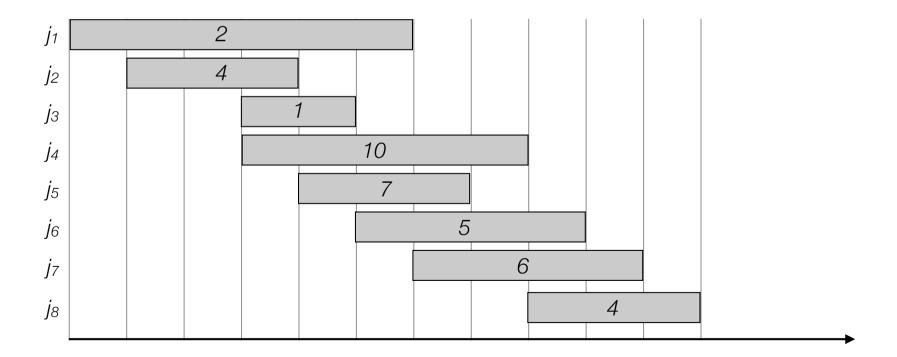
Dynamic Programming

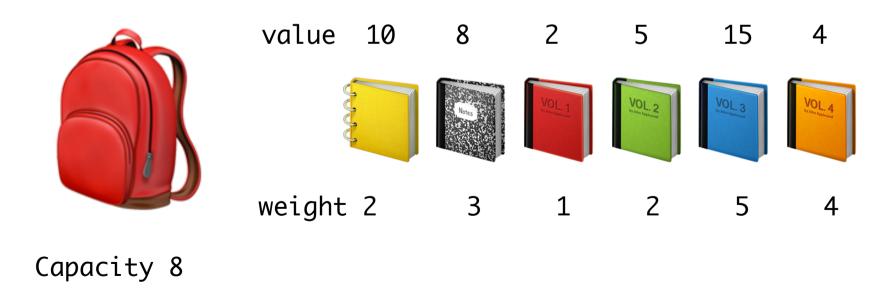
Algorithm Design 6.1, 6.2, 6.4

• In class (today and next time)

- 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



- 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.



- 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?

Dynamic Programming

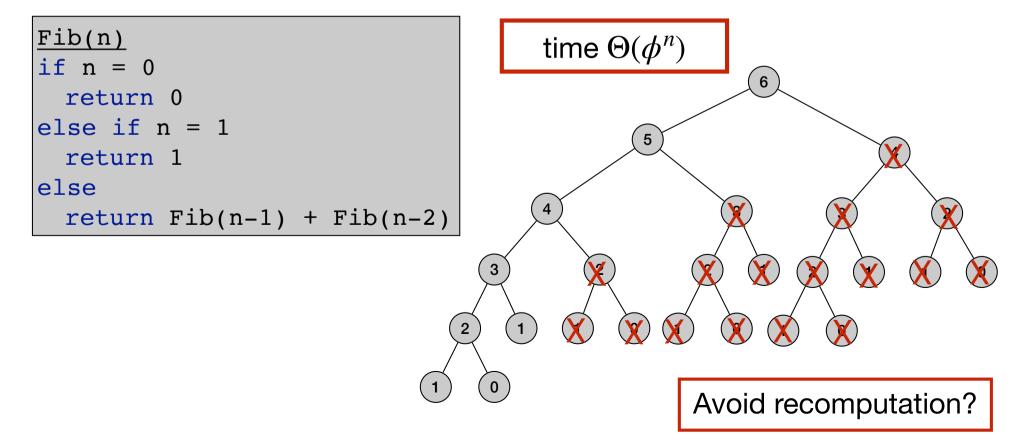
- Greedy. Build solution incrementally, optimizing some local criterion.
- Divide-and-conquer. Break up problem into independent subproblems, solve each subproblem, and combine to get solution to original problem.
- Dynamic programming. Break up problem into overlapping subproblems, and build up solutions 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.

