

# **DESIGN AND ANALYSIS OF ALGORITHMS**

**CS 4120/5120**

**GREEDY STRATEGY – ACTIVITY SELECTION**

# AGENDA

- Activity selection problem
- Greedy method

# WHO GETS TO SEE THE DENTIST AND WHEN

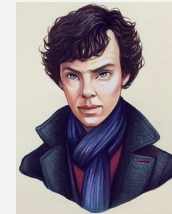
- Dentist Smileyface is very popular in Hollywood. She routinely receives timeslots from celebrities who want to come in on a certain day.
- Being rich, all celebrities pay her a flat \$5000 fee, irrespective of how long they stay.
- Considering they are celebrities, you cannot have more than one in the office at the same time (they bite).



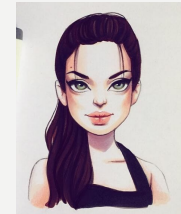
# WHO GETS TO SEE THE DENTIST AND WHEN

- Also, considering they are celebrities, they only come in as per their own schedule.
- Example request (on the same day) are shown on the right
  - Some requests overlap
  - All celebrities need to see Dr. Smileface (obviously)
  - They do not stay for the same amount of time.

- How can we help Dr. Smileyface to earn maximum revenue possible?



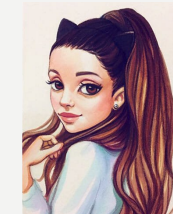
Tom: [8AM – 2PM]



Meg: [9am – 10AM]



Emily: [11AM-4PM]



Britney: [5PM-520PM]

# SCHEDULING FOR RUNNING COMPETITIONS

- The athletic department is trying to schedule *a day of running competitions*. The competitions include potentially the activities shown in the table.

- Some activities have more contestants than other activities, hence they need more hours

Running Competition	Distance (meter)	Time	Running Competition	Distance (meter)	Time
Sprint	100	8am ~ 9 am	Middle distance	800	11 am ~ 1 pm
	200	9:30am ~ 11am		1500	4 pm ~ 5 pm
Hurdles	100	2 pm ~ 2:30 pm	Relays	4 × 100	10:30 am ~ 11:10 am
	400	3 pm ~ 3:30 pm		4 × 400	12 pm ~ 12:40 pm

- Assume two activities can be schedule back-to-back.
- How can we help the Athletic Department schedule **as many activities as possible in one day?**

# EXAMINE THE TWO PROBLEMS

- Fill out the table

Problem	# of Event (Celebs)	Resources required	Goal
Celebrities seeing a dentist	Four celebrities	Dr. SmileyFace	The doctor earns maximum revenue (or see as many patients as possible in a day)
Scheduling running competitions	Eight running competitions	One running track	Schedule maximum-size running competitions in a day
Summary	Several competing events	Exclusive use of a common resource	Select maximum-size activities that do not overlap

# ACTIVITY-SELECTION PROBLEM

- The two stories can be modeled as one *activity-selection problem*.
- In the activity-selection problem, we wish to select a *maximum-size* subset of *mutually compatible* activities.
- Next, we shall see rigorous definition of the problem.

# THE DEFINITION OF ACTIVITY-SELECTION PROBLEM

- Given a set  $S = \{a_1, a_2, \dots, a_n\}$  of  $n$  proposed activities that wish to **use a resource**.
- Each activity  $a_i$  has a **start time  $s_i$**  and a **finish time  $f_i$** , where  $0 \leq s_i < f_i < \infty$ .
  - If selected, activity  $a_i$  takes place during **the half-open time interval  $[s_i, f_i)$** .



# START AND FINISH TIME

- Example

- Given  $S = \{a_1, a_2, a_3\}$  and their start and finish times

- $s_1 = 2, f_1 = 6$

- ~~$s_2 = 5, f_2 = 3$~~   $s_2 = 3, f_2 = 5$

- $s_3 = 5, f_3 = 7$

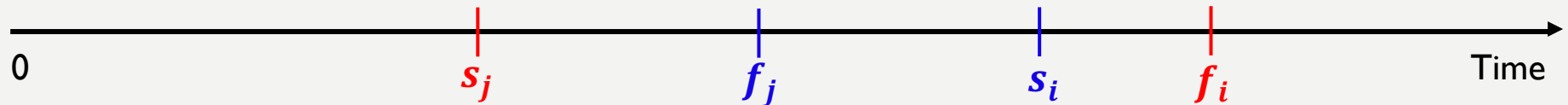
- The start and finish time of activity  $a_2$  is NOT plausible.

- Flip the values of start and finish time of the activity mentioned above.

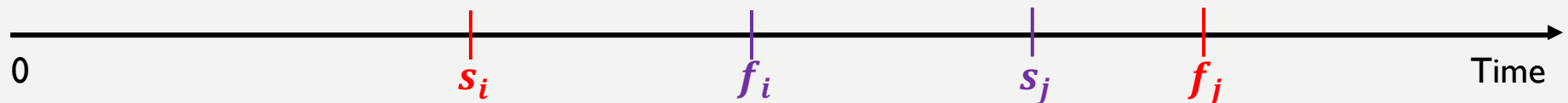
- After the flip, activity  $a_2$  and  $a_3$  can be scheduled back-to-back.

# COMPATIBLE ACTIVITIES

- Activities  $a_i$  and  $a_j$  are **compatible** if the intervals  $[s_i, f_i)$  and  $[s_j, f_j)$  do not overlap.
  - That is,  $a_i$  and  $a_j$  are compatible if
    - $s_i \geq f_j$ , indicating that activity  $a_j$  takes place before  $a_i$ .

