Data Structures in Java

Lecture 4: Introduction to Algorithm Analysis and Recursion

9/21/2015



Algorithms

- An algorithm is a clearly specified set of simple instructions to be followed to solve a problem.
- Algorithm Analysis Questions:
 - Does the algorithm terminate?
 - Does the algorithm solve the problem? (correctness)
 - What resources does the algorithm use?
 - Time / Space

Analyzing Runtime: Basics

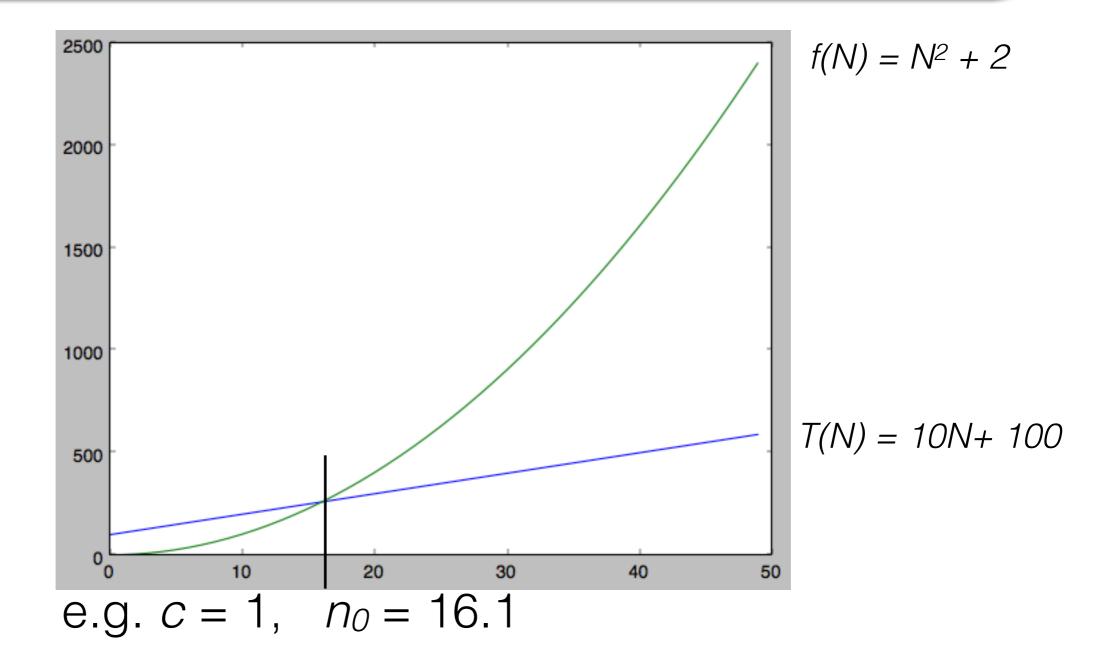
- We usually want to compare several algorithms.
 - Compare between different algorithms how the runtime *T(N)* grows with increasing input sizes *N*.
- We are using Java, but the same algorithms could be implemented in any language on any machine.
- How many basic operations/"steps" does the algorithm take? All operations assumed to have the same time.

Worst and Average case

- Usually the runtime depends on the type of input (e.g. sorting is easy if the input is already sorted).
- *T_{worst}(N): worst case* runtime for the algorithm on ANY input. The algorithm is **at least** this fast.
- T_{average}(N): Average case analysis expected runtime on typical input.
- T_{best}(N): Occasionally we are interested in the best case analysis.

Comparing Function Growth: Big-Oh Notation

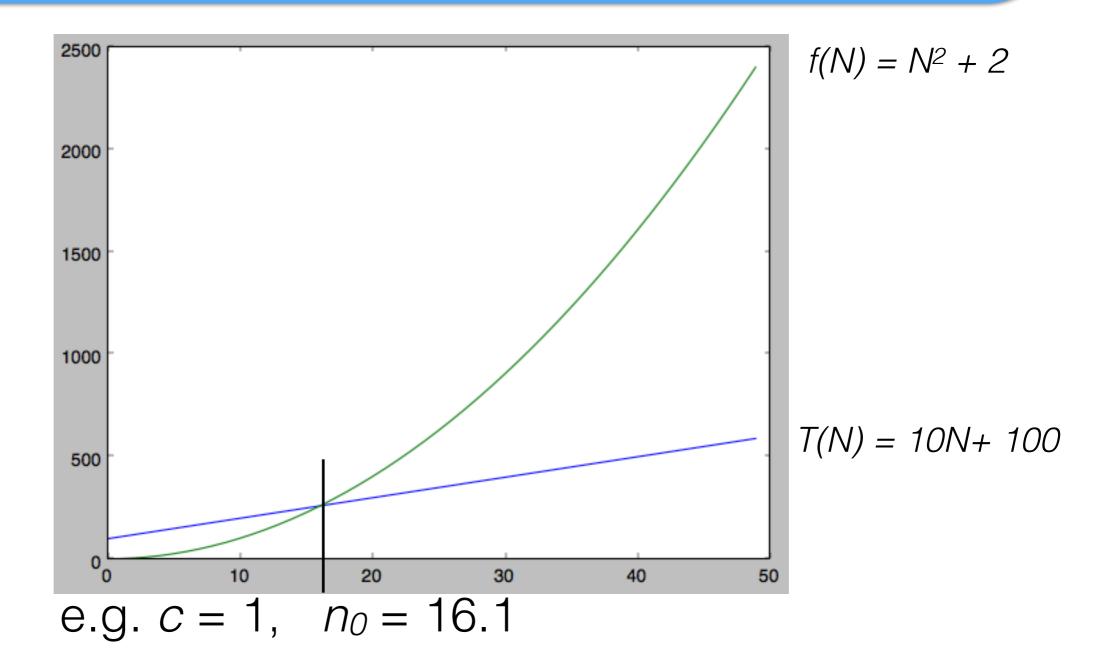
T(N) = O(f(N)) if there are positive constants c and n_0 such that $T(N) \leq cf(N)$ when $N \geq n_0$.



