

Recursive algorithms

Separate and conquer

Lecture 06.01
by Marina Barsky

Recap: recursion

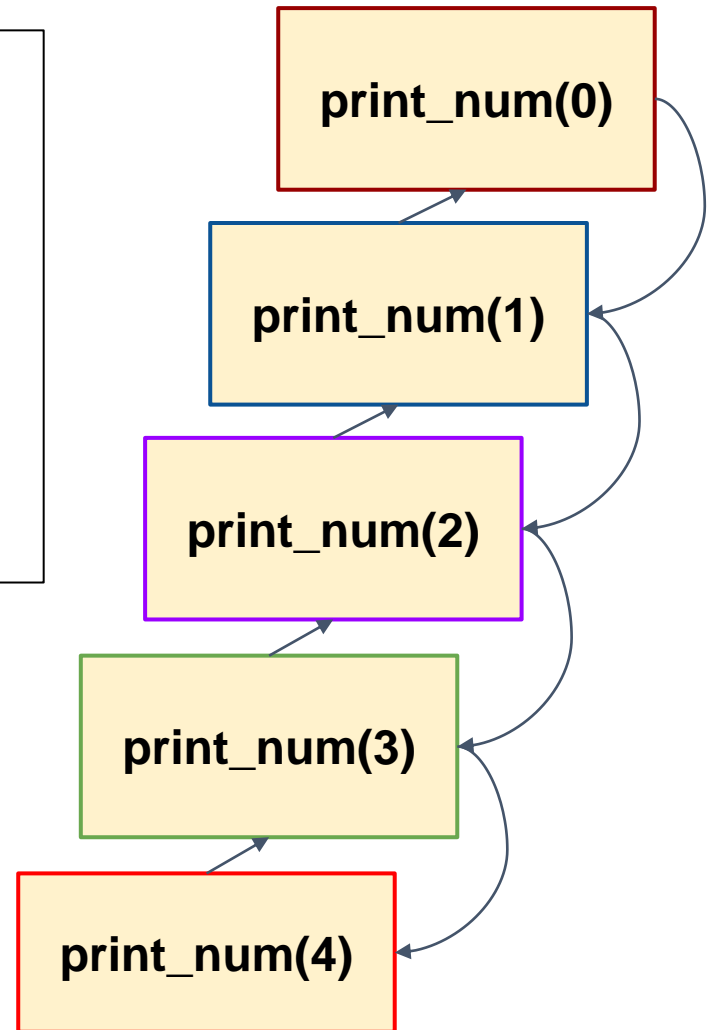
```
algorithm print_num(count) :  
    if count < 1:  
        return  
    print(count)  
    print_num(count-1)  
  
print_num(4)
```

What is printed?

Recap: recursion

```
algorithm print_num(count) :  
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  print(count)  
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What is printed?



Recap: recursion

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    print_num(count-1)  
    print(count)  
  
print_num(4)
```

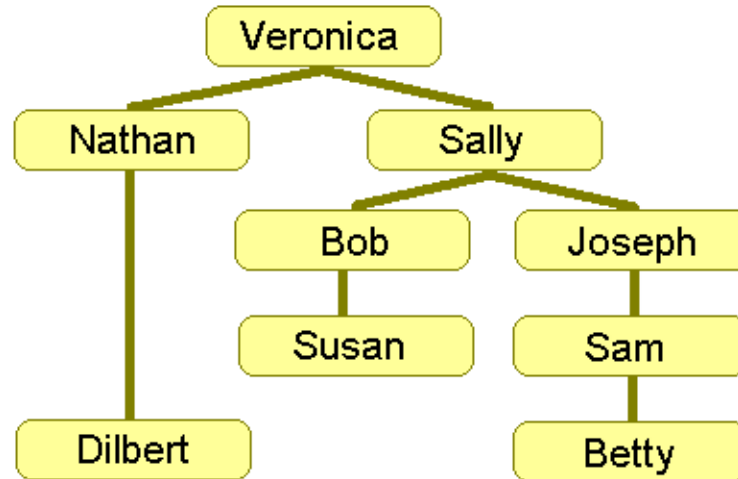
What is printed now?

Solve problems recursively: when?

1. The structure is defined recursively
2. The problem is defined recursively

1. Recursively defined structure: tree

Name	Manager
Betty	Sam
Bob	Sally
Dilbert	Nathan
Joseph	Sally
Nathan	Veronica
Sally	Veronica
Sam	Joseph
Susan	Bob
Veronica	



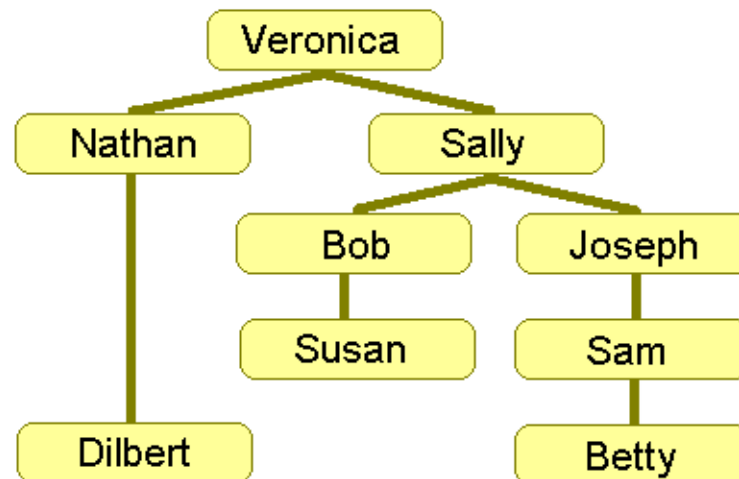
Employee names are stored as pairs of names in list *A*

They represent a hierarchical organizational structure: *X* reports to *Y*

Count Under Problem

Input: list A of pairs (x,y) meaning x reports directly to y , and an employee name S .

