

# COMP 355

## Advanced Algorithms

Dynamic Programming:  
Weighted Interval Scheduling  
KT (Ch.6 Intro, 6.1-6.2)



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## Google Interview Question

Given two sorted arrays with  $N$  elements each, find the median of their union in  $O(\log n)$ .



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## Algorithmic Paradigms

**Greed.** Build up a solution incrementally, myopically optimizing some local criterion.

**Divide-and-conquer.** Break up a problem into two sub-problems, solve each sub-problem independently, and combine solution to sub-problems to form solution to original problem.

**Dynamic programming.** Break up a problem into a series of overlapping sub-problems, and build up solutions to larger and larger sub-problems.

## Dynamic Programming Applications

### Areas.

- Bioinformatics.
- Control theory.
- Information theory.
- Operations research.
- Computer science: theory, graphics, AI, systems, ....

### Some famous dynamic programming algorithms.

- Viterbi for hidden Markov models.
- Unix diff for comparing two files.
- Smith-Waterman for sequence alignment.
- Bellman-Ford for shortest path routing in networks.
- Cocke-Kasami-Younger for parsing context free grammars.

# Dynamic Programming

DP relies are two important structural qualities:

- **Optimal substructure:** (principle of optimality)
  - For the global problem to be solved optimally, each subproblem should be solved optimally.
- **Overlapping Subproblems**
  - The number of distinct subproblems is reasonably small, ideally polynomial in the input size.

# Generating Subproblems

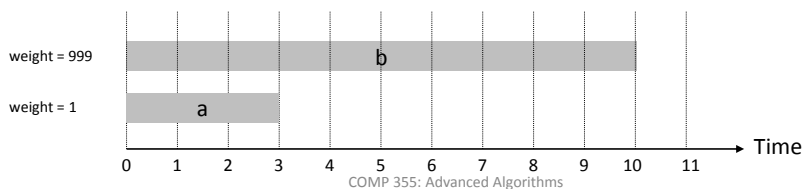
- **Top-Down:**
  - A solution to a DP problem is expressed recursively.
  - Applies recursion directly to solve the problem.
  - The same recursive call is often made many times.
  - Use **memoization** (record the results of recursive calls) so that subsequent calls to a previously solved subproblem are handled by table look-up.
- **Bottom-up:**
  - Formulate problem recursively, but solve iteratively
  - Combine the solutions to small subproblems to obtain the solution to larger subproblems.
  - The results are stored in a table.

# Unweighted Interval Scheduling Review

**Recall.** Greedy algorithm works if all weights are 1.

- Consider jobs in ascending order of finish time.
- Add job to subset if it is compatible with previously chosen jobs.

**Observation.** Greedy algorithm can fail spectacularly if arbitrary weights are allowed.

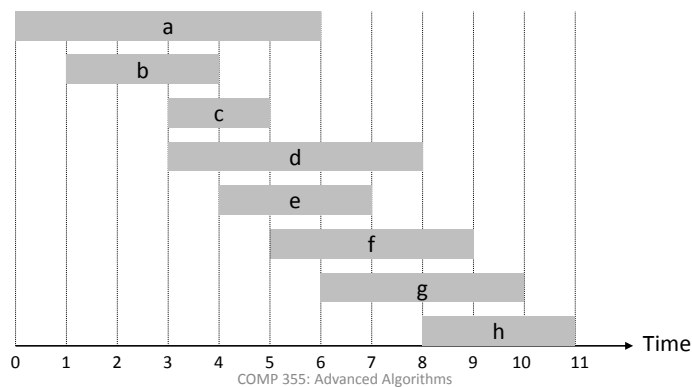


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# Weighted Interval Scheduling

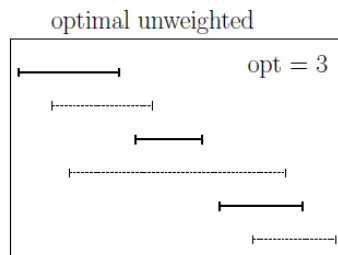
**Weighted interval scheduling problem.**

- Job  $j$  starts at  $s_j$ , finishes at  $f_j$ , and has weight or value  $v_j$ .
- Two jobs **compatible** if they don't overlap.
- Goal: find maximum **weight** subset of mutually compatible jobs.

