

Greedy Algorithm - I

Activity Selection Problem

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Lecture Plan

- Greedy algorithm definition
- Activity selection problem
 - DP solution
 - Greedy solution

Optimization problems

- Algorithms for optimization problems
 - Sequence of steps, with a set of choices at each step
- Types of algorithms we can consider
 - Brute force algorithms
 - Simple recursive algorithms
 - Divide and conquer algorithms
 - Dynamic programming algorithms
 - Greedy algorithms
 - Branch and bound algorithms
 - Randomized algorithms

Greedy Algorithms

- A greedy algorithm works in steps. At each step:
 - You take the best choice, without regard for future consequences
 - You hope that by choosing a *local* optimum at each step, you will end up at a *global* optimum

Activity-selection problem

- Consider a set of activities $S=\{a_1, a_2, \dots, a_n\}$
 - They use resources, such as lecture hall, one lecture at a time
 - Each a_i , has a start time s_i , and finish time f_i , with $0 \leq s_i < f_i < \infty$.
 - a_i and a_j are compatible if $[s_i, f_i)$ and $[s_j, f_j)$ do not overlap
- Goal: select maximum-size subset of mutually compatible activities.
- Start from dynamic programming, then greedy algorithm, see the relation between the two

Activity-selection problem

- The activity-selection problem is to select a maximum-size subset of mutually compatible activities.

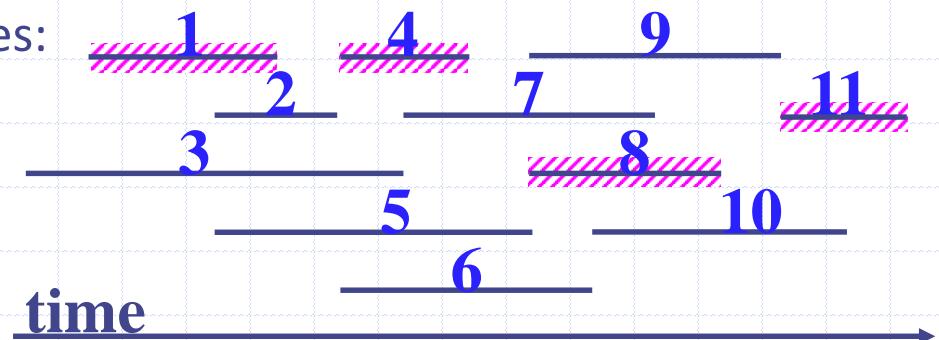
Example:

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

Some mutually compatible activities:

$\{a_3, a_9, a_{11}\}$ --- not the largest

$\{a_1, a_4, a_8, a_{11}\}$



DP solution: Notations

- Optimal substructure of activity-selection problem

$$S_{ij} = \{a_k \in S : f_i \leq s_k < f_k \leq s_j\},$$

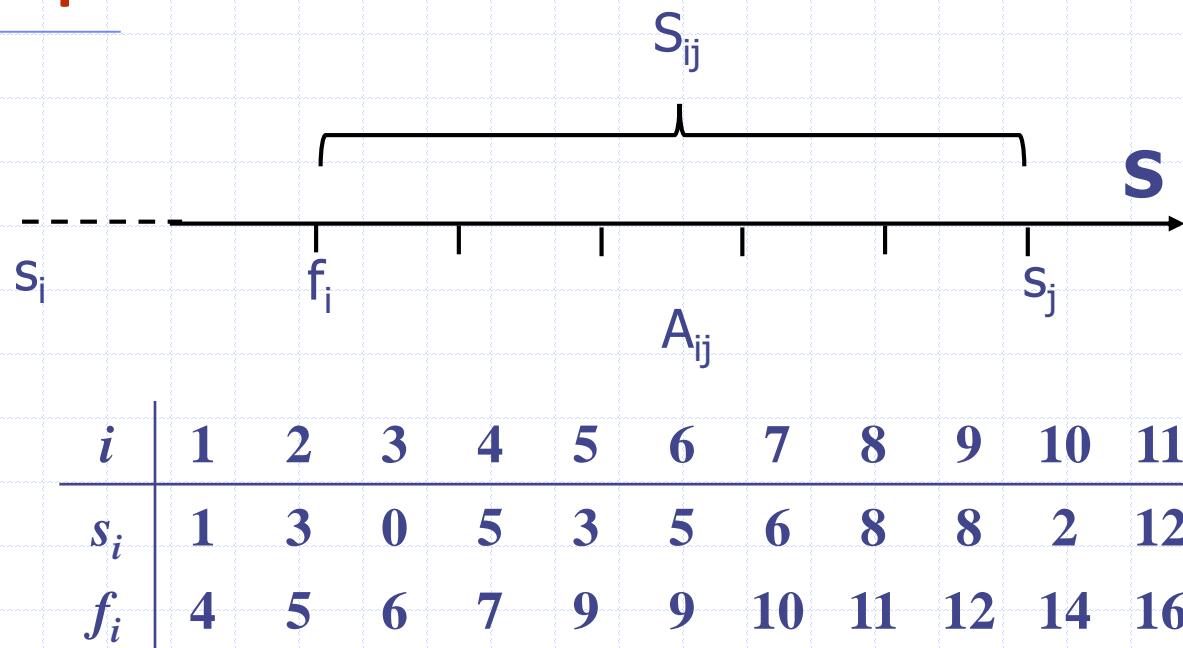