Output: count of all employees who report to S (directly or indirectly)



Designing solution

Name	Manager
Betty	Sam
Bob	Sally
Dilbert	Nathan
Joseph	Sally
Nathan	Veronica
Sally	Veronica
Sam	Joseph
Susan	Bob
Veronica	

- We want to find all people who work under Sally
- We can iterate over list A and count the pairs where Sally is a manager
 - But then we also need to iterate over A again and count and collect people working under Bob and Joseph
 - Then all people who work under Sam
- It seems that we need to have several **nested loops** - **but how many levels?**

The easiest solution: use ***recursion!***

Recursive solution

Algorithm *count_under*(list A of pairs, name S)

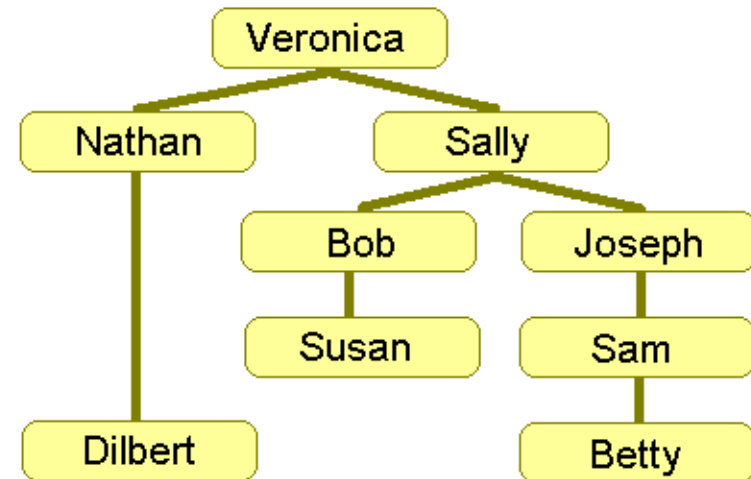
count: = 0

for each pair (name, manager) in A:

if manager = S:

 count: += 1 + *count_under*(name)

return count



2. Recursively defined problem: factorial

Definition

$factorial(1) = 1$

$factorial(n) = n * factorial(n-1)$

$$F_n = n * F_{n-1}$$

Problem: compute factorial

Input: n

Output: $factorial(n)$

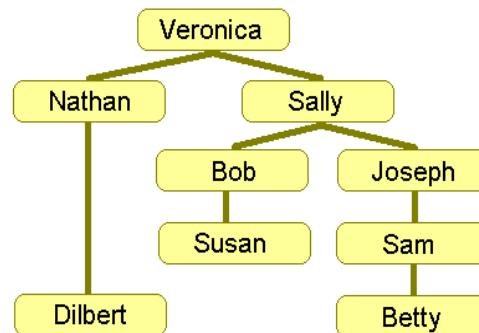
Algorithm $factorial(n)$

if $n = 1$ return 1

return $n * factorial(n - 1)$

When recursion feels natural

Recursive algorithms are particularly appropriate **when the underlying problem or the data to be treated are defined in recursive terms**



$$F_n = n * F_{n-1}$$

Definition

A *recurrence relation* is an equation recursively defining a sequence of values

$$F_n = \begin{cases} 1, & \text{if } n = 0 \\ 1, & \text{if } n = 1 \\ n * F_{n-1} & \text{if } n > 1 \end{cases}$$

1, 1, 2, 6, 24, 120...

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

0, 1, 1, 2, 3, 5, 8, 13, 21, 34

Note how the function $F(n)$ is defined through $F(n-1)$

Fibonacci numbers

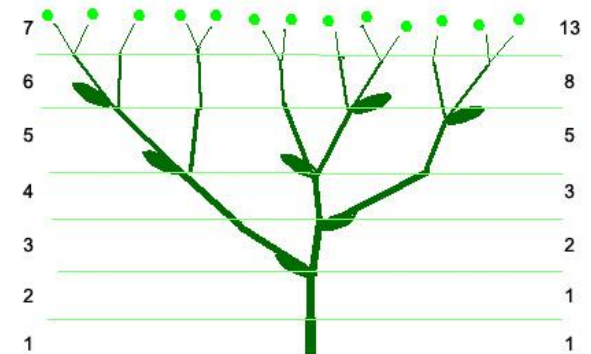
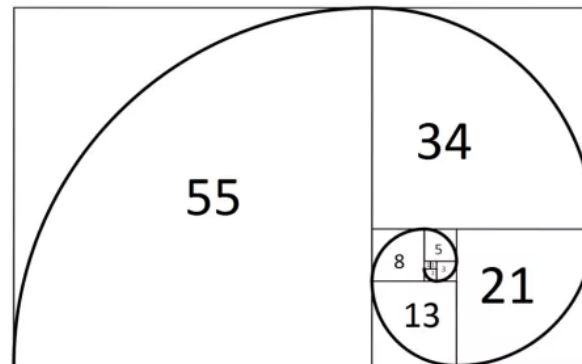
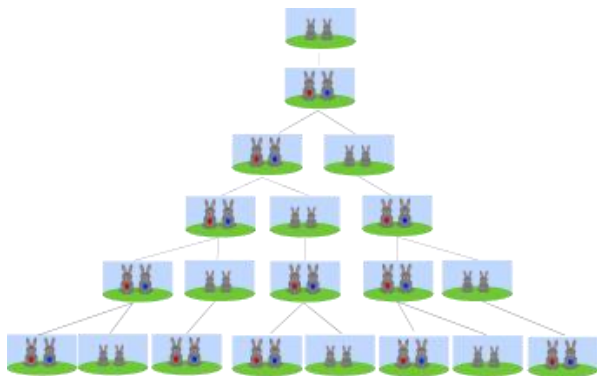
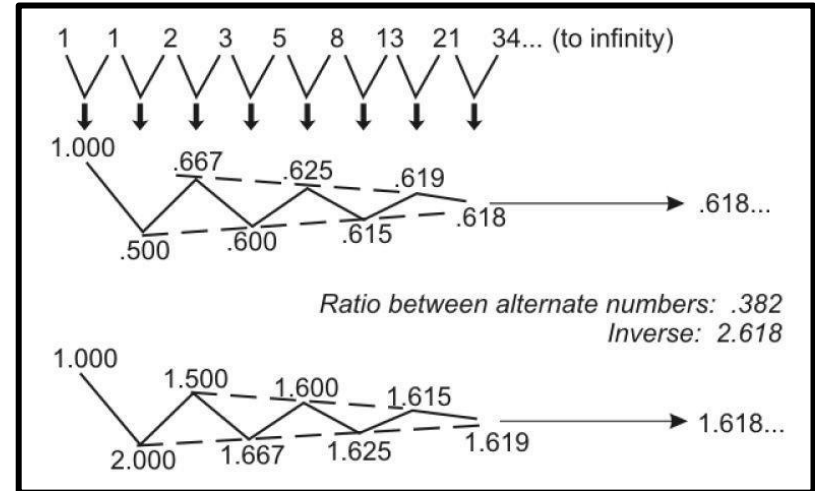
Natural recursive solution?

