Intro: Coin Changing

Coin Changing

Goal. Given currency denominations: 1, 5, 10, 25, 100, give change to customer using *fewest* number of coins.

Ex: 34¢.



Algorithm is "Greedy": One large coin better than two or more smaller ones

Cashier's algorithm. At each iteration, give the *largest* coin valued ≤ the amount to be paid.

Ex: \$2.89.



Coin-Changing: Does Greedy Always Work?

Observation. Greedy algorithm is sub-optimal for US postal denominations: 1, 10, 21, 34, 70, 100, 350, 1225, 1500.

Counterexample. 140¢.

■ Greedy: 100, 34, 1, 1, 1, 1, 1.

Optimal: 70, 70.









Algorithm is "Greedy", but also short-sighted – attractive choice now may lead to dead ends later.

Correctness is key!











Outline & Goals

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"Greedy Algorithms" what they are
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Pros

intuitive often simple often fast

Proof techniques
stay ahead
structural
exchange arguments

Plan

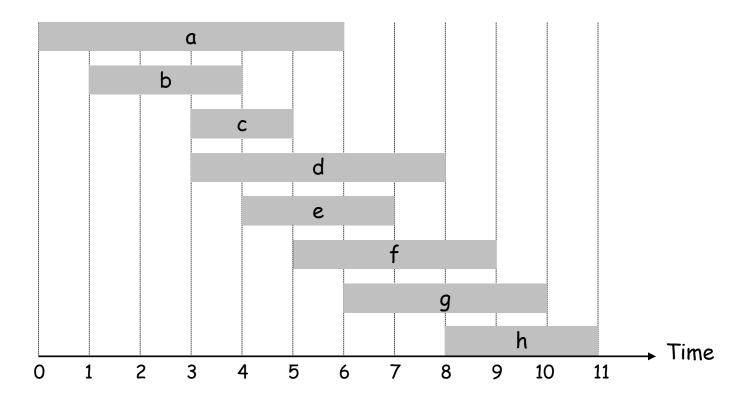
Greed

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Proof Technique 1: "greedy stays ahead"

- Job j starts at s_j and finishes at f_j .
- Two jobs compatible if they don't overlap.
- Goal: find largest subset of mutually compatible jobs.



Greedy template. Consider jobs in some order. Take each job provided it's compatible with the ones already taken.

- What order?
- Does that give best answer?
- Why or why not?
- Does it help to be greedy about order?

Greedy template. Consider jobs in some order. Take each job provided it's compatible with the ones already taken.

[Earliest start time] Consider jobs in ascending order of start time s_j .

[Earliest finish time] Consider jobs in ascending order of finish time f_j .

[Shortest interval] Consider jobs in ascending order of interval length $f_j - s_j$.

[Fewest conflicts] For each job, count the number of conflicting jobs c_j . Schedule in ascending order of conflicts c_j .

Greedy template. Consider jobs in some order. Take each job provided it's compatible with the ones already taken.

counterexample for earliest start time counterexample for shortest interval counterexample for fewest conflicts

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Greedy algorithm. Consider jobs in increasing order of finish time. Take each job provided it's compatible with the ones already taken.

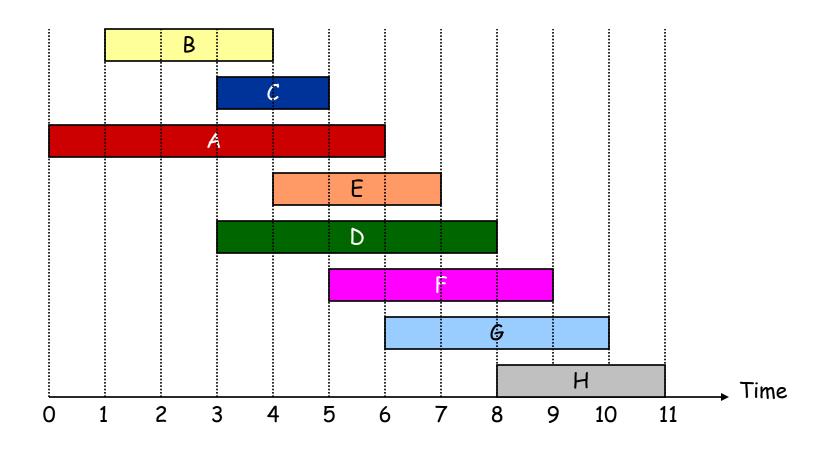
```
Sort jobs by finish times so that f₁ ≤ f₂ ≤ ... ≤ fₙ.

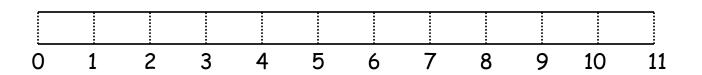
/ jobs selected

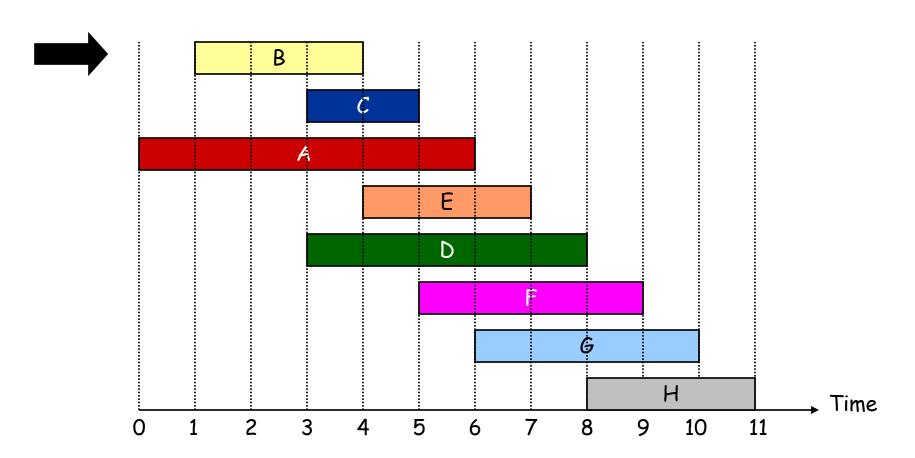
A = {}
for j = 1 to n {
   if (job j compatible with A)
        A = A U {j}
}
return A
```

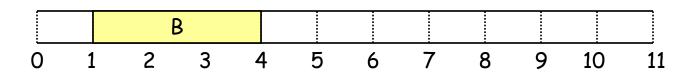
Implementation. O(n log n).

