

#### **Greedy Algorithms**

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# An Advice on Problem Solving

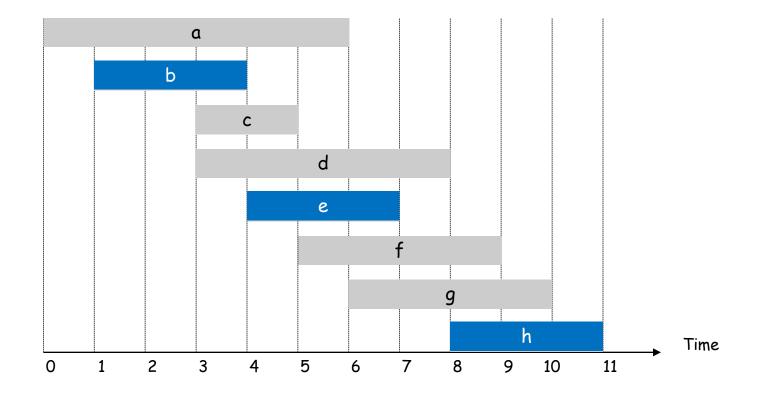
If possible, try not to use arguments of the following type in proofs:

- The Best case is ....
- The worst case is ....

These arguments need rigorous justification, and they are usually the main reason that your proofs can become wrong, or unjustified.

# **Interval Scheduling**

- Job j starts at s(j) and finishes at f(j).
- Two jobs compatible if they don't overlap.
- Goal: find maximum subset of mutually compatible jobs.



# **Possible Approaches for Inter Sched**

Sort the jobs in some order. Go over the jobs and take as much as possible provided it is compatible with the jobs already taken.

[Earliest start time] Consider jobs in ascending order of start time s<sub>i</sub>.

[Earliest finish time] Consider jobs in ascending order of finish time f<sub>i</sub>.

[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 Alg: Earliest Finish Time

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 f(1) \leq f(2) \leq \ldots \leq f(n).

A \leftarrow \emptyset

for j = 1 to n {

    if (job j compatible with A)

