### CS 218 Design and Analysis of Algorithms

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Module 1: Basics of algorithms

Computing Fibonacci numbers

Input: Given  $1^k$ 

Output: *k*th Fibonacci number.

Recurrence f(k) = f(k-1) + f(k-2).

```
Fib(k)
```

- 1: **if** k = 0 **then**
- 2: return 0
- 3: end if
- 4: **if** k = 1 **then**
- 5: return 1
- 6: else if then
- 7: return Fib(k-1) + Fib(k-2).
- 8: end if

Running time:  $T(k) = T(k-1) + T(k-2) + 1 = ? = O(2^k)$ .

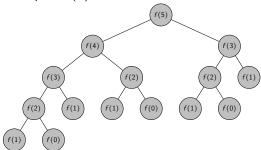
Computing Fibonacci numbers

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Example: f(5).



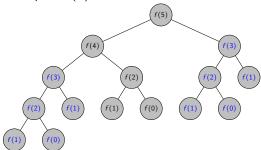
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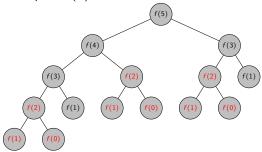
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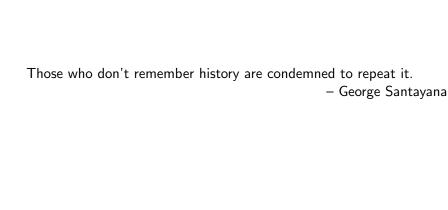
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Recurrence f(k) = f(k-1) + f(k-2).

Example: f(5).



We are recomputing too much!



- 1: **if** Table contains f(k) **then**
- Return the table value.
- 3. end if
- 4: **if** k = 0 or k = 1 **then**
- val  $\leftarrow k$
- 6. else
- $val \leftarrow Fib(k-1) + Fib(k-2)$
- Store val as the kth entry in the table
- return val
- 10: end if f(5)



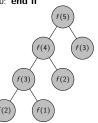
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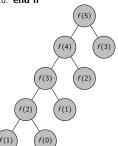
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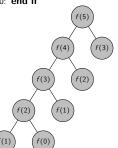
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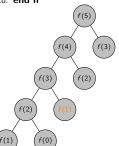
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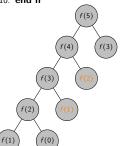
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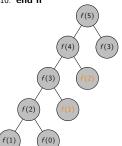
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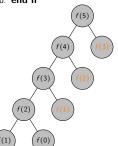
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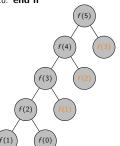
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### Running time analysis

- 1: **if** Table contains f(k) **then**
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The table gets filled from smaller values to larger values.

The recursion for each index i is evaluated at most once.

Look up into the table takes O(1) time.

Overall running time O(k).

# Weighted Interval Scheduling

#### Problem Description

Input: For each  $i \in [n]$ , start and finish times (s(i), f(i))

and a weight for each interval  $w(i) \in \mathbb{N}$ 

Output: Maximum weight non-conflicting jobs

Seen this problem in Problem Sheet 1.

Does not have good greedy heuristic.

What about a recursion-like algorithm for it?

Can we design that?

# Weighted Interval Scheduling

### Problem Description

Input: For each  $i \in [n]$ , start and finish times (s(i), f(i))

and a weight for each interval  $w(i) \in \mathbb{N}$ 

Output: Maximum weight non-conflicting jobs

#### A possible recursive strategy

Let us sort the jobs based on their finish time.  $f(1) \le f(2) \dots \le f(n)$ .

Let p(i) be j if j is the last job (in the above ordering) that is non-conflicting with i.

If the nth job belongs to OPT

for all i > p(n), ith job does not belong to OPT.

Recurse on  $\{1, \ldots, p(n)\}$  jobs.

If the *n*th job does not belong to OPT

recurse on  $\{1, \ldots, n-1\}$  jobs.

# Weighted Interval Scheduling

#### Problem Description

```
Input: For each i \in [n], start and finish times (s(i), f(i)) and a weight for each interval w(i) \in \mathbb{N}
```

Output: Maximum weight non-conflicting jobs

Assume that we have sorted the jobs in non-decreasing order of their finish times.

```
We have also computed p(i) for each i \in [n]. WtIntSc(i)

if i = 0 then
return i

else if Table(i) is non-empty then
return Table(i)

else

Table(i) \leftarrow \max\{w(i) + \text{WtIntSc}(p(i)), \text{WtIntSc}(i-1)\}
end if
```

### Correctness of the recursive algorithm

#### Lemma

For every  $i \in [n]$ , WtIntSc(i) computes the the optimal solution for the sub-problem  $\{1, ..., i\}$ , where the jobs are ordered according to their finish times.

Let  $\mathrm{Opt}(i)$  denote the value of the optimal solution for the sub-problem  $\{0,\ldots,i\}$ .

Trivially  $\mathrm{Opt}(0) = 0$ . Assume that the induction hypothesis holds for  $\forall j < i$ .

#### Proof

Due to this, we know that WtIntSc(p(i)) = Opt(p(i)) and WtIntSc(i-1) = Opt(i-1).

From our previous reasoning, the correct answer for the subproblem  $\{1,\ldots,i\}$  is max of w(i) + Opt(p(i)) and Opt(i-1).

Therefore the algorithm computes it correctly.

# Running time analysis

Running time analysis of the dynamic programming algorithm.

The sorting of the jobs according to their finish time takes time  $O(n \log n)$ .

Assuming writing the values into the table (array) and reading from it takes time O(1)the time taken by the algorithm is O(the number of calls made to WtIntSc) = O(n).

So the overall time taken is  $O(n \log n)$ .