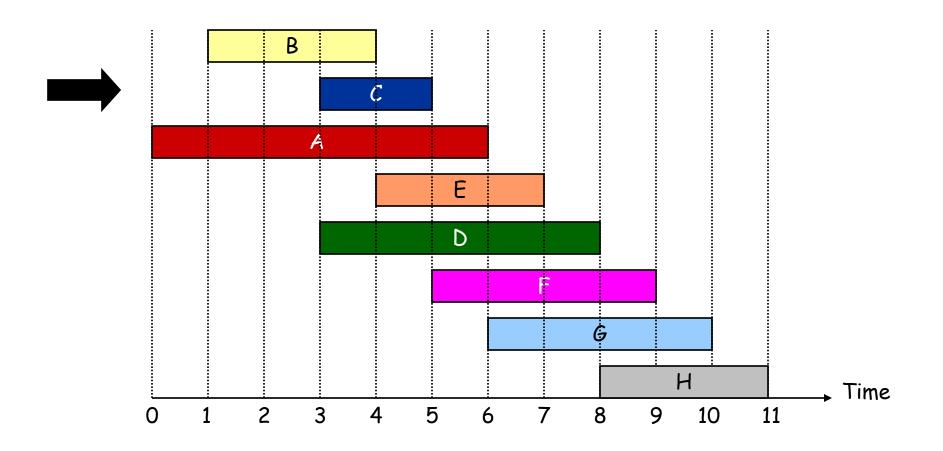
- Remember job j* that was added last to A.
- Job j is compatible with A if $s_j \ge f_{j^*}$.