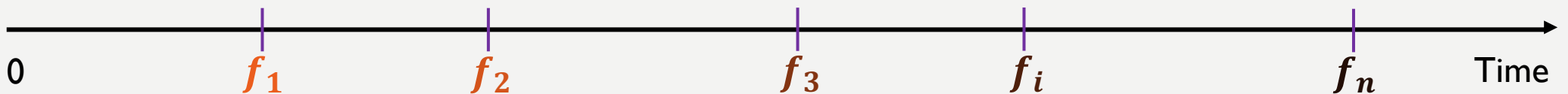


- $s_j \geq f_i$ , indicating that activity  $a_i$  takes place before  $a_j$ .



# THE COMPLETE DEFINITION OF THE ACTIVITY-SELECTION PROBLEM

- Given a set  $S = \{a_1, a_2, \dots, a_n\}$  of  $n$  proposed activities that wish to **use a resource**. We wish to select a **maximum-size subset** of **mutually compatible activities**.
  - By **mutually compatible activities**, we mean no two activities in the subset overlap.
- We assume that the activities are **sorted** in monotonically **increasing** order of finish time, i.e.,  $f_1 \leq f_2 \leq f_3 \leq \dots \leq f_{n-1} \leq f_n$ .



# THE PROBLEM DEFINITION

## PRACTICE #1

- Consider the activities shown in the table.
  - Rearrange (Sort) the activities such that they qualify for the definition of the activity-selection problem.
    - Do not move the indexes when sorting the activities.
    - If two activities finish at the same time, order them by their start times.

$i$	1	2	3	4	5
$s_i$	6	2	1	3	4
$f_i$	9	10	6	9	5
$i$	1	2	3	4	5
$s_i$					
$f_i$					

# THE PROBLEM DEFINITION

## PRACTICE #2

- Consider the activities after the rearrangement.
  - List all **pairs** of activities that are compatible. Write the qualifying pairs **in the form of**  $\langle a_i, a_j \rangle$ .
    - $\langle a_1, a_4 \rangle$
    - $\langle a_2, a_4 \rangle$
    - 
    -

$i$	1	2	3	4	5
$s_i$	4	1	3	6	2
$f_i$	5	6	9	9	10

# AN INSTANCE

- Given a list of activities and their respective start and finish time.

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	5	3	12	5	3	1	6	8	0	2	8
$f_i$	9	5	16	7	9	4	10	11	6	14	12

The activities must be sorted in monotonically **increasing** order of their **finish** times.

- Reorder the activities such that they can be an instance of the activity-selection problem.
  - The indices must remain where they are.
  - If two activities have the same finish time, order them in increasing order of their start times.

# AN INSTANCE REORDERING

- Before

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	5	3	12	5	3	1	6	8	0	2	8
$f_i$	9	5	16	7	9	4	10	11	6	14	12

- After

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16

# SOLVING THE PROBLEM FOR THE GIVEN INSTANCE STEP #1

- Goal
  - We wish to find a maximum-size subset of mutually compatible activities.

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16

- To schedule as many activities as possible, we want to first select activity  $a_1$ .
- The remaining compatible activities are  $a_4, a_6, a_7, a_8, a_9, a_{11}$ .
- The next step is finding a maximum-size subset of mutually compatible activities of  $a_4, a_6, a_7, a_8, a_9, a_{11}$



# SOLVING THE PROBLEM FOR THE GIVEN INSTANCE STEP #2

- Goal of step #2
  - We wish to find a **maximum-size subset** of **mutually compatible** activities of the remaining activities.

$i$	1	4	6	7	8	9	11
$s_i$	1	5	5	6	8	8	12
$f_i$	4	7	9	10	11	12	16

- To schedule as many activities as possible, we want to first select activity  $a_4$ .
- The remaining compatible activities are  $a_8, a_9, a_{11}$ .
- The next step is finding a **maximum-size subset** of **mutually compatible** activities of  $a_8, a_9, a_{11}$

# SOLVING THE PROBLEM FOR THE GIVEN INSTANCE STEP #3

- Goal of step #3
  - We wish to find a **maximum-size subset** of **mutually compatible** activities of the remaining activities.

$i$	1	4	8	9	11
$s_i$	1	5	8	8	12
$f_i$	4	7	11	12	16

- To schedule as many activities as possible, we want to first select activity  $a_8$ .
- The remaining compatible activities are  $a_{11}$ .
- The next step is finding a **maximum-size subset** of **mutually compatible** activities of  $a_{11}$

# SOLVING THE PROBLEM FOR THE GIVEN INSTANCE STEP #4

- Final step
  - Conclude: the **maximum-size subset** of **mutually compatible** activities of  $S = \{a_1, a_2, \dots, a_{11}\}$  is

$i$	1	4	8	11
$s_i$	1	5	8	12
$f_i$	4	7	11	16

- Generally
  - We pick the activity with the earliest finish time.
  - Solve the rising subproblem on the remaining activities that are compatible with the chosen one.

