Greedy algorithms and interval scheduling COMS20010 (Algorithms II)

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Suppose you're running a satellite imaging service.

Taking a satellite picture of an area isn't instant — it takes time proportional to its latitude, and it can only be done at a specific time of day (when your satellite's orbit is lined up correctly).

You have a set of requested images, each of which can only be taken at specific times, and you can only take one picture at once.

How can you satisfy as many requests as possible?

Requested satellite times: 12:00–12:30, 12:05–12:20, 12:15–12:55, 12:20-12:25, 12:25–12:40, 12:38–12:50, and 12:45-13:00.



Requested satellite times: 12:10–12:30, **12:05–12:20**, 12:15–12:55, **12:20-12:25**, **12:25–12:40**, 12:38–12:50, and **12:45-13:00**.



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We might as well give our times integer labels for simplicity, though.

Requested satellite times: 0–30, **5–20**, 15–55, **20–25**, **25–40**, 38–50, and **45-60**.



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Interval scheduling: a formal definition

A **request** is a pair of integers (s, f) with $0 \le s \le f$. We call s the **start time** and f the **finish time**.

A set A of requests is **compatible** if for all distinct (s, f), $(s', f') \in A$, either $s' \ge f$ or $s \ge f'$ — that is, the requests' time intervals don't overlap.



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Interval Scheduling Problem

Input: An array \mathcal{R} of *n* requests $(s_1, f_1), \ldots, (s_n, f_n)$. **Desired Output:** A compatible subset of \mathcal{R} of maximum possible size.

Our satellite problem above is an example of this — the maximum compatible subset is the list of image requests we accept.

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Idea: Use a **greedy** approach: start with an empty output and slowly build it up, making choices that look good in the moment, without worrying about future implications. These algorithms are very common and useful!

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	Algorithm: GREEDYSCHEDULE	
	Input: A	An array ${\mathcal R}$ of n requests.
	Output: A	A maximum compatible subset of $\mathcal R.$
ı begin		
2	Sort \mathcal{R}	's entries so that $\mathcal{R} \leftarrow [(s_1, f_1), \dots, (s_n, f_n)]$ where $f_1 \leq \dots \leq f_n$.
3	Initialis	we $A \leftarrow []$, lastf $\leftarrow 0$.
4	foreac	h $i \in \{1, \ldots, n\}$ do
5	if :	$s_i \geq \texttt{lastf}$ then
6		Append (s_i, f_i) to A and update last $f \leftarrow f_i$.
7	Return	А.

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Time analysis:

Step 2 takes $O(n \log n)$ time, as covered exhaustively in COMS10007.

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Proof of correctness: Next video!

Greedy algorithm is a very informal term, and different people have incompatible definitions. My definition (see e.g. KT's book) is:

- they start with a sub-optimal (often trivial) solution, e.g. A = [], and gradually turn it into the output.
- they look over all possible improvements and pick the one that "looks best at the time", e.g. adding the fastest-ending compatible request.
- they never backtrack in "quality", e.g. the size of A never goes down.

Some people (see e.g. CLRS' book) have stricter definitions. But this will never actually matter to you!

What matters is being able to use and design algorithms like this one.

Greed is... not always good?



Sometimes the locally-optimal choices are the ones we regret the most...

John Lapinskas















E.g. what if instead of choosing the fastest-finishing request to add at each stage, we chose the fastest-starting request?



It doesn't work!

Or we chose the shortest interval?



Or we chose the shortest interval?



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