

V. Greedy Algorithms

Greedy algorithms – Overview

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- ▶ A **greedy algorithm** always makes the choice that *looks best at the moment*, without regard for future consequence, i.e., “*take what you can get now*” strategy
- ▶ Greedy algorithms do not always yield optimal solutions,

Local optimum $\not\Rightarrow$ Global optimum

but *for many problems they do*.

Activity-selection problem

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- ▶ Activities i and j are **compatible** if the intervals $[s_i, f_i)$ and $[s_j, f_j)$ do not overlap.
- ▶ Without loss of generality, assume

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

Activity-selection problem

Example

i	s_i	f_i
1	1	4
2	3	5
3	0	6
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$A = \{1, 4, 8, 11\}$ is an optimal (why?) solution.

$A = \{2, 4, 9, 11\}$ is also an optimal solution.

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- ▶ Intuitively, this choice leaves as much opportunity as possible for the remaining activities to be scheduled
- ▶ That is, the greedy choice is the one that maximizes the amount of unscheduled time remaining.

Activity-selection problem

```
Greedy_Activity_Selector(s,f)
n = length(s)
A = {1}
j = 1
for i = 2 to n
    if s[i] >= f[j]
        A = A U {i}
        j = i
    end if
end for
return A
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Remarks

- ▶ Assume the array f already sorted
- ▶ Complexity: $T(n) = O(n)$

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Solution $A = \{1, 4, 8, 11\}$ by `Greedy_Activity_Selector`.

Activity-selection problem

Question: Does `Greedy_Activity_Selector` work?

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Theorem. Algorithm `Greedy_Activity_Selector` produces a solution of the activity-selection problem.

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- 3. A' is also optimal, since $|A'| = |A|$*

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Then

- 1. the sets $A - \{k_1\}$ and $\{1\}$ are disjoint*
 - 2. the activities in A' are compatible*
 - 3. A' is also optimal, since $|A'| = |A|$*
- ▶ *Therefore, we conclude that there always exists an optimal solution that begins with a greedy choice.*

Activity-selection problem

Property 2.

If A is an optimal solution, then $A' = A - \{1\}$ is an optimal solution to $S' = \{i \in S, s[i] \geq f[1]\}$.

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Proof: By contradiction. If there exists B' to S' such that $|B'| > |A'|$, then let

$$B = B' \cup \{1\},$$

we have

$$|B| > |A|,$$

which is contradicting to the optimality of A .

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Proof of **Theorem**: By Properties 1 and 2, we know that

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- ▶ After each greedy choice is made, we are left with an optimization problem of the same form as the original.
- ▶ *By induction* on the number of choices made, making the greedy choice at every step produces an optimal solution.

Therefore, the `Greedy_Activity_Selector` produces an optimal solution of the activity-selection problem.

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a globally optimal solution can be arrived at by making a locally optimal (greedy) choice.

- ▶ Property 2 is called **the optimal substructure property**, generally casted as

an optimal solution to the problem contains within it optimal solution to subprograms.

These are **two key properties** for the success of greedy algorithms!