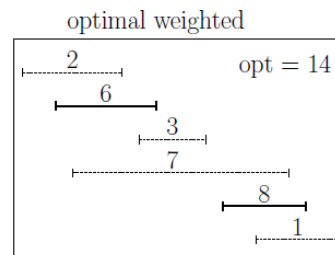


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## Weighted vs. Unweighted



(a)



(b)



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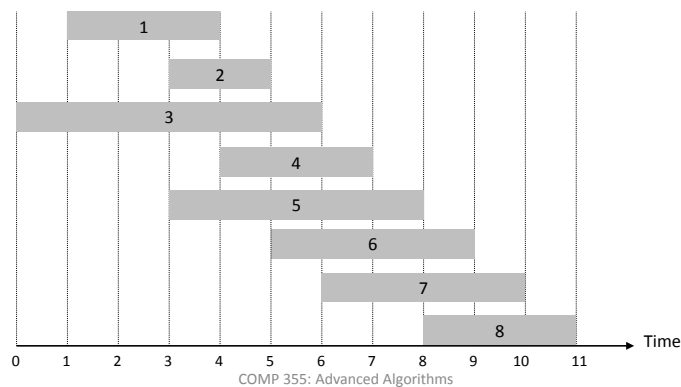
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## Weighted Interval Scheduling

**Notation.** Label jobs by finishing time:  $f_1 \leq f_2 \leq \dots \leq f_n$ .

**Def.**  $p(j)$  = largest index  $i < j$  such that job  $i$  is compatible with  $j$ .

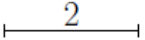
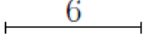
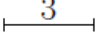
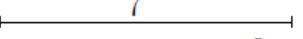
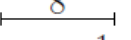
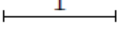
**Ex:**  $p(8) = 5$ ,  $p(7) = 3$ ,  $p(2) = 0$ .



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## Weighted Input and P-Values

$j$	intervals and values	$p(j)$
1		0
2		0
3		1
4		0
5		3
6		3



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## Dynamic Programming: Recursive Formulation

**Notation.**  $OPT(j)$  = value of optimal solution to the problem consisting of job requests 1, 2, ...,  $j$ .

Case 1: OPT selects job  $j$ .

- can't use incompatible jobs  $\{ p(j) + 1, p(j) + 2, \dots, j - 1 \}$
- must include optimal solution to problem consisting of remaining compatible jobs 1, 2, ...,  $p(j)$

↘  
↙ optimal substructure

Case 2: OPT does not select job  $j$ .

- must include optimal solution to problem consisting of remaining compatible jobs 1, 2, ...,  $j-1$

$$OPT(j) = \begin{cases} 0 & \text{if } j = 0 \\ \max\{v_j + OPT(p(j)), OPT(j-1)\} & \text{otherwise} \end{cases}$$

DP Selection Principle:

When given a set of feasible options to choose from, try them all and take the best.

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## Weighted Interval Scheduling: Brute Force

Brute force algorithm.

```

Input:  $n, s_1, \dots, s_n, f_1, \dots, f_n, v_1, \dots, v_n$ 

Sort jobs by finish times so that  $f_1 \leq f_2 \leq \dots \leq f_n$ .

Compute  $p(1), p(2), \dots, p(n)$ 

Compute-Opt( $j$ ) {
    if ( $j = 0$ )
        return 0
    else
        return  $\max(v_j + \text{Compute-Opt}(p(j)), \text{Compute-Opt}(j-1))$ 
}

