

# **COMP108**

# **Algorithmic Foundations**

**Divide and Conquer**

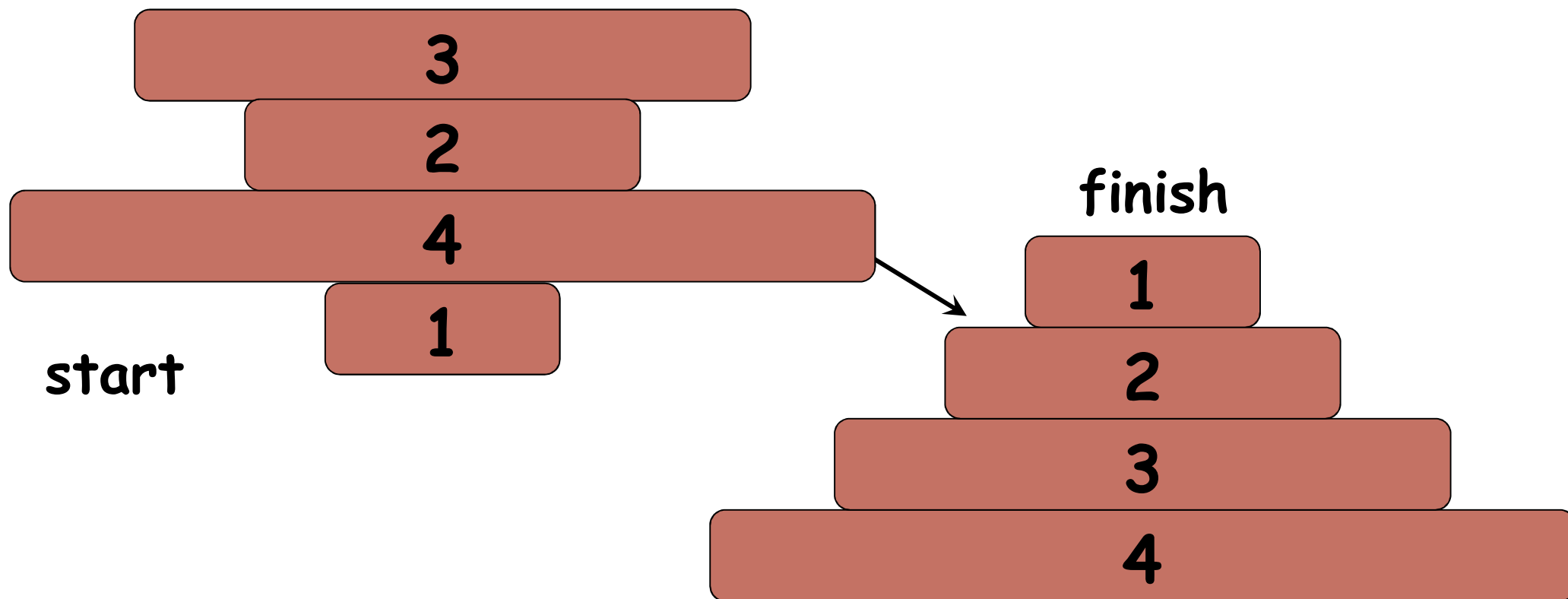
**Prudence Wong**

# Pancake Sorting

**Input:** Stack of pancakes, each of **different** sizes

**Output:** Arrange in **order of size** (smallest on top)

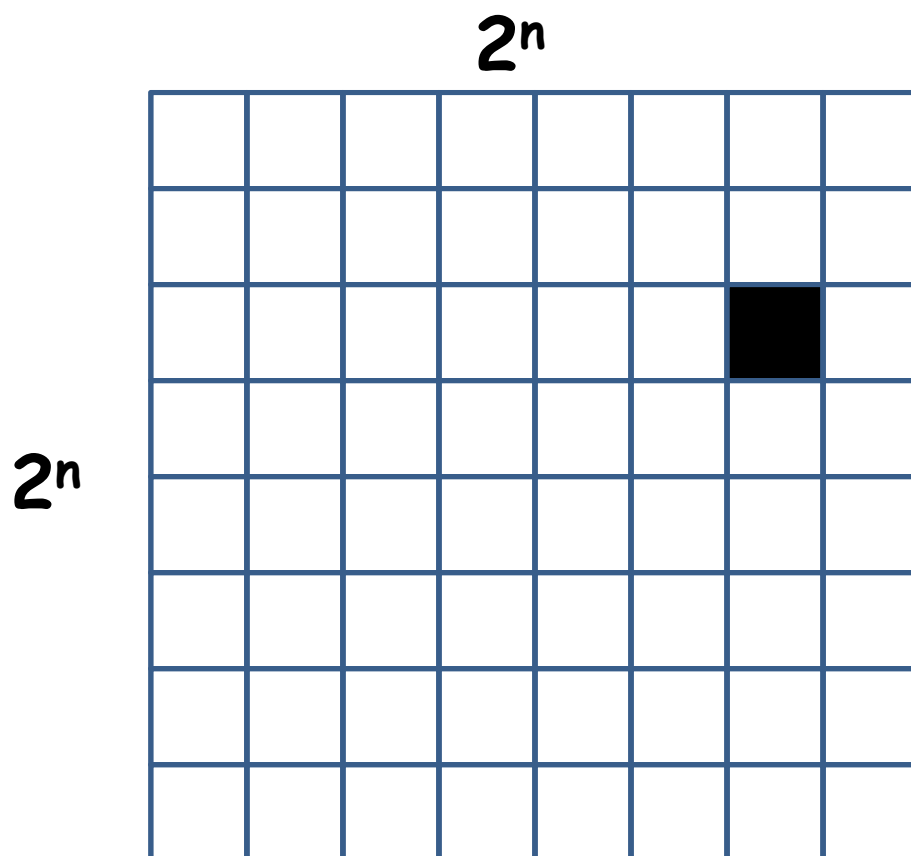
**Action:** Slip a **flipper** under one of the pancakes and **flip** over the **whole stack** above the flipper



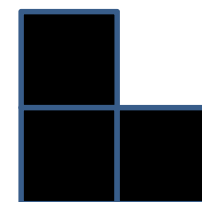
# Triomino Puzzle

**Input:**  $2^n$ -by- $2^n$  chessboard with **one** missing square & many **L-shaped** tiles of **3** adjacent squares

**Question:** **Cover** the chessboard with L-shaped tiles without overlapping



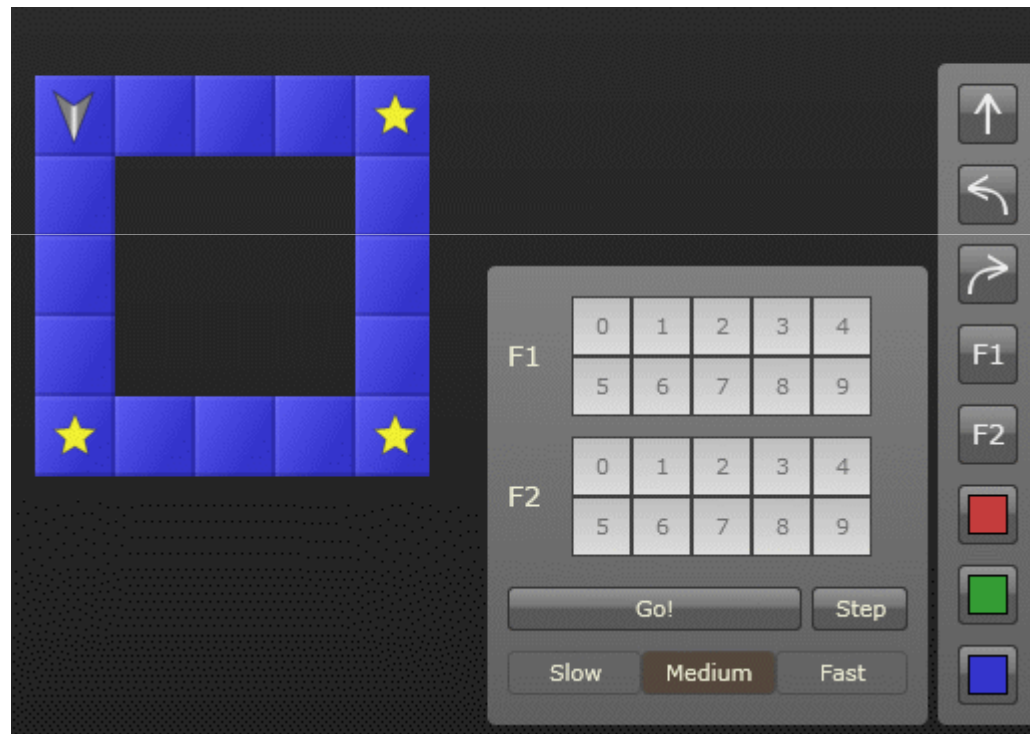
Is it do-able?



# Robozzle - Recursion

**Task:** to program a robot to pick up all stars in a certain area

**Command:** Go straight, Turn Left, Turn Right



# Divide and Conquer ...

# Learning outcomes

- Understand how divide and conquer works and able to analyse complexity of divide and conquer methods by solving recurrence
- See examples of divide and conquer methods

# Divide and Conquer

One of the **best-known** algorithm design techniques

Idea:

- A problem instance is divided into several **smaller** instances of the same problem, ideally of about same size
- The smaller instances are **solved**, typically **recursively**
- The solutions for the smaller instances are combined to get a solution to the large instance

# Merge Sort ...



# Merge sort

- using divide and conquer technique
- divide the sequence of  $n$  numbers into two halves
- **recursively** sort the two halves
- **merge** the two sorted halves into a single sorted sequence

