COMP108 Algorithmic Foundations

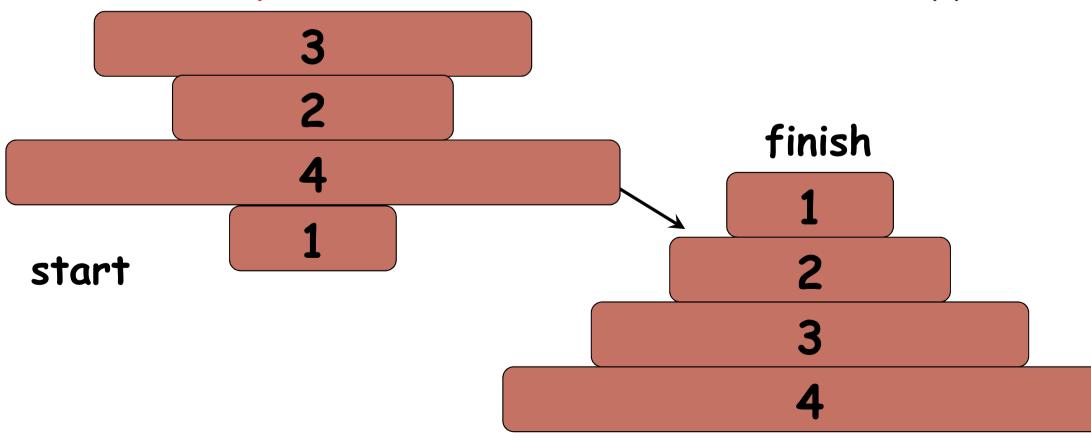
Divide and Conquer

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http://www.csc.liv.ac.uk/~pwong/teaching/comp108/201617

Pancake Sorting

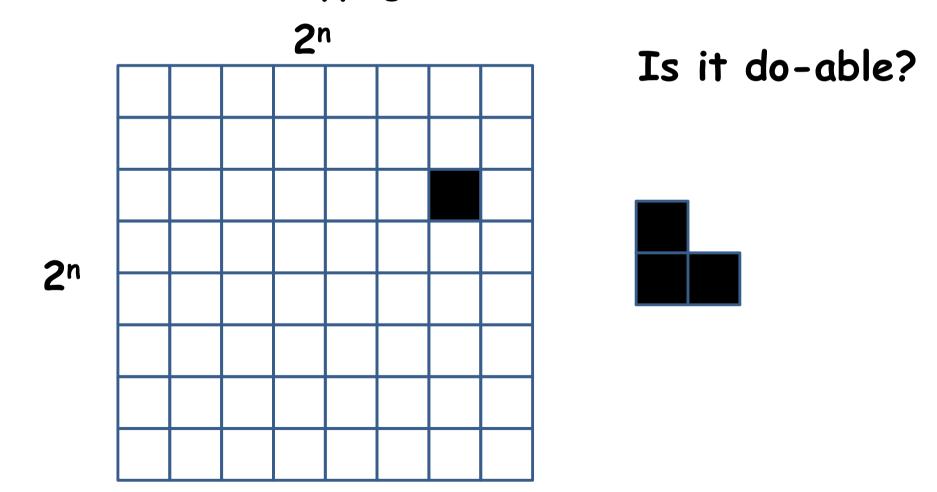
Input: Stack of pancakes, each of different sizes
Output: Arrange in order of size (smallest on top)
Action: Slip a flipper under one of the pancakes and flip over the whole stack above the flipper



Triomino Puzzle

Input:

2ⁿ-by-2ⁿ chessboard with one missing square & many L-shaped tiles of 3 adjacent squares Question: Cover the chessboard with L-shaped tiles without overlapping



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 <u>divided</u> 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 <u>combined</u> 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