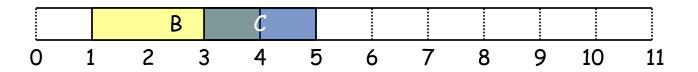


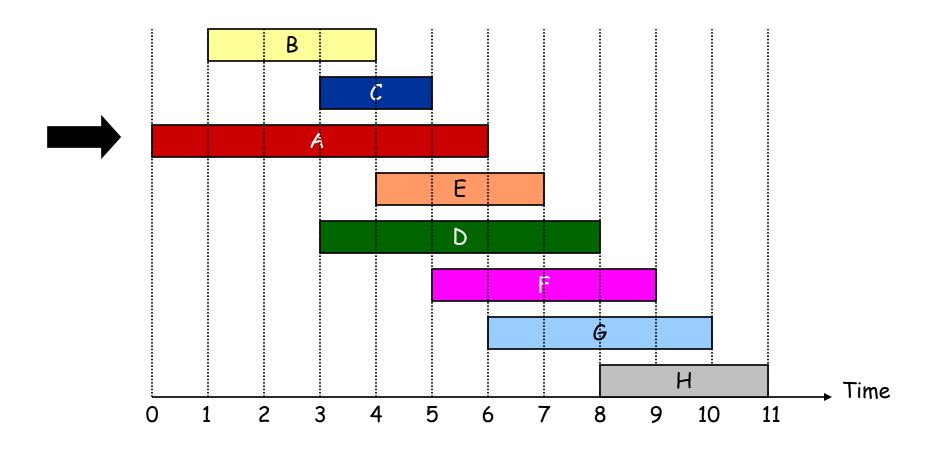


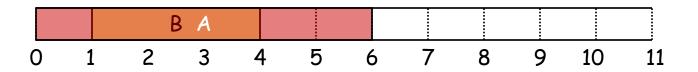


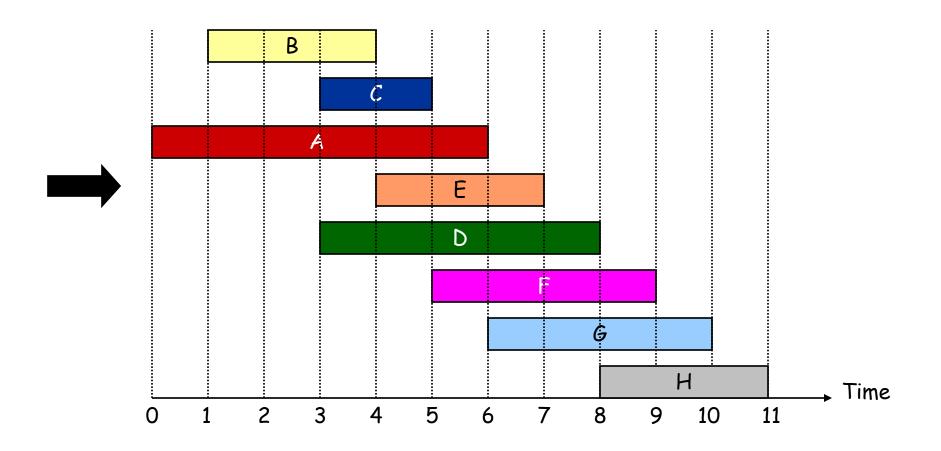




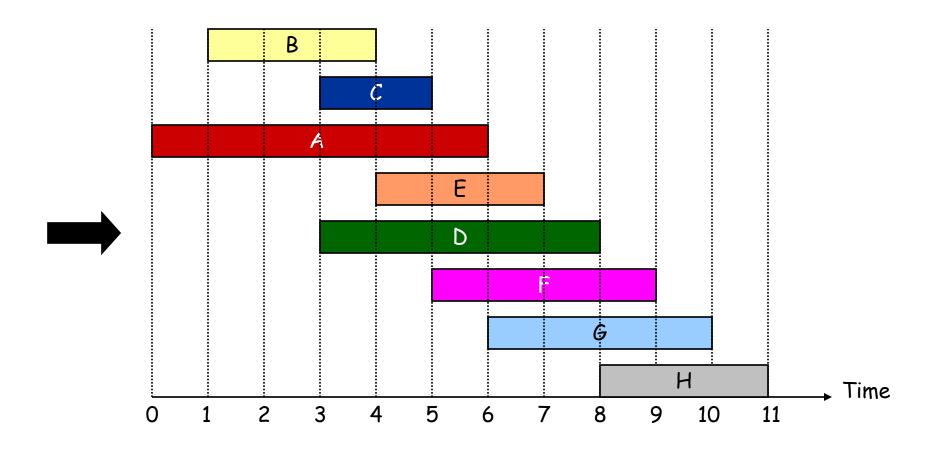


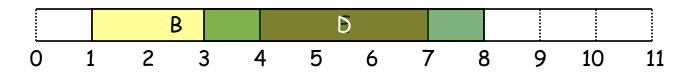


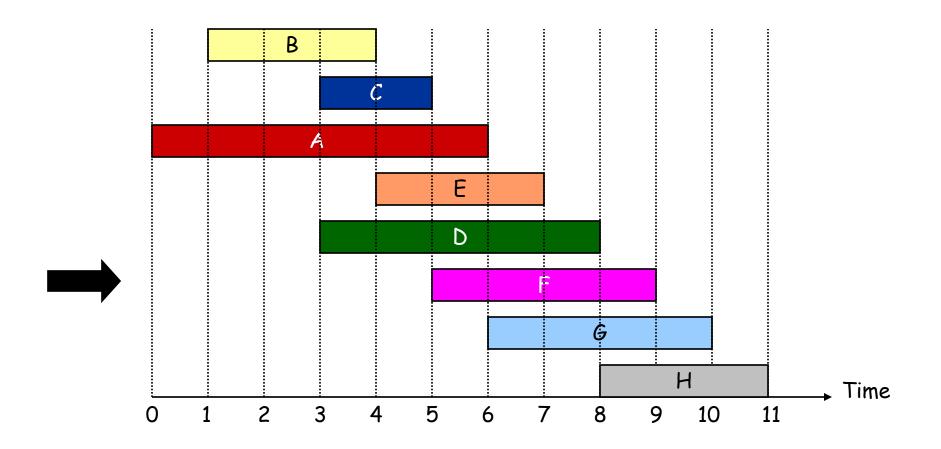




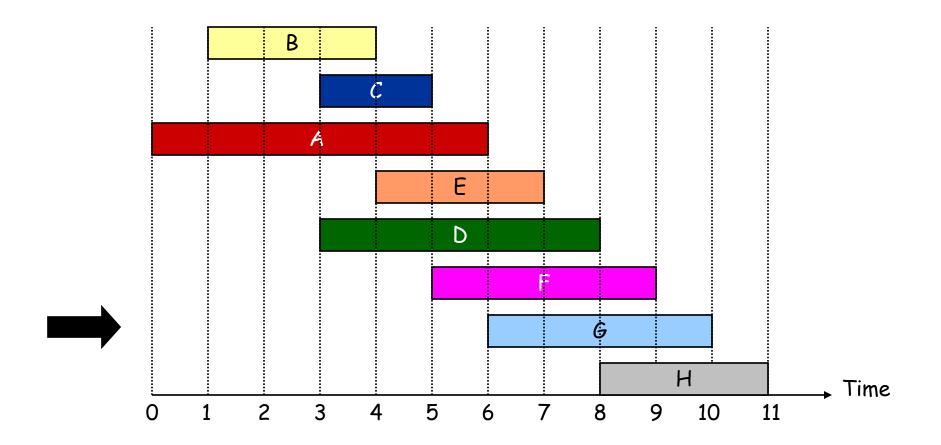


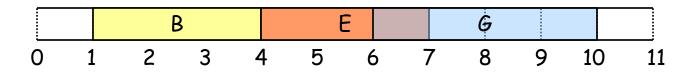


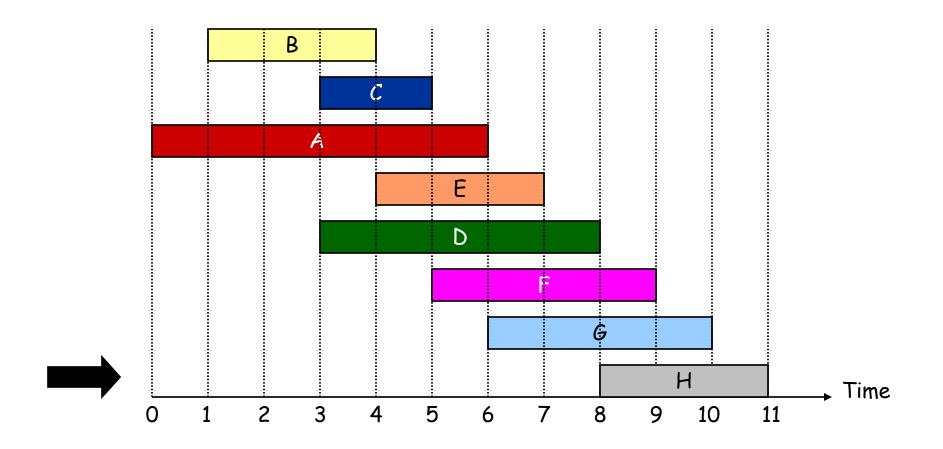














Interval Scheduling: Correctness

Theorem. Greedy algorithm is optimal.