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:



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]
```

time $\Theta(n)$

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 $\Theta(n)$

Bottom-up Fibonacci numbers - save space

Fibonacci numbers:

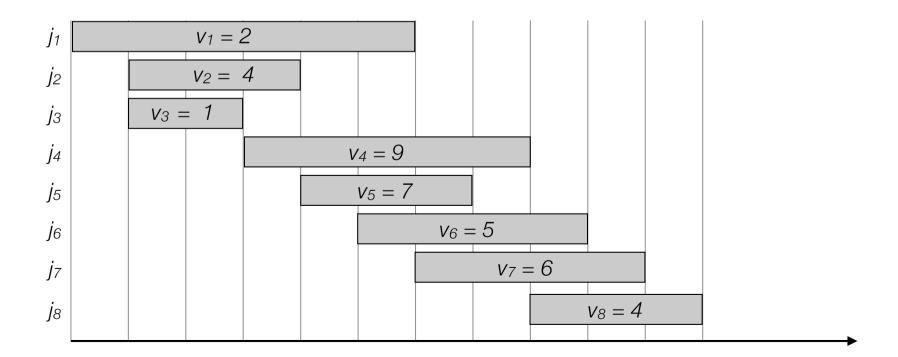
$$F_n = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ F_{n-1} + F_{n-2} & \text{otherwise} \end{cases}$$

Remember last two computed values:

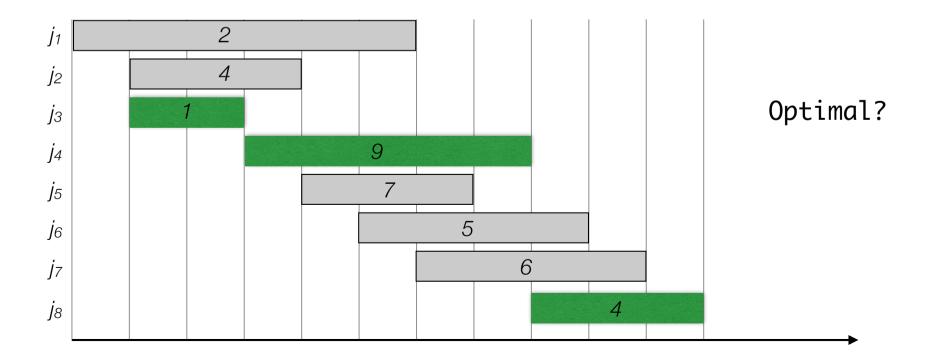
```
Iter-Fib(n)
previous = 0
current = 1
for i = 1 to n
  next = previous + current
  previous = current
  current = next
return current
```

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

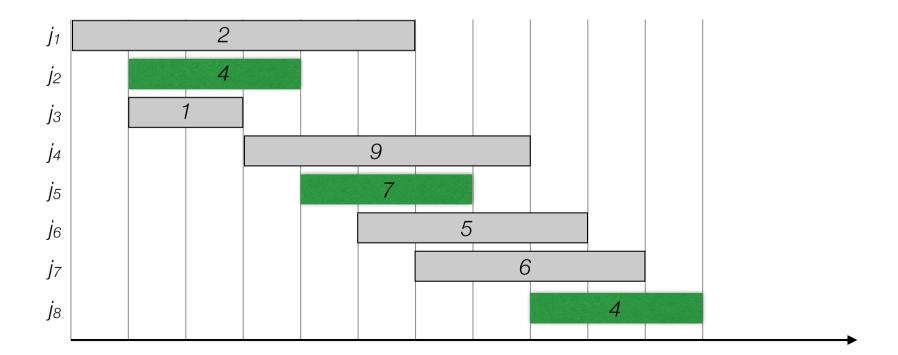
- Weighted interval scheduling problem
 - n jobs (intervals)
