



# Algorithms Theory

## 14 – Dynamic Programming (1)

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# Outline

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- General approach, differences to a recursive solution
- Basic example: Computation of the Fibonacci numbers

# Method of dynamic programming

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**Recursive approach:** Solve a problem by solving several smaller analogous subproblems of the same type. Then combine these solutions to generate a solution to the original problem.

**Drawback:** Repeated computation of solutions

**Dynamic-programming method:** Once a subproblem has been solved, store its solution in a table so that it can be retrieved later by simple table lookup.

## Example: Fibonacci numbers

$$f(0) = 0$$

$$f(1) = 1$$

$$f(n) = f(n-1) + f(n-2), \text{ for } n \geq 2$$

Remark:

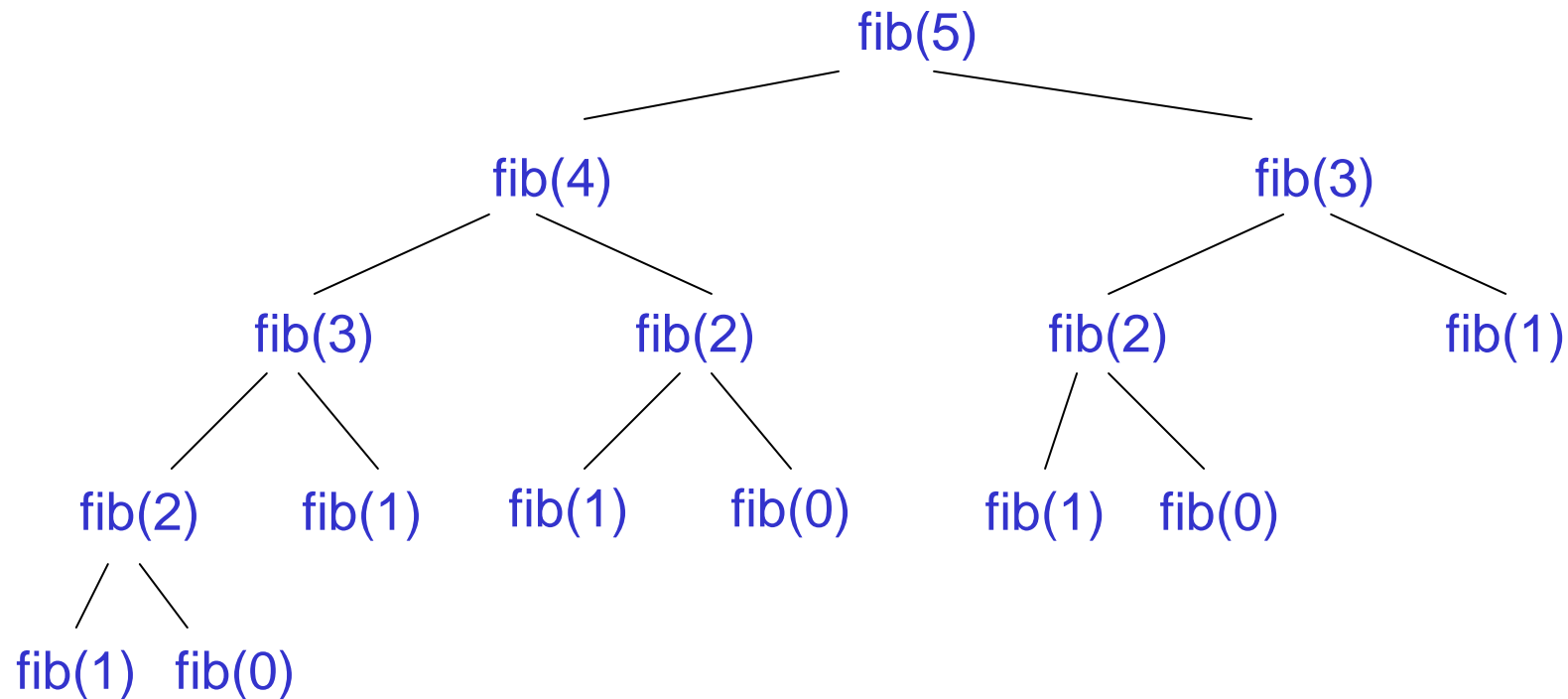
$$f(n) = \left[ \frac{1}{\sqrt{5}} (1.618\dots)^n \right]$$

**Straightforward implementation:**

```
procedure fib (n : integer) : integer  
if (n == 0) or (n == 1)  
  then return n  
  else return fib(n - 1) + fib(n - 2)
```

# Example: Fibonacci numbers

Recursion tree:



**Repeated computation!**

$$T(n) \approx \left[ \left( 1 + \frac{1}{\sqrt{5}} \right) \left( \frac{\sqrt{5} + 1}{2} \right)^n - 1 \right] \approx [1.447 \times 1.618^n - 1]$$

# Dynamic programming

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## Approach:

1. Recursively define problem  $P$ .
2. Determine a set  $T$  consisting of all subproblems that have to be solved during the computation of a solution to  $P$ .
3. Find an order  $T_0, \dots, T_k$  of the subproblems in  $T$  such that during the computation of a solution to  $T_i$  only subproblems  $T_j$  with  $j < i$  arise.
4. Solve  $T_0, \dots, T_k$  in this order and store the solutions.

## Example: Fibonacci numbers

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1. Recursive definition of the Fibonacci numbers, based on the standard equation.
2.  $T = \{ f(0), \dots, f(n-1) \}$
3.  $T_i = f(i), \quad i = 0, \dots, n - 1$
4. Computation of  $fib(i)$ , for  $i \geq 2$ , only requires the results of the last two subproblems  $fib(i - 1)$  and  $fib(i - 2)$ .

# Example: Fibonacci numbers

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Computation by dynamic programming, version 1:

**procedure** *fib*( $n : integer$ ) : *integer*

1  $f_0 := 0; f_1 := 1$

2 **for**  $k := 2$  **to**  $n$  **do**

3      $f_k := f_{k-1} + f_{k-2}$

4 **return**  $f_n$



# Example: Fibonacci numbers

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Computation by dynamic programming, version 2:

```
procedure fib (n : integer) : integer  
1   $f_{secondlast} := 0; f_{last} := 1$   
2  for  $k := 2$  to  $n$  do  
3       $f_{current} := f_{last} + f_{secondlast}$   
4       $f_{secondlast} := f_{last}$   
5       $f_{last} := f_{current}$   
6  if  $n \leq 1$  then return  $n$  else return  $f_{current}$  ;
```

Linear running time, constant space requirement!

# Computation of the Fibonacci numbers using memoization



Compute each number exactly once, store it in an array  $F[0\dots n]$ :

```
procedure fib (n : integer) : integer
1   $F[0] := 0$ ;  $F[1] := 1$ ;
2  for  $i := 2$  to  $n$  do
3       $F[i] := \infty$ ;
4  return lookupfib( $n$ )
```

The procedure *lookupfib* is defined as follows:

```
procedure lookupfib(k : integer) : integer
1  if  $F[k] < \infty$ 
2      then return  $F[k]$ 
3      else  $F[k] := \text{lookupfib}(k - 1) + \text{lookupfib}(k - 2)$ ;
4      return  $F[k]$ 
```