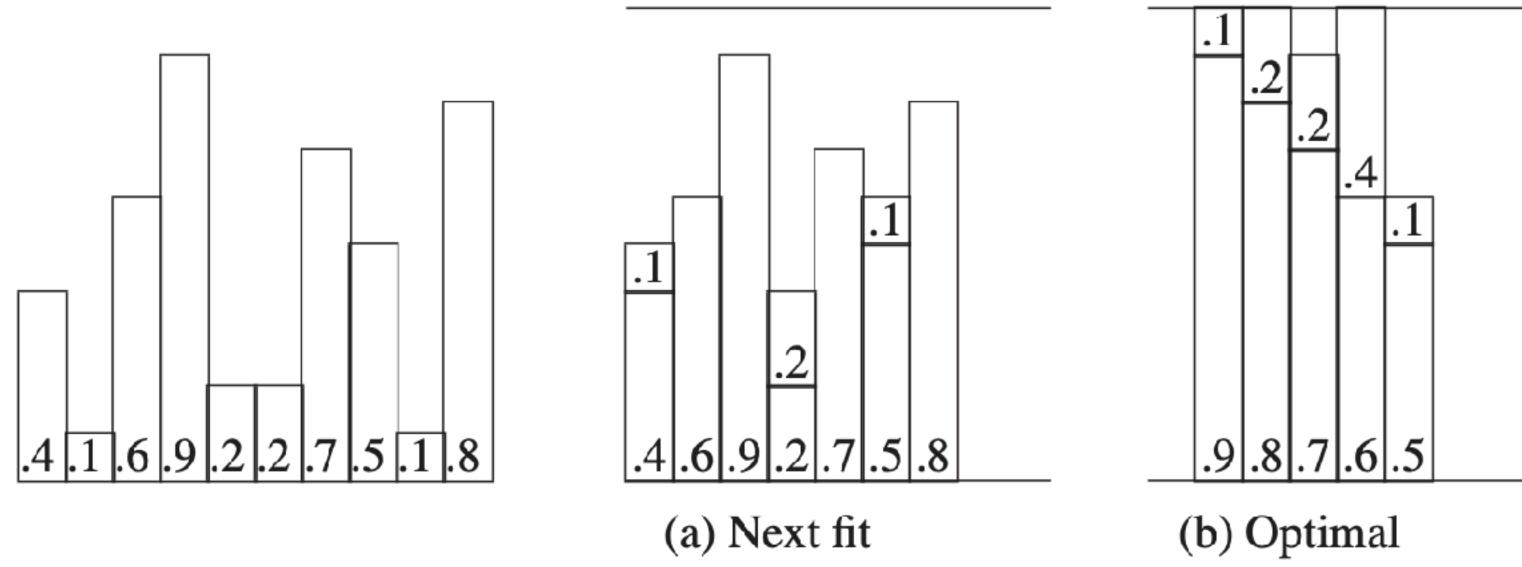


Figure 4.1. Bin packing. Two packings of the same list of 10 weights. The next fit packing uses seven bins, and the optimal packing uses five bins.

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- If the sequence was in the following order, the next fit produce the optimal:
 .9, .1, .8, .2, .7, .2, .6, .4, .5, .1
- So, the issue is to find the correct permutation of the input
- Yes, it is factorial in time.
 However, the solution improves during the search and we can stop when we find a goodenough solution.

- Greedy algorithms cannot always produce the optimal!
- They can even be helpful in the brute force search of the optimum.

Reading assignment

• Read the chapter 4.1 and 4.2 from Kleinberg&Tardos, chapter 16.1 and 16.2 from Cormen.