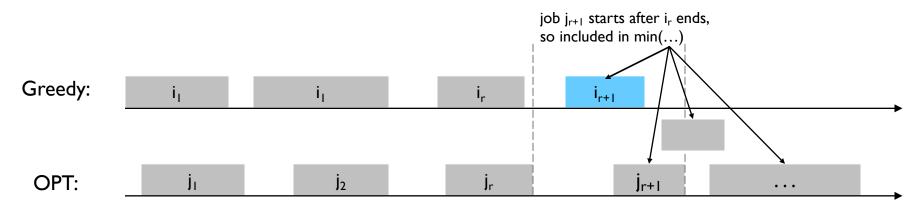
Pf. ("greedy stays ahead")

Let i_1 , i_2 , ... i_k be jobs picked by greedy, j_1 , j_2 , ... j_m those in some optimal solution Show $f(i_r) \le f(j_r)$ by induction on r.

Basis: i_1 chosen to have min finish time, so $f(i_1) \le f(j_1)$

Ind: $f(i_r) \le f(j_r) \le s(j_{r+1})$, so j_{r+1} is among the candidates considered by greedy when it picked i_{r+1} , & it picks min finish, so $f(i_{r+1}) \le f(j_{r+1})$

Similarly, $k \ge m$, else j_{k+1} is among (nonempty) set of candidates for i_{k+1}

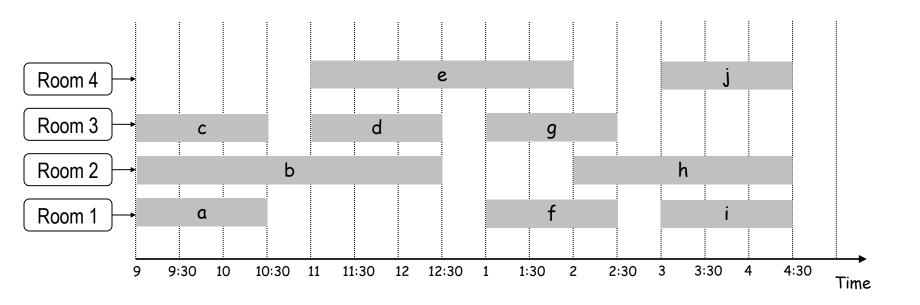


Proof Technique 2: "Structural"

Interval partitioning.

- Lecture j starts at s_j and finishes at f_j .
- Goal: find minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.

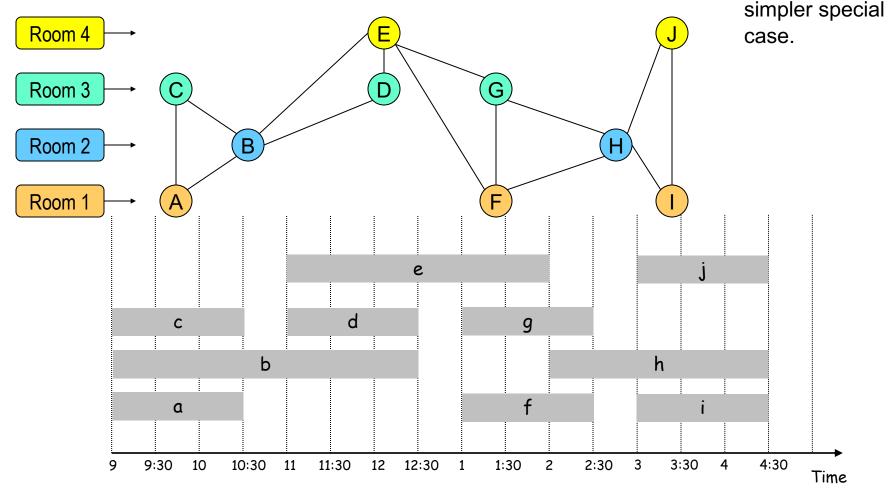
Ex: This schedule uses 4 classrooms to schedule 10 lectures.



Interval Partitioning as Interval Graph Coloring

Vertices = classes; edges = conflicting class pairs; different colors = different assigned rooms

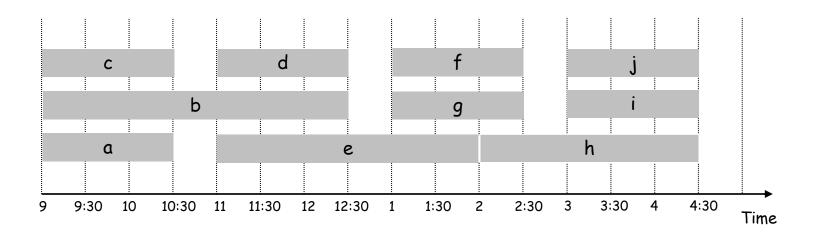
Note: graph coloring is very hard in general, but graphs corresponding to interval intersections are a much



Interval partitioning.

- Lecture j starts at s_j and finishes at f_j .
- Goal: find minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.

Ex: This schedule uses only 3.



Greedy template. Consider lectures in some order. If next lecture fits in the schedule we have, add it to one of the classrooms, otherwise open a new classroom.

[Earliest start time] Consider lectures in ascending order of start time $s_{\rm j}$.

[Earliest finish time] Consider lectures in ascending order of finish time $f_{\rm j}$.

[Shortest interval] Consider lectures in ascending order of interval length f_j - s_j .

[Fewest conflicts] For each lecture, count the number of conflicting lectures c_j . Schedule in ascending order of conflicts c_j .

counterexample for earliest finish time 3 counterexample for shortest interval 3 counterexample for fewest conflicts 3

Greedy template. Consider lectures in some order. If next lecture fits in the schedule we have, add it to one of the classrooms, otherwise open a new classroom.

[Earliest start time] Consider lectures in ascending order of start time s_{j} .