Comparing Function Growth: Big-Oh Notation

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"T(N) is in the order of f(N)"

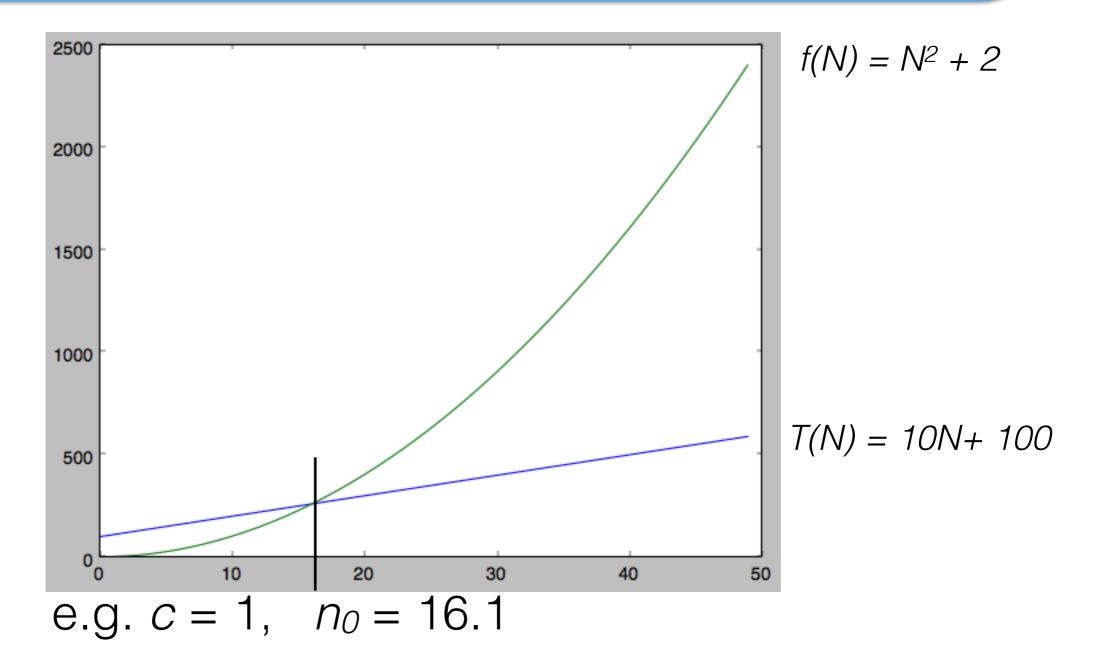


Comparing Function Growth: Big-Oh Notation

T(N) = O(f(N)) if there are positive constants c and n_0 such that $T(N) \leq cf(N)$ when $N \geq n_0$.

"T(N) is in the order of f(N)"

"f(N) is an upper bound on T(N)"



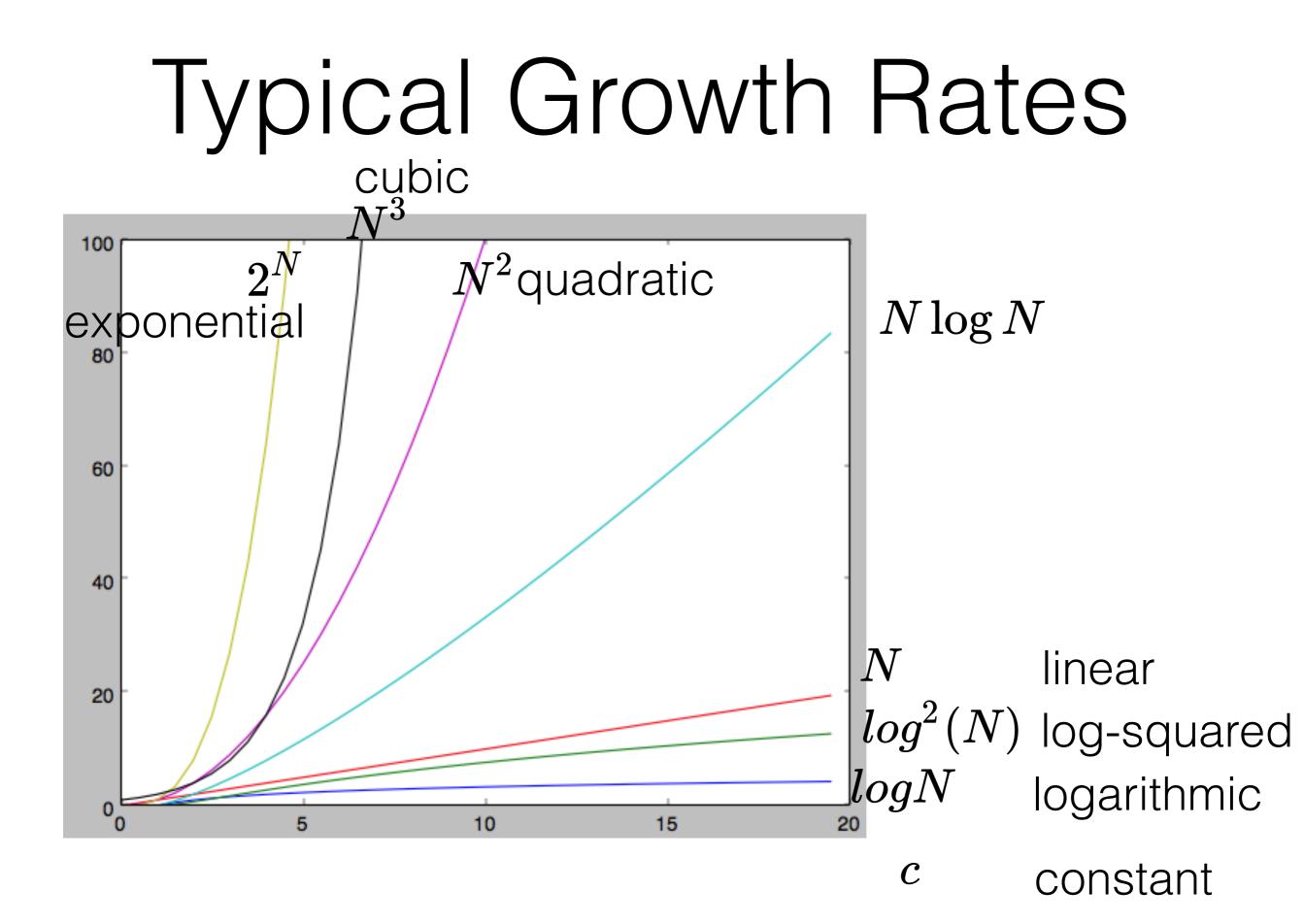
Comparing Function Growth: Additional Notations