Fibonacci numbers

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

0, 1, 1, 2, 3, 5, 8, 13, 21, 34

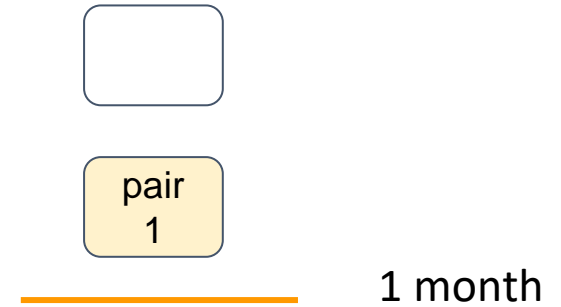
Golden ratio



Fibonacci rabbits



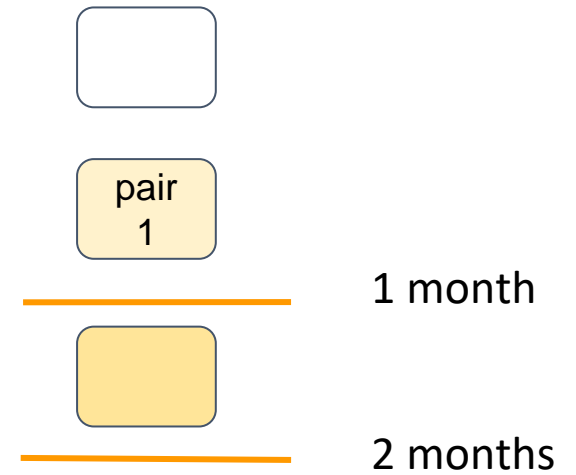
Leonardo
Fibonacci
c1175-1250



Fibonacci rabbits



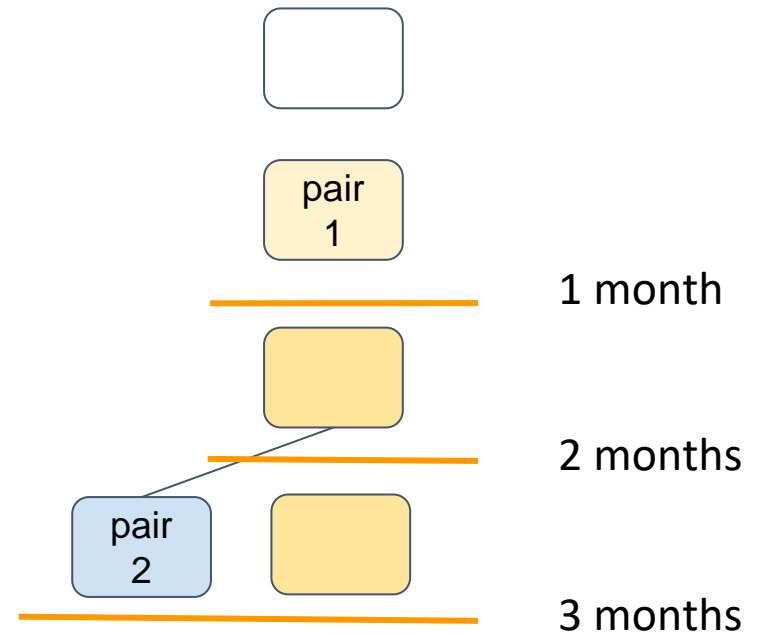
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Fibonacci rabbits



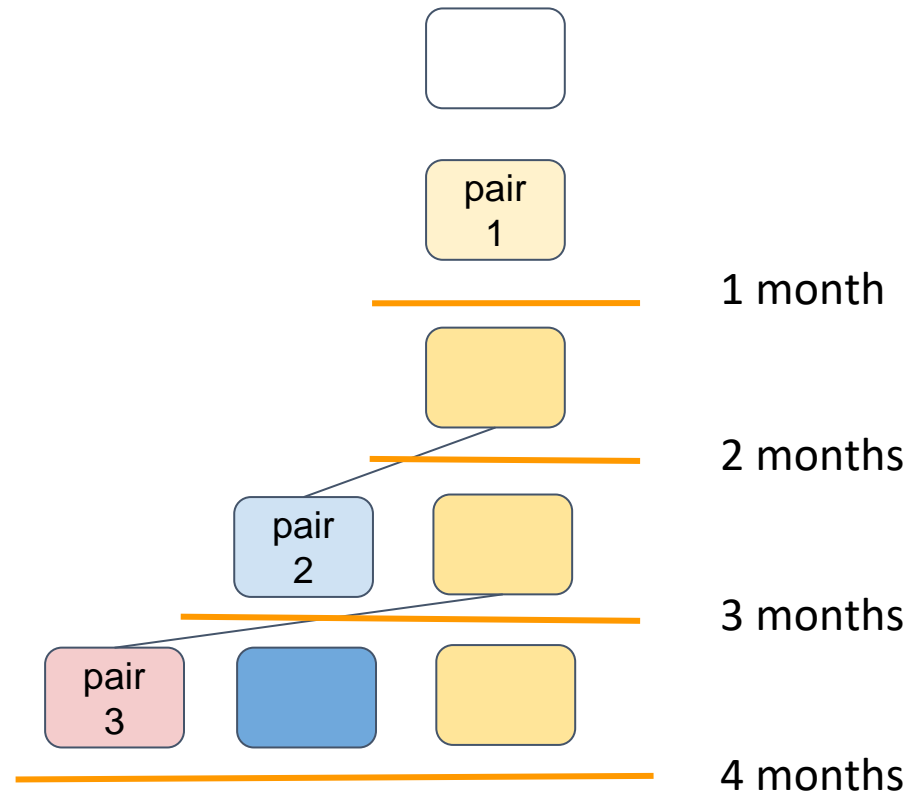
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Fibonacci rabbits