# ATTEMPTING TO SOLVE BY DP

- Here is a checklist of the qualifications of a DP problem.
  - ☒ Optimization problem
    - ☒ Goal is to find a **maximum-size subset** of **mutually compatible** activities
  - ☐ Two key ingredients
    - ☐ Optimal substructure
    - ☐ Overlapping subproblems

# DISCOVERING THE OPTIMAL SUBSTRUCTURE

- **General steps**

- **Step 1:** A solution to the problem consists of making a choice.
- **Step 2:** Suppose that for a given problem, you are given the choice that leads to an optimal solution.
- **Step 3:** Given this choice, you determine which subproblems ensue and how to best characterize the resulting space of subproblems.
- **Step 4:** Show the solutions to the subproblems used within an optimal solution to the problem must themselves be optimal by using a “cut-and-paste” technique.

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 1

- **Step 1:** A solution to the problem consists of making a choice.
- The activity-selection problem can be solved by making a choice.

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 2

- **Step 2:** Suppose that for a given problem, you are given the choice that leads to an optimal solution.
  - At this point, you do not concern yourself with how to determine this choice.
- Previously, we selected activity  $a_1$ . Based off the selection, we proceeded to solve the problem.

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16

- We **DO NOT** know if picking  $a_1$  leads to an optimal solution.

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 2 (CONT'D)

- **Step 2:** Suppose that for a given problem, you are given the choice that leads to an optimal solution.
- Apply abstraction
  - Denote by  $S_{ij}$  the set of activities that start after activity  $a_i$  finishes and that finish before activity  $a_j$  starts.
    - The activities included in  $S_{ij}$  are highlighted in the table below.

$i$	$i$	$i + 1$	...	$k - 1$	$k$	$k + 1$	...	$j - 1$	$j$
$S_i$	$S_i$	$S_{i+1}$	...	$S_{k-1}$	$S_k$	$S_{k+1}$	...	$S_{j-1}$	$S_j$
$f_i$	$f_i$	$f_{i+1}$	...	$f_{k-1}$	$f_k$	$f_{k+1}$	...	$f_{j-1}$	$f_j$



# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 2 (CONT'D)

- **Step 2:** Suppose that for a given problem, you are given the choice that leads to an optimal solution.
- Apply abstraction
  - In general, for a given set of activities  $S_{ij}$ . We suppose that we are given a choice  $a_k, i \leq k \leq j$ , that leads to an optimal solution.

$i$	$i$	$i + 1$	...	$k - 1$	$k$	$k + 1$	...	$j - 1$	$j$
$S_i$	$S_i$	$S_{i+1}$	...	$S_{k-1}$	$S_k$	$S_{k+1}$	...	$S_{j-1}$	$S_j$
$f_i$	$f_i$	$f_{i+1}$	...	$f_{k-1}$	$f_k$	$f_{k+1}$	...	$f_{j-1}$	$f_j$

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 3

- **Step 3:** Given this choice, you determine which subproblems ensue and how to best characterize the resulting space of subproblems.
- After we include activity  $a_k$  in an optimal solution, \_\_\_\_\_ subproblems arise.
  - # 1: Finding the **maximum-size subset** of mutually compatible activities of  $S_{ik}$ .
  - # 2: Finding the **maximum-size subset** of mutually compatible activities of  $S_{kj}$ .

$i$	$i$	$i + 1$	...	$k - 1$	$k$	$k + 1$	...	$j - 1$	$j$
$S_i$	$S_i$	$S_{i+1}$	...	$S_{k-1}$	$S_k$	$S_{k+1}$	...	$S_{j-1}$	$S_j$
$f_i$	$f_i$	$f_{i+1}$	...	$f_{k-1}$	$f_k$	$f_{k+1}$	...	$f_{j-1}$	$f_j$

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 4

- **Step 4:** Show the solutions to the subproblem used within an optimal solution to the problem must themselves be optimal by using a “cut-and-paste” technique.
- We want to show that the **maximum-size subset** of mutually compatible activities of  $S_{ij}$  includes within itself the following.
  - # 1: The **maximum-size subset** of mutually compatible activities of  $S_{ik}$ .
  - # 2: The **maximum-size subset** of mutually compatible activities of  $S_{kj}$ .

$i$	$i$	$i + 1$	...	$k - 1$	$k$	$k + 1$	...	$j - 1$	$j$
$S_i$	$S_i$	$S_{i+1}$	...	$S_{k-1}$	$S_k$	$S_{k+1}$	...	$S_{j-1}$	$S_j$
$f_i$	$f_i$	$f_{i+1}$	...	$f_{k-1}$	$f_k$	$f_{k+1}$	...	$f_{j-1}$	$f_j$

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 4 (CONT'D)

- **Step 4:** Show the solutions to the subproblem used within an optimal solution to the problem must themselves be optimal by using a “cut-and-paste” technique.
- Repeatedly saying the **maximum-size subset** of mutually compatible activities of  $S_{ij}$  can be a hassle. Let us denote such a **maximum-size subset** by  $A_{ij}$ .
  - In other words,  $A_{ij}$  is an optimal solution to  $S_{ij}$ .

$i$	$i$	$i + 1$	...	$k - 1$	$k$	$k + 1$	...	$j - 1$	$j$
$S_i$	$S_i$	$S_{i+1}$	...	$S_{k-1}$	$S_k$	$S_{k+1}$	...	$S_{j-1}$	$S_j$
$f_i$	$f_i$	$f_{i+1}$	...	$f_{k-1}$	$f_k$	$f_{k+1}$	...	$f_{j-1}$	$f_j$

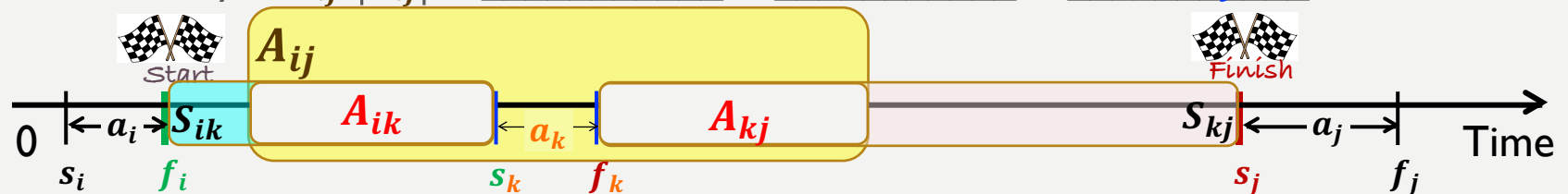
# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 4 (CONT'D)

- **Step 4:** Notations.