        A \leftarrow A \cup \{j\}

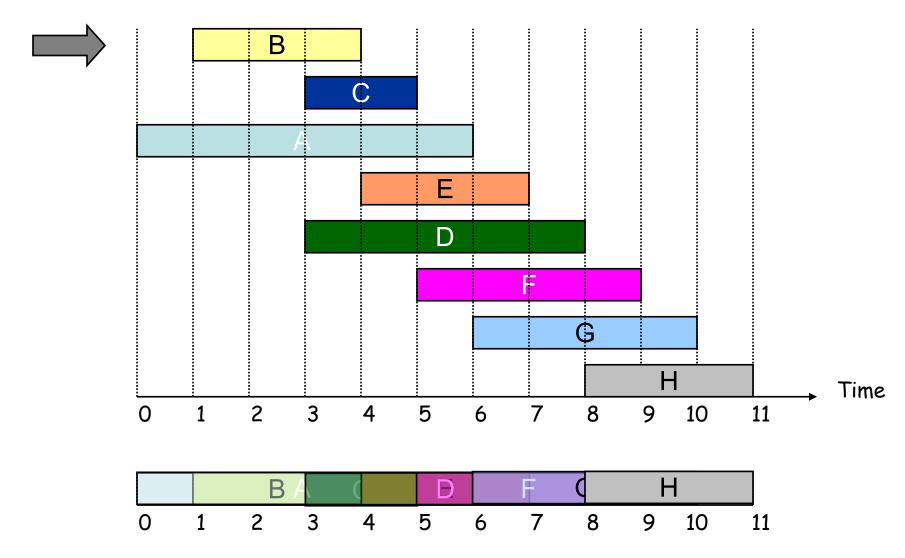
}

return A
```

Implementation. O(n log n).

- Remember job j\* that was added last to A.
- Job j is compatible with A if  $s(j) \ge f(j^*)_*$ .

## Greedy Alg: Example



#### Correctness

Theorem: Greedy algorithm is optimal.

#### Pf: (technique: "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 in order.

We show  $f(i_r) \le f(j_r)$  for all r, by induction on r.

Base Case:  $i_1$  chosen to have min finish time, so  $f(i_1) \le f(j_1)$ . IH:  $f(i_r) \le f(j_r)$  for some r IS: Since  $f(i_r) \le f(j_r) \le s(j_{r+1})$ ,  $j_{r+1}$  is among the candidates considered by greedy when it picked  $i_{r+1}$ , & it picks min finish, so  $f(i_{r+1}) \le f(j_{r+1})$ 

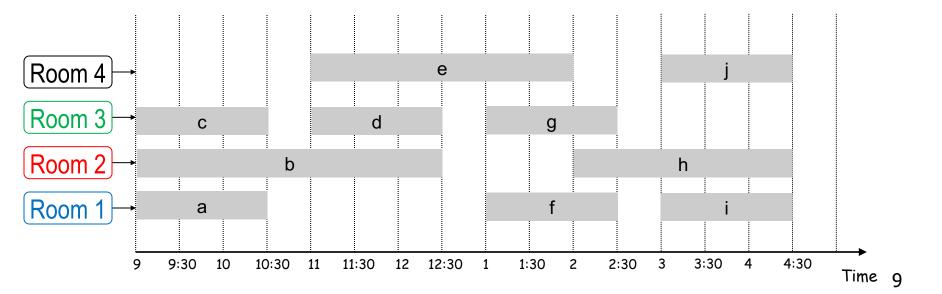
Observe that we must have  $k \ge m$ , else  $j_{k+1}$  is among (nonempty) set of candidates for  $i_{k+1}$ 

Interval Partitioning Technique: Structural