51, 13, 10, 64, 34, 5, 32, 21

we want to sort these 8 numbers,  
**divide** them into two halves

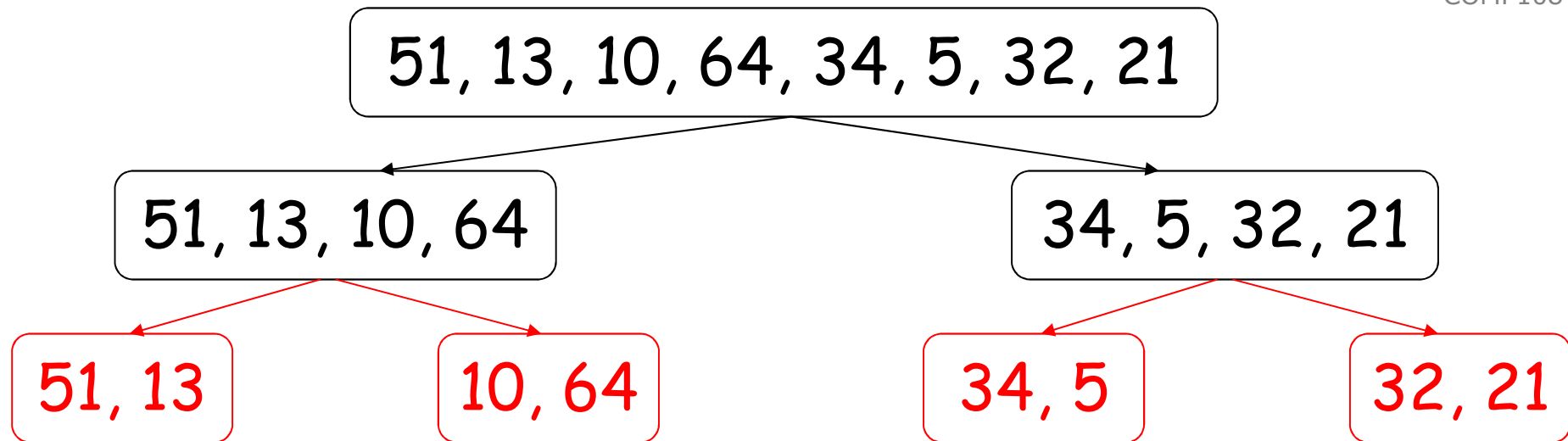
51, 13, 10, 64, 34, 5, 32, 21

51, 13, 10, 64

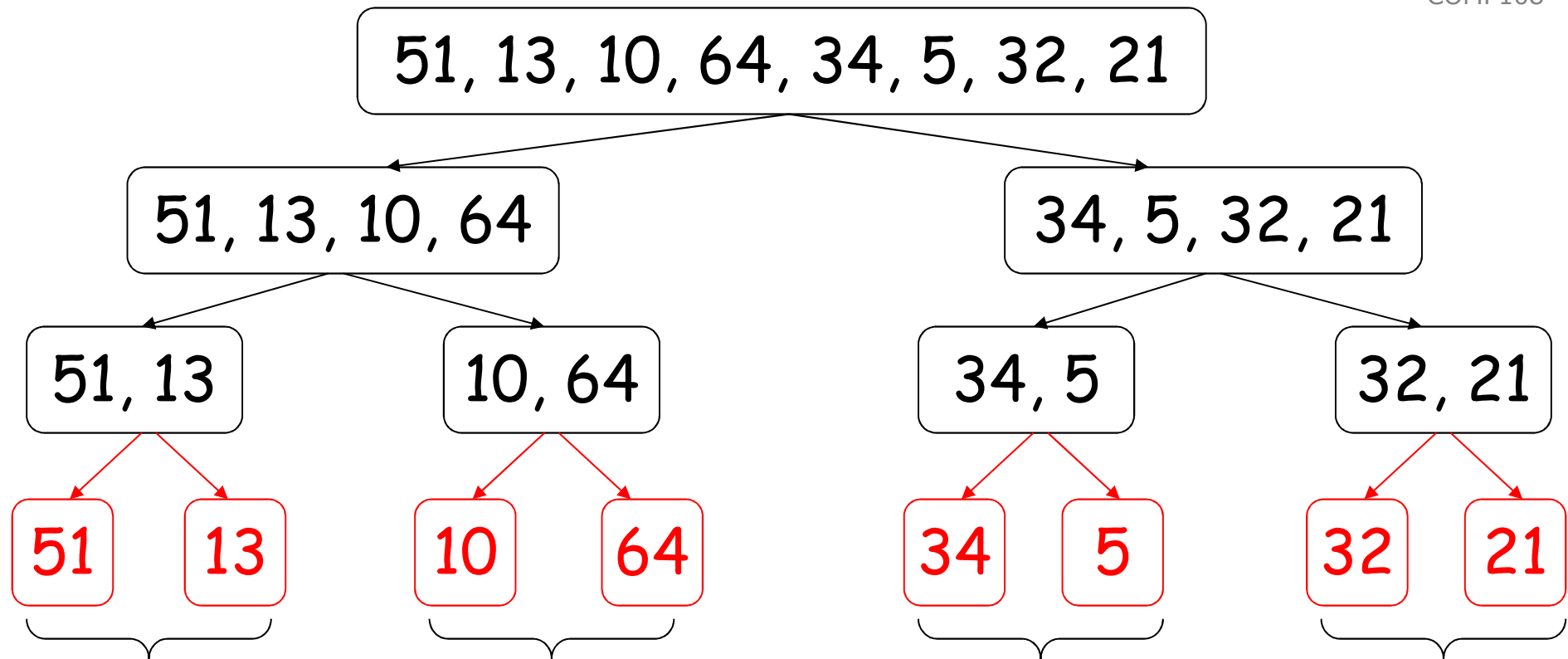
34, 5, 32, 21

**divide** these 4  
numbers into  
halves

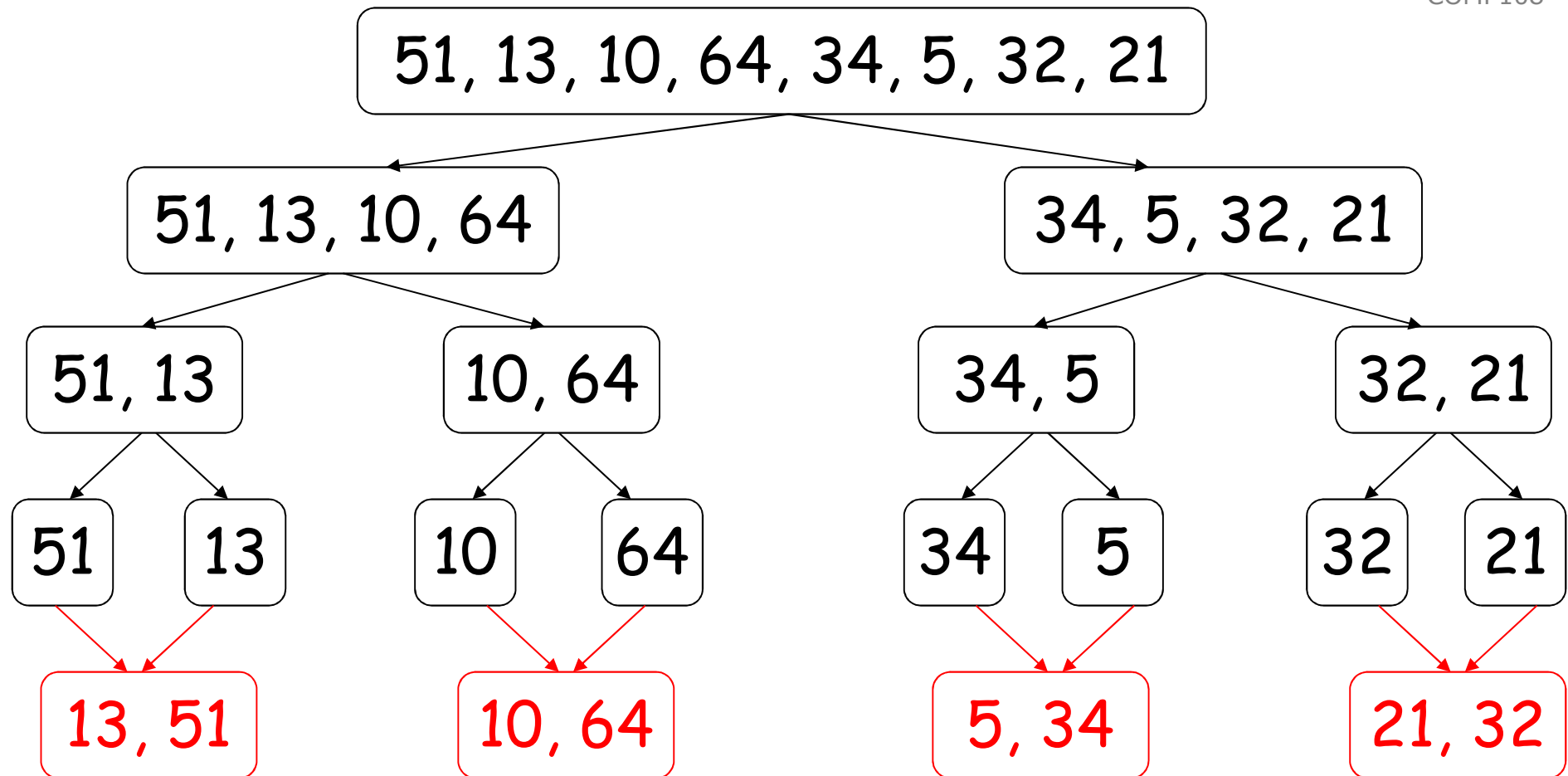
similarly for  
these 4



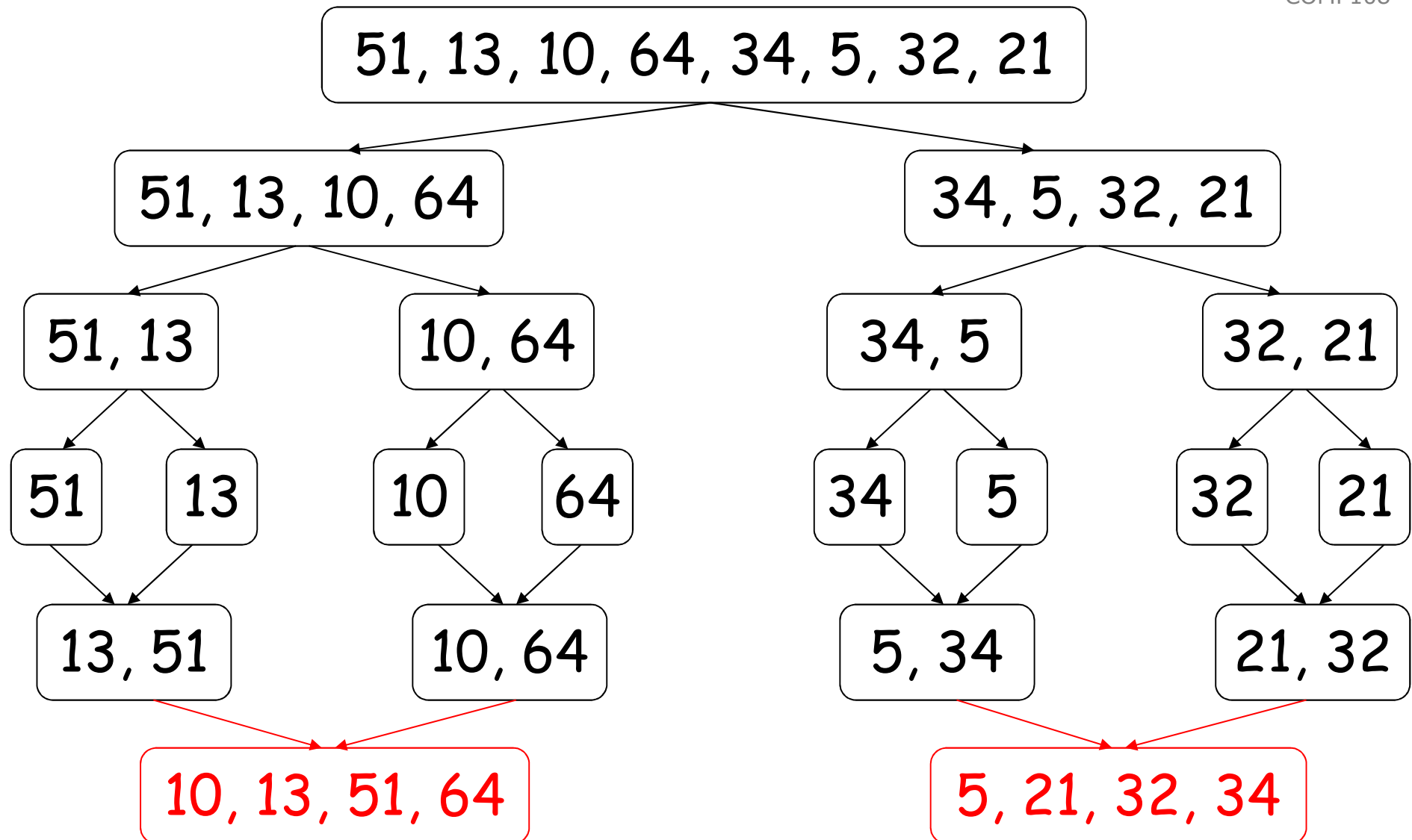
further divide each shorter sequence ...  
until we get sequence with only **1** number