• Lower Bound:

 $T(N) = \Omega(f(N))$ if there are positive constants c and n_0 such that $T(N) \ge cf(N)$ when $N \ge n_0$.

- Tight Bound: T(N) and f(N) grow at the same rate $T(N) = \Theta(f(N))$ if $T(N) = \Omega(f(N))$ and T(N) = O(f(N))
- Strict Upper Bound:

T(N) = o(f(N)) if for all positive constants c there is some n_0 such that T(N) < cf(N) when $N > n_0$.



Rules for Big-Oh (1)

If $T_1(N) = O(f(N))$ and $T_2(N) = O(g(N))$ then

1. $T_1(N) + T_2(N) = O(f(N) + g(N))$ $O(\max{(f(N), g(N))})$

2. $T_1(N) * T_2(N) = O(f(N) * g(N))$

Rules for Big-Oh (2)

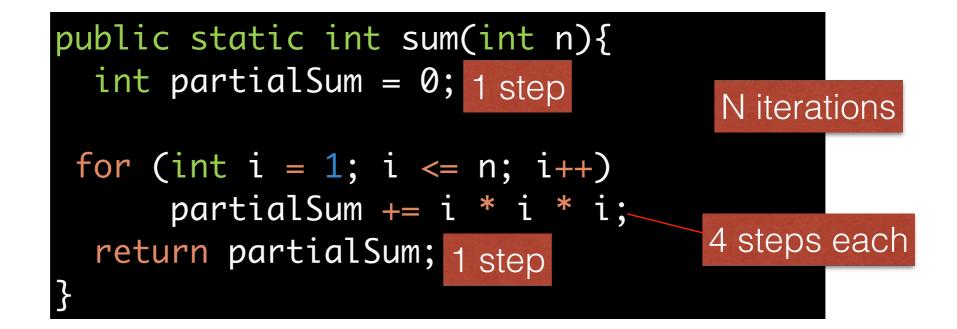
If T(N) is a polynomial of degree k then $T(N) = \Theta(N^k)$

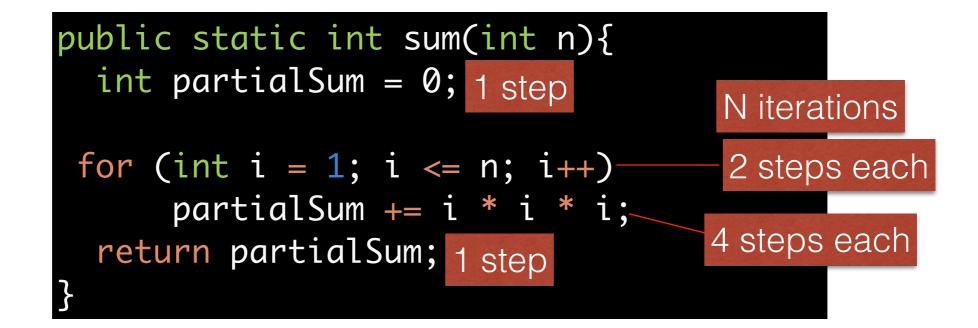
For instance:
$$9N^3 + 12N^2 - 5 = \Theta(N^3)$$

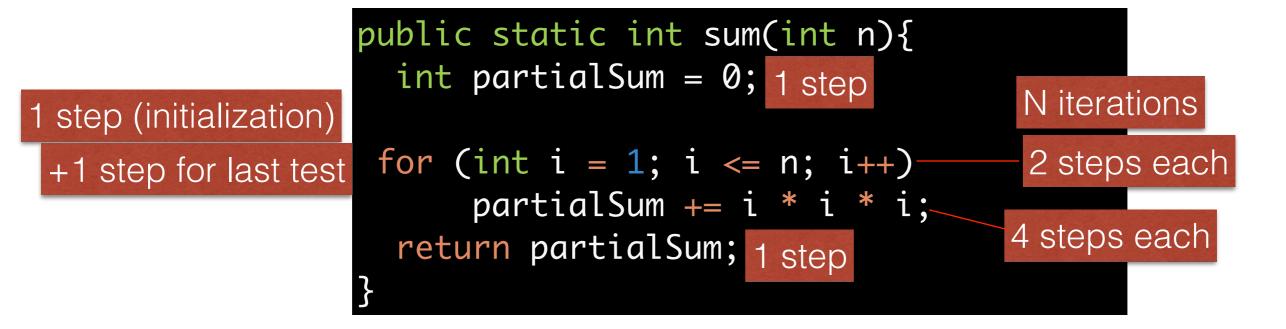
$$log^k(N) = O(N)$$
 for any k .