[Earliest finish time] Consider lectures in ascending order of finish time $f_{j\bar{l}}$

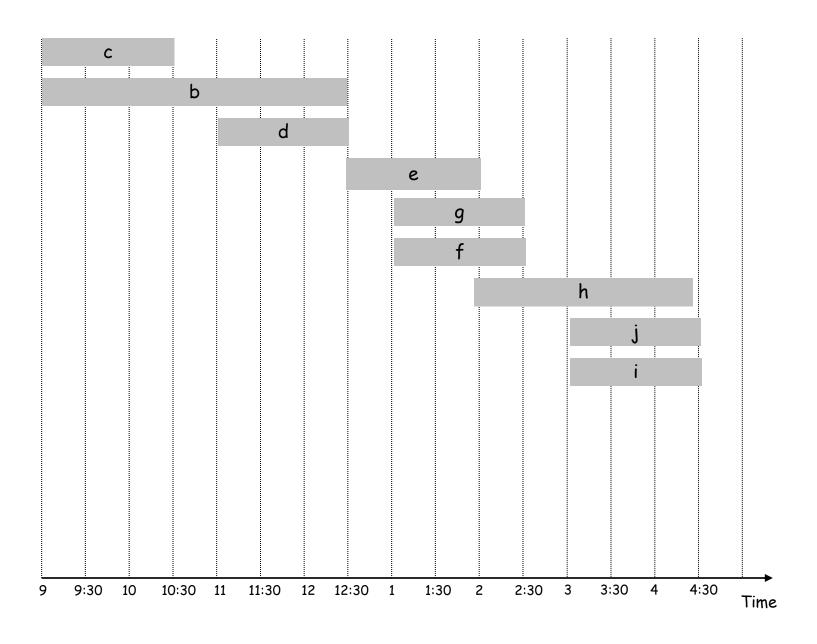
[Shortest interval] Consider lectures in ascending order of interval length $f_i - s_i$.

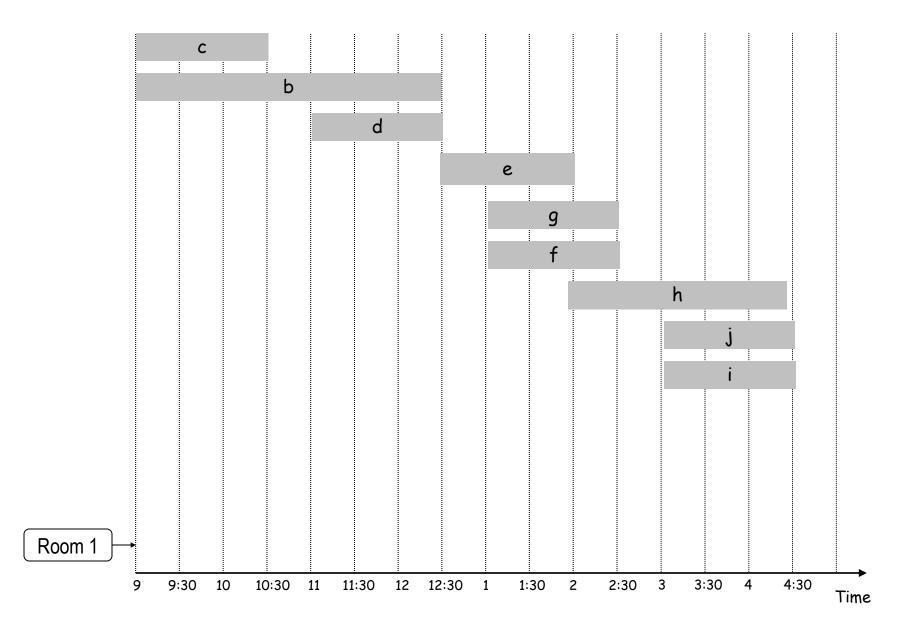
[Fewest conflicts] For each lecture, count the number of conflicting lectures \mathbf{c}_{j} . Schedule in ascending order of conflicts \mathbf{c}_{j} .