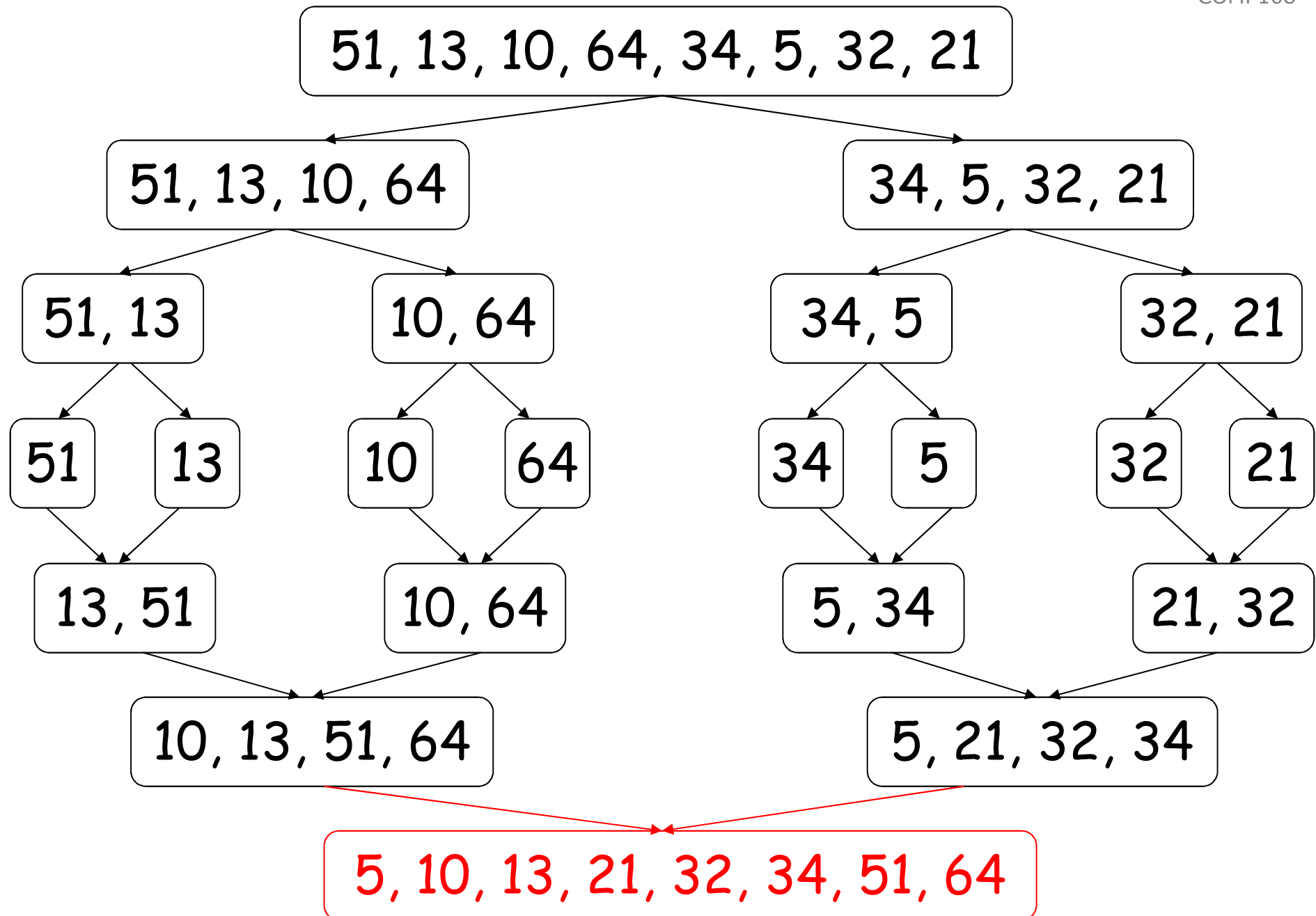
**merge** pairs of  
single number into  
a sequence of 2  
sorted numbers



then **merge** again into sequences of  
4 sorted numbers



one more merge give the **final** sorted sequence





# Summary

## Divide

- dividing a sequence of  $n$  numbers into **two** smaller sequences is straightforward

## Conquer

- merging two sorted sequences of **total length  $n$**  can also be done easily, at most  **$n-1$**  comparisons

10, 13, 51, 64

5, 21, 32, 34

Result:

To merge two sorted sequences,  
we keep two **pointers**, one to each sequence

Compare the two numbers pointed,  
copy the **smaller** one to the result  
and **advance** the corresponding pointer

10, 13, 51, 64

5, 21, 32, 34

Result: 5,

Then compare again the two numbers  
pointed to by the pointer;  
copy the smaller one to the result  
and advance that pointer

10, 13, 51, 64

5, 21, 32, 34

Result: 5, 10,

Repeat the same process ...

10, 13, 51, 64

5, 21, 32, 34

Result: 5, 10, 13

Again ...

10, 13, 51, 64

5, 21, 32, 34

Result: 5, 10, 13, **21**

and again ...

10, 13, 51, 64

5, 21, 32, 34

Result: 5, 10, 13, 21, **32**

...

10, 13, 51, 64

5, 21, 32, 34

Result: 5, 10, 13, 21, 32, 34

When we reach the **end** of one sequence,  
simply copy the **remaining** numbers in the other  
sequence to the result



10, 13, 51, 64

5, 21, 32, 34

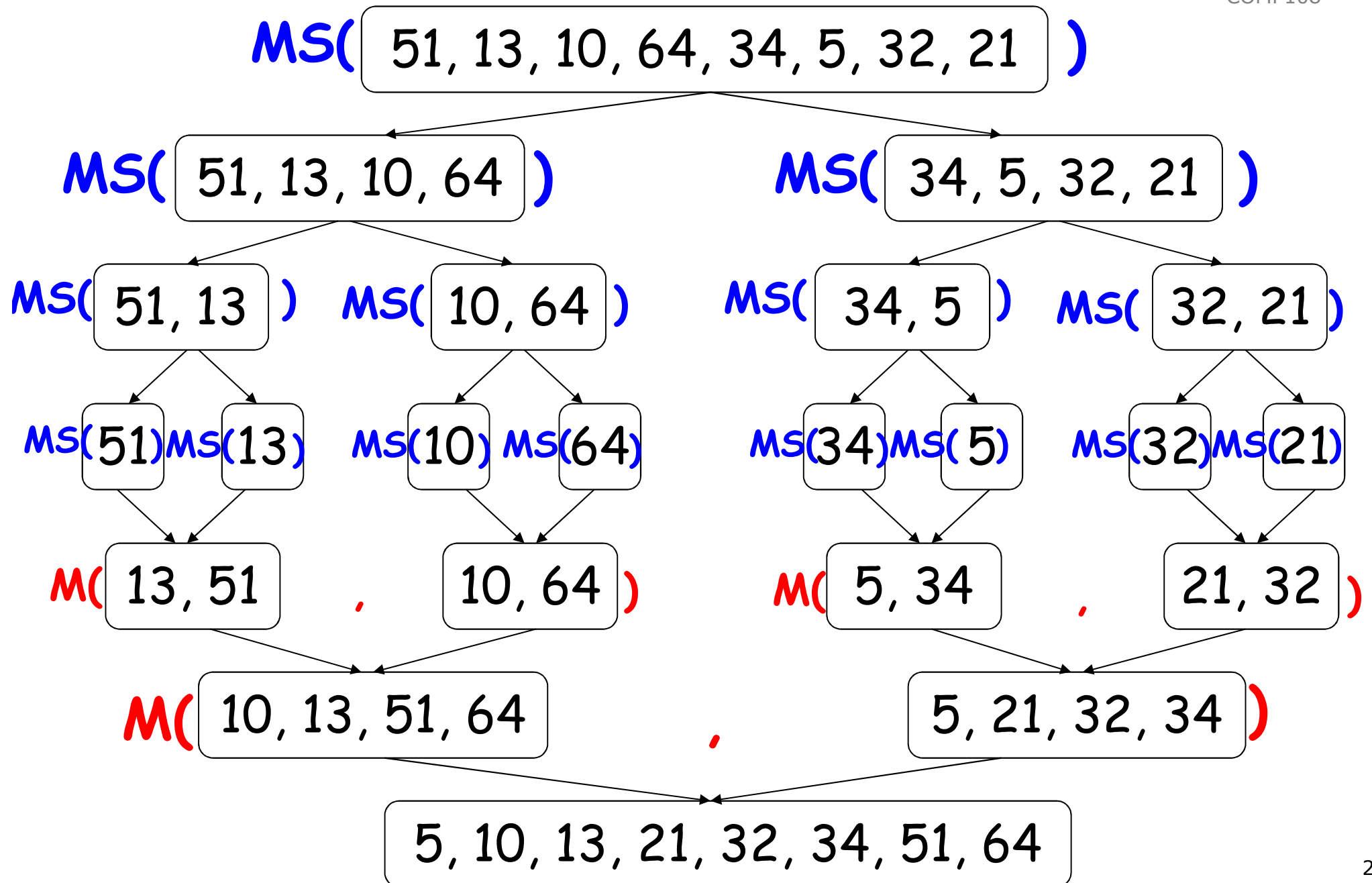


Result: 5, 10, 13, 21, 32, 34, 51, 64

Then we obtain the final sorted sequence

# Pseudo code

```
Algorithm Mergesort( $A[1..n]$ )  
  if  $n > 1$  then begin  
    copy  $A[1..\lfloor n/2 \rfloor]$  to  $B[1..\lfloor n/2 \rfloor]$   
    copy  $A[\lfloor n/2 \rfloor + 1..n]$  to  $C[1..\lceil n/2 \rceil]$   
    Mergesort( $B[1..\lfloor n/2 \rfloor]$ )  
    Mergesort( $C[1..\lceil n/2 \rceil]$ )  
    Merge( $B, C, A$ )  
  end
```



# Pseudo code

```

Algorithm Merge(B[1..p], C[1..q], A[1..p+q])
  set i=1, j=1, k=1
  while i<=p and j<=q do
  begin
    if B[i]≤C[j] then
      set A[k] = B[i] and i = i+1
    else set A[k] = C[j] and j = j+1
    k = k+1
  end
  if i==p+1 then copy C[j..q] to A[k..(p+q)]
  else copy B[i..p] to A[k..(p+q)]

```

**p=4**

**B:** 10, 13, 51, 64

**q=4**

**C:** 5, 21, 32, 34

	i	j	k	A[ ]
Before loop	1	1	1	empty
End of 1st iteration	1	2	2	5
End of 2nd iteration	2	2	3	5, 10
End of 3rd	3	2	4	5, 10, 13
End of 4th	3	3	5	5, 10, 13, 21
End of 5th	3	4	6	5, 10, 13, 21, 32
End of 6th	3	5	7	5, 10, 13, 21, 32, 34
				5, 10, 13, 21, 32, 34, 51, 64

# Time complexity

Let  $T(n)$  denote the time complexity of running merge sort on  $n$  numbers.

$$T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(n/2) + n & \text{otherwise} \end{cases}$$

We call this formula a recurrence.

A recurrence is an equation or inequality that describes a function in terms of its value on smaller inputs.

To solve a recurrence is to derive *asymptotic bounds* on the solution

# Time complexity

Prove that  $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(n/2) + n & \text{otherwise} \end{cases}$  is  $O(n \log n)$

**Make a guess:**  $T(n) \leq 2 n \log n$  (We prove by MI)

For the base case when  $n=2$ ,  
 L.H.S =  $T(2) = 2 \times T(1) + 2 = 4$ ,  
 R.H.S =  $2 \times 2 \log 2 = 4$   
 L.H.S  $\leq$  R.H.S

## Substitution method

# Time complexity

Prove that  $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(n/2) + n & \text{otherwise} \end{cases}$  is  $O(n \log n)$

**Make a guess:**  $T(n) \leq 2 n \log n$  (We prove by MI)

Assume true for all  $n' < n$  [assume  $T(n/2) \leq 2 (n/2) \log(n/2)$ ]

$$T(n) = 2 \times T(n/2) + n \quad \text{by hypothesis}$$

$$\leq 2 \times (2 \times (n/2) \times \log(n/2)) + n$$

$$= 2 n (\log n - 1) + n$$

$$= 2 n \log n - 2n + n$$

$$\leq 2 n \log n$$

$$\text{i.e., } T(n) \leq 2 n \log n$$



# Example

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

Guess:  $T(n) \leq 2 \log n$

For the base case when  $n=2$ ,

$$\text{L.H.S} = T(2) = T(1) + 1 = 2$$

$$\text{R.H.S} = 2 \log 2 = 2$$

$$\text{L.H.S} \leq \text{R.H.S}$$

# Example

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

Guess:  $T(n) \leq 2 \log n$

Assume true for all  $n' < n$  [assume  $T(n/2) \leq 2 \times \log(n/2)$ ]

$$T(n) = T(n/2) + 1$$

$$\leq 2 \times \log(n/2) + 1 \leftarrow \text{by hypothesis}$$

$$= 2 \times (\log n - 1) + 1 \leftarrow \log(n/2) = \log n - \log 2$$

$$< 2 \log n$$

$$\text{i.e., } T(n) \leq 2 \log n$$

# More example

Prove that

$$T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(n/2) + 1 & \text{otherwise} \end{cases}$$

is  $O(n)$

**Guess:**  $T(n) \leq 2n - 1$

For the base case when  $n=1$ ,

$$\text{L.H.S} = T(1) = 1$$

$$\text{R.H.S} = 2 \times 1 - 1 = 1$$

$$\text{L.H.S} \leq \text{R.H.S}$$

# More example

Prove that  $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(n/2) + 1 & \text{otherwise} \end{cases}$  is  $O(n)$

