

## 4. Greedy Algorithms

# Optimization problems

- For the next four worksheets, we will consider **optimization problems**:
  - We have a problem with several valid solutions.
  - There is an **objective function** that tells us how good or bad each valid solution is.
  - We are looking for the valid solution that minimizes or maximizes the value of the objective function.
- We will look at two algorithm design paradigms:
  - Greedy algorithms
  - Dynamic programming

# Optimization problems

- Examples:
  - Given a set of intervals, find the largest set of intervals that don't overlap.
    - ◆ The objective function is the cardinality of the set.
  - Given a set of jobs with deadlines, order the jobs so as to minimize their total lateness.
    - ◆ The objective function is the total lateness.
  - Given a weighted connected graph, find a spanning tree with the smallest total edge weight.
    - ◆ The objective function is the sum of the weights of the edges in the spanning tree.

# Defining greedy algorithms

- A greedy algorithm proceeds by:
  - Making a choice based on a simple, local criterion.
  - Solving the subproblem that results from that choice.
  - Combining the choice and the subproblem solution.
- We can think of a greedy algorithm as making a **sequence of choices**.
- There is no precise definition of **greedy**.

# Defining greedy algorithms

- Examples of choice:
  - interval with the earliest finishing time.
  - job with the earliest deadline.
  - item needed further in the future.
  - smallest weight edge that does not create a cycle.

# Defining greedy algorithms

- Does a greedy algorithm always give the correct solution?
  - Sometimes yes, sometimes no.
  - There are some classes of problems (e.g. matroids) for which there exists a greedy algorithm that always returns the correct solution.
  - There are other problems where no one knows any greedy algorithm with this property.
    - ◆ e.g. weighted interval scheduling.

# Proving a greedy algorithm correct

- Method 1: “the greedy algorithm stays ahead”
  - It’s basically a proof by induction.
  - You compare
    - ◆ The list of choices made by the greedy algorithm, to
    - ◆ A similar list for an optimal solution
  - You show that at each stage, the greedy choice is *at least as good* as the choice in the optimal solution.
  - Examples:
    - ◆ The algorithm for the interval scheduling problem (4.1).

# Proving a greedy algorithm correct

- Method 2: exchange arguments
  - Prove that if  $S$  is an *arbitrary* solution, and  $G$  is the greedy solution, then you can modify  $S$  slightly to get  $S'$  such that:
    - ◆  $S'$  is more similar to  $G$  than  $S$ .
    - ◆  $S'$  is at least as good a solution as  $S$ .
  - Examples of what “more similar to” might mean:
    - ◆ Has more edges in common with.
    - ◆ Has a longer initial sequence that’s the same as.



# Proving a greedy algorithm correct

- Method 2: exchange arguments
  - Why this works:
    - ◆ You start from any arbitrary solution  $S_0$ .
    - ◆ You get  $S_1$  which is at least as good as  $S_0$  and closer to  $G$ .
    - ◆ You get  $S_2$  which is at least as good as  $S_1$  and closer to  $G$ .
    - ◆ You get  $S_3$  which is at least as good as  $S_2$  and closer to  $G$ .
    - ◆ ...
    - ◆ You get  $G$  which is at least as good as  $S_t$  and closer to  $G$ .

# Proving a greedy algorithm correct

- Method 2: exchange arguments
  - By induction (or transitivity):
    - ◆  $G$  is “at least as good” a solution as  $S_0$ .
  - This works no matter what  $S_0$  is
    - ◆ Even if  $S_0$  is an optimal solution.
  - So  $G$  is at least as good as an optimal solution.
  - Therefore  $G$  is optimal.