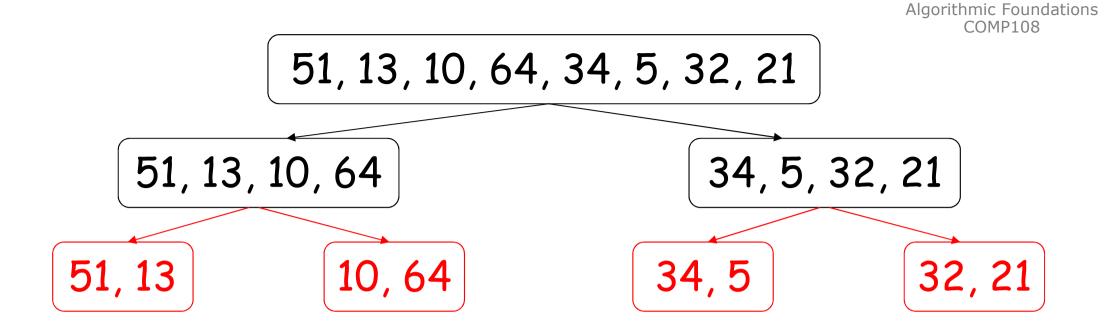
51, 13, 10, 64, 34, 5, 32, 21

51, 13, 10, 64

34, 5, 32, 21

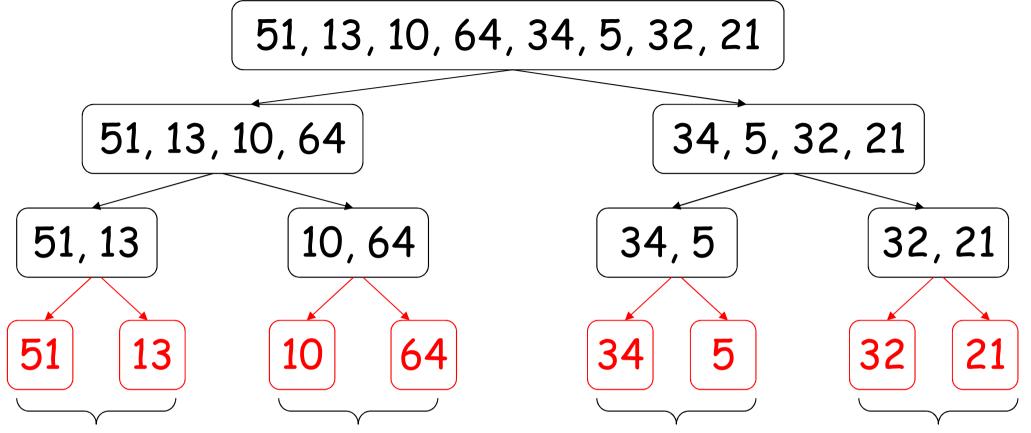
divide these 4 numbers into halves

similarly for these 4

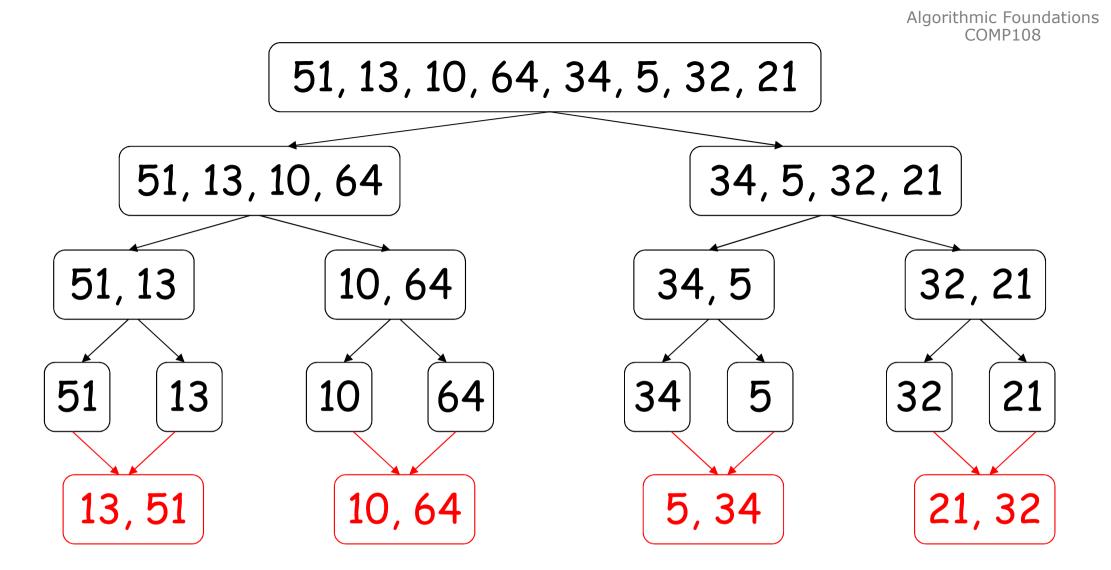


further divide each shorter sequence ... until we get sequence with only 1 number

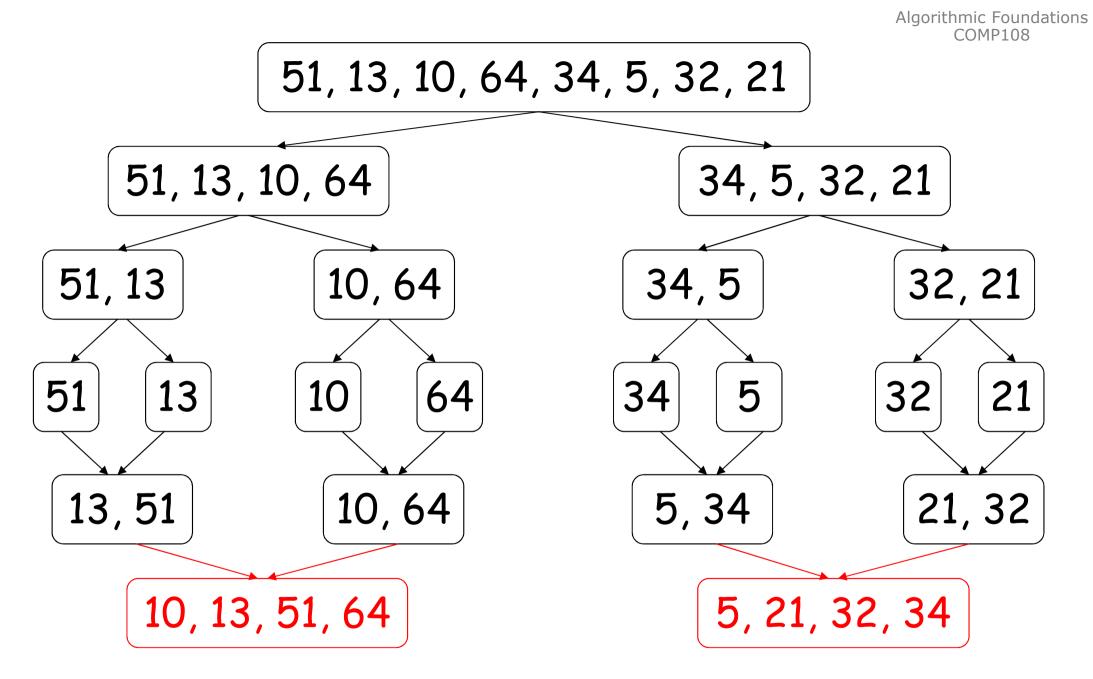




merge pairs of single number into a sequence of 2 sorted numbers



then merge again into sequences of 4 sorted numbers



one more merge give the final sorted sequence

COMP108 51, 13, 10, 64, 34, 5, 32, 21 51, 13, 10, 64 34, 5, 32, 21 51, 13 32, 21 10,64 34,5 51 13 10 64 34 5 32 21 5,34 21, 32 13, 51 10,64 5, 21, 32, 34 10, 13, 51, 64 5, 10, 13, 21, 32, 34, 51, 64 15