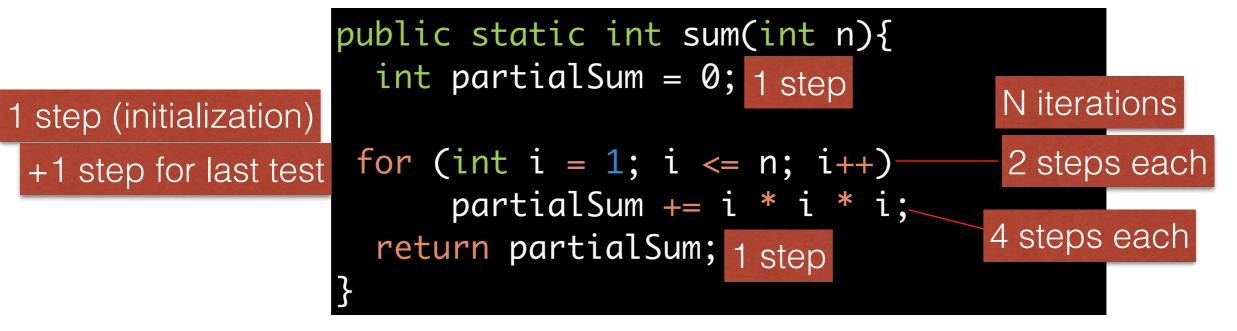
public static int sum(int n){
 int partialSum = 0;
 for (int i = 1; i <= n; i++)
 partialSum += i * i * i;
 return partialSum;
}</pre>

public static int sum(int n){
 int partialSum = 0; 1 step
 for (int i = 1; i <= n; i++)
 partialSum += i * i * i;
 return partialSum; 1 step
}</pre>

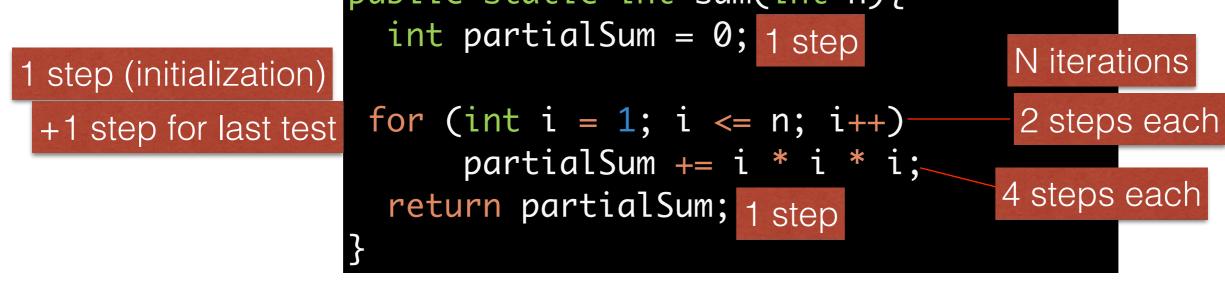








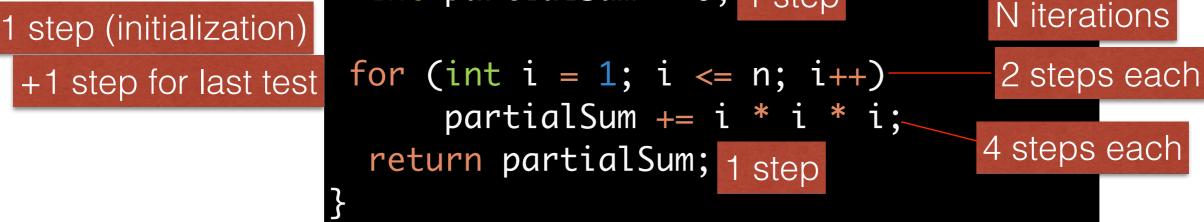
T(N) = 6 N + 4 = O(N)



T(N) = 6 N + 4 = O(N)

(running time of statements in the loop) X (iterations)

General Rules: Basic for-loops Compute $\sum_{i=1}^{N} i^{3}$ public static int sum(int n){ int partialSum = 0; 1 step



T(N) = 6 N + 4 = O(N)

(running time of statements in the loop) X (iterations)

If loop runs a constant number of times: O(c)





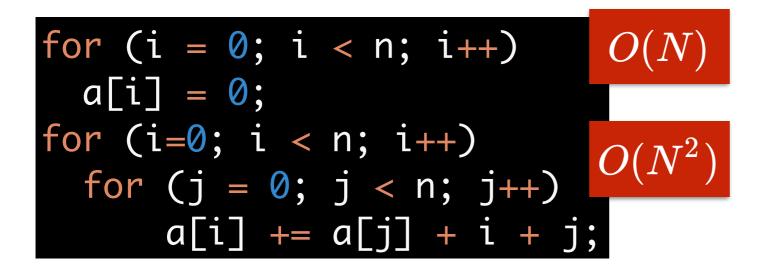
N iterations
$$O(N) * O(N) = O(N^2)$$

N iterations $O(N)$
1 step each $O(c)$

General Rules: Consecutive Statements

for (i = 0; i < n; i++) a[i] = 0; for (i=0; i < n; i++) for (j = ∅; j < n; j++)</pre> a[i] += a[j] + i + j;