**Guess:**  $T(n) \leq 2n - 1$

Assume true for all  $n' < n$  [assume  $T(n/2) \leq 2(n/2) - 1$ ]

$$\begin{aligned} T(n) &= 2 \times T(n/2) + 1 \\ &\leq 2 \times (2 \times (n/2) - 1) + 1 && \leftarrow \text{by hypothesis} \\ &= 2n - 2 + 1 \\ &= 2n - 1 \end{aligned}$$

$$\text{i.e., } T(n) \leq 2n - 1$$

# Summary

Depending on the recurrence, we can guess the order of growth

$$T(n) = T(n/2) + 1 \quad T(n) \text{ is } O(\log n)$$

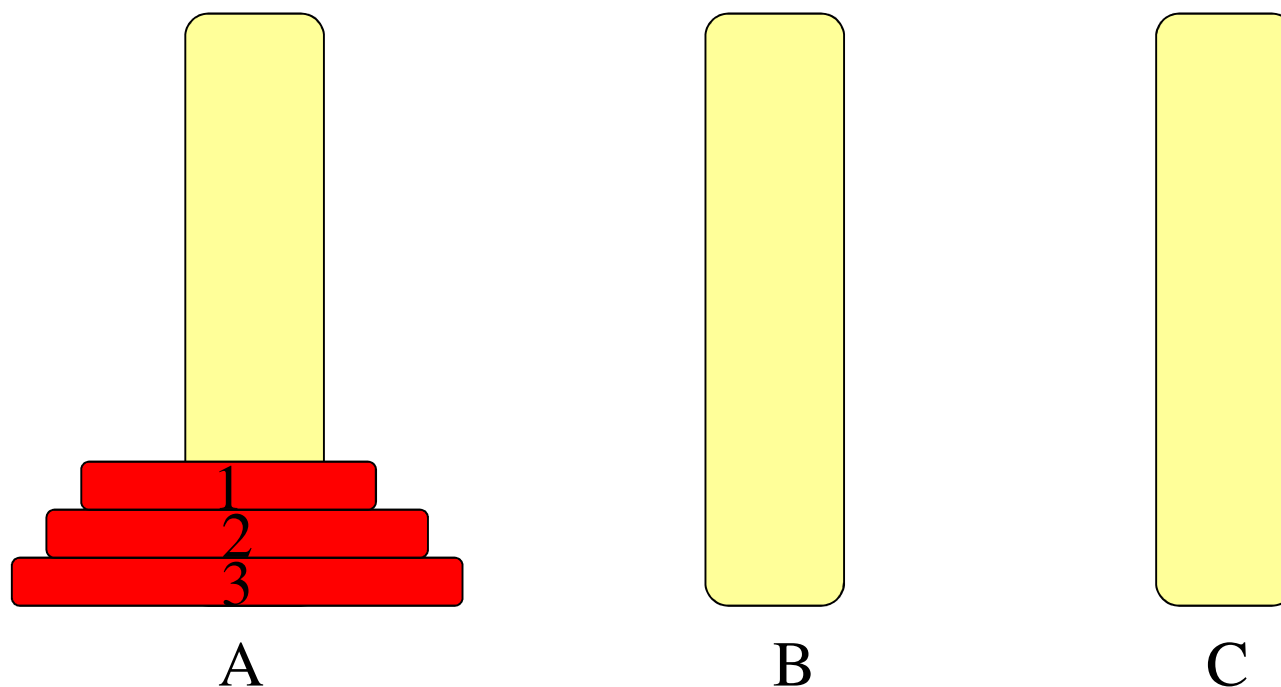
$$T(n) = 2 \times T(n/2) + 1 \quad T(n) \text{ is } O(n)$$

$$T(n) = 2 \times T(n/2) + n \quad T(n) \text{ is } O(n \log n)$$

# Tower of Hanoi ...

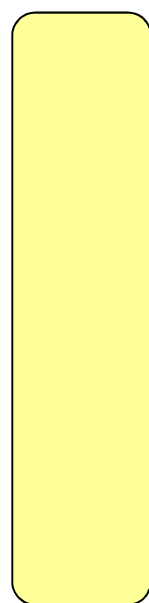
# Tower of Hanoi - Initial config

There are three pegs and some discs of different sizes are on Peg A



# Tower of Hanoi - Final config

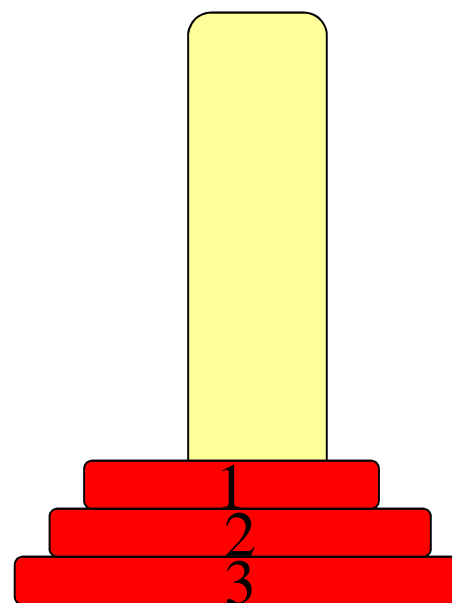
Want to move the discs to Peg C



A



B



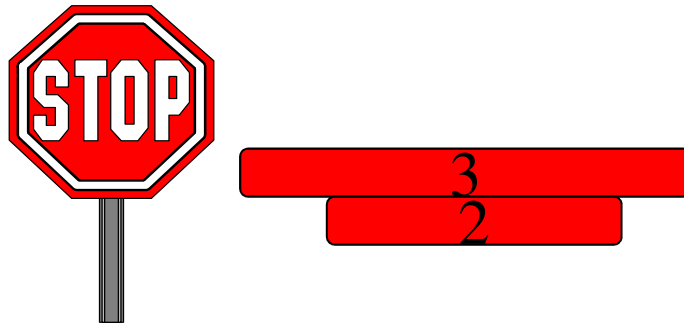
C



# Tower of Hanoi - Rules

Only 1 disk can be moved at a time

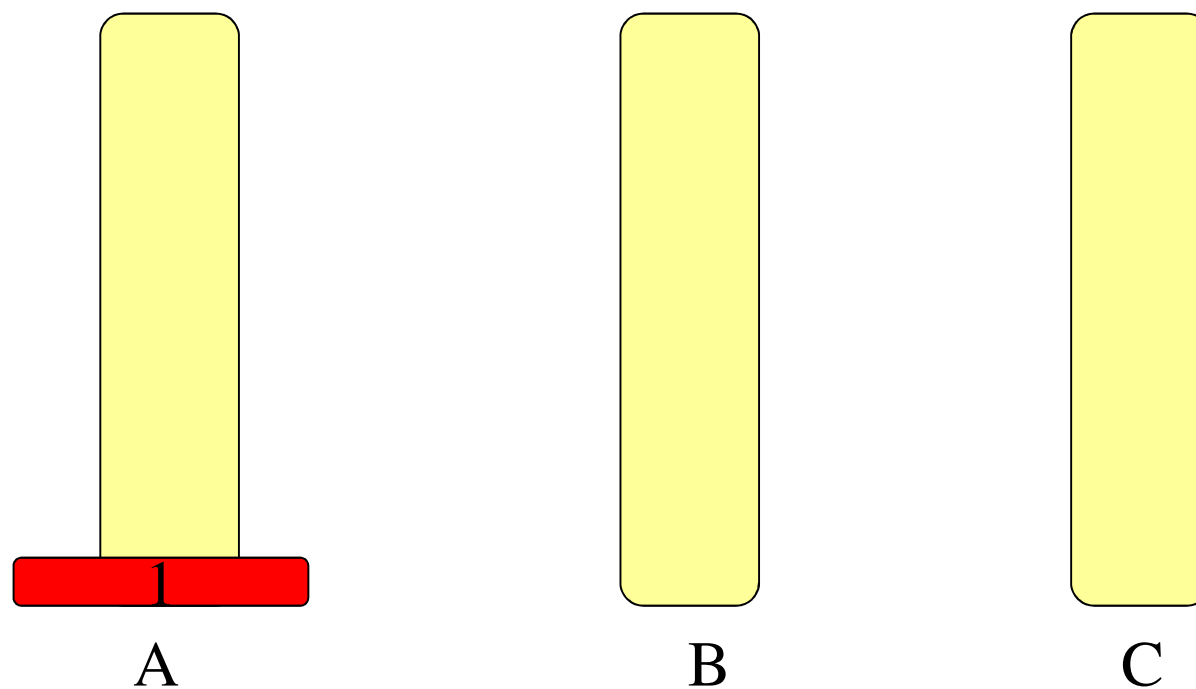
A disc cannot be placed on top of other discs that are smaller than it



**Target: Use the smallest number of moves**

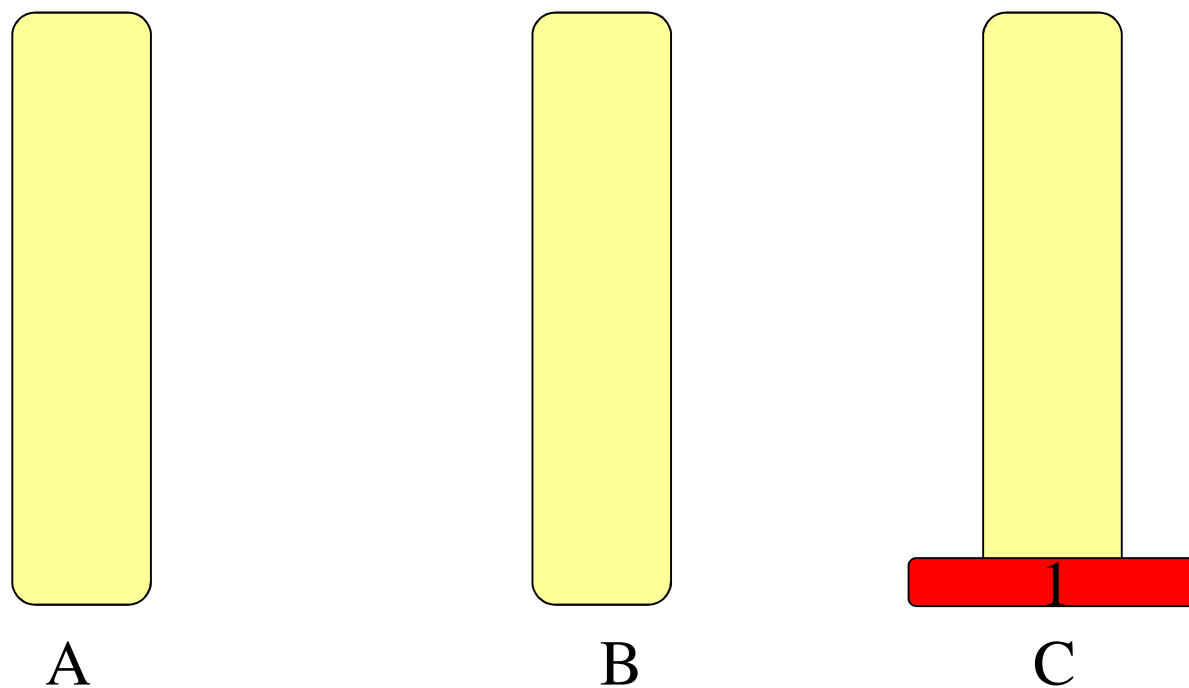
# Tower of Hanoi - One disc only

Easy!