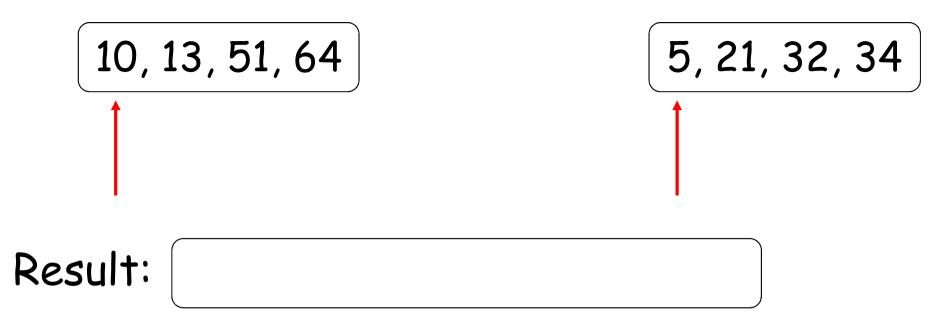
Summary

Divide

> dividing a sequence of n numbers into two smaller sequences is straightforward

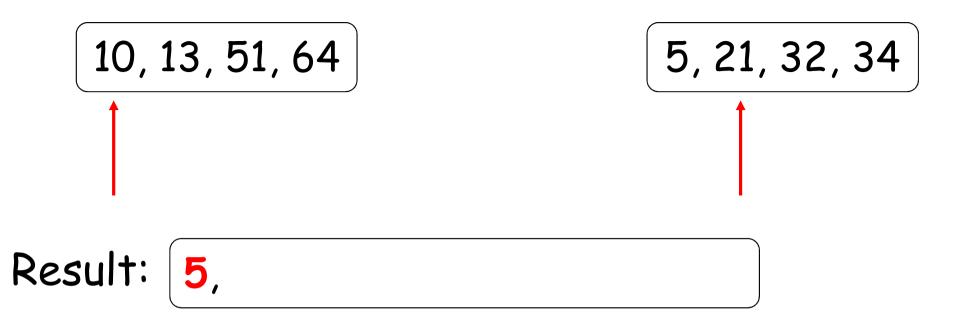
Conquer

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

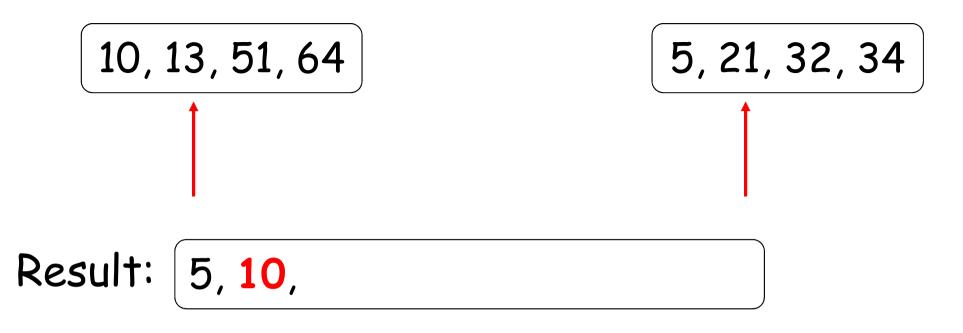


To merge two sorted sequences, we keep two **pointers**, one to each sequence

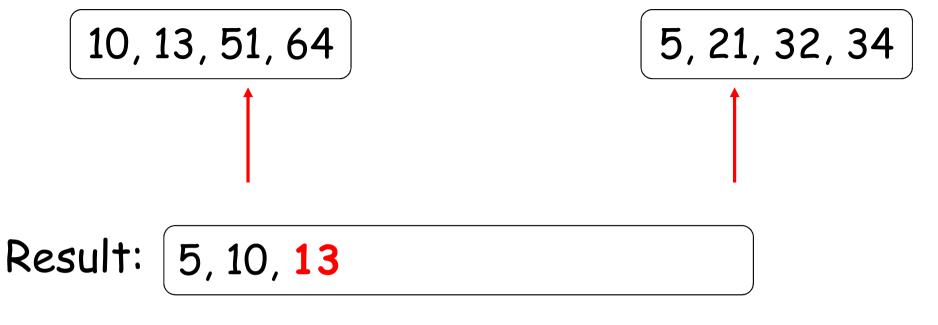
Compare the two numbers pointed, copy the smaller one to the result and advance the corresponding pointer



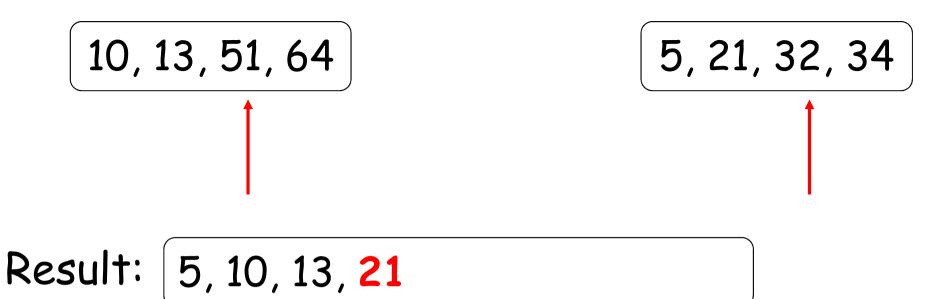
Then compare again the two numbers pointed to by the pointer; copy the smaller one to the result and advance that pointer



Repeat the same process ...

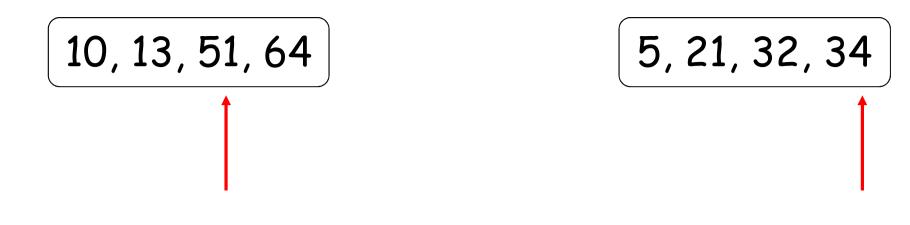


Again ...



and again ...

21 (Divide & Conquer)



...

