# CSCI3160 Design and Analysis of Algorithms (2025 Fall)

Greedy 1: Activity Selection

Instructor: Xiao Liang<sup>1</sup>

Department of Computer Science and Engineering Chinese University of Hong Kong

<sup>&</sup>lt;sup>1</sup>These slides are primarily based on materials prepared by Prof. Yufei Tao (please refer to Prof. Tao's version from 2024 Fall for the original content). Some modifications have been made to better align with this year's teaching progress, incorporating student feedback, in-class interactions, and my own teaching style and research perspective.

In this lecture, we will commence our discussion of **greedy** algorithms, which enforce a simple strategy: make the **locally optimal** decision at each step. Although this strategy does not always guarantee finding a **globally optimal** solution, sometimes it does. The nontrivial part is to prove (or disprove) the global optimality.

## Activity Selection — Real-World Example

**Story:** You're planning a one-day mini-conference. You have many candidate talks, each with a start time and an end time. You only have one room, so talks can't overlap.

Goal: Fit in as many talks as possible without time conflicts.

What does "overlap" mean? Two talks overlap if their time intervals share any time. Example:

- [9:00, 10:00] and [9:30, 10:30] overlap;
- [9:00, 10:00] and [10:00, 11:00] do *not* (if we allow back-to-back).

#### Your choices:

- You can include a talk or skip it.
- If you include it, you must avoid any other talk that overlaps with it.

## Activity Selection: Formal Setup

### Activity Selection

**Input:** A set S of n intervals of the form [s, f] where s and f are integers. **Output:** A subset T of disjoint intervals in S with the largest size |T|.

**Remark:** You can think of [s, f] as the duration of an activity, and consider the problem as picking the largest number of activities that do not have time conflicts.

Activity Selection

#### **Example:** Suppose

$$S = \{[1, 9], [3, 7], [6, 20], [12, 19], [15, 17], [18, 22], [21, 24]\}.$$

 $T = \{[3, 7], [15, 17], [18, 22]\}$  is an optimal solution, and so is  $T = \{[1, 9], [12, 19], [21, 24]\}$ .

## A Greedy Algorithm for Activity Selection

## Activity Selection

**One-line summary:** Pick the earliest-finishing interval, keep it, and discard all intervals that overlap it. Repeat until none remain.

### **Algorithm** (formal)

Repeat until *S* becomes empty:

- Add to T the interval  $\mathcal{I} \in S$  with the smallest finish time.
- ullet Remove from S all the intervals intersecting  ${\mathcal I}$  (including  ${\mathcal I}$  itself)

**Time Complexity:** Convince yourself that this algorithm can be implemented in  $O(n \log n)$  time.

**Example:** Suppose  $S = \{[1, 9], [3, 7], [6, 20], [12, 19], [15, 17], [18, 22], [21, 24]\}.$ 

Let us rearrange the intervals in S in ascending order of finish time:  $S = \{[3,7], [1,9], [15,17], [12,19], [6,20], [18,22], [21,24]\}.$ 

We first add [3,7] to T, after which intervals [3,7], [1,9] and [6,20] are removed. Now S becomes  $\{[15,17],[12,19], [18,22], [21,24]\}$ . The next interval added to T is [15,17], which shrinks S further to  $\{[18,22],[21,24]\}$ . After [18,22] is added to T, S becomes empty and the algorithm terminates.

# Why the Greedy Choice?

### **Greedy Choice Property:**

• Picking the earliest finishing activity leaves the most room for future choices.

### **Key Insight:**

• If we always choose the earliest finishing compatible activity, we never miss out on a better solution.

# Proof Outline: Exchange Argument

**Note:** there may be more than one optimal solution (think: can you construct an example for this fact?). Therefore, it is not accurate to say that our algorithm outputs "the" optimal solution. It is sufficient to show that our algorithm outputs an optimal solution.

#### Let:

- $G = \{g_1, g_2, ..., g_k\}$  be the greedy solution
- $O = \{o_1, o_2, ..., o_m\}$  be an optimal solution, ordered by increasing finish time.

#### Remarks:

- Each  $g_i$  (or  $o_i$ ) is an interval!
- Since there is no overlap among the  $o_i$ 's, it holds for each i that  $f_{o_{i-1}}$  is earlier than  $s_{o_i}$ , which is the start time of  $o_i$ .

**Goal:** Show that k = m (greedy is optimal)

## Some Simple Facts

#### Let:

- $G = \{g_1, g_2, ..., g_k\}$  be the greedy solution
- $O = \{o_1, o_2, ..., o_m\}$  be an optimal solution, ordered by increasing finish time.

#### Facts:

- Each  $g_i$  (or  $o_i$ ) is an interval!
- ② Since there is no overlap among the  $o_i$ 's, it holds for each i that  $f_{o_{i-1}}$  is earlier than  $s_{o_i}$ , which is the start time of  $o_i$ .
- **1** It must be that  $k \le m$ ; Otherwise, it violates the optimality of the optimal solution O.
- So, to prove k = m (our goal), it suffices to show that k is no less than m.

## Exchange Argument in Action

**Strategy:** Transform O into G without reducing the number of activities using an **exchange** argument.

#### Assume:

- Greedy picks  $g_1$  (earliest finishing)
- Optimal picks o<sub>1</sub> (maybe different)
- $f_{g_1} \leq f_{o_1}$  (where  $f_{g_1}$  denotes the finish time of activity  $g_1$ . Similar for  $f_{o_1}$ .)

### **Exchange Step:**

- Replace  $o_1$  with  $g_1$  in O
- Since  $g_1$  finishes no later than  $o_1$ , it's compatible with all following activities in O

Repeat the above steps to replace  $o_2, o_3, \ldots$ 

## Finishing the Proof

If we repeat the above exchange argument, we can gradually update O in the following order:

$$O = \{o_1, o_2, o_3, \dots, o_m\}$$

$$O_1 = \{g_1, o_2, o_3, \dots, o_m\}$$

$$O_2 = \{g_1, g_2, o_3, \dots, o_m\}$$

$$\vdots$$

Assuming for the sake of contradiction that k < m, the following sequence would stop at

$$O_k = \{g_1, g_2, \dots, g_k, o_{k+1}, o_{k+2}, \dots, o_m\}.$$

# Finishing the Proof

Such an  $O_k$  contradicts the description of the greedy algorithm:

• Note that the finish time of  $g_k$  is earlier than the finish time of all intervals in

$$\{o_{k+1},o_{k+2}\ldots,o_m\},\,$$

and  $g_k$  do not over lap with any one of them.

- Given the above, by definition, the greedy algorithm would **not** stop at  $g_k$ , because there are still intervals satisfying the algorithm's condition to be added to the solution set G.
- Contradiction: However, by our assumption,  $g_k$  is the last element in G. This gives a contradiction.

Thus, our assumption that k < m is incorrect. It must be that k = m.

This finish the proof