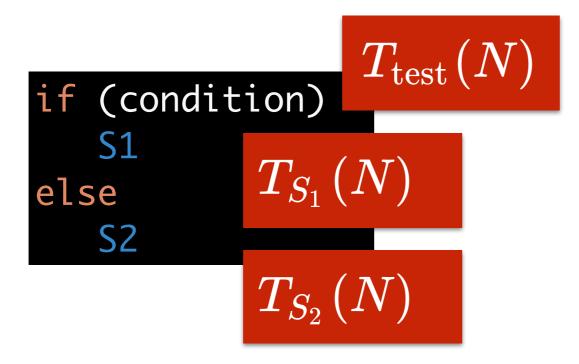
General Rules: Consecutive Statements



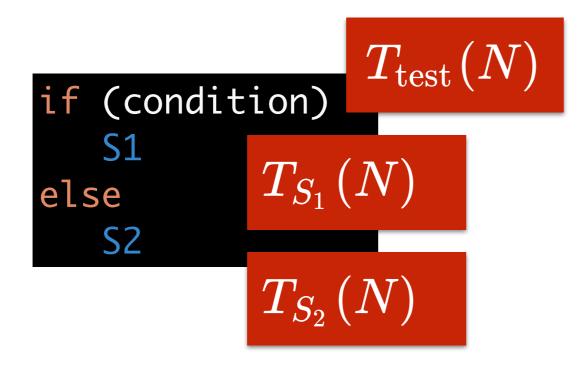
General Rules: Consecutive Statements

 $O(N) + O(N^2) = O(N^2)$

Basic Rules: *if/else* conditionals



Basic Rules: *if/else* conditionals



$T(N)=O(\max(T_{S_1}(N),T_{S_2}(N))+T_{ ext{test}}(N))$

Logarithms in the Runtime

```
public static int binarySearch(int[] a, int x) {
  int low = 0;
  int high = a.length - 1;
  while ( low <= high) {</pre>
    int mid = (low + high) / 2;
    if (a[mid] < x)
       low = mid + 1;
     else if(a[mid] > x)
       high = mid - 1;
     else
       return mid; // found
  }
  return -1; // Not found.
```

How many iterations of the *while* loop? Every iteration cuts remaining partition in half.

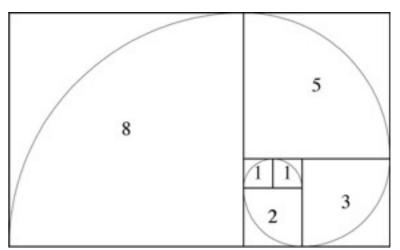
Recursion

- A recursive algorithm uses a function (or method) that calls itself.
- Need to make sure there is some base case (otherwise causing an infinite loop).
- The recursive call needs to *make progress* towards the base case.
 - Reduces the problem to a simpler subproblem.

Recursive Binary Search

• 1, 1, 2, 3, 5, 8, 13, 21, ...

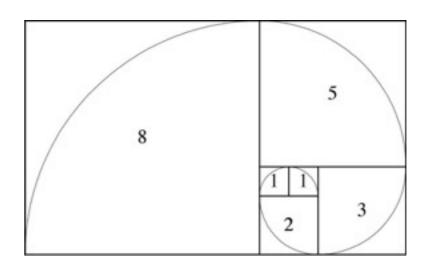
• 1, 1, 2, 3, 5, 8, 13, 21, ...



Recursive Definition

 $F_{k+1} = F_k + F_{k-1}$

• 1, 1, 2, 3, 5, 8, 13, 21, ...



- Closed form solution is complicated.
- Instead easier to compute this algorithmically.

Fibonacci Sequence in Java

public class Fibonacci {

```
public static void main(String[] args) {
  Fibonacci fib = new Fibonacci();
  int k = Integer.parseInt(args[0]);
  System.out.println(fib.fibonacci(k));
}
public int fibonacci(int k) throws IllegalArgumentException{
  if (k < 1) {
    throw new IllegalArgumentException("Expecting a positive integer.");
  }
  if (k == 1 | k == 2) {
    return 1;
 } else {
    return fibonacci(k-1) + fibonacci(k-2);
```

Fibonacci in Java

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 if (k == 1 | k == 2) { Base case
   return 1;
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Fibonacci in Java

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                         Recursive call - making progress
```

How many steps does the algorithm need?

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```

```
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    }
    if (k == 1 | k == 2) { Base case: 1 step T(1) = O(c), T(2) = O(c)
        return 1;
    } else {
        return fibonacci(k-1) + fibonacci(k-2);
        }
    }
}</pre>
```

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```