- The goal of step 4 can be rephrased as showing that the  $A_{ij}$  ~~the maximum size subset of mutually compatible activities of  $S_{ij}$~~  includes within itself the following.
  - # 1: ~~The maximum size subset of mutually compatible activities of  $S_{ik}$ .~~  $A_{ik}$
  - # 2: ~~The maximum size subset of mutually compatible activities of  $S_{kj}$ .~~  $A_{kj}$
- In other words, we want to prove that  $A_{ij} = \underline{A_{ik}} \cup \underline{\{a_k\}} \cup \underline{A_{kj}}$ .
  - The  $\cup$  operator indicates these are set operations.

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 4 (CONT'D)

- **Step 4:** Proof of optimality of sub-solutions using a “cut-and-paste” technique.
- $A_{ij} = A_{ik} \cup \{a_k\} \cup A_{kj}$
- The goal is to prove that  $A_{ik}$  and  $A_{kj}$  are the **maximum-size subset** of mutually compatible of  $S_{ik}$  and  $S_{kj}$ , respectively.
  - Assume** that  $A_{ik}$  is **NOT** an optimal solution in  $S_{ik}$ , and that  $A_{kj}$  is **NOT** an optimal solution in  $S_{kj}$ .
    - Instead, let  $A_{ik} = S_{ik} \cap A_{ij}$  and  $A_{kj} = S_{kj} \cap A_{ij}$ , yielding  $A_{ij} = A_{ik} \cup \{a_k\} \cup A_{kj}$ .
    - The cardinality of  $A_{ij}$ ,  $|A_{ij}| = |A_{ik}| + 1 + |A_{kj}|$ .



# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 4 (CONT'D)

- **Step 4:** Proof of optimality of sub-solutions using a “cut-and-paste” technique. (Continued)
- The goal is to prove that  $A_{ik}$  and  $A_{kj}$  are the **maximum-size subset** of mutually compatible of  $S_{ik}$  and  $S_{kj}$ , respectively.
  - ii. **There exist an optimal solution**, denoted by  $A_{ik}^*$  in  $S_{ik}$ , and  $A_{kj}^*$  in  $S_{kj}$ .
    - a. The following relationships hold  $|A_{ik}^*| > |A_{ik}|$  and  $|A_{kj}^*| > |A_{kj}|$ .

# DISCOVERING THE OPTIMAL SUBSTRUCTURE STEP 4 (CONT'D)

- **Step 4:** Proof of optimality of sub-solutions using a “cut-and-paste” technique. (Continued)
- The goal is to prove that  $A_{ik}$  and  $A_{kj}$  are the **maximum-size subset** of mutually compatible of  $S_{ik}$  and  $S_{kj}$ , respectively.
  - iii. We can **construct a** subset  $A_{ij}^*$  as follows,  $A_{ij}^* = A_{ik}^* \cup \{a_k\} \cup A_{kj}^*$
  - iv. Obviously,  $A_{ij}^*$  contains the mutually compatible activities of set  $S_{ij}$ , and the cardinality of  $A_{ij}^*$ ,  $|A_{ij}^*| = |A_{ik}^*| + 1 + |A_{kj}^*| > |A_{ij}|$ , **contradicting the supposition** that  $A_{ij}$  is the **maximum-size subset** of mutually compatible activities of  $S_{ij}$ .
  - v. **Therefore**,  $A_{ik}$  is an optimal solution in  $S_{ik}$  and  $A_{kj}$  is an optimal solution in  $S_{kj}$ .



# DISCOVERING THE OPTIMAL SUBSTRUCTURE CONCLUSION

- Given a set  $S = \{a_1, a_2, \dots, a_n\}$  of  $n$  proposed activities that wish to **use a resource**. We wish to select a **maximum-size subset** of mutually compatible activities.
- Denote by  $S_{ij}$  the **set of activities** that **start after** activity  $a_i$  finishes and that **finish before** activity  $a_j$  starts.
- Let  $A_{ij}$  denote the **maximum-size subset** of mutually compatible activities of  $S_{ij}$ .
- The **optimal substructure**
  - The **optimal** solution  $A_{ij}$  must also include **optimal** solutions to the two subproblems for  $S_{ik}$  and  $S_{kj}$ , i.e.,  $A_{ij} = A_{ik} \cup \{a_k\} \cup A_{kj}$ .