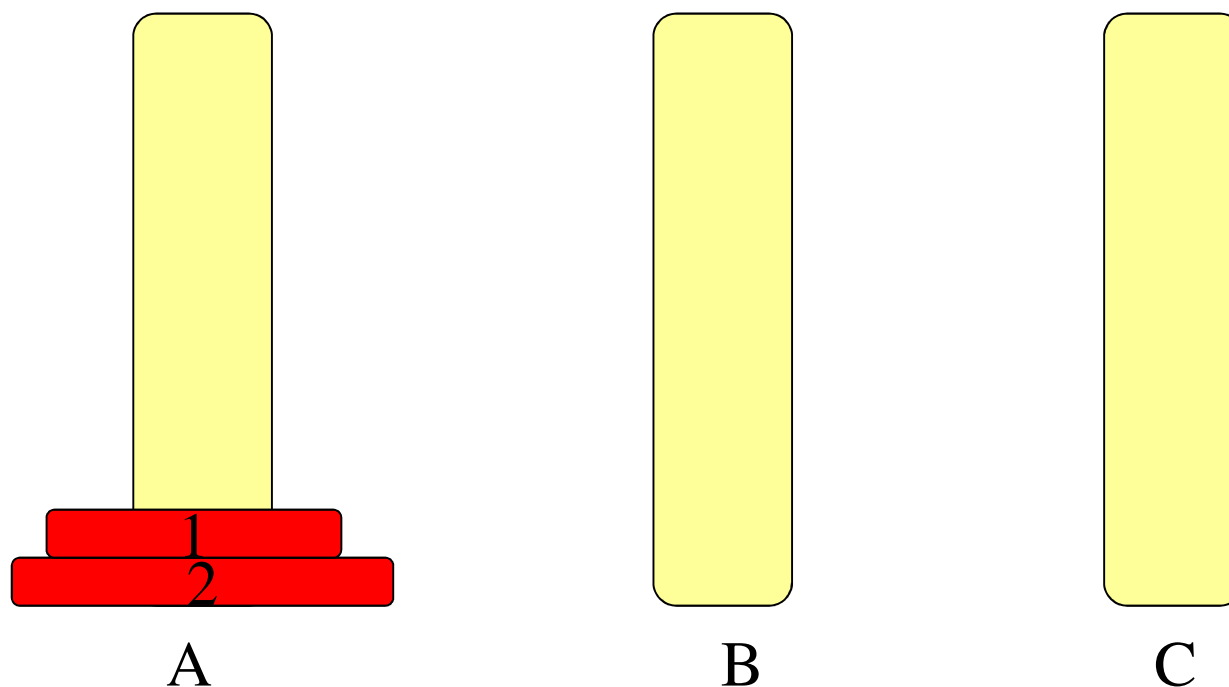
# Tower of Hanoi - One disc only

Easy! Need one move only.



# Tower of Hanoi - Two discs

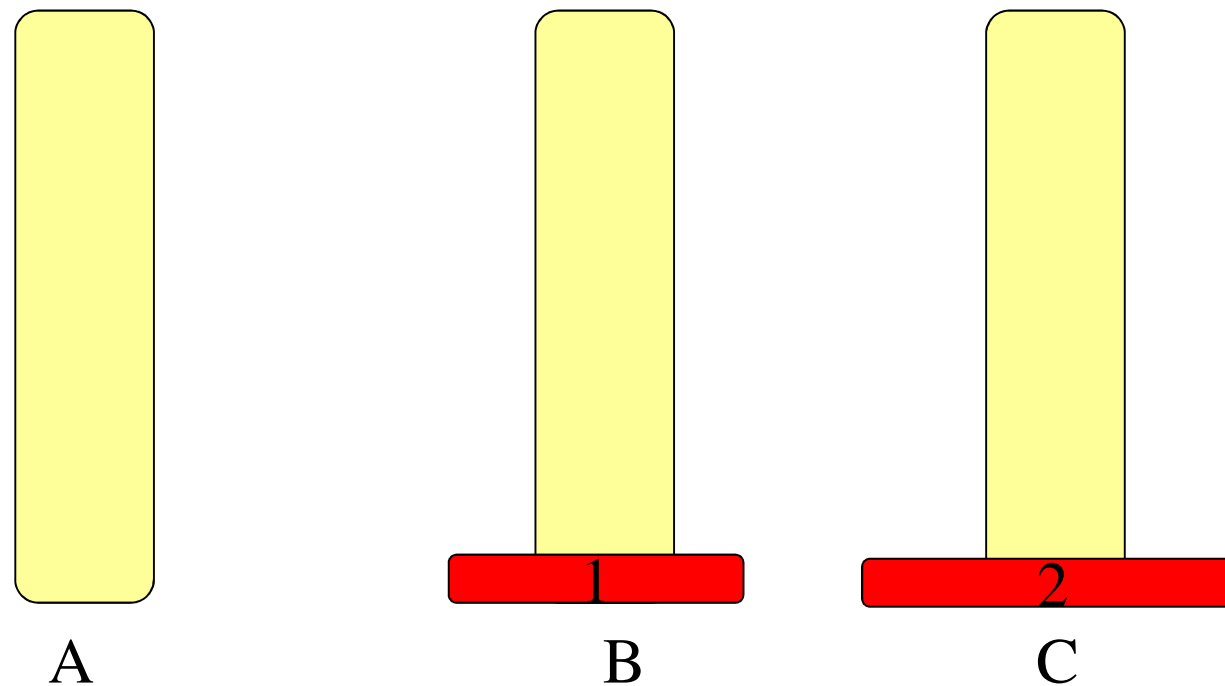
We first need to move Disc-2 to C, How?  
by moving Disc-1 to B first, then Disc-2 to C



# Tower of Hanoi - Two discs

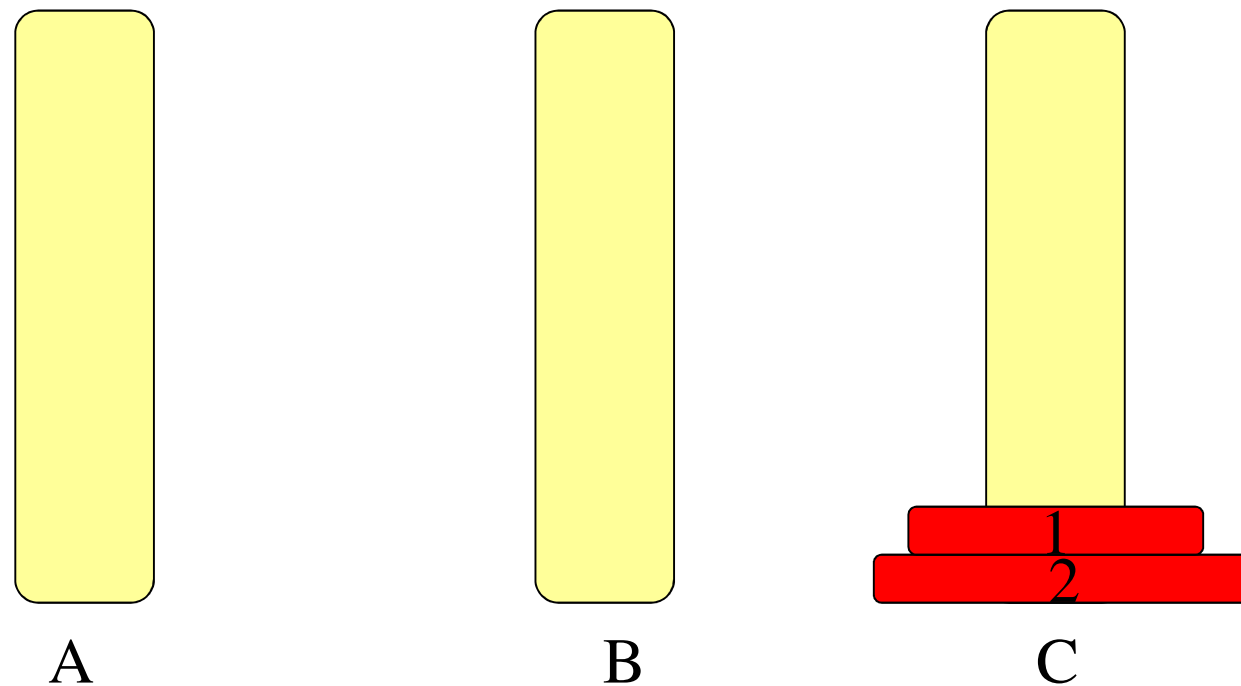
Next?

Move Disc-1 to C



# Tower of Hanoi - Two discs

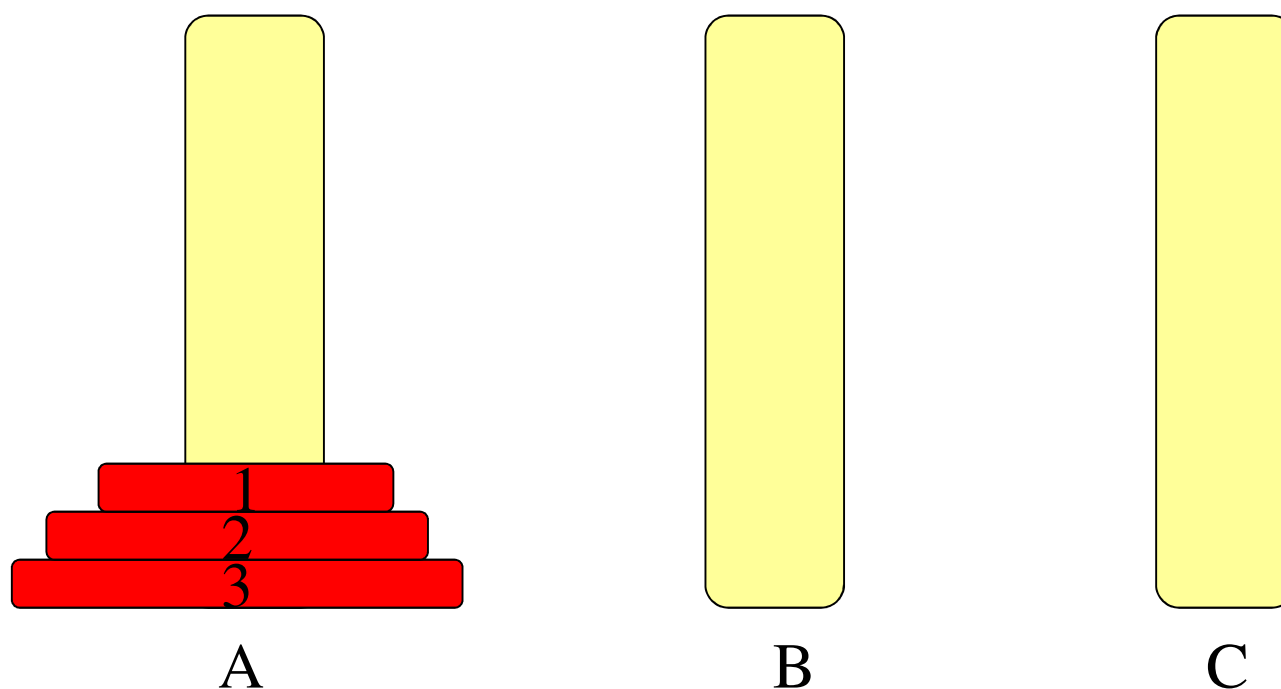
Done!



# Tower of Hanoi - Three discs

We first need to move Disc-3 to C, How?

