

# Dynamic Programming

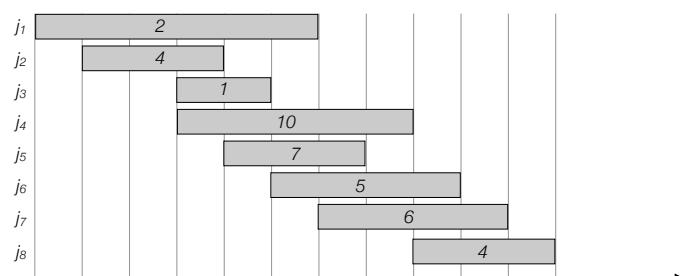
Algorithm Design 6.1, 6.2, 6.4

Thank you to Kevin Wayne for inspiration to slides

3

## Applications

- In class (today and next time)
  - Weighted interval scheduling
    - Set of weighted intervals with start and finishing times
    - Goal: find maximum weight subset of non-overlapping intervals



## Applications

- In class (today and next time)

2

## Applications

- Today and next time
  - Weighted interval scheduling
  - Subset Sum and Knapsack
    - Set of items each having a weight and a value
    - Knapsack with a bounded capacity
    - Goal: fill knapsack so as to maximise the total value.



value	10	8	2	5	15	4
weight	2	3	1	2	5	4

Capacity 8

4

## Applications

- Today and next time
  - Weighted interval scheduling
  - Subset Sum and Knapsack
  - Sequence alignment
    - Given two strings A and B how many edits (insertions, deletions, relabelings) is needed to turn A into B?

A C A **A** G T C  
- C A **T** G T -  
1 mismatch, 2 gaps

A C A A - G T C  
- C A - T G T -  
0 mismatches, 4 gaps

5

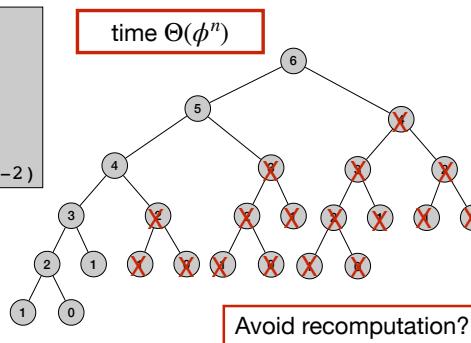
## Computing Fibonacci numbers

- Fibonacci numbers:

$$F_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ F_{n-1} + F_{n-2} & \text{otherwise} \end{cases}$$

- First try:

```
Fib(n)
if n = 0
  return 0
else if n = 1
  return 1
else
  return Fib(n-1) + Fib(n-2)
```



## Dynamic Programming

- **Greedy.** Build solution incrementally, optimizing some local criterion.
- **Divide-and-conquer.** Break up problem into **independent** subproblems, solve each subproblem, and combine to get solution to original problem.
- **Dynamic programming.** Break up problem into **overlapping** subproblems, and build up solutions to larger and larger subproblems.
  - Can be used when the problem have “optimal substructure”:
    - Solution can be constructed from optimal solutions to subproblems
    - Use dynamic programming when subproblems overlap.

6

## Memoized Fibonacci numbers

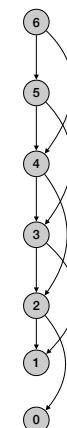
- Fibonacci numbers:

$$F_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ F_{n-1} + F_{n-2} & \text{otherwise} \end{cases}$$

- Remember already computed values:

```
for j=1 to n
  F[j] = null
Mem-Fib(n)

Mem-Fib(n)
if n = 0
  return 0
else if n = 1
  return 1
else
  if F[n] is empty
    F[n] = Mem-Fib(n-1) + Mem-Fib(n-2)
  return F[n]
```



## Bottom-up Fibonacci numbers

- Fibonacci numbers:

$$F_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ F_{n-1} + F_{n-2} & \text{otherwise} \end{cases}$$

- Remember already computed values:

```
Iter-Fib(n)
F[0] = 0
F[1] = 1
for i = 2 to n
    F[i] = F[i-1] + F[i-2]
return F[n]
```

time  $\Theta(n)$   
space  $\Theta(n)$

## Bottom-up Fibonacci numbers - save space

- Fibonacci numbers:

$$F_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ F_{n-1} + F_{n-2} & \text{otherwise} \end{cases}$$