```

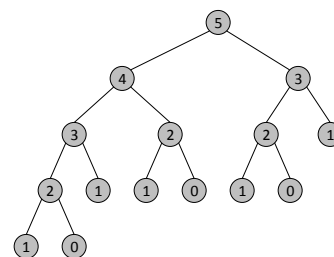
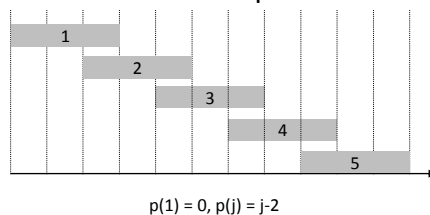
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## Weighted Interval Scheduling: Brute Force

**Observation.** Recursive algorithm fails spectacularly because of redundant sub-problems  $\Rightarrow$  exponential algorithms.

**Ex.** Number of recursive calls for family of "layered" instances - grows like Fibonacci sequence.



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## Weighted Interval Scheduling: Memoization

**Memoization.** Store results of each sub-problem in a cache; lookup as needed.

```

Input:  $n, s_1, \dots, s_n, f_1, \dots, f_n, v_1, \dots, v_n$ 

Sort jobs by finish times so that  $f_1 \leq f_2 \leq \dots \leq f_n$ .
Compute  $p(1), p(2), \dots, p(n)$ 

for  $j = 1$  to  $n$        $\leftarrow$  global array
     $M[j] = \text{empty}$ 
 $M[0] = 0$ 

M-Compute-Opt( $j$ ) {
    if ( $M[j]$  is empty)
         $M[j] = \max(v_j + \text{M-Compute-Opt}(p(j)), \text{M-Compute-Opt}(j-1))$ 
    return  $M[j]$ 
}
  
```

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## Weighted Interval Scheduling: Running Time

**Claim.** Memoized version of algorithm takes  $O(n \log n)$  time.

- Sort by finish time:  $O(n \log n)$ .
- Computing  $p(\cdot)$ :  $O(n)$  after sorting by start time.
- $\text{M-Compute-Opt}(j)$ : each invocation takes  $O(1)$  time and either
  - (i) returns an existing value  $M[j]$
  - (ii) fills in one new entry  $M[j]$  and makes two recursive calls
- Overall running time of  $\text{M-Compute-Opt}(n)$  is  $O(n)$ .

**Remark.**  $O(n)$  if jobs are pre-sorted by start and finish times.

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# Weighted Interval Scheduling: Bottom-Up

Bottom-up dynamic programming. Unwind recursion.

```

Input:  $n, s_1, \dots, s_n, f_1, \dots, f_n, v_1, \dots, v_n$ 

Sort jobs by finish times so that  $f_1 \leq f_2 \leq \dots \leq f_n$ .

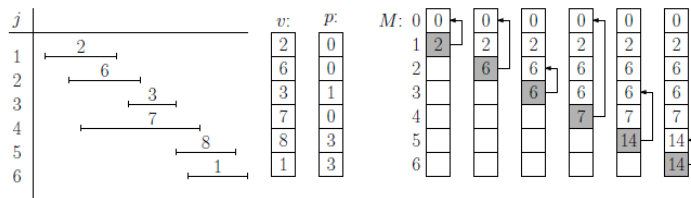
Compute  $p(1), p(2), \dots, p(n)$ 

Iterative-Compute-Opt {
   $M[0] = 0$ 
  for  $j = 1$  to  $n$ 
    if  $M[j-1] > v_j + M[p(j)]$ :
       $M[j] = M[j-1]; \text{pred}[j] = j-1;$ 
    else:
       $M[j] = v_j + M[p(j)]; \text{pred}[j] = p[j];$ 
  }
  
```

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## Computing the Final Schedule



Example of iterative construction and predecessor values. The final optimal value is 14. By following the predecessor pointers back from  $M[6]$  we see that the requests that are in the schedule are 5 and 2.

Computing Weighted Interval Scheduling Schedule

```

get-schedule() {
  j = n
  sched = (empty list)
  while (j > 0) {
    if (pred[j] == p[j]) {
      prepend j to the front of sched
    }
    j = pred[j]
  }
}
  
```



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