is the subset of activities in S that can start after activity a_i finishes and finish before activity a_j starts.

- Problem:

- Find maximum set of mutually compatible activities in S_{ij}

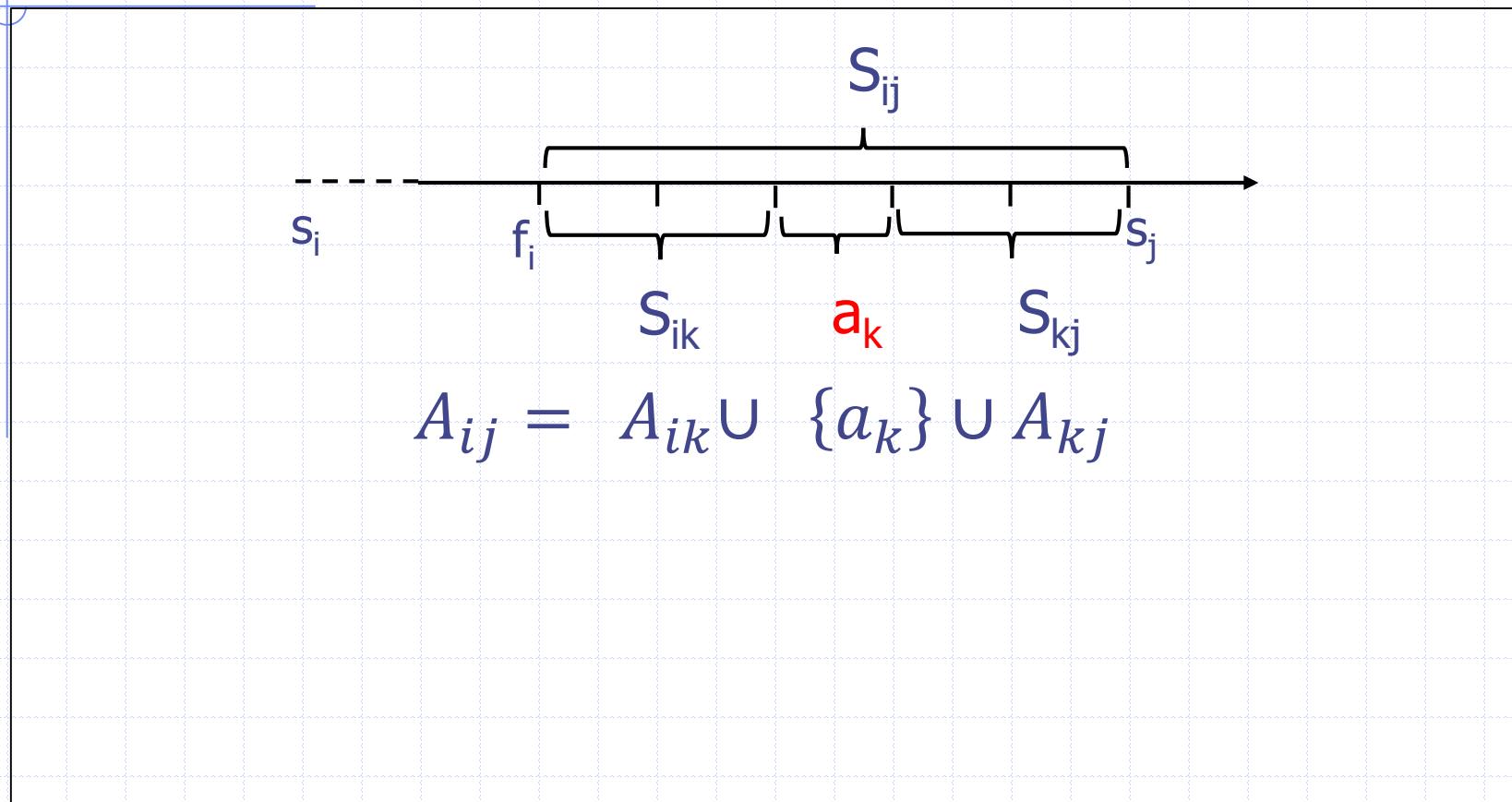
A_{ij} : Maximum set of mutually compatible activities in S_{ij}