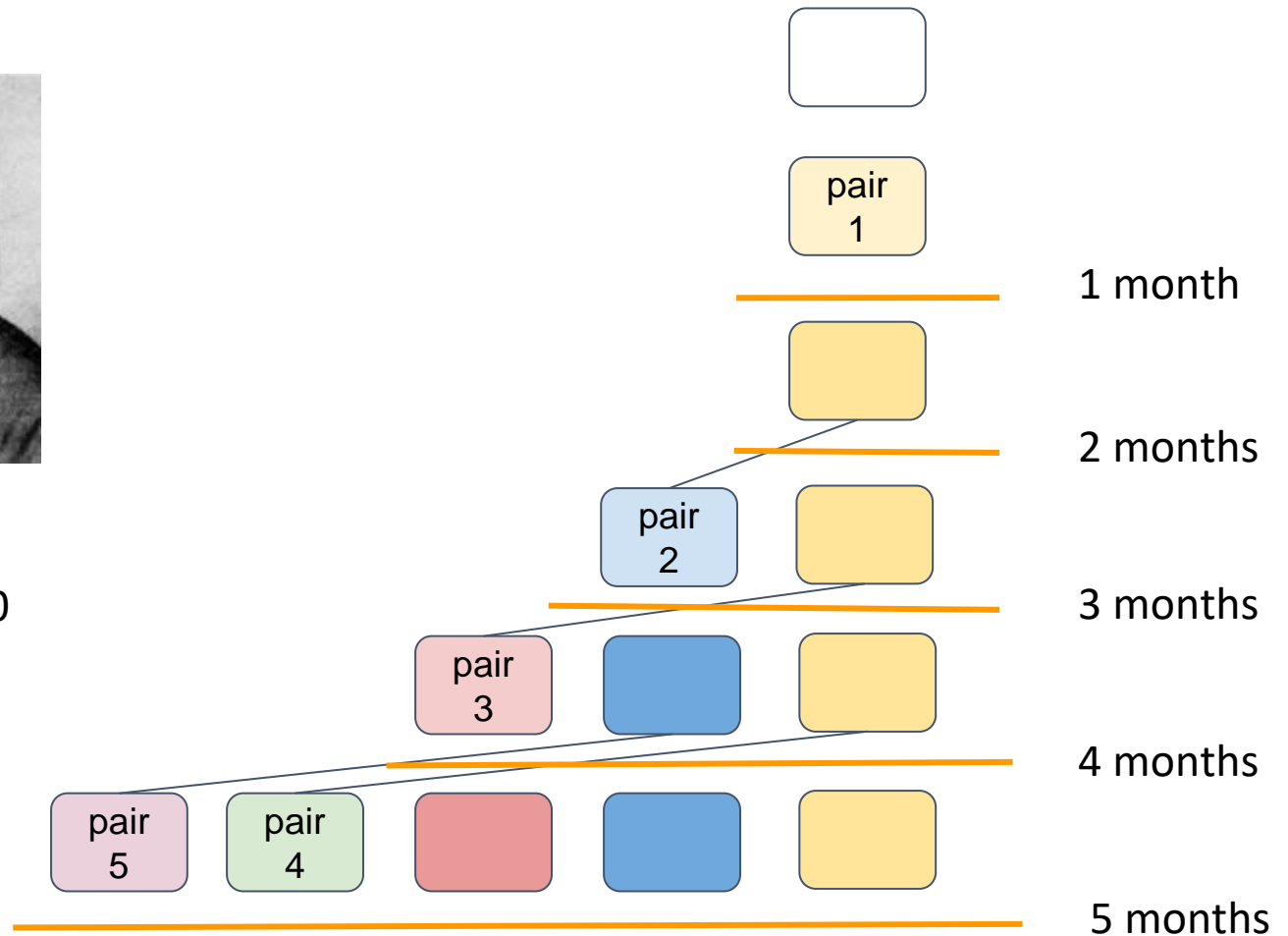
Leonardo
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Fibonacci rabbits