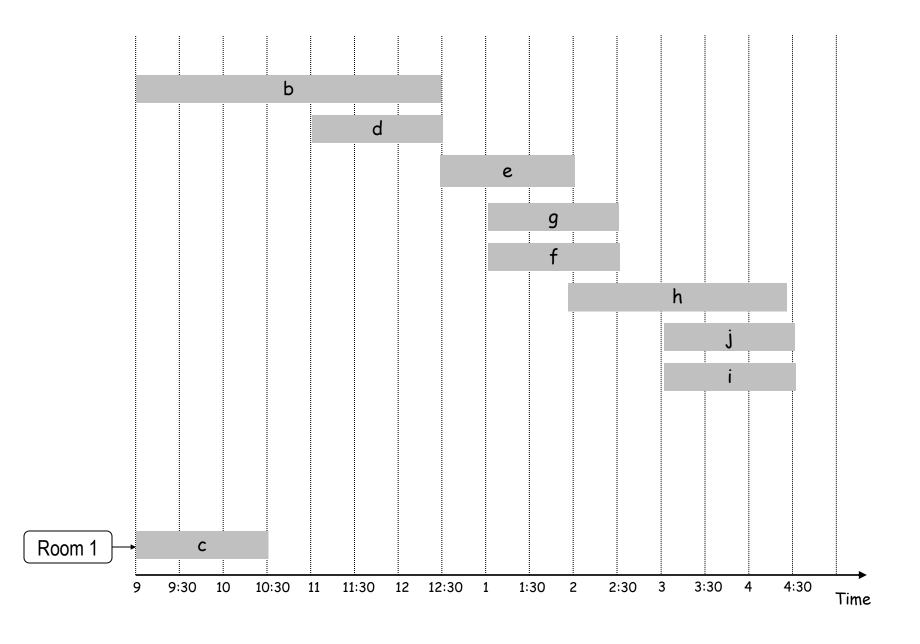
Interval Partitioning: Greedy Algorithm

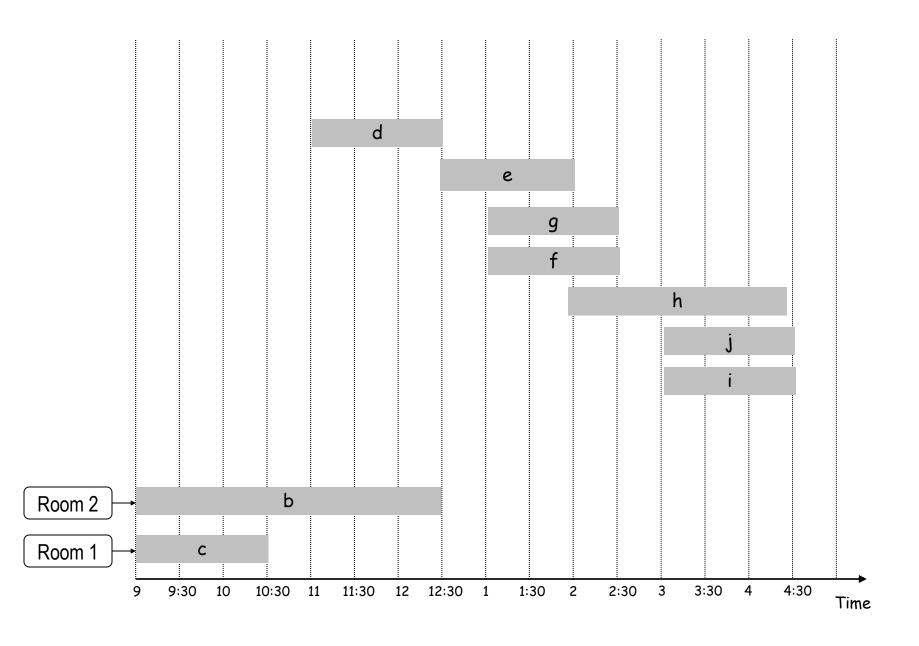
Greedy algorithm. Consider lectures in increasing order of start time: assign lecture to any compatible classroom.