 - Job *i* starts at s_i , finishes at f_i and has weight/value v_i .
 - Goal: Find maximum weight subset of non-overlapping (compatible) jobs.



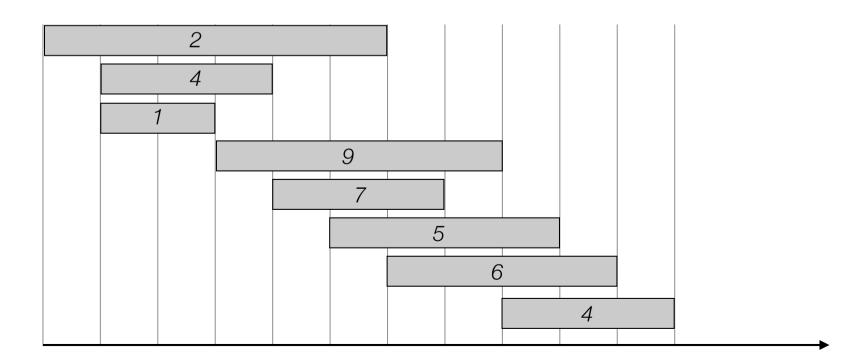
- Weighted interval scheduling problem
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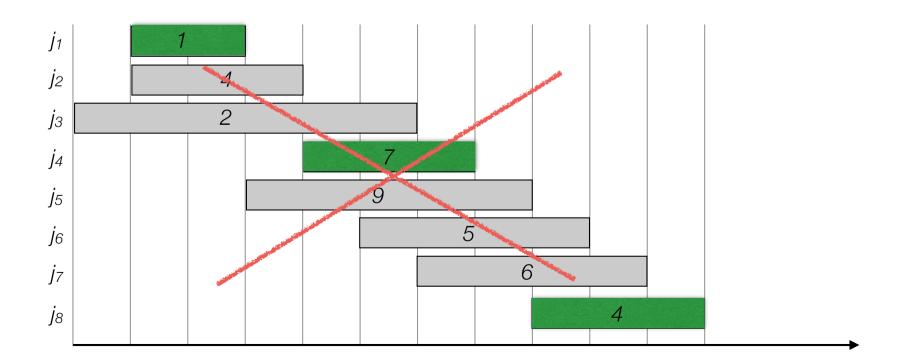
- Weighted interval scheduling problem
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• Label/sort jobs by finishing time: $f_1 \le f_2 \le ... \le f_n$



- Label/sort jobs by finishing time: $f_1 \le f_2 \le ... \le f_n$
- Greedy?

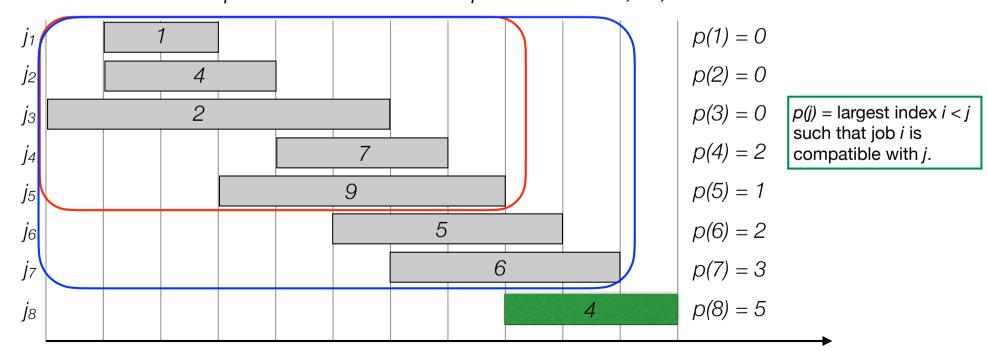


- Label/sort jobs by finishing time: $f_1 \le f_2 \le ... \le f_n$
- Optimal solution OPT:
 - Case 1. OPT selects last job

 $OPT = v_n + optimal solution to subproblem on the subset of jobs ending before job n starts$

Case 2. OPT does not select last job