```
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```

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    }
    if (k == 1 | k == 2) {
        Base case: 1 step T(1) = O(c), T(2) = O(c)
        return 1;
    } else {
        return fibonacci(k-1) + fibonacci(k-2);
        }
        Recursive calls: T(k) = O(T(k-1) + T(k-2))
```

Analyzing the Recursive Fibonacci Solution

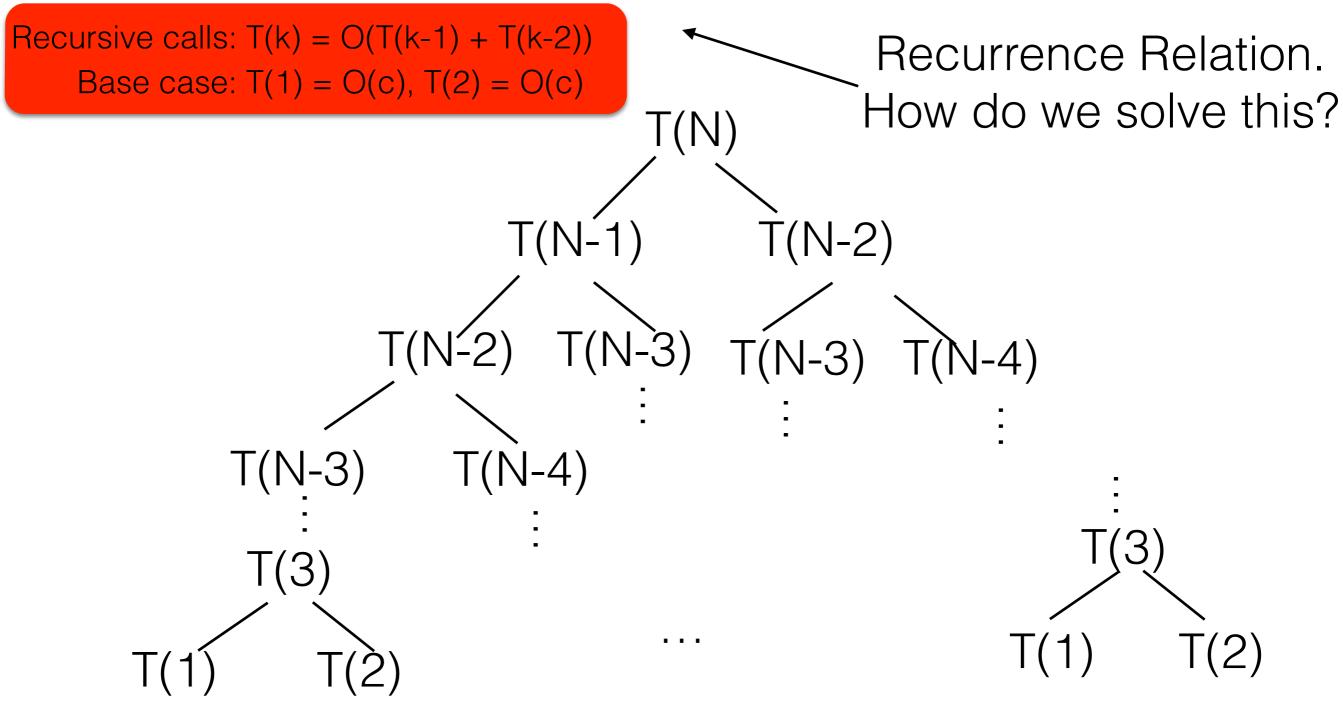
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Analyzing the Recursive Fibonacci Solution

Recursive calls: T(k) = O(T(k-1) + T(k-2))Base case: T(1) = O(c), T(2) = O(c)

Recurrence Relation. How do we solve this?

Analyzing the Recursive Fibonacci Solution



Fibonacci Sequence v.2

public int fibonacci(int k) throws IllegalArgumentException{