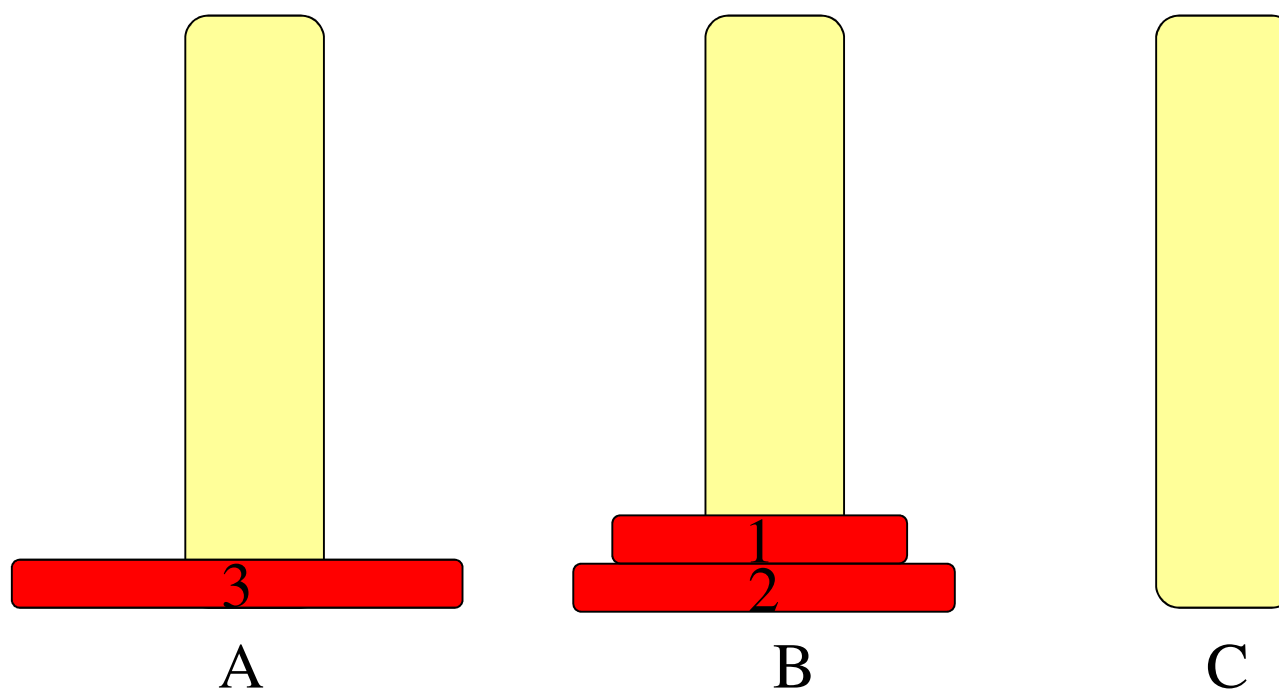
- Move Disc-1&2 to B (recursively)



# Tower of Hanoi - Three discs

We first need to move Disc-3 to C, How?

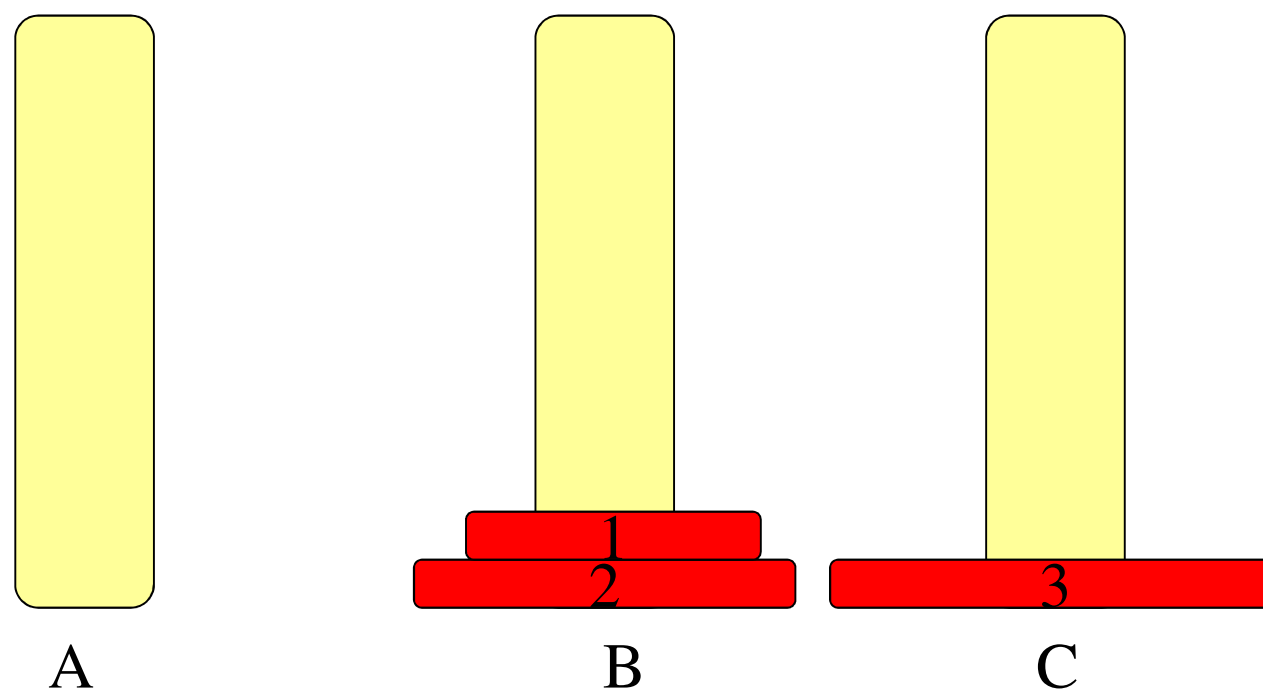
- Move Disc-1&2 to B (recursively)
- Then move Disc-3 to C





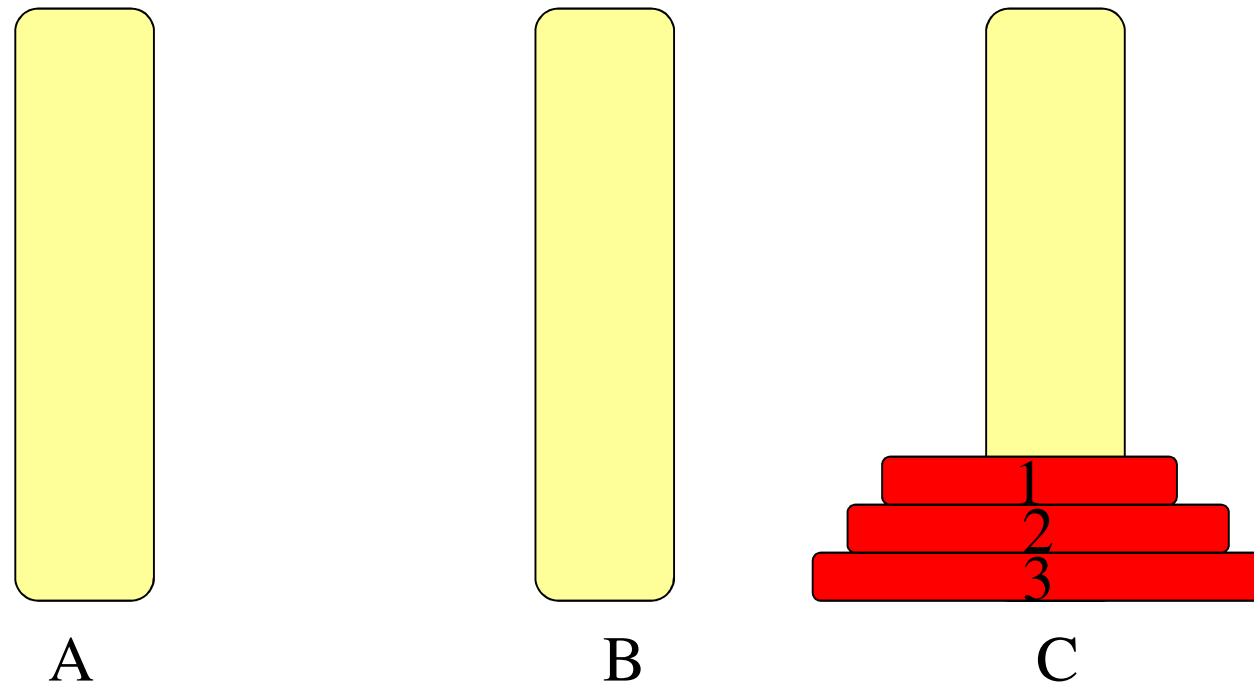
# Tower of Hanoi - Three discs

Only task left: move Disc-1&2 to C (similarly as before)



# Tower of Hanoi - **Three** discs

Done!



# Tower of Hanoi

ToH(num\_disc, source, dest, spare)

begin

if (num\_disc > 1) then

ToH(num\_disc-1, source, spare, dest)