# ATTEMPTING TO SOLVE BY DP

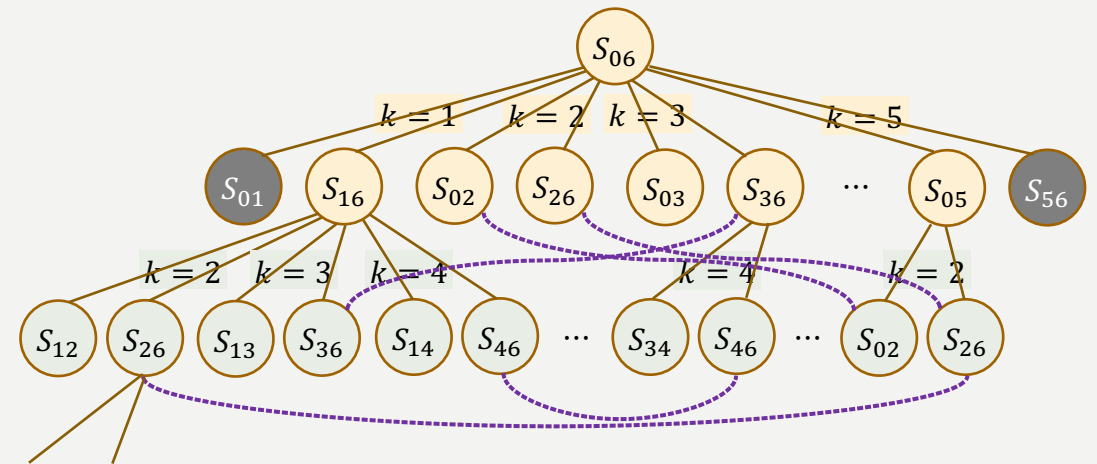
- Here is a checklist of the qualifications of a DP problem.
  - ☒ Optimization problem
    - ☒ Goal is to find a **maximum-size subset** of **mutually compatible** activities
  - ☐ Two key ingredients
    - ☒ Optimal substructure
    - ☐ Overlapping subproblems

# DISCOVER OVERLAPPING SUBPROBLEMS

- Draw the corresponding subproblem graph to the input instance

$i$	1	2	3	4	5
$s_i$	1	3	0	5	3
$f_i$	4	5	6	7	9

- Let a vertex  $S_{ij}$  represent the size of the (sub)problem.
  - Insert two dummy activities  $a_0$  and  $a_6$



# ATTEMPTING TO SOLVE BY DP

- Here is a checklist of the qualifications of a DP problem.
  - ✓ ☒ Optimization problem
    - ✓ ☒ Goal is to find a **maximum-size subset** of **mutually compatible** activities
  - ✓ ☒ Two key ingredients
    - ✓ ☒ Optimal substructure
    - ✓ ☒ Overlapping subproblems

# MOVING FORWARD WITH DYNAMIC PROGRAMMING

- **General steps**

- **Step 1:** Characterize the structure of an optimal solution
- **Step 2:** Recursively define the **value** of an optimization.
  - The **value** of an optimization in the activity-selection problem is the size of an optimal solution for the set  $S_{ij}$ . We will denote the size by  $c[i, j]$ , then we would have the following recurrence

$$c[i, j] = \begin{cases} 0 & \text{if } S_{ij} = \emptyset, \\ \max_{a_k \in S_{ij}} \{c[i, k] + c[k, j] + 1\} & \text{if } S_{ij} \neq \emptyset. \end{cases}$$

- **Step 3:** Compute the **value** of an optimal solution. (Top-down or Bottom-up with **memoization**)
- **Step 4:** Construct the optimal solution from the computed information.

# SEEKING A FASTER WAY

- From previous analysis, we see that a DP solution can be developed.
- However, we would be overlooking another important characteristic of the activity-selection problem.
- Consider the input activities.
  - They are ordered in **monotonically increasing order** of their finish times.

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16



# SEEKING A FASTER WAY

## THE BEST CHOICE

- Consider the input activities ordered in **monotonically increasing order** of their finish times.

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16

- Which of the activity would be **the best** activity to be **scheduled first**?
  - We want to choose an activity that leaves the resource **available for as many other activities as possible** as all activities share the same resource.
    - If we pick  $a_1$  (**finishes the earliest**), we will have a longer period of time available.
- Which of the remaining to schedule after that?

# MAKING A GREEDY CHOICE

- In discovering the **optimal substructure**, we suppose that *we are given a choice* that leads to an optimal solution.
- If picking the activity that finishes the earliest is included in the **optimal** solution, then we no longer need such a supposition.
- We need to prove the optimality of the greedy choice.



# THE GREEDY CHOICE NOTATIONS

- Let us use  $S_k = \{a_i \in S: s_i \geq f_k\}$  to denote the activities that start after  $a_k$  finishes.
- Let  $a_m$  be an activity in  $S_k$  with the **earliest finish time**.
- Consider the instance below

$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16

- $S_1 = \{a_4, a_6, a_7, a_8, a_9, a_{11}\}$ ,  $a_m = a_4$
- $S_3 = \{a_7, a_8, a_9, a_{11}\}$ ,  $a_m = a_7$
- $S_{10} = \emptyset$

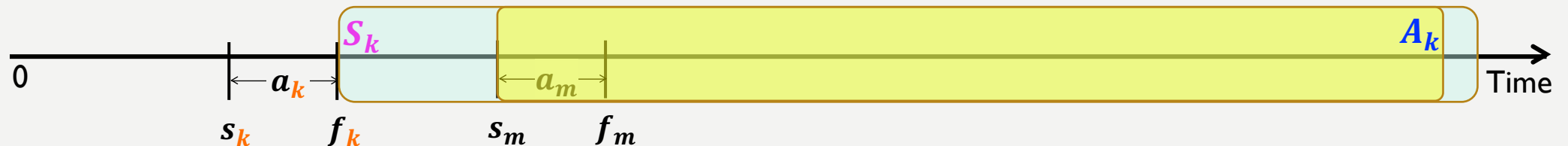
# THE GREEDY CHOICE THEOREM

- Theorem 16.1

- Consider any nonempty subproblem  $S_k$ , and let  $a_m$  be an activity in  $S_k$  with the **earliest finish time**. Then  $a_m$  **is included** in some **maximum-size subset** of **mutually compatible** activities of  $S_k$ .

