Announcements:

• midterm next tuesday, in class.

Reading: 6.0-6.3

Last time:

- Polynomial Multiplication
- Fast Fourier Transform

Today:

- Dynamic Programming
- Weighted interval scheduling

Dynamic Programming

"divide problem into small number of subproblems and **memoize** solution to avoid redundant computation"

Example: Weighted Interval Scheduling

input:

- $n \text{ jobs } J = \{1, \dots, n\}$
- $s_i = \text{start time of job } i$
- $f_i = \text{finish time of job } i$
- v_i = value of job i

output: Schedule $S \subseteq J$ of compatible jobs with maximum total value.

Recall Greedy: "earliest finish time"

Idea: job i is either in OPT(J) or not.

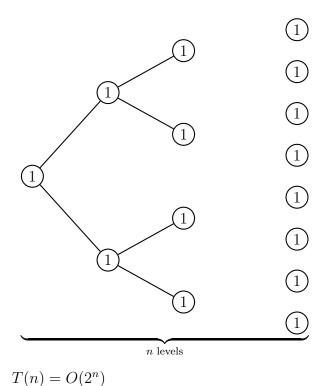
- 1. let J' = jobs compatible with i in J.
- 2. let $V = \text{value of OPT if } "i \in OPT(j)"$.

$$= v_i + \mathrm{OPT}(J')$$

- 3. let $V' = \text{vale of OPT if "} i \notin \text{OPT}(j)$ " $= \text{OPT}(J \setminus \{i\}).$
- 4. return OPT(J) = max(V, V').

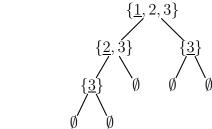
Note: subproblems: schedule J' and $J \setminus \{i\}$.

Recurrence: T(n) = 2T(n-1) + 1



Challenge 1: redundant computation

Example:



Note: $OPT({3})$ called twice!

Solution: memoize

"when computing the value of a subproblem save the answer to avoid computing it again"

Result: runtime = # of subproblems \times cost to combine.

Challenge 2: could have too many subproblems.

(could be exponential!)

Solution: require "succinct description" of subproblems.

Idea: for interval scheduling, process jobs in order of start time so subproblems suffixes of order.

- sort jobs by increasing start time, $s_1 \leq s_2 \leq \cdots \leq s_n$.
- let next[i] denote job with earliest start time after i finishes. (if none, set next[i] = n + 1)
- subproblems when processing job 1:
 - $J' = {\text{next}[i], \text{next}[i] + 1, \dots, n}$
 - $J \setminus \{i\} = \{2, 3, \dots, n\}$
- suffix $\{j,...,n\}$ is succinctly described by "j".

could have too many sub- Algorithm: Weighted Interval Scheduling:

- 1. sort jobs by increasing start time.
- 2. initialize array next[i].
- 3. initialize memo $[i] = \emptyset$ for all i.
- 4. initialize memo[n+1] = 0.
- 5. compute OPT(1).

Subroutine: OPT(i)

- 1. if memo[i] $\neq \emptyset$, return memo[i].
- 2. $\text{memo}[i] \leftarrow \max(v_i + \text{OPT}(\text{next}[i]), \text{OPT}(i+1)).$
- 3. return memo(i).

Correctness

"OPT(i)" is correct (by induction on i)

Runtime Analysis

- \bullet *n* subproblems
- constant time to combine
- initialization: sorting & precomputing next array

Runtime: O(n)+ initialization = $O(n \log n)$

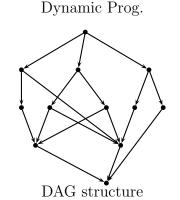
Key Ideas of Dynamic Program- Finding Optimal Schedule ming

Subproblems must be:

- 1. succinct (only a polynomial number of them)
- 2. efficiently combinable.
- 3. partially ordered (avoid infinite loops), e.g.,
 - process elements "once and for all"
 - "measure of progress/size".

Comparison to Divide and Conquer





Iterative DPs

"fill in memoization table from bottom to top"

Algorithm: iterative weighted interval scheduling

- 1. memo[n+1] = 0.
- 2. for i = n down to 1.

$$memo[i] = max(v_i + memo[next(i)], memo[i + 1]).$$

"traverse memoization table to find schedule"

Algorithm: schedule

$$i=1$$
 while $i < n$
$$\text{if memo}[i+1] < v_i + \text{memo}[\text{next}(i)]$$

$$\text{schedule } i; \ i \leftarrow \text{next}(i).$$

else

$$i \leftarrow i + 1$$
.

endif

endwhile

Suggested Approach

- I. identify subproblem in english
 - "optimal schedule OPT(i) $\{i, \ldots, n\}$ (sorted by increasing start time)"
- II. specify sumbroblem recurrence

$$OPT(i) = max(OPT(i + 1), v_i + OPT(next(i)))$$

III. identify base case

$$OPT(n+1) = 0$$

IV. write iterative DP.