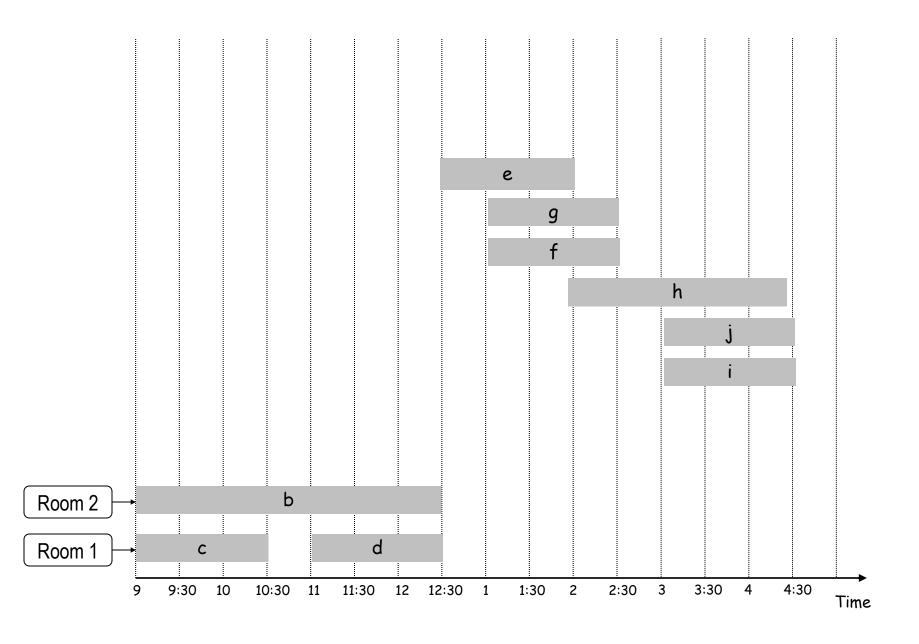
Implementation? Run-time? Exercises

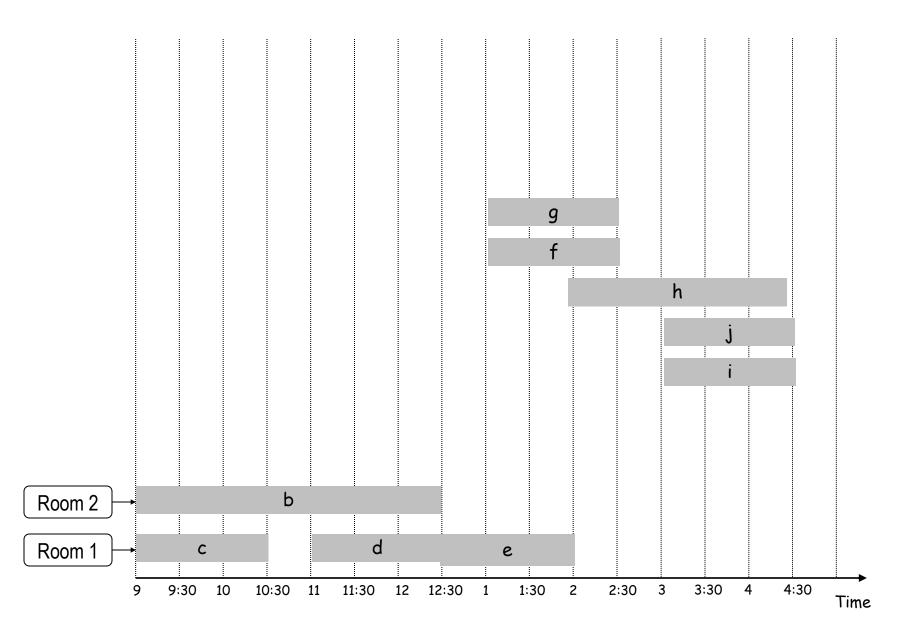


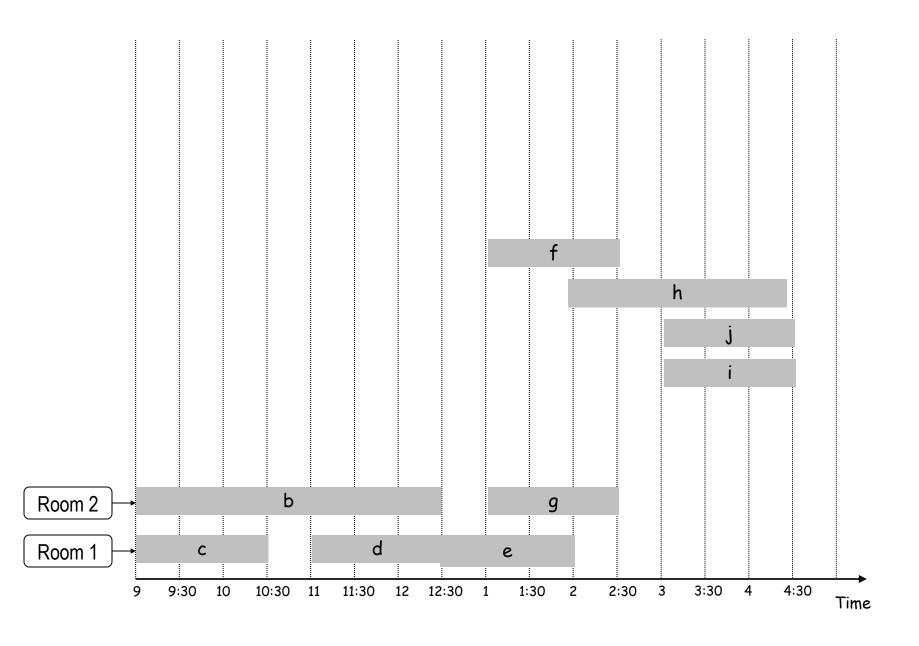


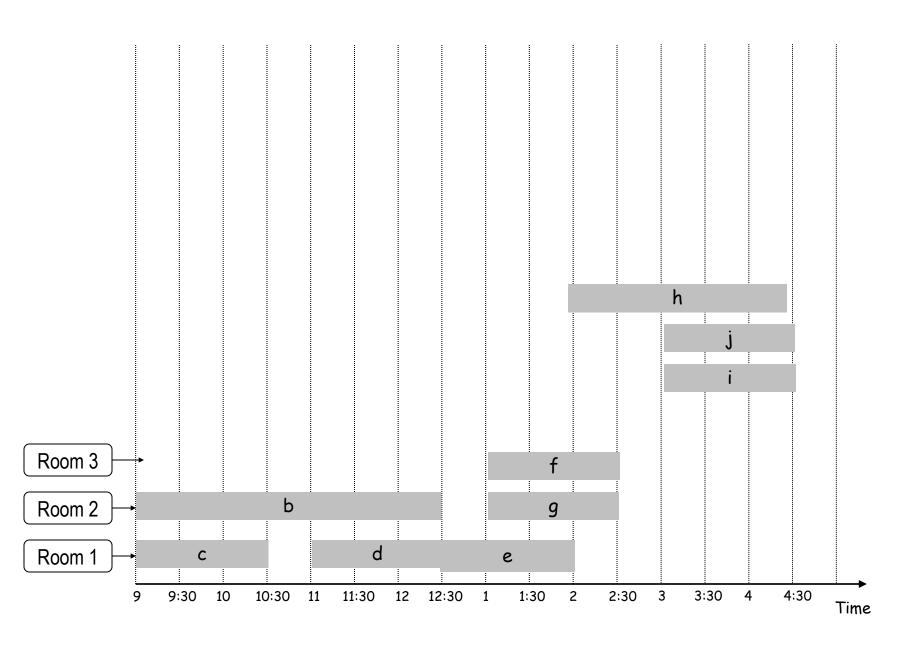


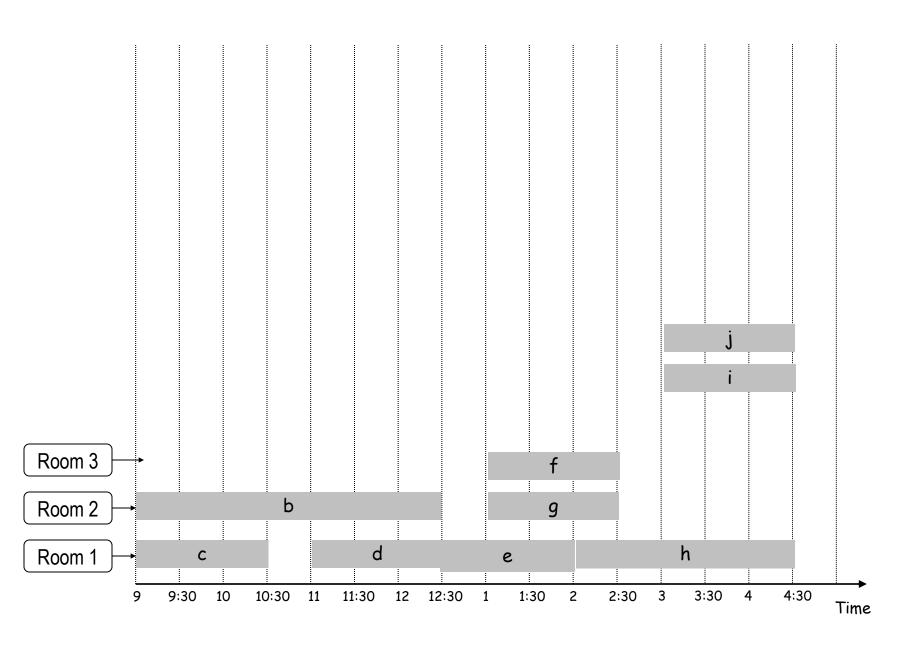


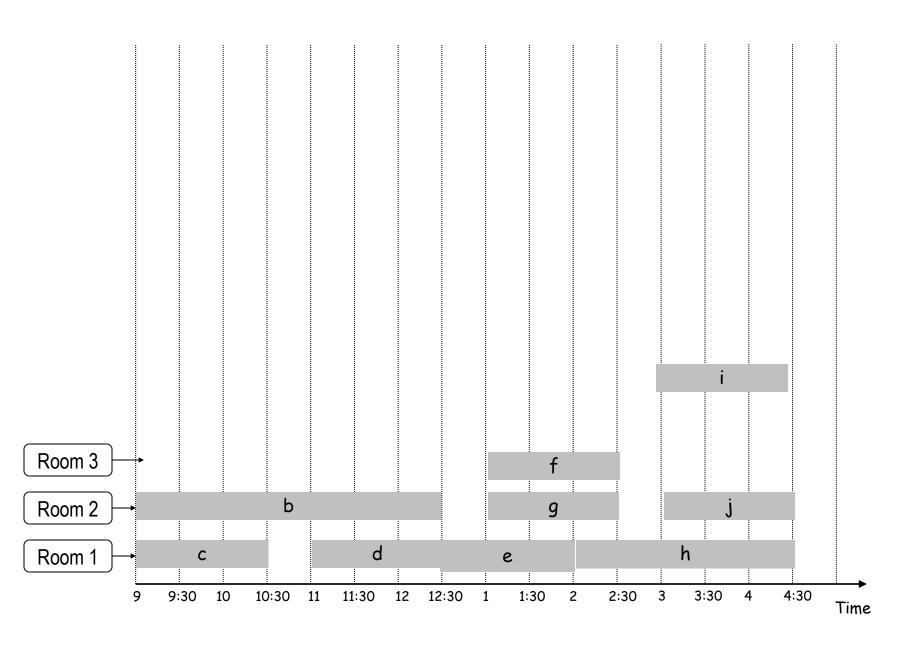


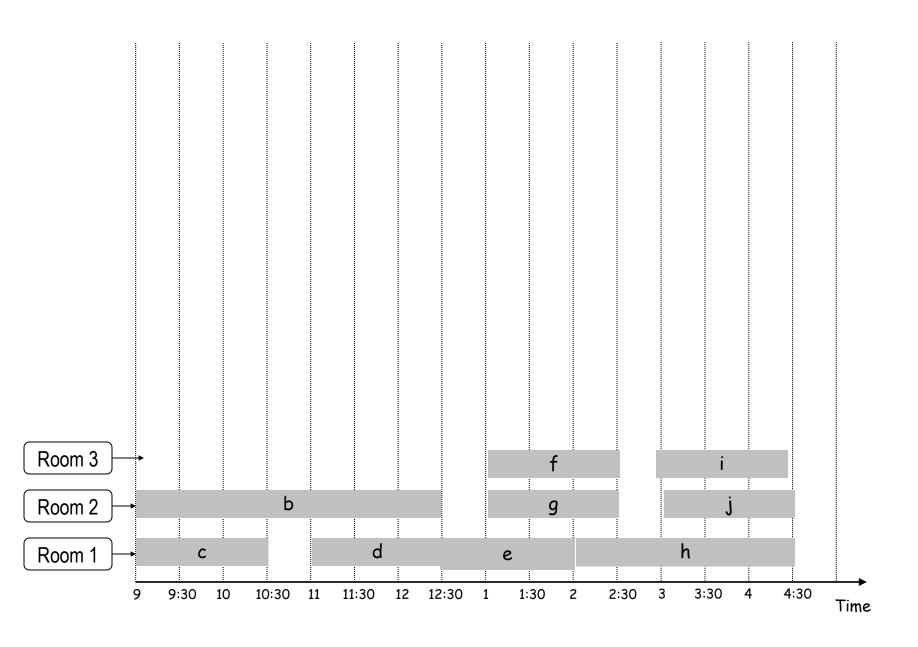












Interval Partitioning: A "Structural" Lower Bound on Optimal Solution

Def. The <u>depth</u> of a set of open intervals is the maximum number that contain any given time.

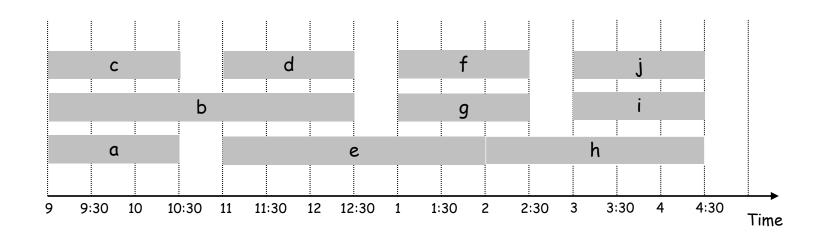
no collisions at ends

Key observation. Number of classrooms needed ≥ depth.

Ex: Depth of schedule below = 3 ⇒ schedule below is optimal.

a, b, c all contain 9:30

Q. Does there always exist a schedule equal to depth of intervals?



Interval Partitioning: Greedy Analysis

Theorem. Greedy algorithm is optimal.

Pf (exploit structural property).

- Let d = number of classrooms that the greedy algorithm allocates.
- Classroom d is opened because we needed to schedule a job, say j, that is incompatible with all d-1 previously used classrooms.
- Since we sorted by start time, all these incompatibilities are caused by lectures that start no later than s_j .
- \blacksquare Thus, we have d lectures overlapping at time s_j , i.e. depth $\ge d$
- "Key observation" all schedules use ≥ depth classrooms, so d = depth and greedy is optimal -