- Visualization

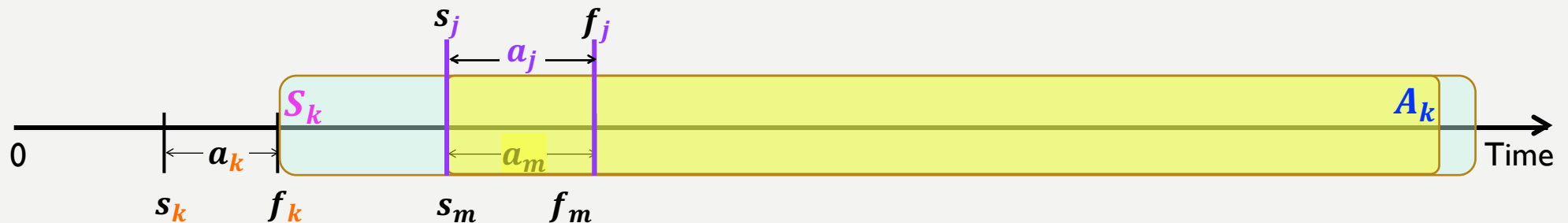
- Let  $A_k$  be a **maximum-size** subset of mutually compatible activities in  $S_k$ .
  - Then we have  $a_m \in A_k$ . We shall prove this theorem by the “cut-and-paste” technique.



# PROOF OF THE GREEDY CHOICE

## CASE #1

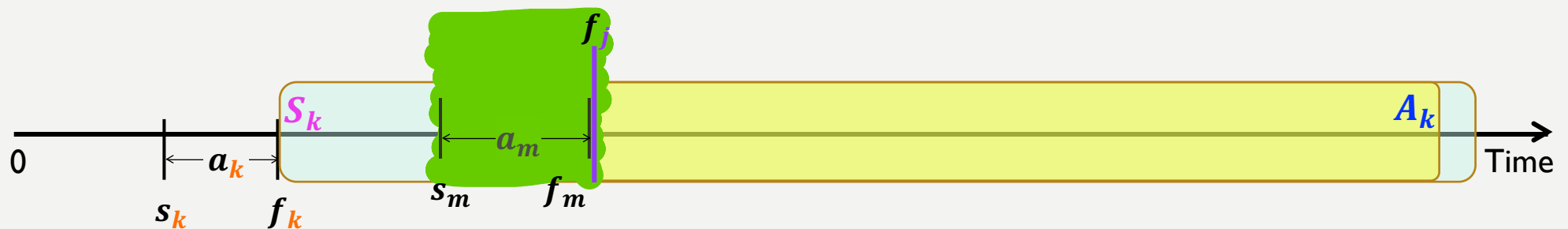
- Our goal is to show that  $a_m$  is included in the optimal solution for  $S_k$ .
- Let  $a_j$  be the activity in  $A_k$  with the **earliest finish time**. Two cases arise
  - Case #1: If  $a_j = a_m$  (activity  $a_j$  and  $a_m$  overlap), done  $a_m \in A_k$ .



# PROOF OF THE GREEDY CHOICE

## CASE #2

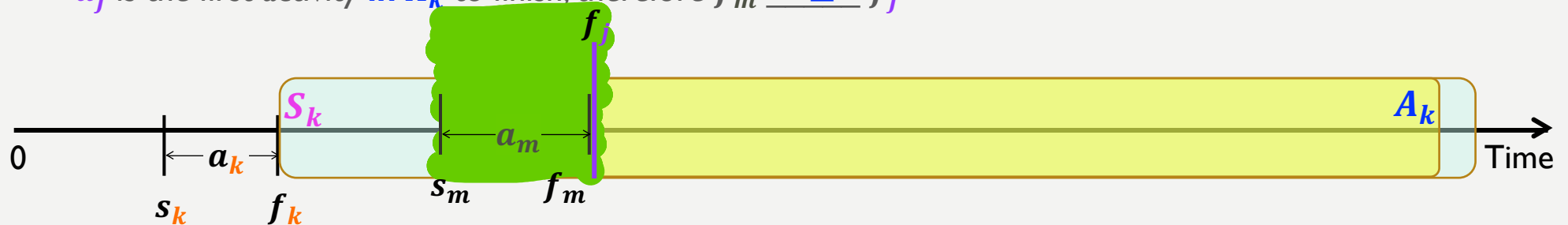
- Our goal is to show that  $a_m$  is included in the optimal solution for  $S_k$ .
- Let  $a_j$  be the activity in  $A_k$  with the **earliest finish time**. Two cases arise.
  - Case #2: Else  $a_j \neq a_m$ . Let us **construct a new** set of activities  $A'_k = (A_k - \{a_j\}) \cup \{a_m\}$ .
    - Cut  $\{a_j\}$  out of  $A_k$  and paste  $a_m$ , i.e., substitute  $a_m$  for  $a_j$  in  $A_k$ .