## **Interval Partitioning**

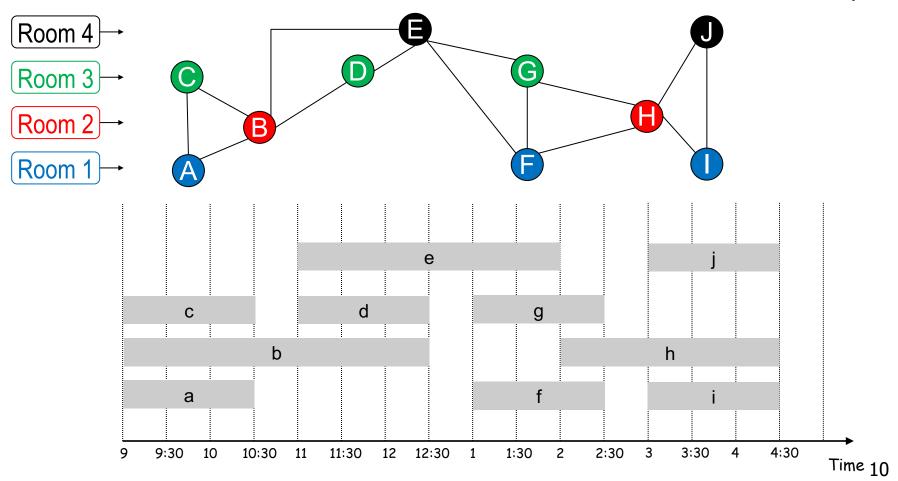
Lecture j starts at s(j) and finishes at f(j).

Goal: find minimum number of classrooms to schedule all lectures so that no two occur at the same time in the same room.



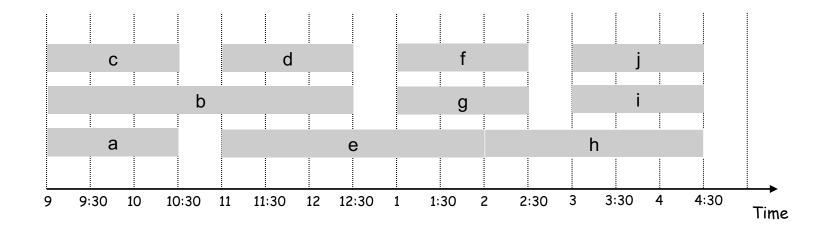
## **Interval Partitioning**

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



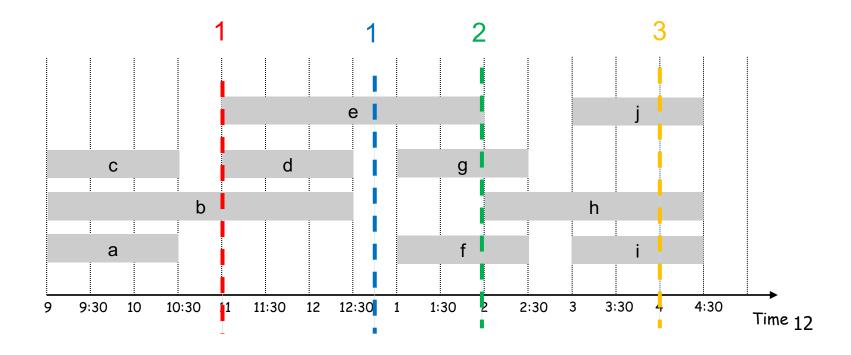
#### A Better Schedule

This one uses only 3 classrooms



#### A Structural Lower-Bound on OPT

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



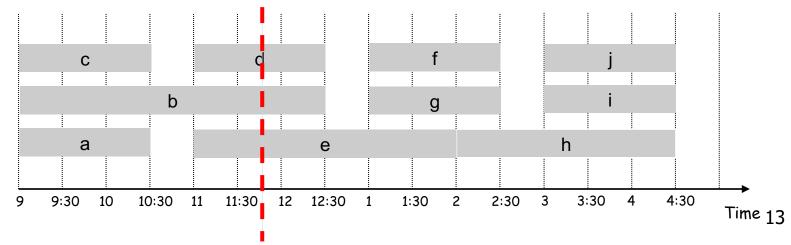
## A Structural Lower-Bound on OPT

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

Key observation. Number of classrooms needed  $\geq$  depth.

**Ex**: Depth of schedule below =  $3 \Rightarrow$  schedule below is optimal.

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



# A Greedy Algorithm

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