 $OPT = optimal \ solution \ to \ subproblem \ on \ 1,...,n-1$

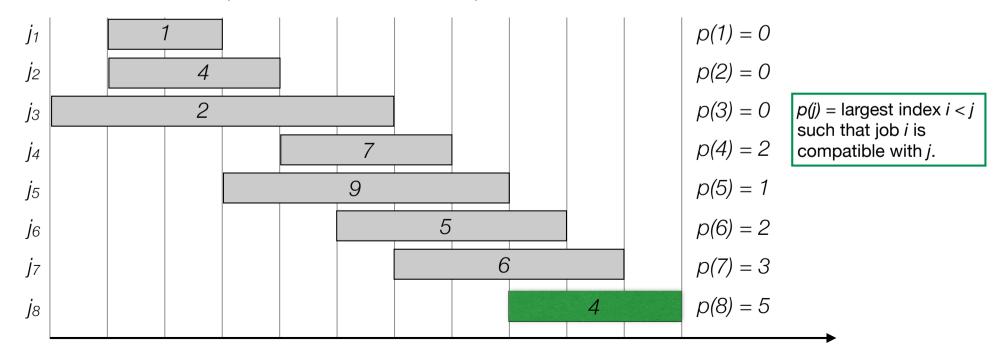


- Label/sort jobs by finishing time: $f_1 \le f_2 \le ... \le f_n$
- Optimal solution OPT:
 - Case 1. OPT selects last job

$$OPT = v_n + optimal solution to subproblem on 1,...,p(n)$$

Case 2. OPT does not select last job

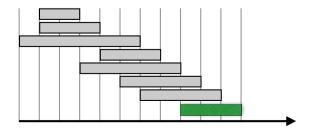
 $OPT = optimal \ solution \ to \ subproblem \ on \ 1, ..., n-1$



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

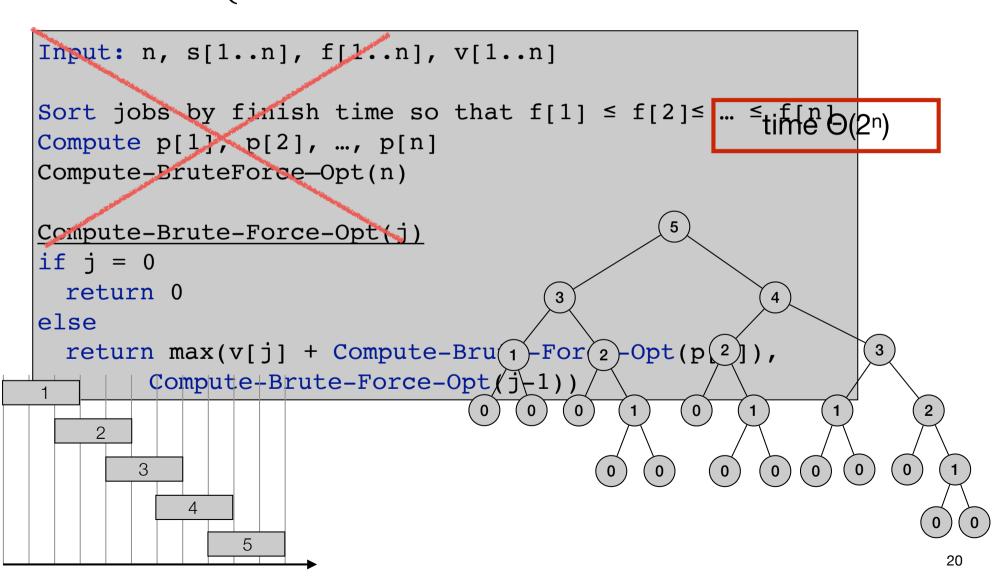
Recurrence:

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



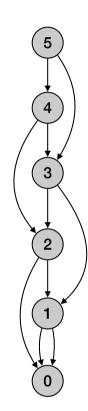
Weighted interval scheduling: brute force

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



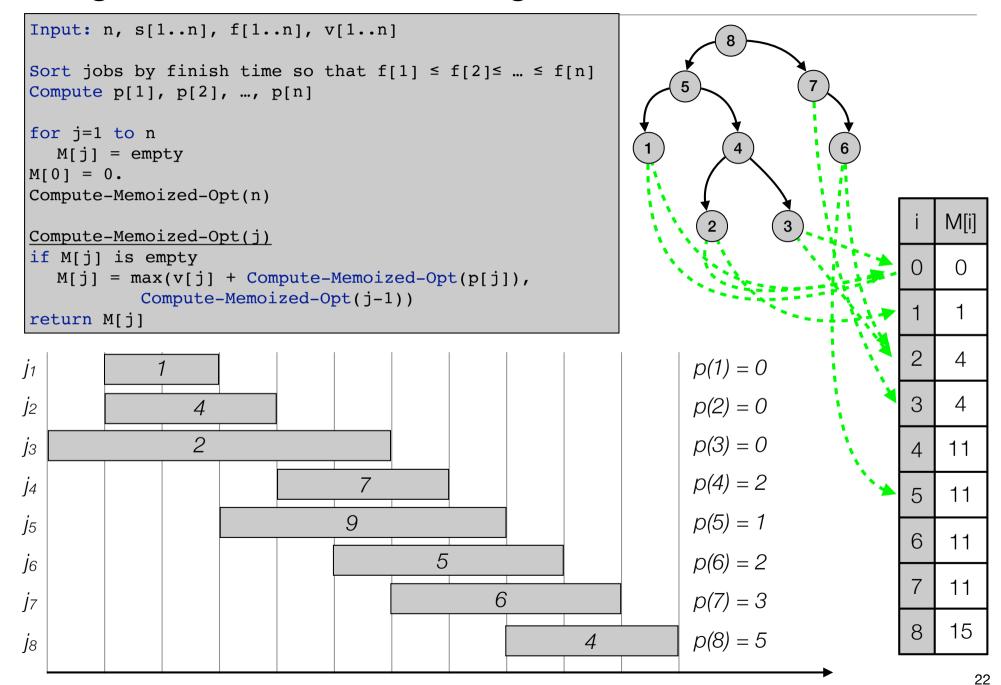
Weighted interval scheduling: memoization