- Remember last two computed values:

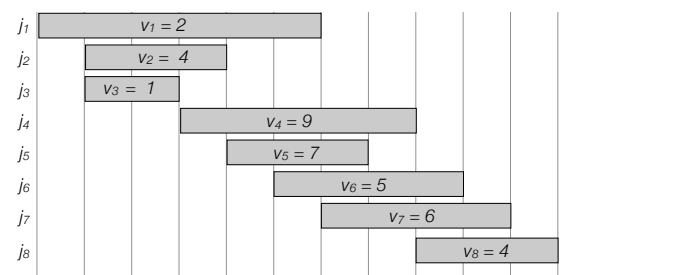
```
Iter-Fib(n)
previous = 0
current = 1
for i = 1 to n
    next = previous + current
    previous = current
    current = next
return current
```

time  $\Theta(n)$   
space  $\Theta(1)$

## Weighted Interval Scheduling

## Weighted interval scheduling

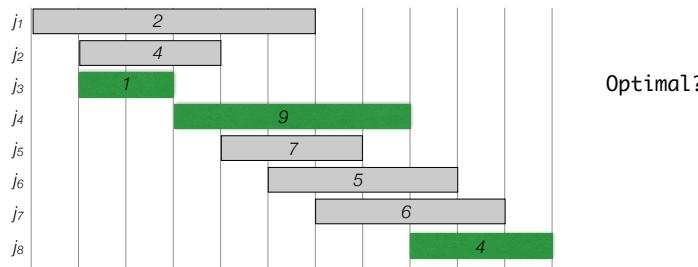
- Weighted interval scheduling problem
  - $n$  jobs (intervals)
  - Job  $i$  starts at  $s_i$ , finishes at  $f_i$  and has weight/value  $v_i$ .
  - Goal: Find maximum weight subset of non-overlapping (compatible) jobs.



## Weighted interval scheduling

- Weighted interval scheduling problem

- n jobs (intervals)
- Job  $i$  starts at  $s_i$ , finishes at  $f_i$  and has weight/value  $v_i$ .
- Goal: Find maximum weight subset of non-overlapping (compatible) jobs.

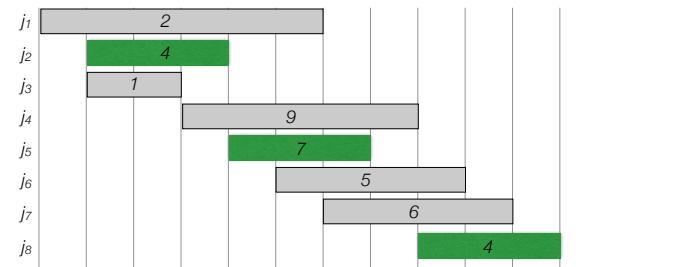


13

## Weighted interval scheduling

- Weighted interval scheduling problem

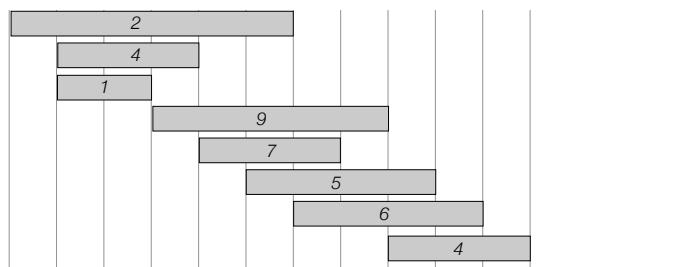
- n jobs (intervals)
- Job  $i$  starts at  $s_i$ , finishes at  $f_i$  and has weight/value  $v_i$ .
- Goal: Find maximum weight subset of non-overlapping (compatible) jobs.