Result: 5, 10, 13, 21, 32



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

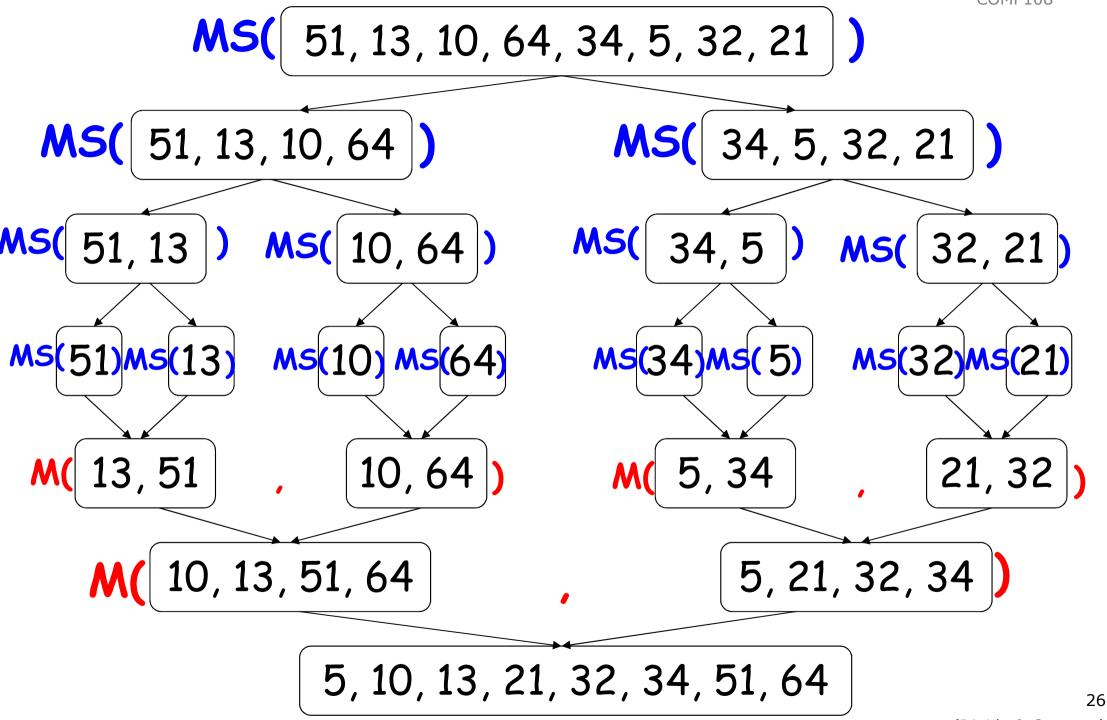


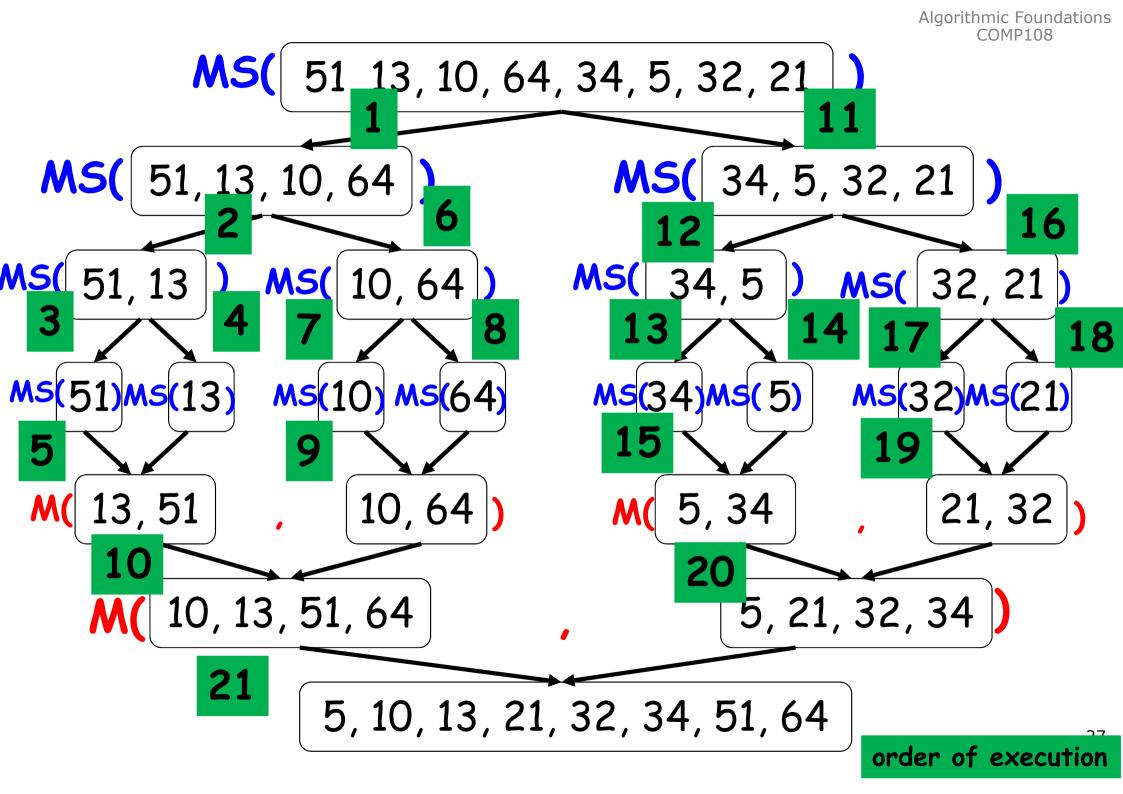
Result: 5, 10, 13, 21, 32, 34, **51, 64**

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[[n/2]+1..n] to C[1..[n/2]] **Mergesort**(B[1.. $\lfloor n/2 \rfloor$]) Mergesort(C[1..[n/2]]) Merge(B, C, A)end