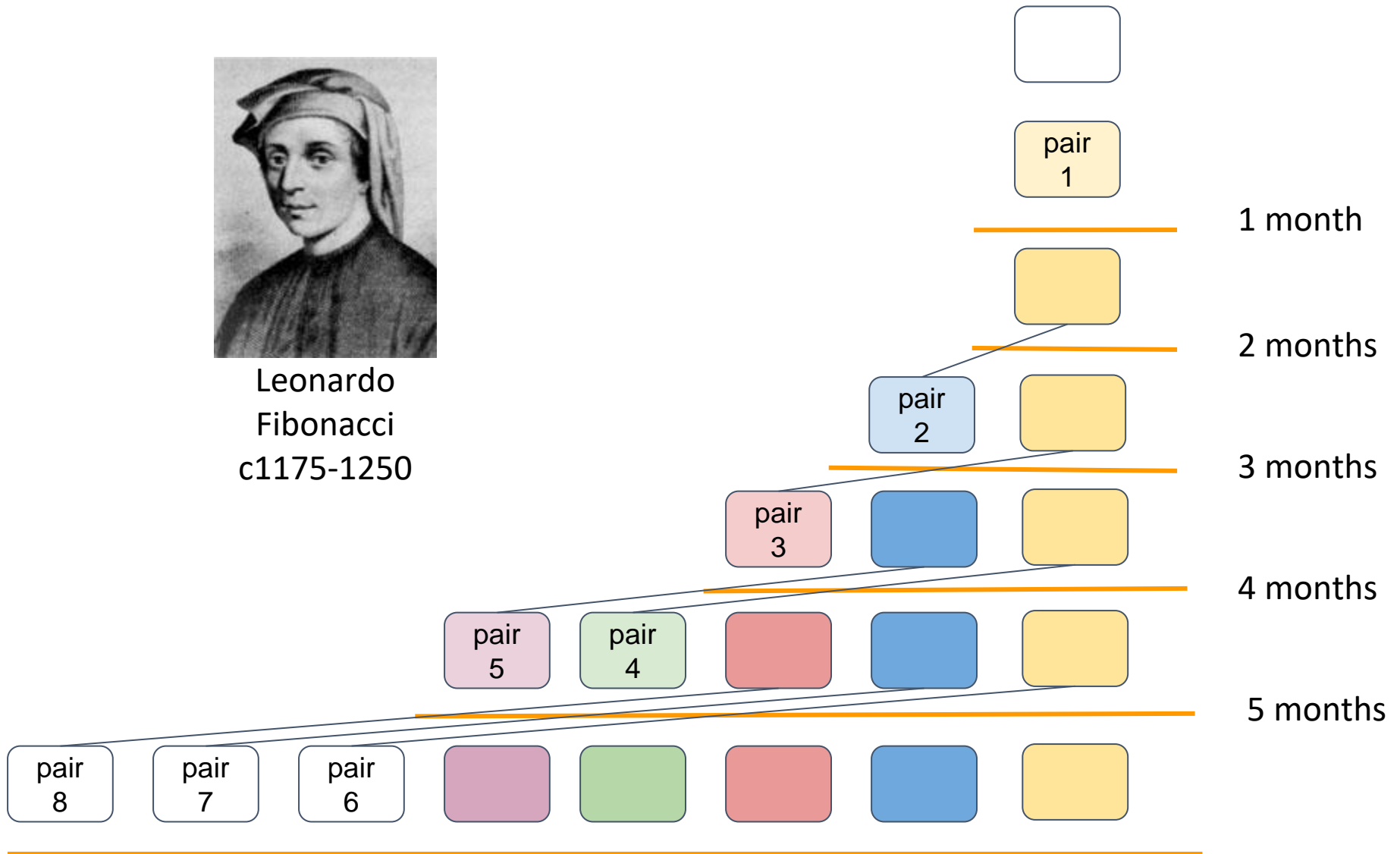
Leonardo
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c1175-1250



Fibonacci rabbits



Leonardo
Fibonacci
c1175-1250





It all started with a single pair...

University of Victoria, BC, Canada, 2010

Fibonacci numbers grow exponentially

Lemma

$$F_n \geq 2^{n/2} \text{ for } n \geq 6$$

Proof: By induction

Base case: $n = 6, 7$ (by direct computation).

Inductive step:

Assume that it is true for F_{n-1} : $F_{n-1} \geq 2^{(n-1)/2}$.

Let's show that it is true for F_n

$$F_n = F_{n-1} + F_{n-2}$$

$$\geq 2^{(n-1)/2} + 2^{(n-2)/2} \geq 2 \cdot 2^{(n-2)/2} = 2^{n/2} \quad \blacksquare$$

Theorem:

$$F_n = \frac{\phi^n - (1 - \phi)^n}{\sqrt{5}}$$

$$\phi = \frac{1 + \sqrt{5}}{2}$$

$$\phi = 1.618034\dots$$

$$F_{20} = 6765$$

$$F_{50} = 12586269025$$

$$F_{100} = 354224848179261915075$$

$$F_{500} = 1394232245616978801397243828$$

$$7040728395007025658769730726$$

$$4108962948325571622863290691$$

$$557658876222521294125$$

Recursive algorithm for computing the n -th Fibonacci number

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

Problem: Compute F_n

Input: integer $n \geq 0$

Output: F_n

Algorithm $\text{Fib_recurs}(n)$

if $n \leq 1$: return n

return $\text{Fib_recurs}(n - 1) + \text{Fib_recurs}(n - 2)$

What is the running time?

Recursive Fibonacci: running time

Algorithm `Fib_rekurs(n)`

```
if  $n \leq 1$ :  
    return  $n$   
else:  
    return Fib_rekurs( $n - 1$ ) + Fib_rekurs( $n - 2$ )
```

Let $T(n)$ denote the count of **lines of code** executed by `Fib_rekurs`(n).

$$\text{if } n \leq 1: T(n) = 2$$

$$\text{if } n \geq 2: T(n) = 3 + T(n - 1) + T(n - 2)$$

Number of operations

$$T(n) = \begin{cases} 2 & \text{if } n \leq 1 \\ 3 + T(n-1) + T(n-2) & \text{if } n > 1 \end{cases}$$

n -th Fibonacci number

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

Therefore $T(n) \geq F_n$

Recursive algorithm: running time

Let $T(n)$ denote the count of **lines of code** executed by `Fib_rekurs(n)`.