Example



- $S_{1,11} = ??$
- $A_{1,11} = ??$

Illustration: DP solution – Optimal Substructure



DP solution –step 1

- Optimal substructure
 - Suppose an optimal solution A_{ij} includes a_k
 - We are left with two subproblems - S_{ik} and S_{kj}
- Let $A_{ik} = A_{ij} \cap S_{ik}$
 - Contain the activities in A_{ij} that finish before a_k
- Let $A_{kj} = A_{ij} \cap S_{kj}$
 - Contain the activities in A_{ij} that start after a_k
- $A_{ij} = A_{ik} \cup \{a_k\} \cup A_{kj}$
 $|A_{ij}| = |A_{ik}| + |A_{kj}| + 1$

DP Solution – step 2

- Recursive cost

$c[i,j]$: size of optimal solution of S_{ij}

$$c[i, j] = c[i, k] + c[k, j] + 1$$

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

- Need to solve all sub-problems

Greedy Choice

- Choose an activity to add to the optimal solution without having to solve all the subproblems
- What is a greedy choice?
 - One possibility: Choose an activity such that it will leave maximum resources available for others
 - Choose an activity a_1 with the earliest finish time
- Making a greedy choice leaves us with only **one** subproblem to choose
 - Find activities that start after a_1

Greedy choice (2)

- Set of activities that start after a_k finishes

$$S_k = \{a_i \in S : s_i \geq f_k\}$$

- Ex: If a_1 is greedy choice, only subproblem is S_1

A_k is optimal solution of S_k

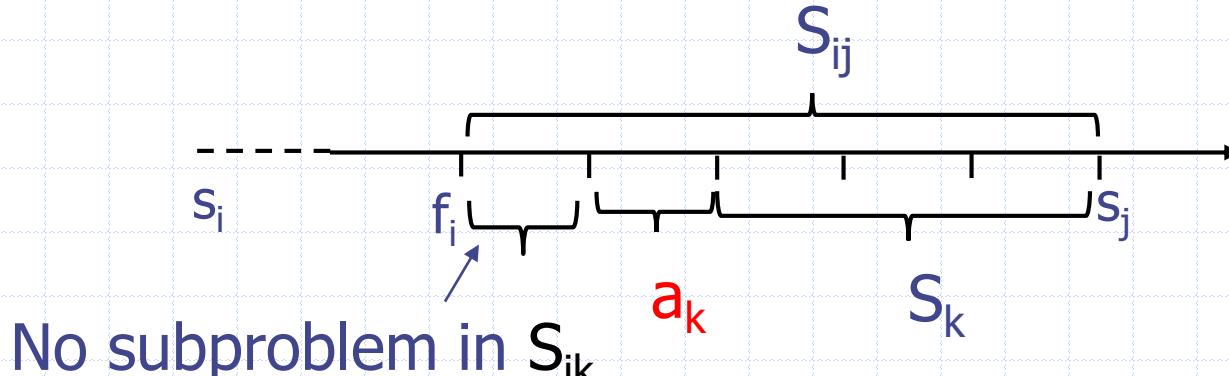
- We have shown that Activity-Selection problem exhibit optimal substructure