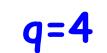




Pseudo code

```
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
    \mathbf{k} = \mathbf{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



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.

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

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 <u>solve</u> a recurrence is to derive *asymptotic bounds* on the solution

Time complexity

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

Make a guess: $T(n) \le 2 n \log n$ (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 $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(\frac{n}{2}) + n & \text{otherwise} \end{cases}$ is O(n log n)

Make a guess: $T(n) \le 2 n \log n$ (We prove by MI) Assume true for all n'<n [assume $T(\frac{n}{2}) \le 2 \times (\frac{n}{2}) \times \log(\frac{n}{2})$]

 $T(n) = 2 \times T(\frac{n}{2}) + n$ $\leq 2 \times (2 \times (\frac{n}{2}) \times \log(\frac{n}{2})) + n$

 \leq 2 n log n

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 $2 n \log$

Example

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

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

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

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

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

Assume true for all n' < n [assume $T(\frac{n}{2}) \le 2 \times \log(\frac{n}{2})$] $T(n) = T(\frac{n}{2}) + 1$ $\le 2 \times \log(\frac{n}{2}) + 1 \quad \leftarrow by \ hypothesis$ $= 2x(\log n - 1) + 1 \leftarrow \log(\frac{n}{2}) = \log n - \log 2$ $< 2\log n$ i.e., $T(n) \le 2 \log n$

(Divide & Conquer)

More example

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

is O(n)

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

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

More example

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

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