Algorithm `Fib_rekurs(n)`

if $n \leq 1$:

 return n

else:

 return `Fib_rekurs(n - 1) + Fib_rekurs(n - 2)`

$$T(n) = \begin{cases} 2 & \text{if } n \leq 1 \\ 3 + T(n-1) + T(n-2) & \text{otherwise} \end{cases}$$

$$T(n) \geq F_n$$

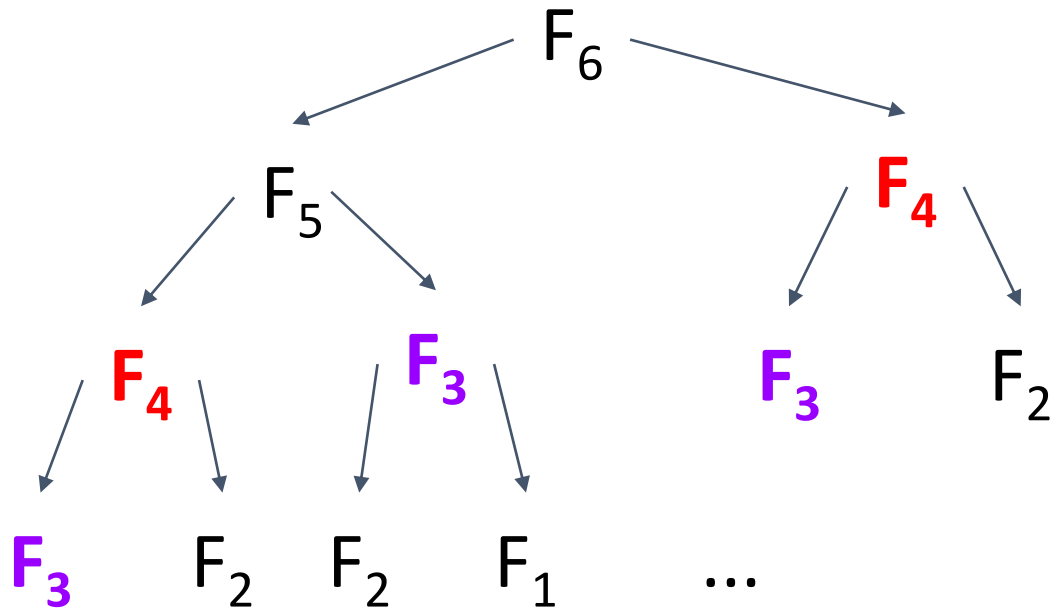
but $F_n \geq 2^{n/2}$ for $n \geq 6$!!!

Running time **$\Omega(2^n)$**

$T(100) \approx 1.77 \cdot 10^{21}$
(1.77 sextillion)

Takes 56,000 years at 1GHz

Why so slow?



Recursion tree

Note the repeating calls with the same arguments

Efficient iterative algorithm

Algorithm Fib_list(n)

create an array $F [0 \dots n]$

$F [0] \leftarrow 0$

$F [1] \leftarrow 1$

for i from 2 to n :

$F [i] \leftarrow F [i - 1] + F [i - 2]$

return $F [n]$

Running time

$$T(n) = 2n + 2$$

$$\text{So } T(100) = 202$$

Recursion or not recursion?

Recursive algorithms are particularly appropriate **when the underlying problem or the data to be treated are defined in recursive terms**

- Such recursive definitions **do not guarantee** that a recursive algorithm is the best way to solve the problem
- The use of recursive algorithms by inappropriate examples created apprehension and antipathy toward the use of recursion in programming:

recursion = inefficiency

Recursion vs. iteration

→ Recursion

- ◆ Each recursive call **requires extra space** on the stack
- ◆ If we get **infinite recursion**, the program will eventually run out of memory, cause stack overflow, and **the program will terminate**
- ◆ Solutions to some problems are **easier to formulate** recursively

→ Iteration

- ◆ Each iteration **does not require extra space**
- ◆ An **infinite loop could loop forever** since there is no extra memory being created
- ◆ Iterative **solutions** to a problem may **not** always be as **obvious** as a recursive solution

Generally, recursive solutions are less efficient than iterative solutions due to the overhead of function calls

Recursive algorithms: running time

Steps:

- ❑ Draw recursion tree
- ❑ Estimate the depth of the tree
- ❑ Estimate work done at each level of the tree
- ❑ Add all level work together

Example: recursive max

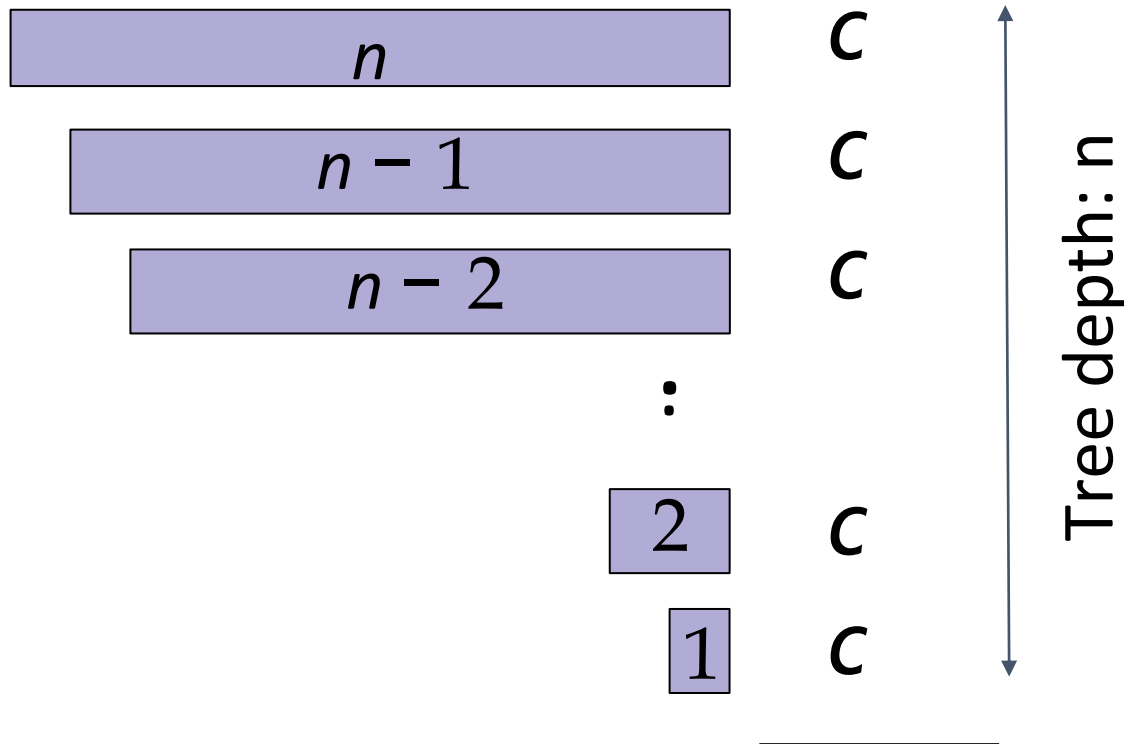
Algorithm Recurs_max (non-empty linked list A)

```
if len(A) = 1:  
    return A[1]  
else:  
    if A[1] < A[2]:  
        remove A[1] from A  
    else:  
        remove A[2] from A  
    return Recurs_max (A)
```

What is time complexity?