```
if (k < 1) {
   throw new IllegalArgumentException("Expecting a positive integer.");
}
int b = 1; //k-2
int a = 1; //k-1
for (int i=3; i<=k; i++) {
   int new_fib = a + b;
   b = a;
   a = new_fib;
}
return a;</pre>
```

Dynamic programming: Cache intermediate solutions so they can be re-used.

Fibonacci Sequence v.2

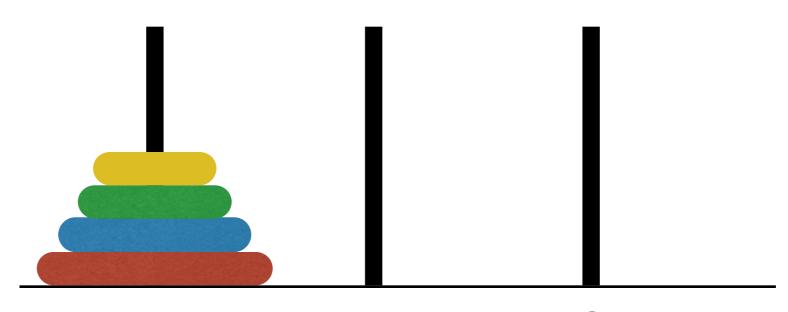
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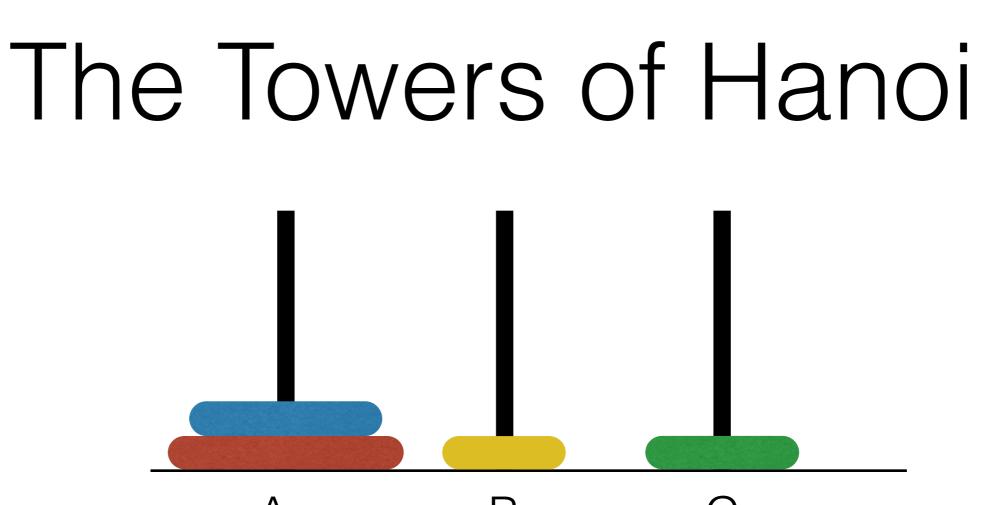
```
if (k < 1) {
   throw new IllegalArgumentException("Expecting a positive integer.");
}
int b = 1; //k-2
int a = 1; //k-1
for (int i=3; i<=k; i++) {
   int new_fib = a + b;