Assume true for all n' < n [assume $T(\frac{n}{2}) \le 2x(\frac{n}{2})-1$]

$$T(n) = 2 \times T(\frac{n}{2}) + 1$$

$$\leq 2 \times (2 \times (\frac{n}{2}) - 1) + 1 \quad \leftarrow by \ hypothesis$$

$$= 2n - 2 + 1$$

$$= 2n - 1$$
i.e., $T(n) \leq 2n - 1$

Summary

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

$$T(n) = T(\frac{n}{2})+1$$

$$T(n) \text{ is } O(\log n)$$

$$T(n) = 2 \times T(\frac{n}{2})+1$$

$$T(n) \text{ is } O(n)$$

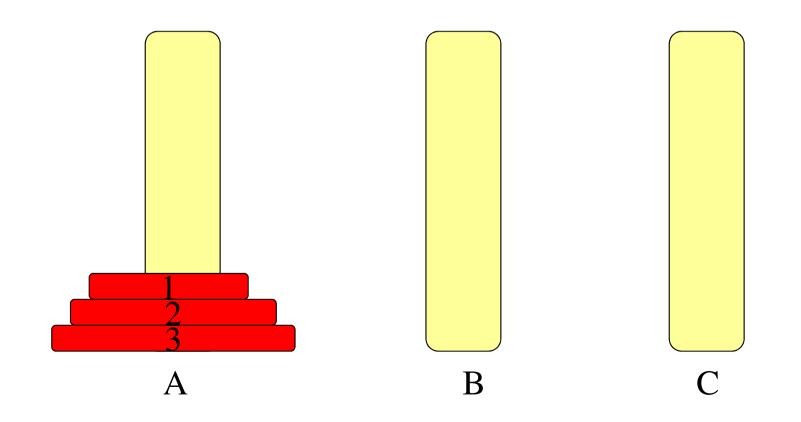
$$T(n) = 2 \times T(\frac{n}{2})+n$$

$$T(n) \text{ is } O(n \log n)$$

Tower of Hanoi ...

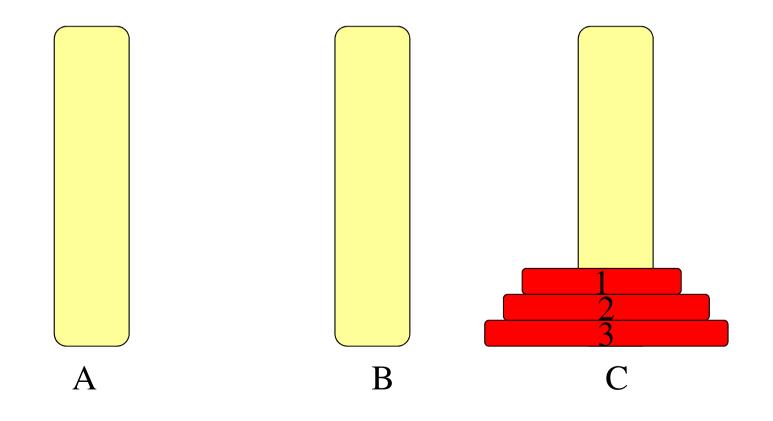
Tower of Hanoi - Initial config

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

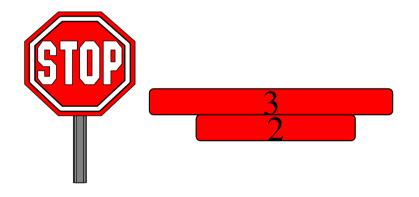


(Divide & Conquer)

Tower of Hanoi - Rules

Only 1 disk can be moved at a time

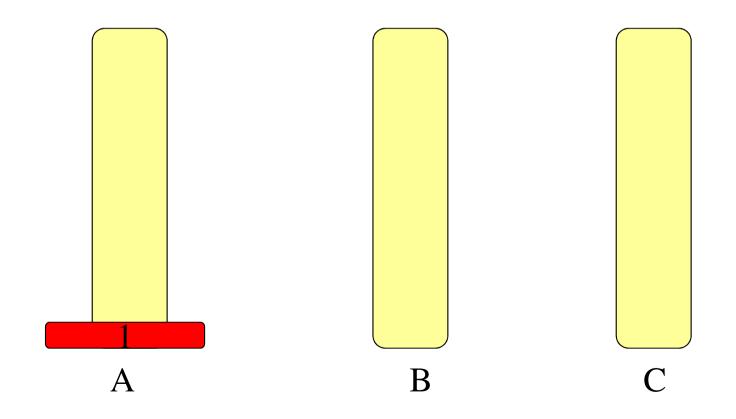
A disc cannot be placed on top of other discs that are smaller than it



Target: Use the smallest number of moves

Tower of Hanoi - One disc only

Easy!



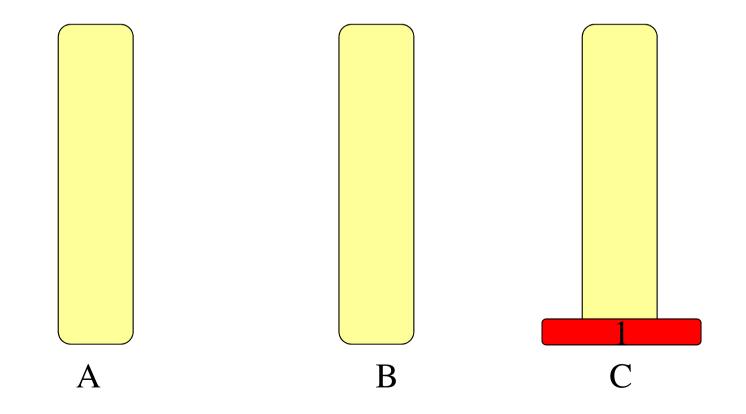
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(Divide & Conquer)

Algorithmic Foundations COMP108

Tower of Hanoi - One disc only

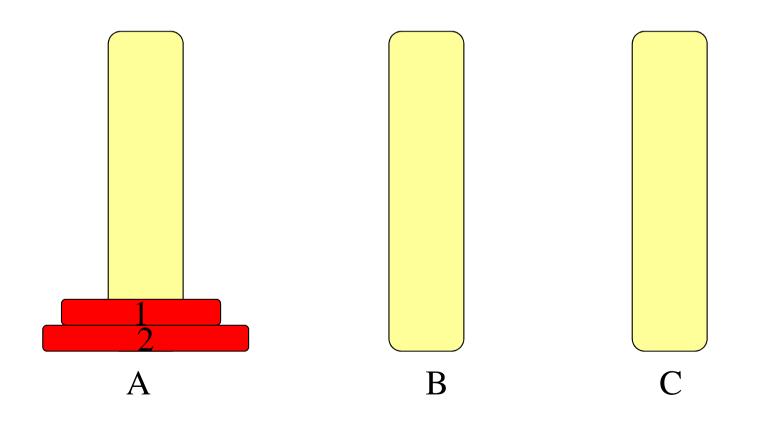
Easy! Need one move only.



Algorithmic Foundations COMP108

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

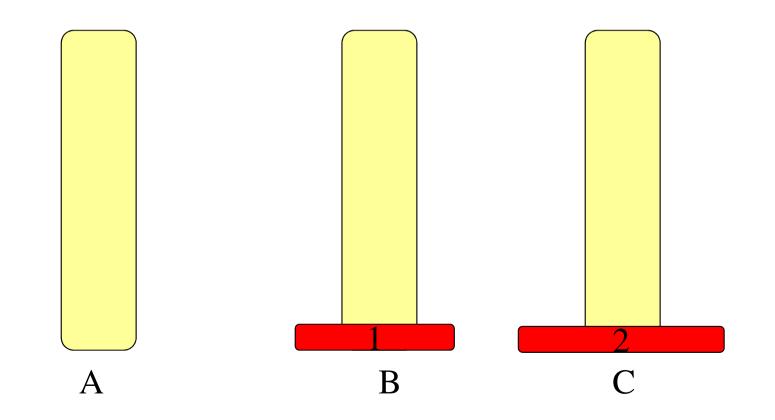


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Tower of Hanoi - Two discs

Next?

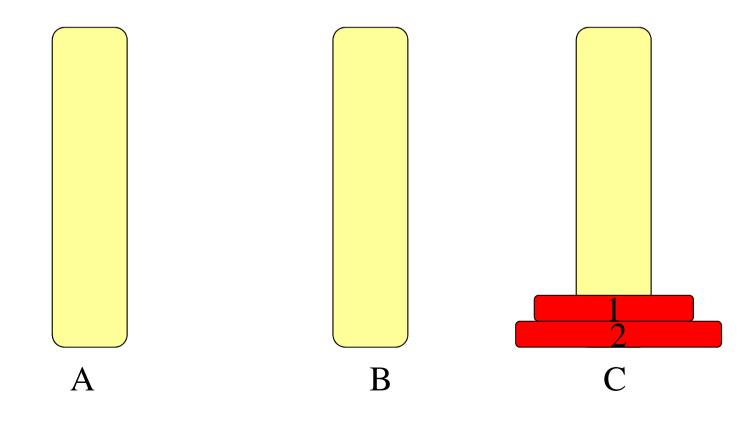
Move Disc-1 to C



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Tower of Hanoi - Two discs

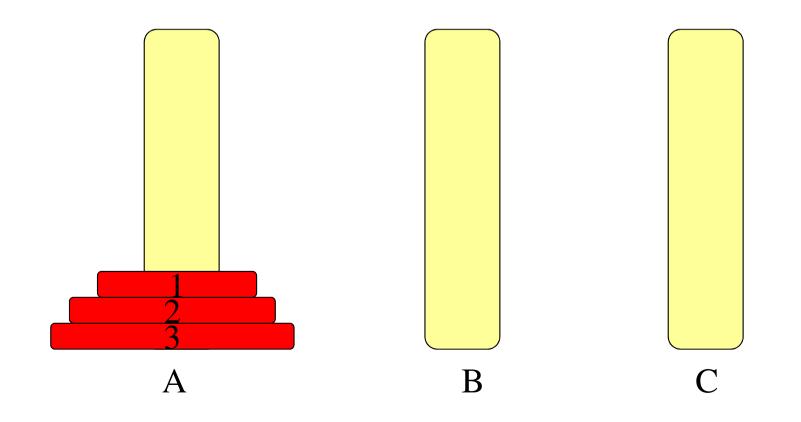
Done!



Tower of Hanoi - Three discs

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

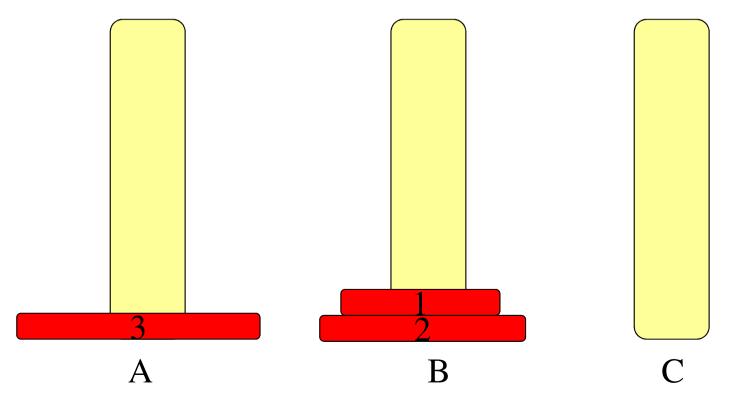
> Move Disc-1&2 to B (recursively)



Tower of Hanoi - Three discs

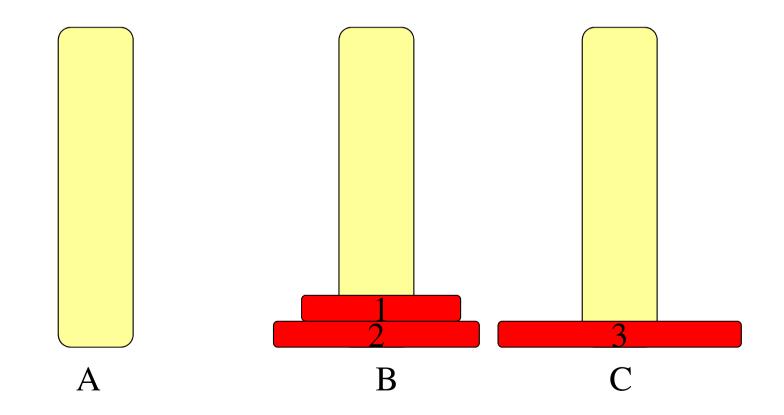
We first need to move Disc-3 to C, How?