# PROOF OF THE GREEDY CHOICE

## CASE #2 (CONT'D)

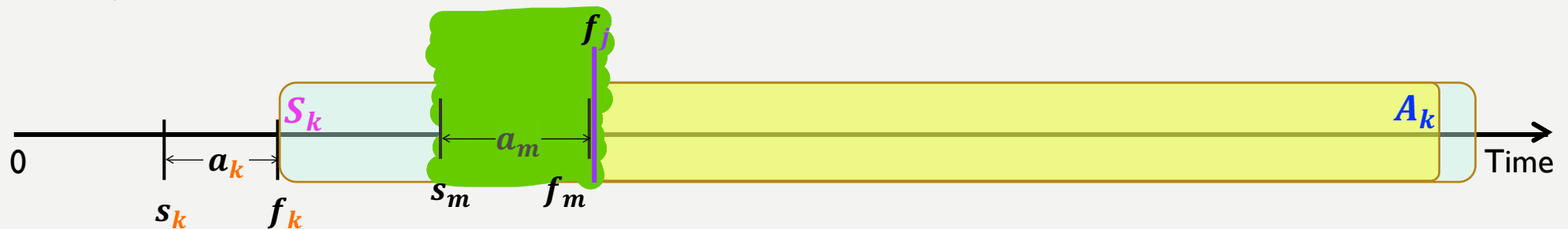
- Our goal is to show that  $a_m$  is included in the optimal solution for  $S_k$ .
- Let  $a_j$  be the activity in  $A_k$  with the **earliest finish time**. Two cases arise.
  - Case #2: Else  $a_j \neq a_m$ . After constructing  $A'_k = (A_k - \{a_j\}) \cup \{a_m\}$ .
    - The activities in  $A'_k$  are disjoint as the activities in  $A_k$  are disjoint.
      - Activities in a set being disjoint means that no two activities overlap.
    - $a_j$  is the first activity in  $A_k$  to finish, therefore  $f_m \leq f_j$



# PROOF OF THE GREEDY CHOICE

## CASE #2 (CONT'D)

- Our goal is to show that  $a_m$  is included in the optimal solution for  $S_k$ .
- Let  $a_j$  be the activity in  $A_k$  with the **earliest finish time**. Two cases arise.
  - Case #2: Else  $a_j \neq a_m$ . After constructing  $A'_k = (A_k - \{a_j\}) \cup \{a_m\}$ .
    - $|A'_k| = \underline{|A_k| - 1 + 1} = \underline{|A_k|}$
    - Conclude case #2,  $A'_k$  is a **maximum-size subset of mutually compatible activities** of  $S_k$ , and it includes  $a_m$ .



# CONTINUE TO SOLVE THE PROBLEM

- Design an algorithm that makes the greedy choice every time.
  - Unlike a normal DP solution where the algorithms can be in **bottom-up** fashion, a greedy algorithm **does not need to work bottom-up**.
- Generally, the algorithm can work as follows
  - **make a greedy choice** to put into the optimal solution and then
  - **solve the subproblem** of choosing activities from those that are compatible with those already chosen.

# RECURSIVE-ACTIVITY-SELECTOR

- Input

- An array  $s$  that stores the start times of the activities
- An array  $f$  that stores the finish times of the activities
- An index  $k$  that defines the subproblem  $S_k$  it is to solve. Use 0 for initial call.
- The size  $n$  of the original problem

- Example: RECURSIVE-ACTIVITY-SELECTOR ( $s, f, 0, 5$ )

- $s = \{1, 3, 0, 5, 3\}, f = \{4, 5, 6, 7, 9\}$

RECURSIVE-ACTIVITY-SELECTOR ( $s, f, k, n$ )	
1	$m = k + 1$
2	<b>while</b> $m \leq n$ <b>and</b> $s[m] < f[k]$ // find the first activity in $S_k$ to finish
3	$m = m + 1$
4	<b>if</b> $m \leq n$
5	<b>return</b> $\{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)$
6	<b>else return</b> $\emptyset$

$i$	1	2	3	4	5
$s_i$	1	3	0	5	3
$f_i$	4	5	6	7	9



# RECURSIVE-ACTIVITY-SELECTOR

## THE USE OF FICTITIOUS ACTIVITY $a_0$

- The initial call to solve a given problem is  
RECURSIVE-ACTIVITY-SELECTOR ( $s, f, 0, 5$ )
  - Line 2 accesses  $f[k] = f[0]$  to make a choice
- We add a **fictitious activity**  $a_0$  with  $f_0 = 0$  to avoid index-out-of-bound exception.

RECURSIVE-ACTIVITY-SELECTOR ( $s, f, k, n$ )	
1	$m = k + 1$
2	<b>while</b> $m \leq n$ <b>and</b> $s[m] < f[k]$ // find the first activity in $S_k$ to finish
3	$m = m + 1$
4	<b>if</b> $m \leq n$
5	<b>return</b> $\{a_m\} \cup$ RECURSIVE-ACTIVITY-SELECTOR ( $s, f, m, n$ )
6	<b>else return</b> $\emptyset$

- The input activities **must be sorted in non-decreasing order** by their finish times
  - If the given input is NOT sorted, we can spend  $O(n \lg n)$  time to sort it.

# RECURSIVE-ACTIVITY-SELECTOR IN ACTION

- Run the algorithm on the input sequence.