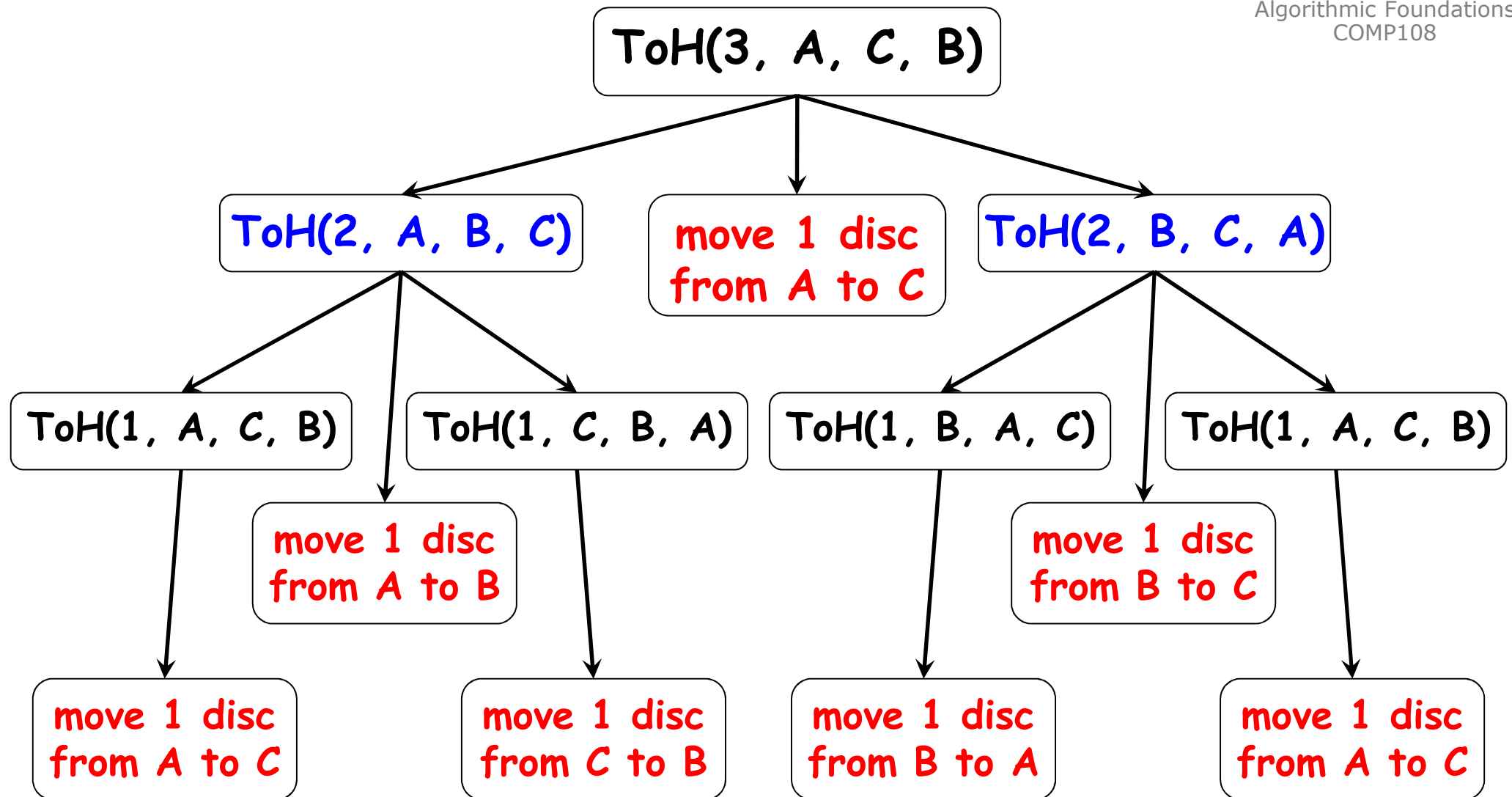
Move the disc from source to dest

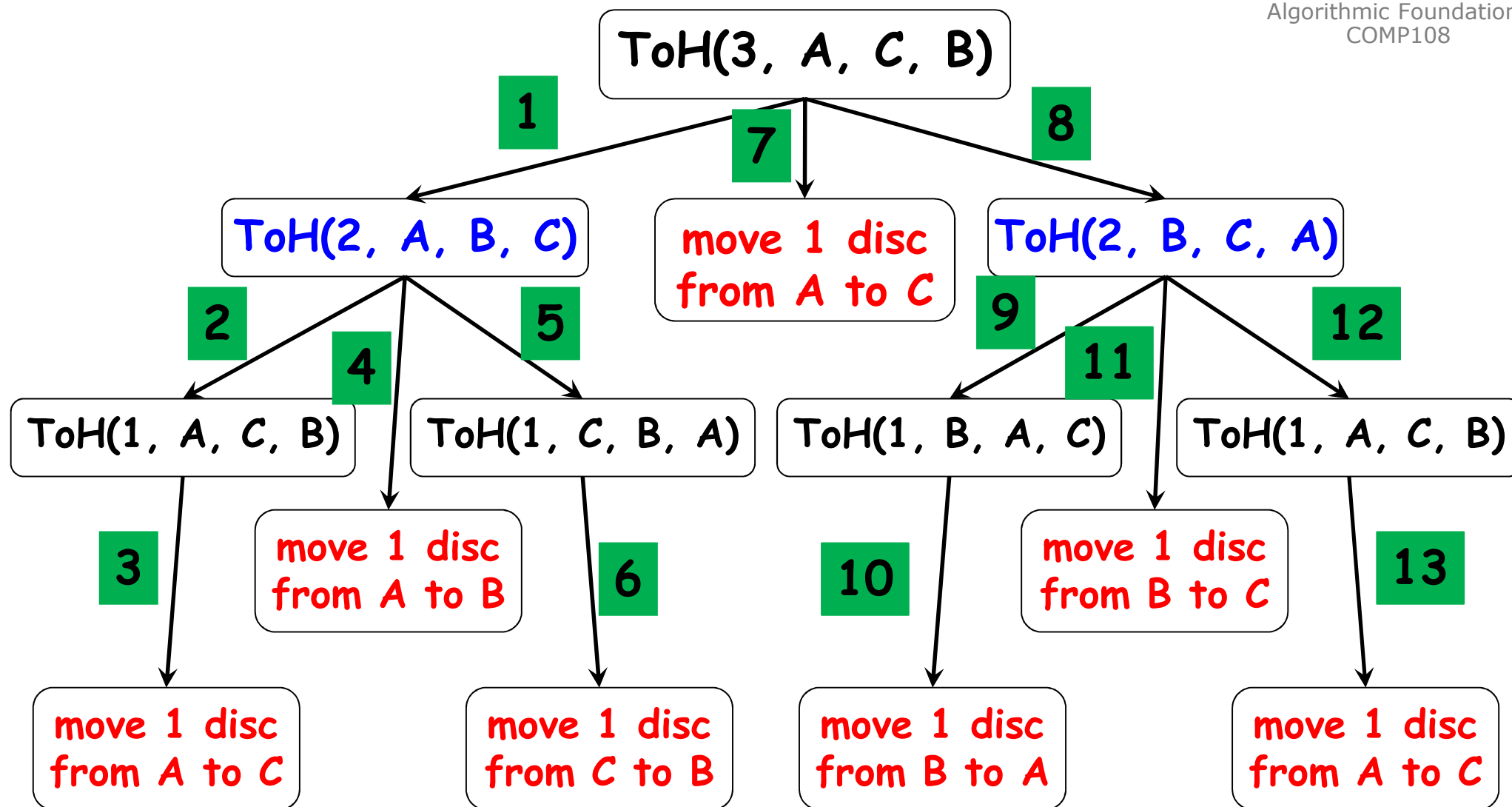
if (num\_disc > 1) then

ToH(num\_disc-1, spare, dest, source)

end

invoke by calling  
ToH(3, A, C, B)





from A to C; from A to B; from C to B;  
from A to C;  
from B to A; from B to C; from A to C;

# Time complexity

Let  $T(n)$  denote the time complexity of running the Tower of Hanoi algorithm on  $n$  discs.

$$T(n) = \underbrace{T(n-1)}_{\substack{\text{move } n-1 \\ \text{discs from} \\ \text{A to B}}} + \underset{\substack{\text{move Disc-}n \\ \text{from A to C}}}{\uparrow} 1 + \underbrace{T(n-1)}_{\substack{\text{move } n-1 \\ \text{discs from} \\ \text{B to C}}}$$

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

# Time complexity (2)

$$\begin{aligned}
 T(n) &= 2 \times T(n-1) + 1 \\
 &= 2[2 \times T(n-2) + 1] + 1 \\
 &= 2^2 T(n-2) + 2 + 1 \\
 &= 2^2 [2 \times T(n-3) + 1] + 2^1 + 2^0 \\
 &= 2^3 T(n-3) + 2^2 + 2^1 + 2^0
 \end{aligned}$$

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

$$\begin{aligned}
 &\dots \\
 &= 2^k T(n-k) + 2^{k-1} + 2^{k-2} + \dots + 2^2 + 2^1 + 2^0
 \end{aligned}$$

$$\begin{aligned}
 &\dots \\
 &= 2^{n-1} T(1) + 2^{n-2} + 2^{n-3} + \dots + 2^2 + 2^1 + 2^0
 \end{aligned}$$

$$= 2^{n-1} + 2^{n-2} + 2^{n-3} + \dots + 2^2 + 2^1 + 2^0$$

iterative  
method

$$= 2^{n-1}$$

i.e.,  $T(n)$  is  $O(2^n)$

In Tutorial 2, we prove by MI that  
 $2^0 + 2^1 + \dots + 2^{n-1} = 2^n - 1$

# Summary - continued

Depending on the recurrence, we can guess the order of growth

$$T(n) = T(n/2) + 1 \quad T(n) \text{ is } O(\log n)$$

$$T(n) = 2 \times T(n/2) + 1 \quad T(n) \text{ is } O(n)$$

$$T(n) = 2 \times T(n/2) + n \quad T(n) \text{ is } O(n \log n)$$

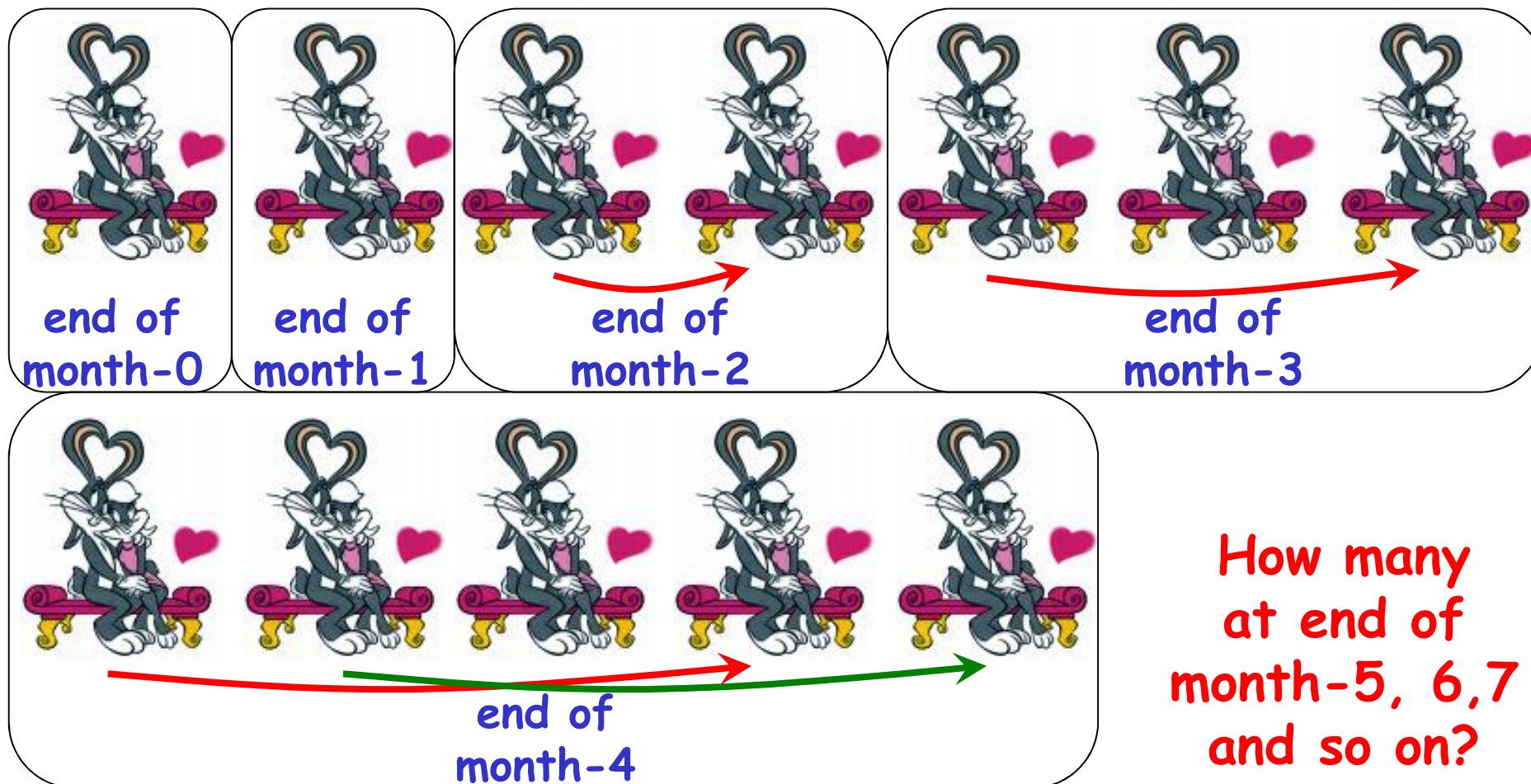
$$T(n) = 2 \times T(n-1) + 1 \quad T(n) \text{ is } O(2^n)$$



# Fibonacci number ...

# Fibonacci's Rabbits

A pair of rabbits, one month old, is too young to reproduce.  
Suppose that in their second month, and every month thereafter, they produce a new pair.



How many  
at end of  
month-5, 6, 7  
and so on?