Recursive Max: time complexity

Work at each level



Total: $cn = O(n)$



Searching Sorted Data

Separate and conquer

<https://www.khanacademy.org/computing/computer-science/algorithms/binary-search/a/binary-search>

Warm-up: find the fake coin

- There are 8 identical-looking coins
- One of these coins is counterfeit and is known to be lighter than the genuine coins
- What is the minimum number of weighings needed to identify the fake coin with a two-pan balance scale without weights?



Problem: Searching in a sorted array

Input: A sorted array $A[low \dots high]$
($\forall low \leq i < high : A[i] \leq A[i + 1]$).
A value *key* to search for.

Output: An index, i , ($low \leq i \leq high$) where
 $A[i] = key$.
Otherwise, return -1 (NOT_FOUND).

binary_search(*A, low, high, key*)

if *high* < *low* :
 return -1

mid ← *low* + $\lfloor \frac{\textit{high}-\textit{low}}{2} \rfloor$

if *key* = *A*[*mid*]:
 return *mid*

else if *key* < *A*[*mid*]:
 return binary_search(*A, low, mid* - 1, *key*)

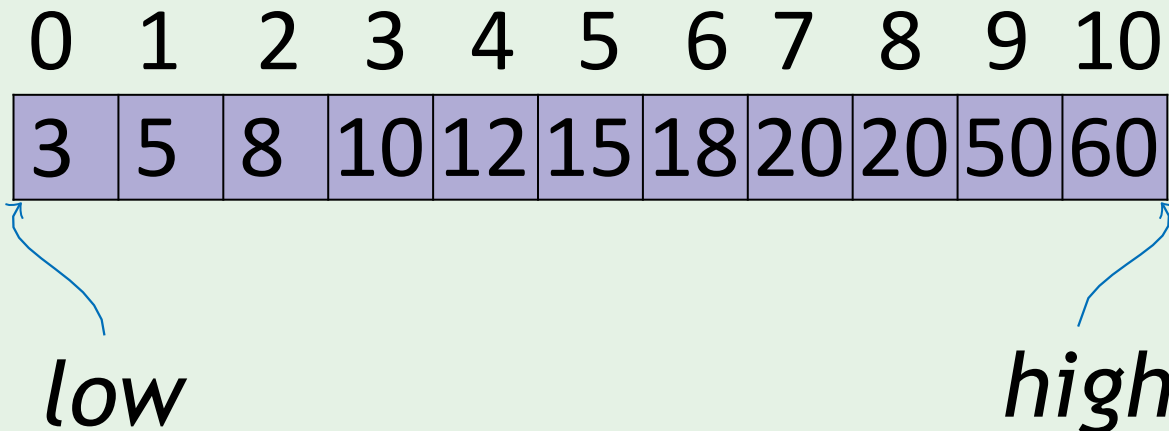
else:
 return binary_search(*A, mid* + 1, *high, key*)