```
Sort intervals by starting time so that s_1 \le s_2 \le \ldots \le s_n.

d \leftarrow 0

for j = 1 to n {

    if (lect j is compatible with some classroom k, 1 \le k \le d)

        schedule lecture j in classroom k

    else

        allocate a new classroom d + 1

        schedule lecture j in classroom d + 1

        d \leftarrow d + 1

}
```

Implementation: Exercise!

#### Correctness

Observation: Greedy algorithm never schedules two incompatible lectures in the same classroom.

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).

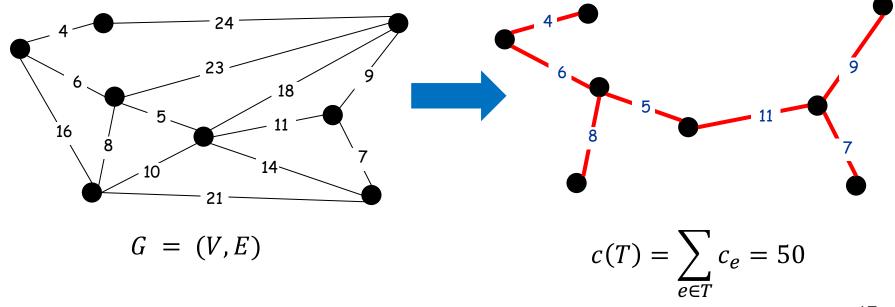
Thus, we have d lectures overlapping at time  $s(j) + \epsilon$ , i.e. depth  $\geq$  d

"OPT Observation"  $\Rightarrow$  all schedules use  $\geq$  depth classrooms, so d = depth and greedy is optimal •

## Minimum Spanning Tree Problem

# Minimum Spanning Tree (MST)

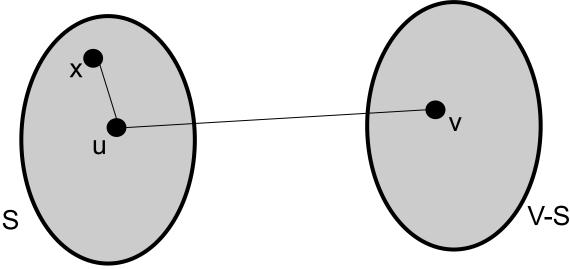
Given a connected graph G = (V, E) with real-valued edge weights  $c_e$ , an MST is a subset of the edges  $T \subseteq E$  such that T is a spanning tree whose sum of edge weights is minimized.



#### Cuts

In a graph G = (V, E) a cut is a bipartition of V into sets S, V - S for some  $S \subseteq V$ . We show it by (S, V - S)

An edge  $e = \{u, v\}$  is in the cut (S, V - S) if exactly one of u,v is in S.

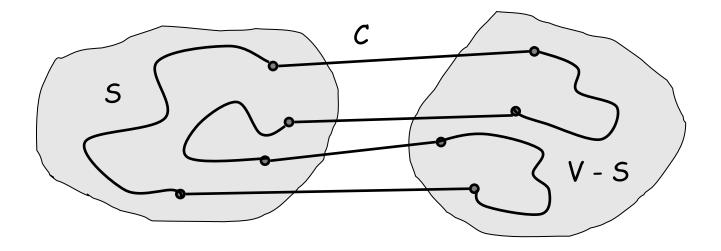


Obs: If G is connected then there is at least one edge in every cut.

#### **Cycles and Cuts**

Claim. A cycle crosses a cut (from S to V-S) an even number of times.

Pf. (by picture)

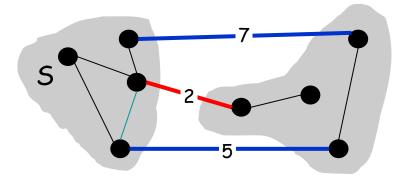


# Properties of the OPT

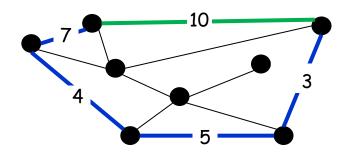
Simplifying assumption: All edge costs  $c_e$  are distinct.

Cut property: Let S be any subset of nodes (called a cut), and let e be the min cost edge with exactly one endpoint in S. Then every MST contains e.

Cycle property. Let C be any cycle, and let f be the max cost edge belonging to C. Then no MST contains f.



red edge is in the MST



Green edge is not in the MST

# Cut Property: Proof

Simplifying assumption: All edge costs  $c_e$  are distinct.

Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then T\* contains e.

Pf. By contradiction

Suppose  $e = \{u, v\}$  does not belong to T\*.

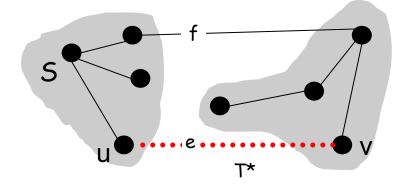
Adding e to T\* creates a cycle C in T\*.

C crosses S even number of times $\Rightarrow$  there exists another edge, say f, that leaves S.

 $T = T^* \cup \{e\} - \{f\}$  is also a spanning tree.

Since  $c_e < c_f$ ,  $c(T) < c(T^*)$ .

This is a contradiction.



# Cycle Property: Proof

Simplifying assumption: All edge costs  $c_e$  are distinct.

Cycle property: Let C be any cycle in G, and let f be the max cost edge belonging to C. Then the MST T\* does not contain f.

Pf. (By contradiction)

Suppose f belongs to T\*.

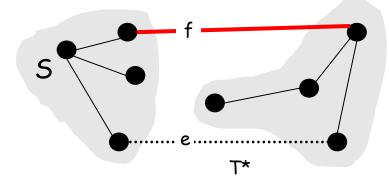
Deleting **f** from T\* cuts T\* into two connected components.

There exists another edge, say e, that is in the cycle and connects the components.

 $T = T^* \cup \{e\} - \{f\}$  is also a spanning tree.

Since  $c_e < c_f$ ,  $c(T) < c(T^*)$ .

This is a contradiction.



## Kruskal's Algorithm [1956]

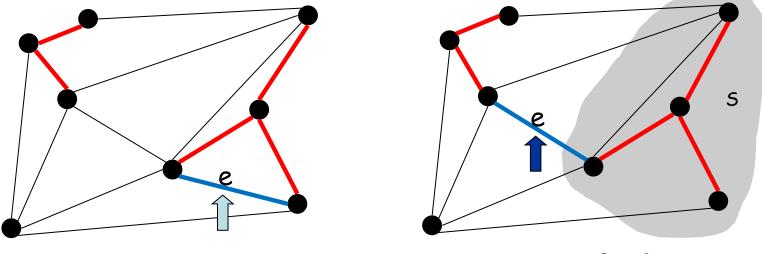
```
Kruskal(G, c) {
   Sort edges weights so that c_1 \leq c_2 \leq \ldots \leq c_m.
   T \leftarrow \emptyset
   foreach (u \in V) make a set containing singleton {u}
   for i = 1 to m
    Let (u, v) = e_i
        if (u and v are in different sets) {
            T \leftarrow T \cup \{e_i\}
            merge the sets containing u and v
        }
    return T
}
```

# Kruskal's Algorithm: Pf of Correctness

Consider edges in ascending order of weight.

Case 1: If adding e to T creates a cycle, discard e according to cycle property.

Case 2: Otherwise, insert e = (u, v) into T according to cut property where S = set of nodes in u's connected component.



Case 1

Case 2

## Implementation: Kruskal's Algorithm

Implementation. Use the union-find data structure.

- Build set *T* of edges in the MST.
- Maintain a set for each connected component.
- O(m log n) for sorting and O(m log n) for union-find

```
Kruskal(G, c) {

Sort edges weights so that c_1 \leq c_2 \leq \ldots \leq c_m.

T \leftarrow \emptyset

for each (u \in V) make a set containing singleton \{u\}

for i = 1 to m

Let (u, v) = e_i

if (u and v are in different sets) {

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}

return T

}
```