14

## Weighted interval scheduling

- Label/sort jobs by finishing time:  $f_1 \leq f_2 \leq \dots \leq f_n$

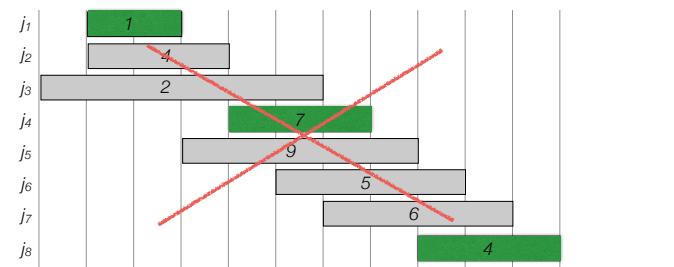


15

## Weighted interval scheduling

- Label/sort jobs by finishing time:  $f_1 \leq f_2 \leq \dots \leq f_n$

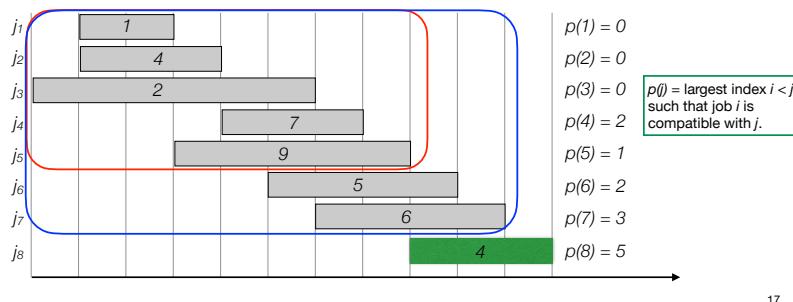
• Greedy?



16

## Weighted interval scheduling

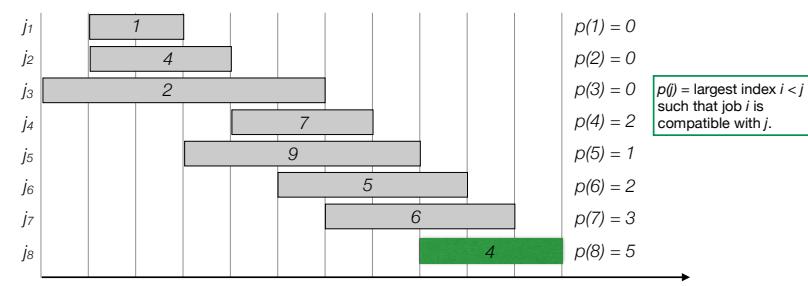
- Label/sort jobs by finishing time:  $f_1 \leq f_2 \leq \dots \leq f_n$
- Optimal solution OPT:
  - Case 1.** OPT selects last job  
 $OPT = v_n + \text{optimal solution to subproblem on the subset of jobs ending before job } n \text{ starts}$
  - Case 2.** OPT does not select last job  
 $OPT = \text{optimal solution to subproblem on } 1, \dots, n-1$



17

## Weighted interval scheduling

- Label/sort jobs by finishing time:  $f_1 \leq f_2 \leq \dots \leq f_n$
- Optimal solution OPT:
  - Case 1.** OPT selects last job  
 $OPT = v_n + \text{optimal solution to subproblem on } 1, \dots, p(n)$
  - Case 2.** OPT does not select last job  
 $OPT = \text{optimal solution to subproblem on } 1, \dots, n-1$

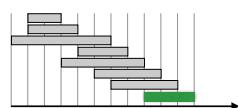


18

## Weighted interval scheduling

- $OPT(j)$  = value of optimal solution to the problem consisting job requests  $1, 2, \dots, j$ .
  - Case 1.**  $OPT(j)$  selects job  $j$   
 $OPT(j) = v_j + \text{optimal solution to subproblem on } 1, \dots, p(j)$
  - Case 2.**  $OPT(j)$  does not select job  $j$   
 $OPT = \text{optimal solution to subproblem } 1, \dots, j-1$
- Recurrence:

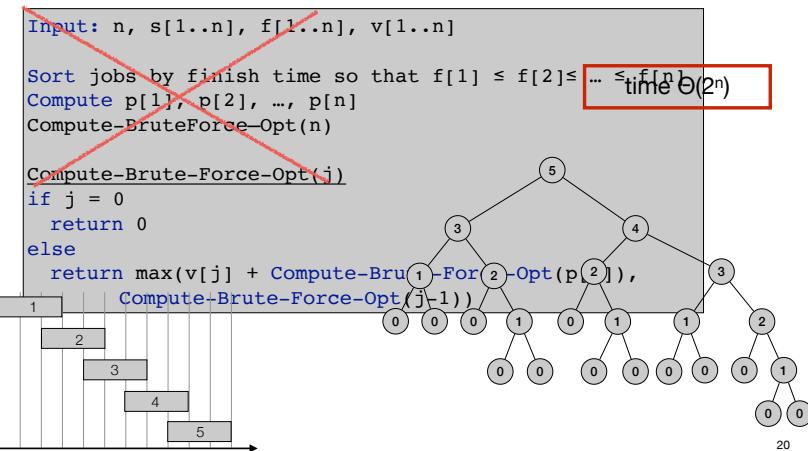
$$OPT(j) = \begin{cases} 0 & \text{if } j = 0 \\ \max\{v_j + OPT(p(j)), OPT(j-1)\} & \text{otherwise} \end{cases}$$



19

## Weighted interval scheduling: brute force

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



20

## Weighted interval scheduling: memoization

```

Input: n, s[1..n], f[1..n], v[1..n]

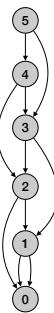
Sort jobs by finish time so that f[1] ≤ f[2] ≤ ... ≤ f[n]
Compute p[1], p[2], ..., p[n]

for j=1 to n
    M[j] = null
M[0] = 0.
Compute-Memoized-Opt(n)

Compute-Memoized-Opt(j)
if M[j] is empty
    M[j] = max(v[j] + Compute-Memoized-Opt(p[j]),
                Compute-Memoized-Opt(j-1))
return M[j]

```

- Running time  $O(n \log n)$ :
  - Sorting takes  $O(n \log n)$  time.
  - Computing  $p(n)$ :  $O(n \log n)$  - use  $\log n$  time to find each  $p(i)$ .
  - Each subproblem solved once.
  - Time to solve a subproblem constant.
- Space  $O(n)$



21

## Weighted interval scheduling: memoization

```

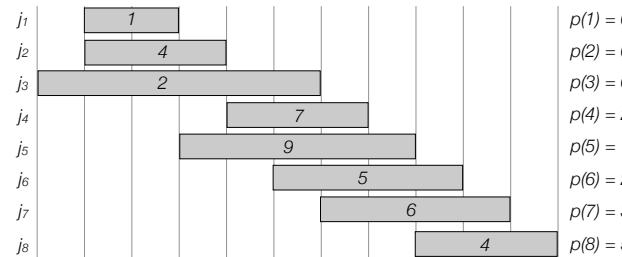
Input: n, s[1..n], f[1..n], v[1..n]

Sort jobs by finish time so that f[1] ≤ f[2] ≤ ... ≤ f[n]
Compute p[1], p[2], ..., p[n]

for j=1 to n
    M[j] = empty
M[0] = 0.
Compute-Memoized-Opt(n)

Compute-Memoized-Opt(j)
if M[j] is empty
    M[j] = max(v[j] + Compute-Memoized-Opt(p[j]),
                Compute-Memoized-Opt(j-1))
return M[j]

```



22

## Weighted interval scheduling: bottom-up

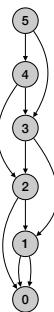
```

Compute-Bottom-Up-Opt(n, s[1..n], f[1..n], v[1..n])

Sort jobs by finish time so that f[1] ≤ f[2] ≤ ... ≤ f[n]
Compute p[1], p[2], ..., p[n]

M[0] = 0.
for j=1 to n
    M[j] = max(v[j] + M(p[j]), M(j-1))
return M[n]

```



23

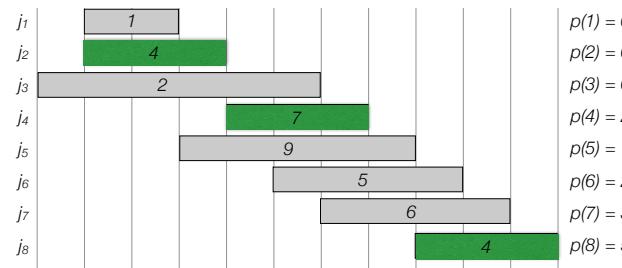
## Weighted interval scheduling: find solution

```

Find-Solution(j)
if j=0
    Return emptyset
else if M[j] > M[j-1]
    return {j} ∪ Find-Solution(p[j])
else
    return Find-Solution(j-1)

```

Solution = 8, 4, 2



24