- That is , if a_1 is in the optimal solution, then

$$A_{ij} = a_1 \cup A_1$$

- We need to prove a_1 in the optimal solution

Illustration: Greedy choice



$$A_{ij} = \{a_k\} \cup A_k$$

a_k : Activity with earliest finish time

S_k : Activities that start after a_k finishes

Illustration of 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 .

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

- Consider S_0 [all activities that start after a_0 , $f_0=0$]
 - The activity with earliest finish time is a_1
 - a_1 must be included in some optimal solution
- Optimal solutions: $\{a_1, a_4, a_8, a_{11}\}, \{a_2, a_4, a_9, a_{11}\}$

Recursive Greedy Algorithm

- Assume that n input activities are already ordered by monotonically increasing finish times

$$f_1 \leq f_2 \leq \dots \leq f_n$$

- Add fictitious activities a_0 with $f_0=0$
 - Subproblem S_0 is the entire set of activities

Recursive Greedy Algorithm

- Input: $s[]$: start time of a_i ; $f[]$: finish time of a_i
- Initial call: Recursive-Activity-Selector($s, f, 0, n$)

RECURSIVE-ACTIVITY-SELECTOR(s, f, k, n)

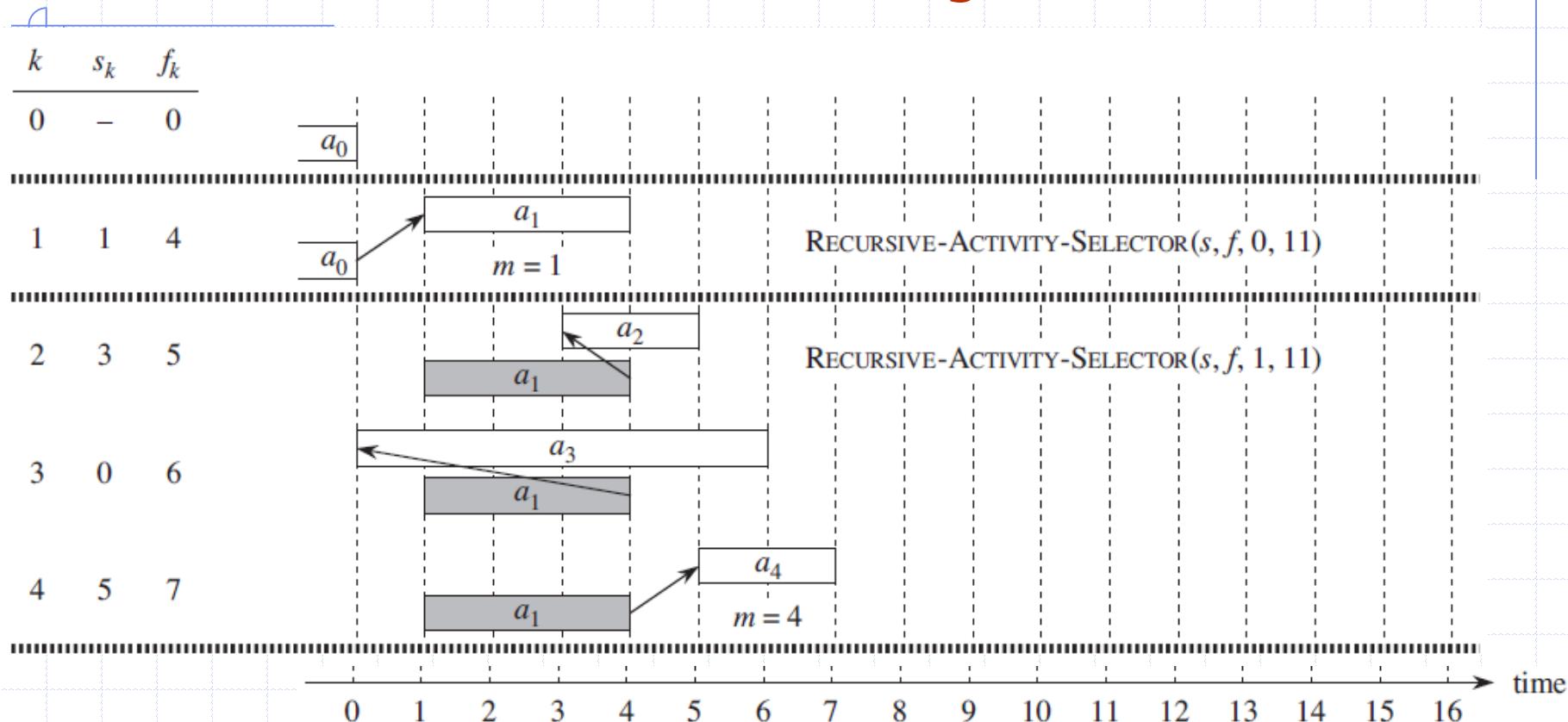
```
1   $m = k + 1$ 
2  while  $m \leq n$  and  $s[m] < f[k]$       // find the first activity in  $S_k$  to finish
3       $m = m + 1$ 
4  if  $m \leq n$ 
5      return  $\{a_m\} \cup$  RECURSIVE-ACTIVITY-SELECTOR( $s, f, m, n$ )
6  else return  $\emptyset$ 
```