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



Tower of Hanoi - Three discs

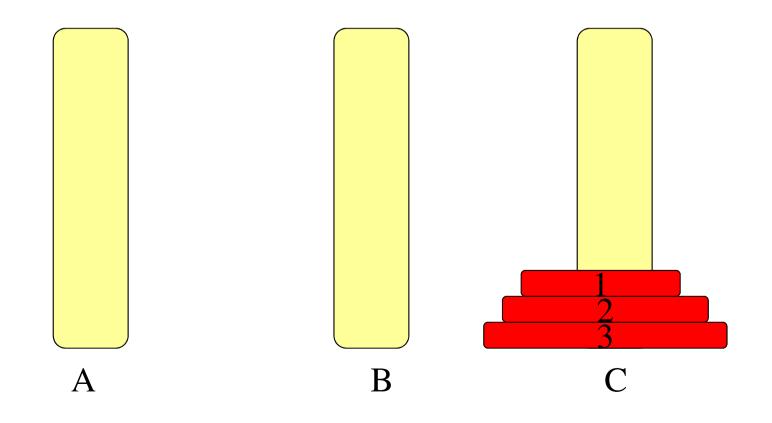
Only task left: move Disc-1&2 to C (similarly as before)



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Tower of Hanoi - Three discs

Done!



(Divide & Conquer)

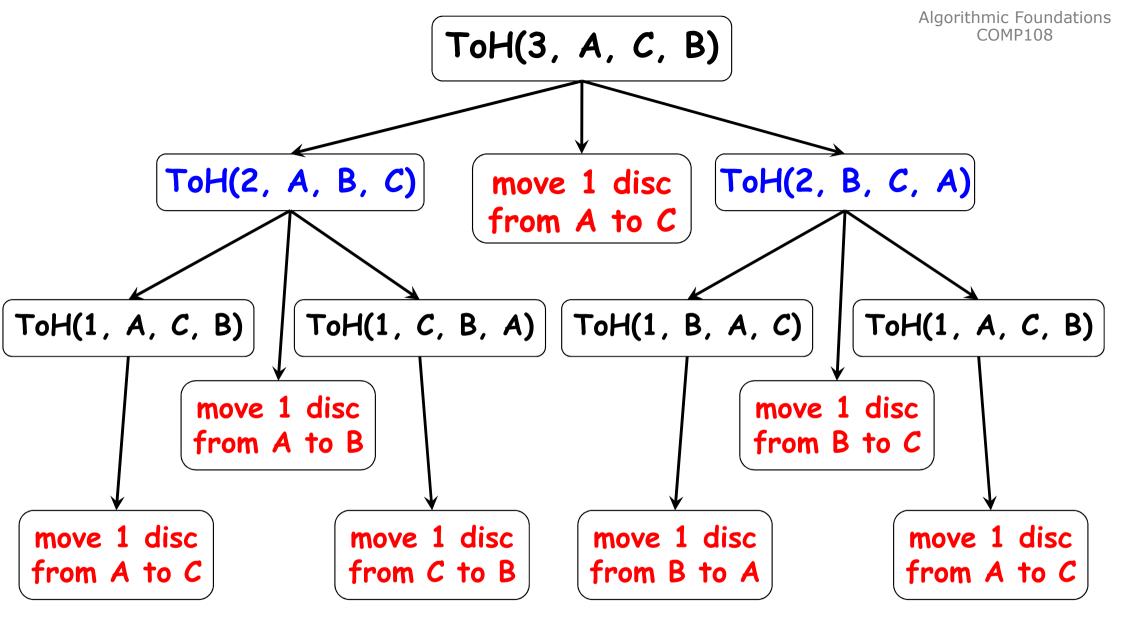
invoke by calling

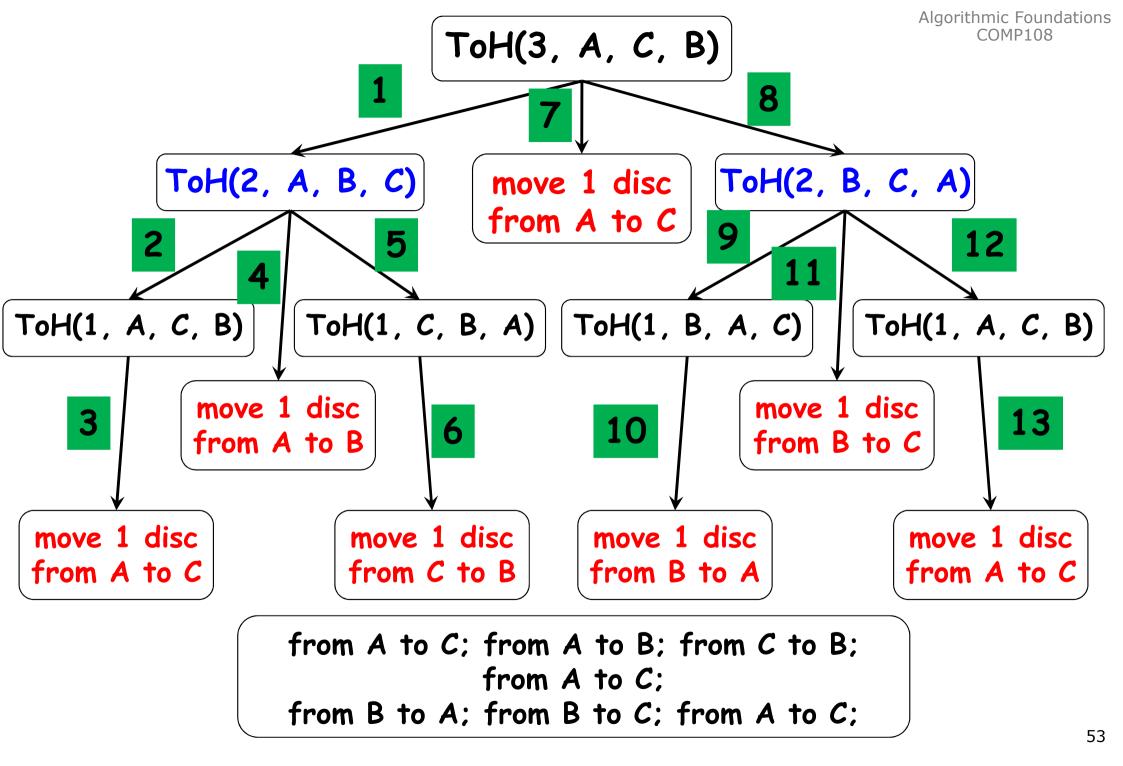
ToH(3, A, C, B)

Tower of Hanoi

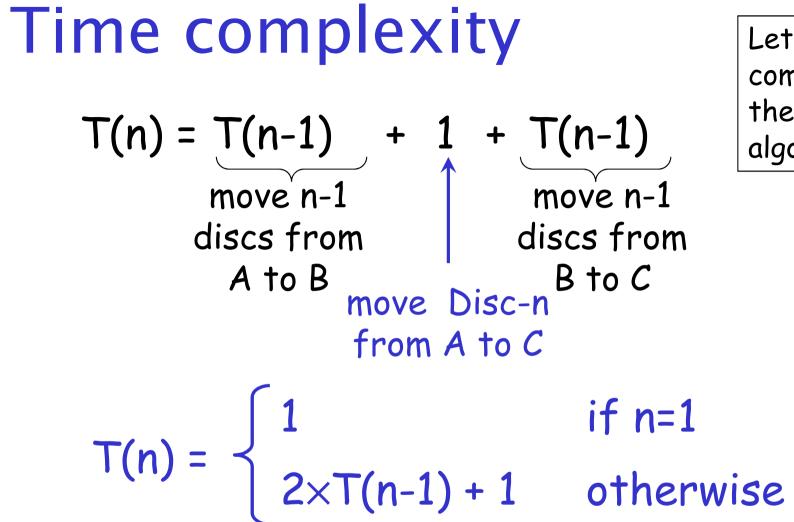
ToH(num_disc, source, dest, spare)<

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





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



Time complexity (2)

- $T(n) = 2 \times T(n-1) + 1$ = 2[2×T(n-2) + 1] + 1 $T(n) = \begin{cases} 1 & \text{if } n=1 \\ 2 \times T(n-1) + 1 & \text{otherwise} \end{cases}$
 - = $2^{2} T(n-2) + 2 + 1$ = $2^{2} [2 \times T(n-3) + 1] + 2^{1} + 2^{0}$

$$= 2^{3} T(n-3) + 2^{2} + 2^{1} + 2^{0}$$

$$= 2^{k} T(n-k) + 2^{k-1} + 2^{k-2} + ... + 2^{2} + 2^{1} + 2^{0}$$

$$= 2^{n-1}T(1) + 2^{n-2} + 2^{n-3} + ... + 2^2 + 2^1 + 2^0$$

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

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

i.e., T(n) is O(2ⁿ)

iterative method ...

...

 $= 2^{n} - 1$

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Summary - continued

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

T(n) =
$$T(\frac{n}{2})+1$$
 T(n) is $O(\log n)$

 T(n) = $2 \times T(\frac{n}{2})+1$
 T(n) is $O(n)$

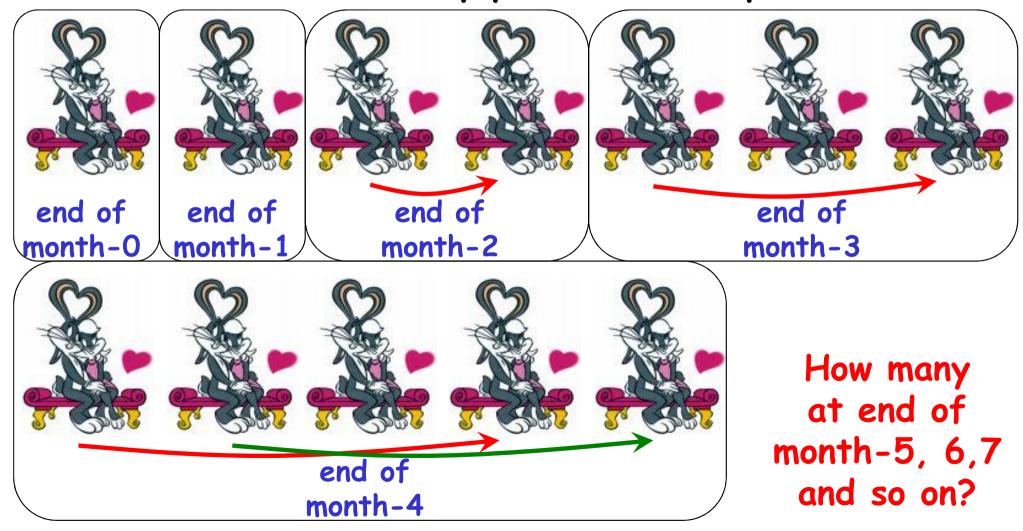
 T(n) = $2 \times T(\frac{n}{2})+n$
 T(n) is $O(n \log n)$

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

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.



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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

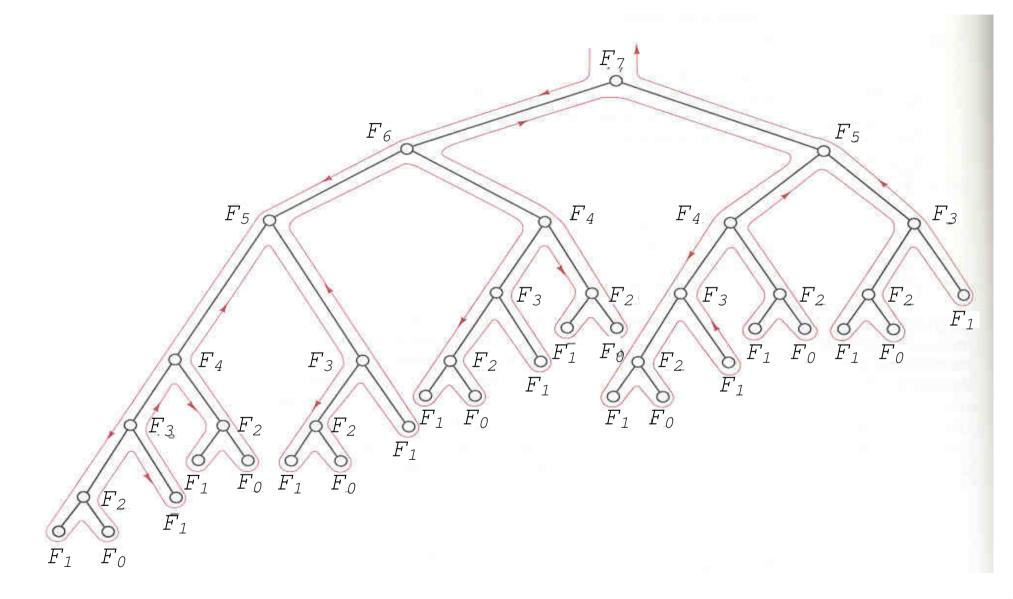
Fibonacci number F(n)

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

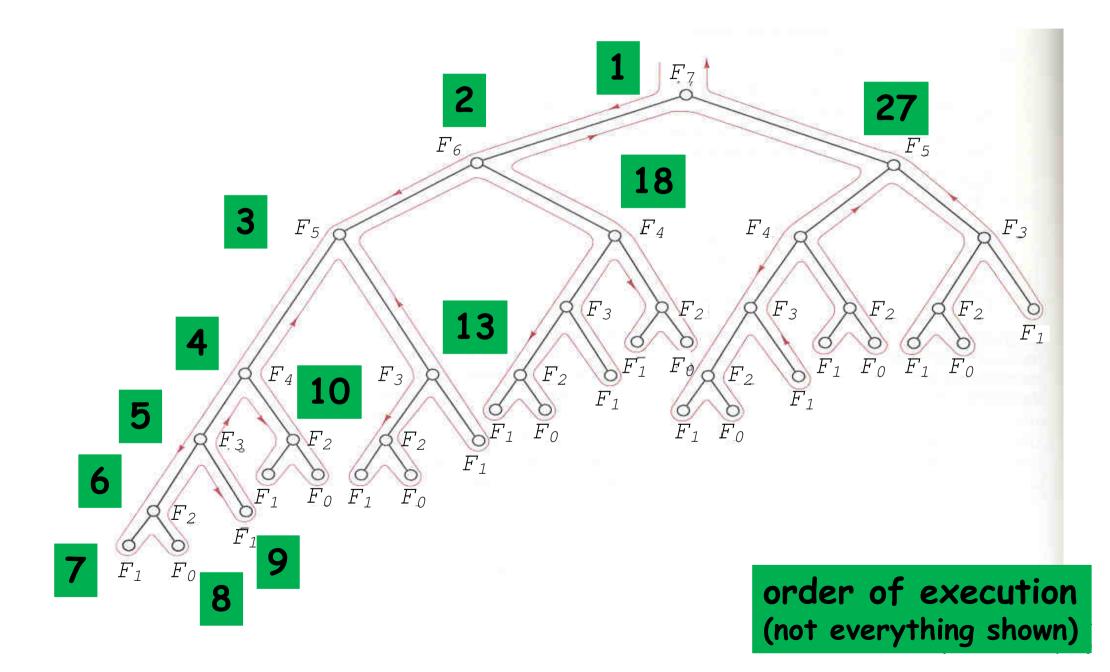
n	0	1	2	3	4	5	6	7	8	9	10
F(n)	1	1	2	3	5	8	13	21	34	55	89