Example: Searching for key 50

0	1	2	3	4	5	6	7	8	9	10
3	5	8	10	12	15	18	20	20	50	60

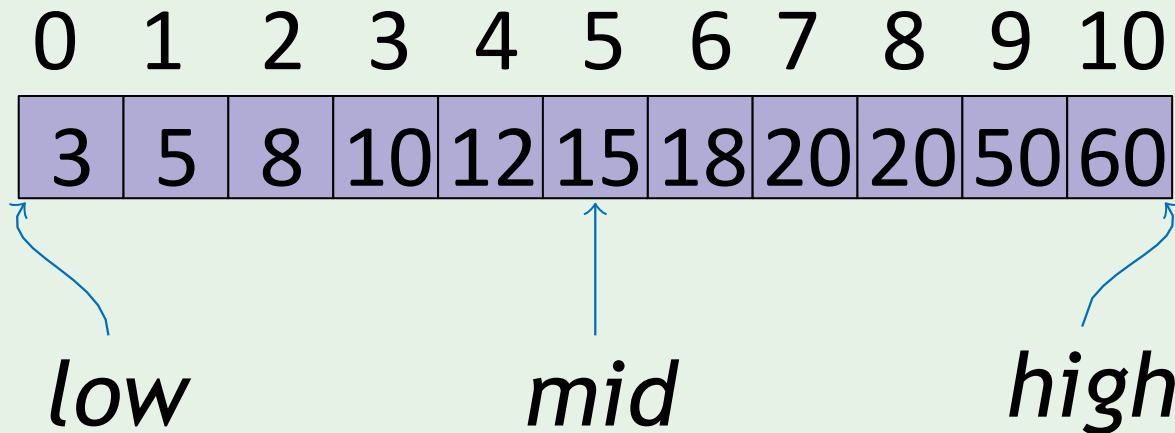
Example: Searching for key 50

`binary_search(A, 0, 10, 50)`



Example: Searching for key 50

`binary_search(A, 0, 10, 50)`



Example: Searching for key 50

`binary_search(A, 0, 10, 50)`

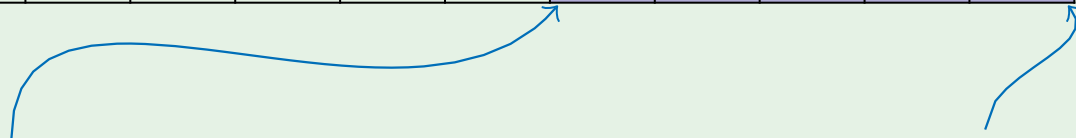
`binary_search(A, 6, 10, 50)`

0	1	2	3	4	5	6	7	8	9	10
3	5	8	10	12	15	18	20	20	50	60

low

mid

high



Example: Searching for key 50

`binary_search(A, 0, 10, 50)`

`binary_search(A, 6, 10, 50)`

0	1	2	3	4	5	6	7	8	9	10
3	5	8	10	12	15	18	20	20	50	60

low

mid

high



Example: Searching for key 50

`binary_search(A, 0, 10, 50)`

`binary_search(A, 6, 10, 50)`

`binary_search(A, 9, 10, 50)`

0	1	2	3	4	5	6	7	8	9	10
3	5	8	10	12	15	18	20	20	50	60

low

mid

high

Example: Searching for key 50

`binary_search(A, 0, 10, 50)`

`binary_search(A, 6, 10, 50)`

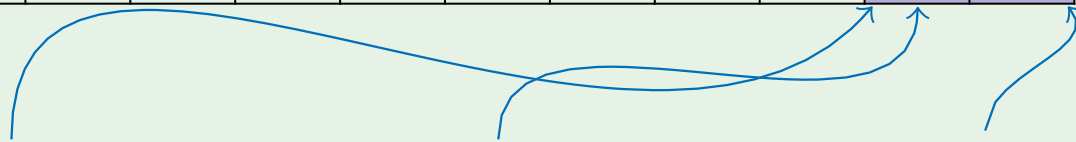
`binary_search(A, 9, 10, 50)`

0	1	2	3	4	5	6	7	8	9	10
3	5	8	10	12	15	18	20	20	50	60

low

mid

high



Example: Searching for key 50

binary_search(A, 0, 10, 50)

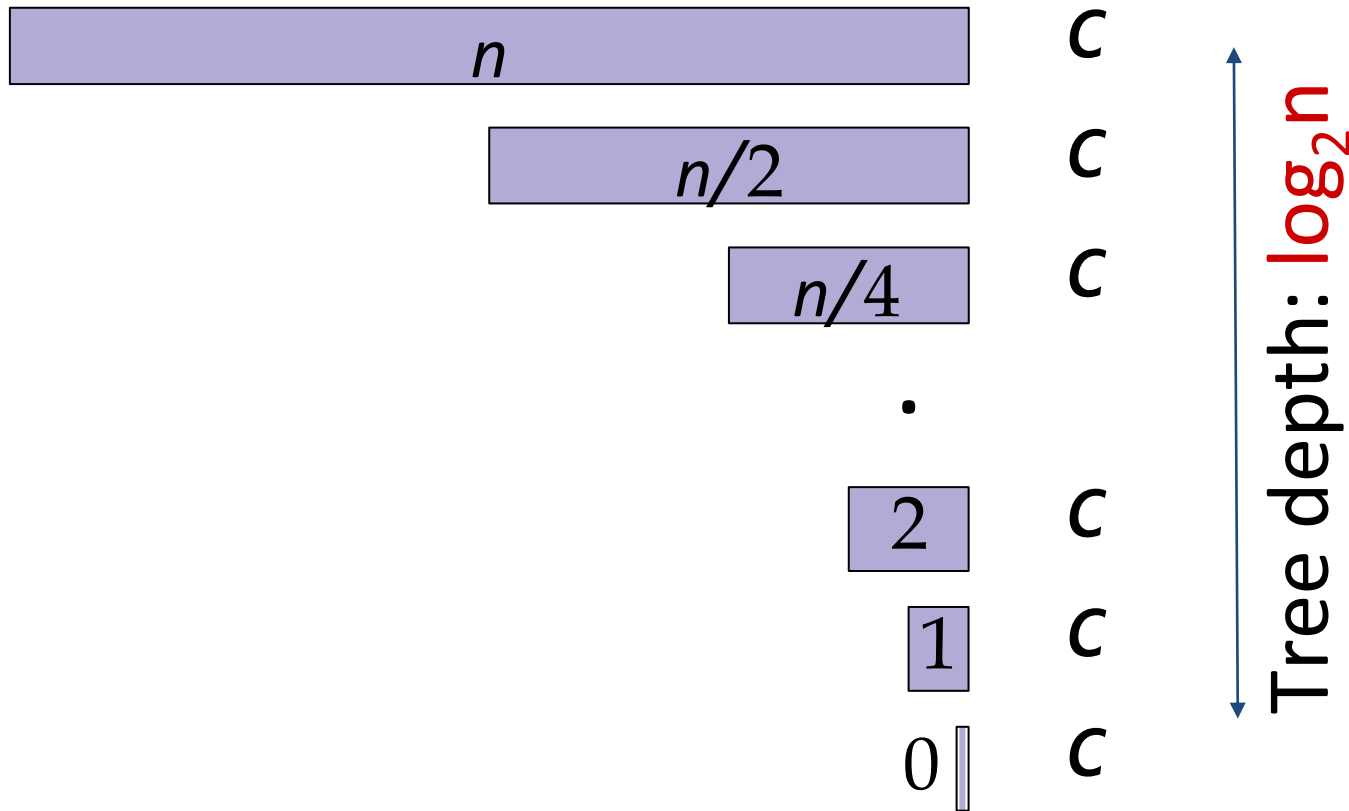
binary_search(A, 6, 10, 50)

binary_search(A, 9, 10, 50) → 9

0	1	2	3	4	5	6	7	8	9	10
3	5	8	10	12	15	18	20	20	50	60

Running time of Binary Search

Work at each level



$$\text{Total: } \sum_{i=0}^{\log_2 n} c = O(\log n)$$

Iterative Version

binary_search_it(A, low, high, key)

while *low* ≤ *high*:

mid ← *low* + $\lfloor \frac{\textit{high} - \textit{low}}{2} \rfloor$

if *key* = *A*[*mid*]:

 return *mid*

else if *key* < *A*[*mid*]:

high = *mid* - 1

else:

low = *mid* + 1

return -1

Linear search

$O(n)$

Binary search

$O(\log n)$

Calculating runtime of recursive algorithms
is not always that easy