- While loop returns first activity a_m compatible with a_k

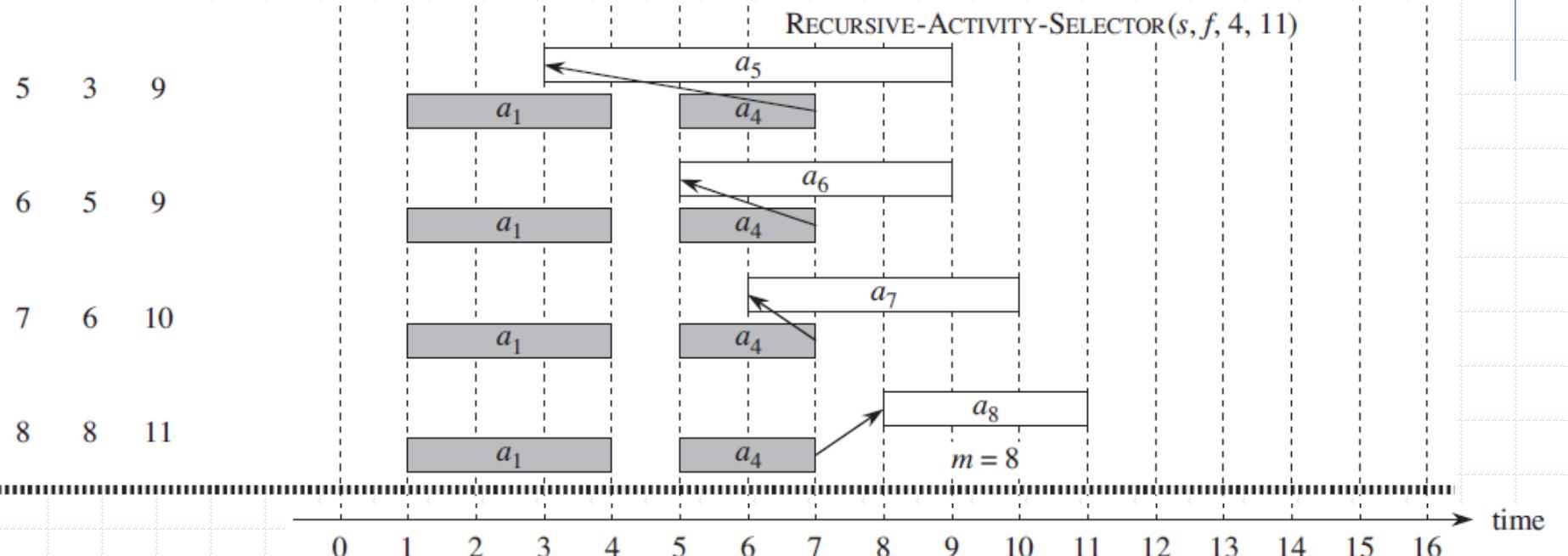
Running time: $\Theta(n)$

RECURSIVE-ACTIVITY-SELECTOR:

Illustration on the 11 activities given earlier

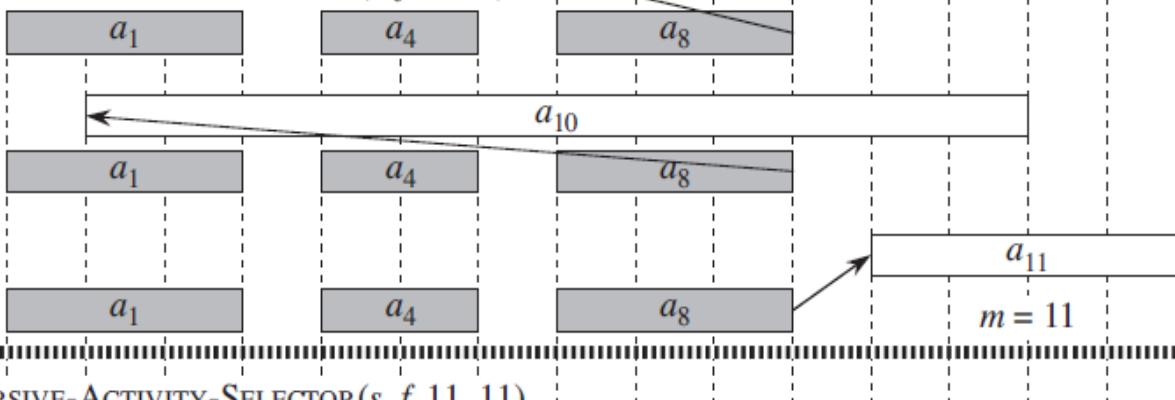


\leftarrow
 $k \quad s_k \quad f_k$

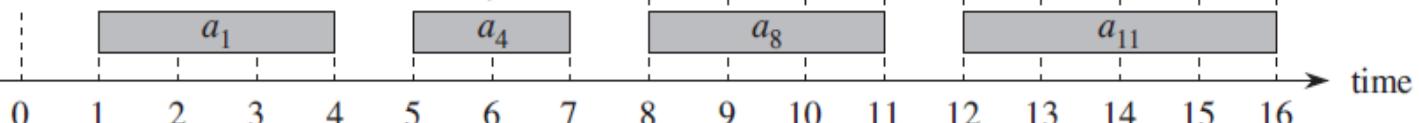


k	s_k	f_k
9	8	12
10	2	14
11	12	16

RECURSIVE-ACTIVITY-SELECTOR($s, f, 8, 11$)



RECURSIVE-ACTIVITY-SELECTOR($s, f, 11, 11$)



Iterative version: Greedy Algo

- Write an iterative version of the Greedy algorithm (recursive) given below.

RECURSIVE-ACTIVITY-SELECTOR(s, f, k, n)

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

Exercise

16.1-1

Give a dynamic-programming algorithm for the activity-selection problem, based on recurrence (16.2). Have your algorithm compute the sizes $c[i, j]$ as defined above and also produce the maximum-size subset of mutually compatible activities.