RECURSIVE-ACTIVITY-SELECTOR ( <i>s</i> , <i>f</i> , 11, 11)
RECURSIVE-ACTIVITY-SELECTOR ( <i>s</i> , <i>f</i> , 8, 11)
RECURSIVE-ACTIVITY-SELECTOR ( <i>s</i> , <i>f</i> , 4, 11)
RECURSIVE-ACTIVITY-SELECTOR ( <i>s</i> , <i>f</i> , 1, 11)
RECURSIVE-ACTIVITY-SELECTOR ( <i>s</i> , <i>f</i> , 0, 11)

- Return:  $\{a_1, a_4, a_8, a_{11}\}$

RECURSIVE-ACTIVITY-SELECTOR ( <i>s</i> , <i>f</i> , <i>k</i> , <i>n</i> )	
1	$m = k + 1$
2	<b>while</b> $m \leq n$ <b>and</b> $s[m] < f[k]$ // find the first activity in $S_k$ to finish
3	$m = m + 1$
4	<b>if</b> $m \leq n$
5	<b>return</b> $\{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)$
6	<b>else return</b> $\emptyset$

<i>i</i>	1	2	3	4	5	6	7	8	9	10	11
<i>s<sub>i</sub></i>	1	3	0	5	3	5	6	8	8	2	12
<i>f<sub>i</sub></i>	4	5	6	7	9	9	10	11	12	14	16

# RECURSIVE-ACTIVITY-SELECTOR

## RUNNING TIME

- The algorithm examines each activity EXACTLY once.
- The running time function of RECURSIVE-ACTIVITY-SELECTOR algorithm is  $T(n) = \underline{\Theta(n)}$ .

RECURSIVE-ACTIVITY-SELECTOR ( $s, f, k, n$ )	
1	$m = k + 1$
2	<b>while</b> $m \leq n$ <b>and</b> $s[m] < f[k]$ // find the first activity in $S_k$ to finish
3	$m = m + 1$
4	<b>if</b> $m \leq n$
5	<b>return</b> $\{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)$
6	<b>else return</b> $\emptyset$

# RECURSIVE-ACTIVITY-SELECTOR

## TAIL RECURSIVE

- The algorithm is almost “tail recursive.”
  - It ends with a recursive call to itself followed by a union operation.
- A **tail-recursive** procedure is easy to transform to an **iterative** form.

RECURSIVE-ACTIVITY-SELECTOR ( $s, f, k, n$ )	
1	$m = k + 1$
2	<b>while</b> $m \leq n$ <b>and</b> $s[m] < f[k]$ // find the first activity in $S_k$ to finish
3	$m = m + 1$
4	<b>if</b> $m \leq n$
5	<b>return</b> $\{a_m\} \cup \text{RECURSIVE-ACTIVITY-SELECTOR}(s, f, m, n)$
6	<b>else return</b> $\emptyset$



# ITERATIVE-ACTIVITY-SELECTOR

- Input
  - An array  $s$  that stores the start times of the activities
  - An array  $f$  that stores the finish times of the activities
- The algorithm takes advantage of the greedy property by including the first activity in set  $A$ .
  - The input must be pre-sorted.
  - This way activity  $a_1$  is always the one **with the earliest finish time**.

GREEDY-ACTIVITY-SELECTOR ( $s, f$ )	
1	$n = s.length$
2	$A = \{a_1\}$
3	$k = 1$
4	<b>for</b> $m = 2$ <b>to</b> $n$
5	<b>if</b> $s[m] \geq f[k]$
6	$A = A \cup \{a_m\}$
7	$k = m$
8	<b>return</b> $A$

# ITERATIVE-ACTIVITY-SELECTOR IN ACTION

- Run the algorithm on the input sequence.

											
$i$	1	2	3	4	5	6	7	8	9	10	11
$s_i$	1	3	0	5	3	5	6	8	8	2	12
$f_i$	4	5	6	7	9	9	10	11	12	14	16
											
		$k$									

- $A = \{ a_1 \} \cup \{ a_4 \} \cup \{ a_8 \} \cup \{ a_{11} \} = \{ a_1, a_4, a_8, a_{11} \}$

GREEDY-ACTIVITY-SELECTOR ( $s, f$ )	
1	$n = s.length$
2	$A = \{a_1\}$
3	$k = 1$
4	<b>for</b> $m = 2$ <b>to</b> $n$
5	<b>if</b> $s[m] \geq f[k]$
6	$A = A \cup \{a_m\}$
7	$k = m$
8	<b>return</b> $A$

# ITERATIVE-ACTIVITY-SELECTOR

## RUNNING TIME

- The code of the GREEDY-ACTIVITY-SELECTOR  $(s, f)$  algorithm is structured as follows.

```
for  $m = 2$  to  $n$ 
    if  $s[m] \geq f[k]$ 
        Union
```

- Therefore, the running time  $T(n) = \underline{\hspace{2cm}}$ 
  - Assuming the input activities are sorted and union operation takes  $\Theta(1)$ .

GREEDY-ACTIVITY-SELECTOR $(s, f)$	
1	$n = s.length$
2	$A = \{a_1\}$
3	$k = 1$
4	for $m = 2$ to $n$
5	if $s[m] \geq f[k]$
6	$A = A \cup \{a_m\}$
7	$k = m$
8	return $A$

# **NEXT UP HUFFMAN CODES**



# REFERENCE

- [https://www.netclipart.com/isee/hRwxRh\\_kids-clipart-nurse-cute-female-doctor-cartoon/](https://www.netclipart.com/isee/hRwxRh_kids-clipart-nurse-cute-female-doctor-cartoon/)
- <https://listposts.com/lera-kiryakova-celebrities-cartoon-characters/>