```
Input: n, s[1..n], f[1..n], v[1..n]
Sort jobs by finish time so that f[1] \le f[2] \le ... \le f[n]
Compute p[1], p[2], ..., p[n]
for j=1 to n
 M[i] = 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)

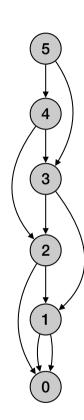
Weighted interval scheduling: memoization



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]
```

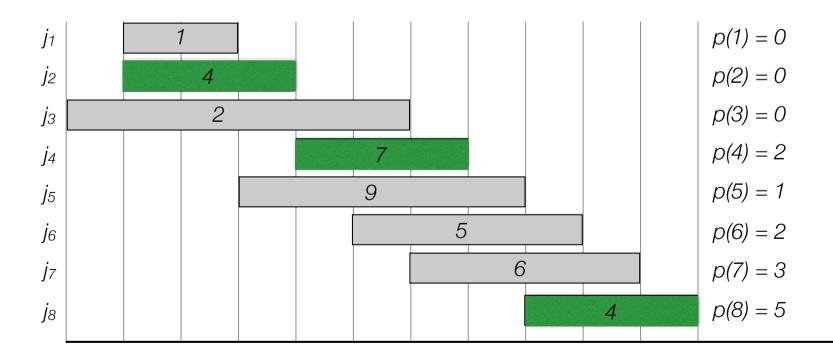


- Running time O(n log n):
 - Sorting takes O(n log n) time.
 - Computing p(n): O(n log n)
 - For loop: O(n) time
 - Each iteration takes constant time.
- Space O(n)

Weighted interval scheduling: find solution

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

Solution = 8, 4, 2



i	M[i]
0	0
1	1
2	4
3	4
4	11
5	11
6	11
7	11
8	15