# Petals on flowers



**1 petal:  
white calla lily**



**2 petals:  
euphorbia**



**3 petals:  
trillium**



**5 petals:  
columbine**



**8 petals:  
bloodroot**



**13 petals:  
black-eyed susan**



**21 petals:  
shasta daisy**



**34 petals:  
field daisy**

**Search: Fibonacci Numbers in Nature**

# Fibonacci number

Fibonacci number  $F(n)$

$$F(n) = \begin{cases} 1 & \text{if } n = 0 \text{ or } 1 \\ F(n-1) + F(n-2) & \text{if } n > 1 \end{cases}$$

n	0	1	2	3	4	5	6	7	8	9	10
F(n)	1	1	2	3	5	8	13	21	34	55	89

Pseudo code for the recursive algorithm:

**Algorithm**  $F(n)$

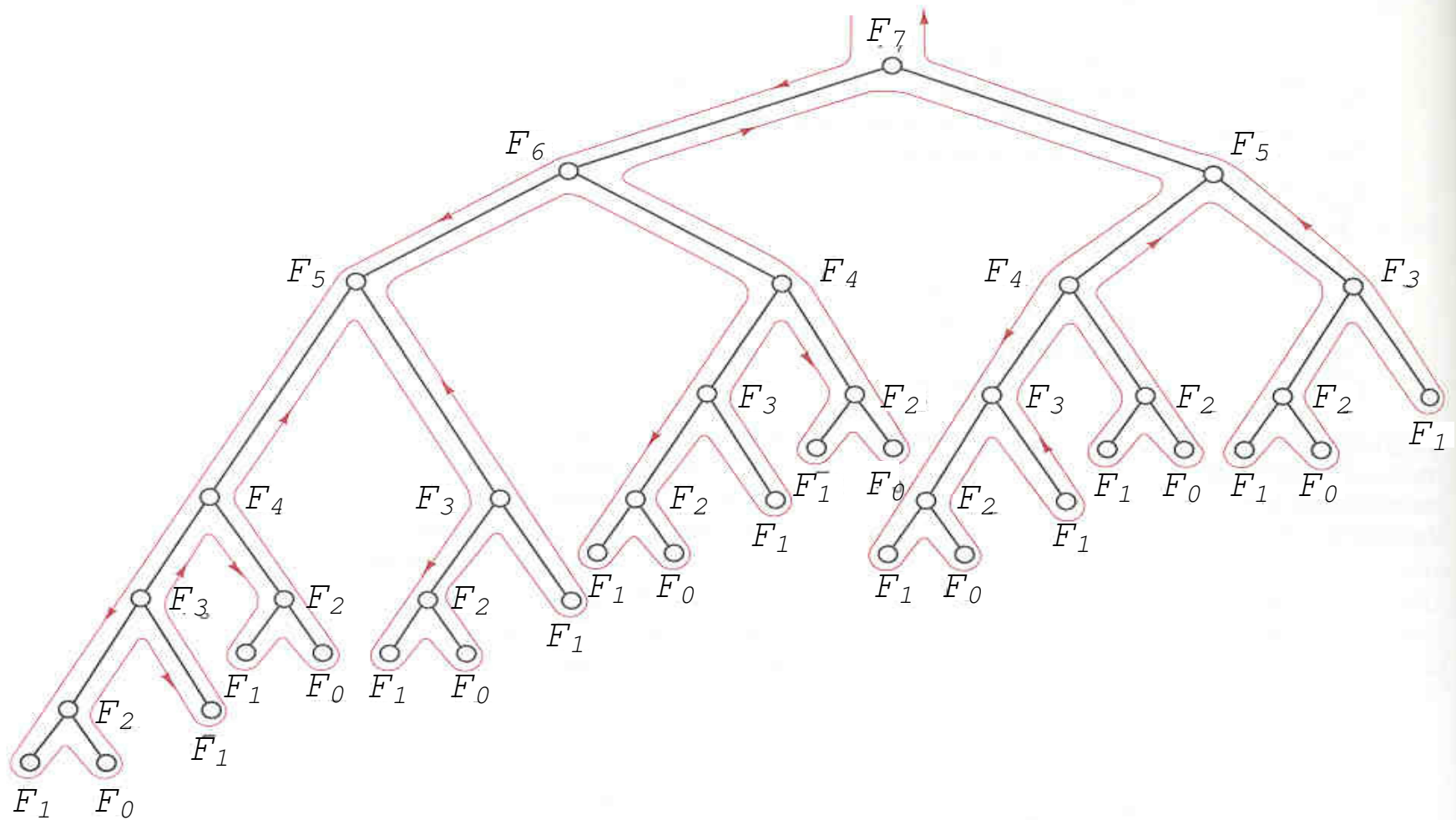
**if**  $n==0$  or  $n==1$  **then**

**return** 1

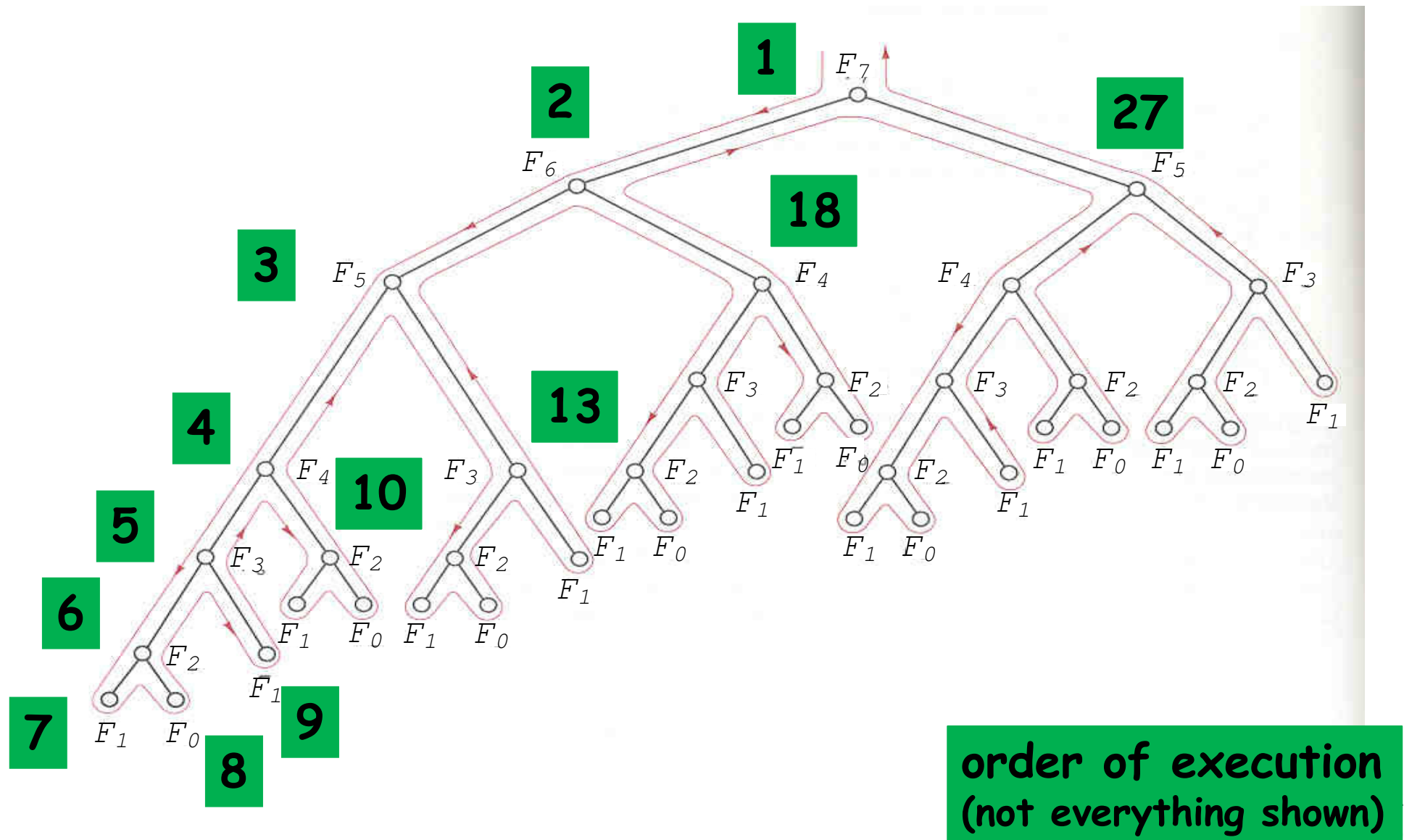
**else**

**return**  $F(n-1) + F(n-2)$

# The execution of $F(7)$

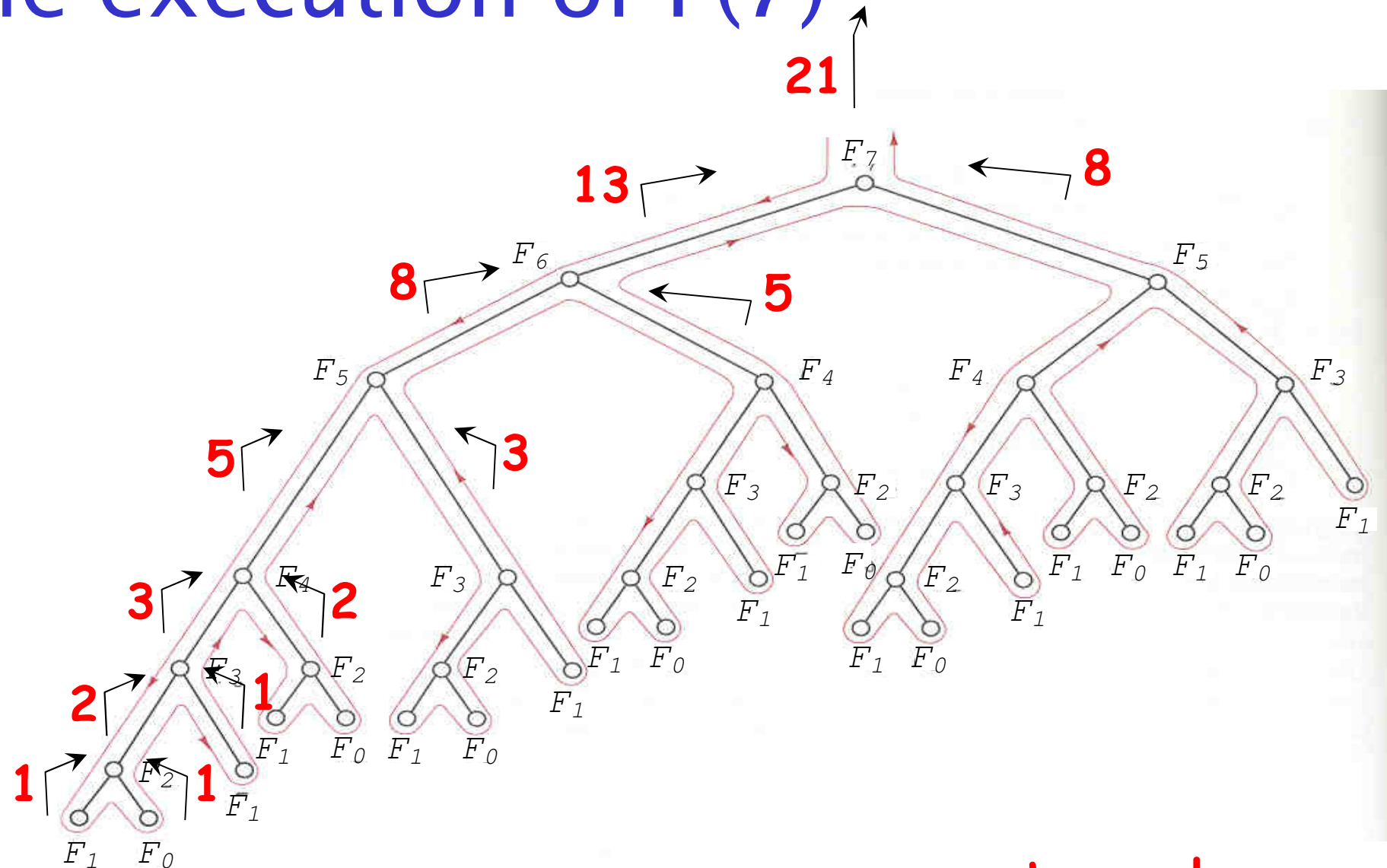


# The execution of $F(7)$





# The execution of $F(7)$



return value  
(not everything shown)

# Time complexity - exponential

$$\begin{aligned}
 f(n) &= \underbrace{f(n-1)} + f(n-2) + 1 \\
 &= [\textcolor{red}{f(n-2)} + \textcolor{red}{f(n-3)} + \textcolor{red}{1}] + f(n-2) + 1 \\
 &> 2 \underbrace{f(n-2)} \\
 &> 2 [\textcolor{red}{2} \times \textcolor{red}{f(n-2-2)}] = 2^2 \underbrace{f(n-4)} \\
 &> 2^2 [\textcolor{red}{2} \times \textcolor{red}{f(n-4-2)}] = 2^3 \underbrace{f(n-6)} \\
 &> 2^3 [\textcolor{red}{2} \times \textcolor{red}{f(n-6-2)}] = 2^4 \underbrace{f(n-8)} \\
 &\dots \\
 &> 2^k f(n-2k)
 \end{aligned}$$

Suppose  $f(n)$   
denote the time  
complexity to  
compute  $F(n)$

exponential in  $n$

If  $n$  is even,  $f(n) > 2^{n/2} f(0) = 2^{n/2}$

If  $n$  is odd,  $f(n) > f(n-1) > 2^{(n-1)/2}$