```
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)
```

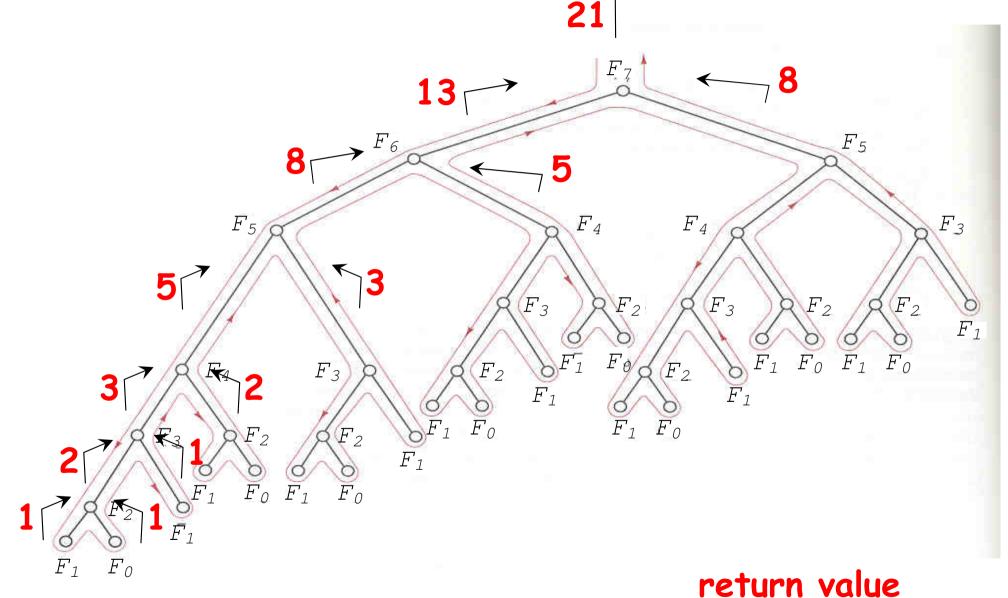
The execution of F(7)



The execution of F(7)



The execution of F(7)



(not everything shown)

Time complexity - exponential

- f(n) = f(n-1) + f(n-2) + 1
 - = [f(n-2)+f(n-3)+1] + f(n-2) + 1
- Suppose f(n) denote the time complexity to compute F(n)

- > 2 <u>f(n-2)</u>
- > 2 [2×f(n-2-2)] = 2² f(n-4)
- > $2^{2} [2 \times f(n-4-2)] = 2^{3} f(n-6)$
- > 2³ [2×f(n-6-2)] = 2⁴ f(n-8)

...

If n is even, $f(n) > 2^{n/2} f(0) = 2^{n/2}$ If n is odd, $f(n) > f(n-1) > 2^{(n-1)/2}$

exponential in n

(Dynamic Programming)

Robozzle - Recursion

Task:

to program a robot to pick up all stars in a certain area

Command: Go straight, Turn Left, Turn Right