   b = a;
   a = new_fib;
}
return a;</pre>
T(N) = O(N)
```

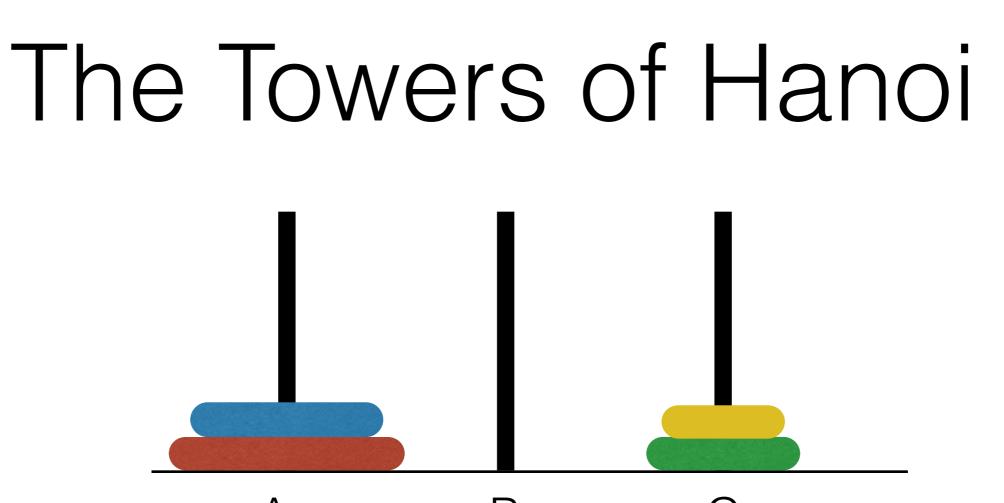
Dynamic programming: Cache intermediate solutions so they can be re-used.

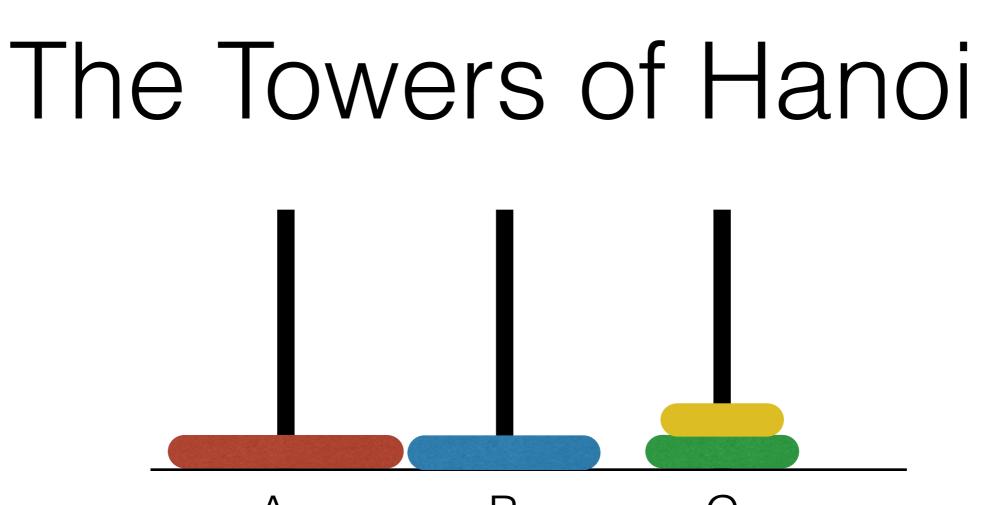
Rules for Recursion

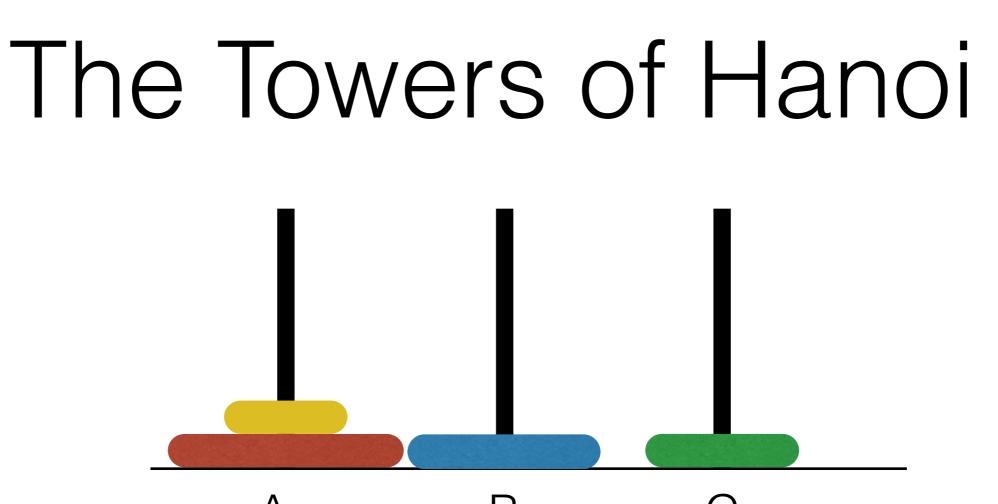
- 1. Base Case
- 2. Making Progress
- 3. Design Rule Assume all recursive calls work.
- *4. Compound Interest Rules -* Never duplicate work by solving the same instance of a problem in separate recursive calls.

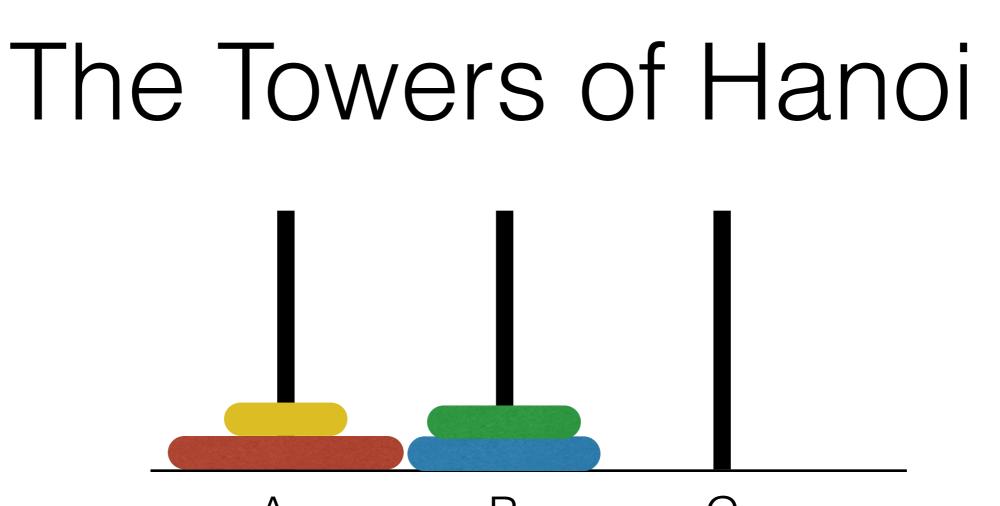


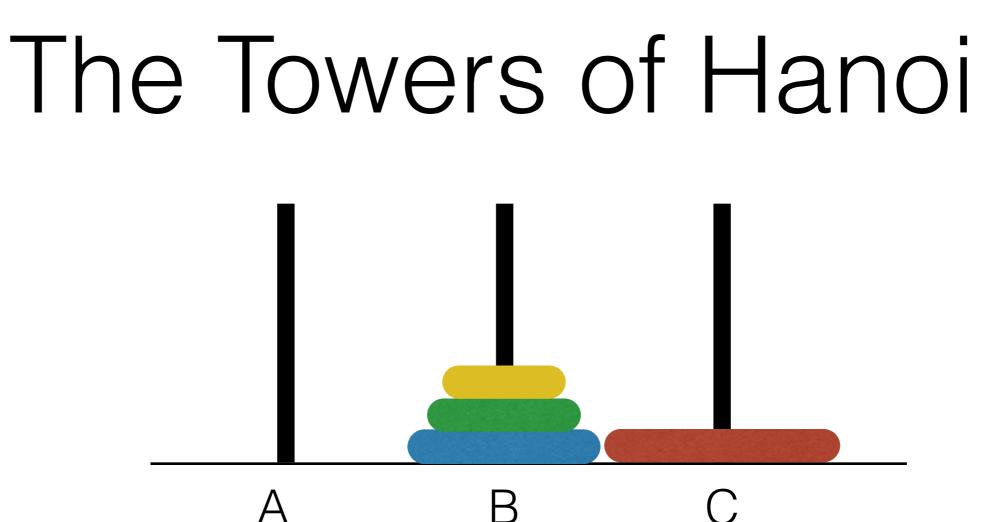


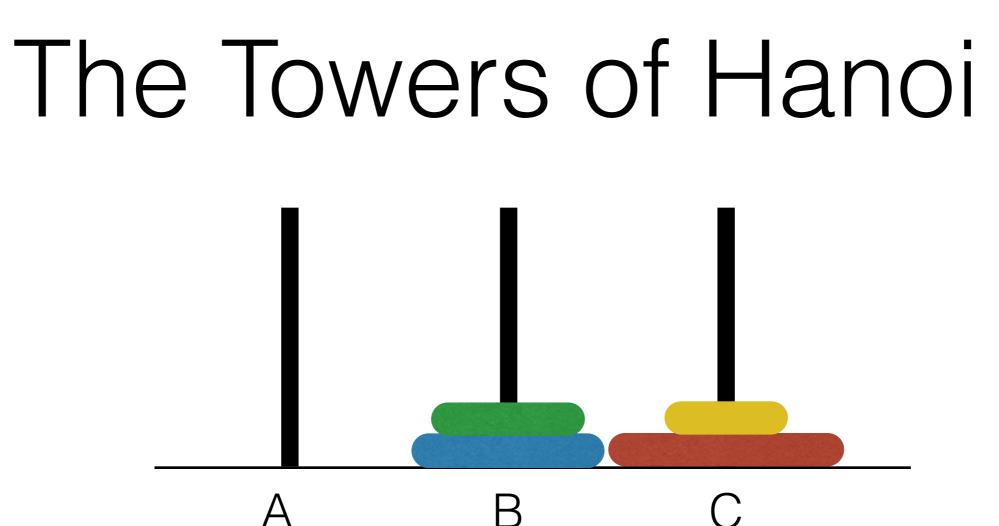


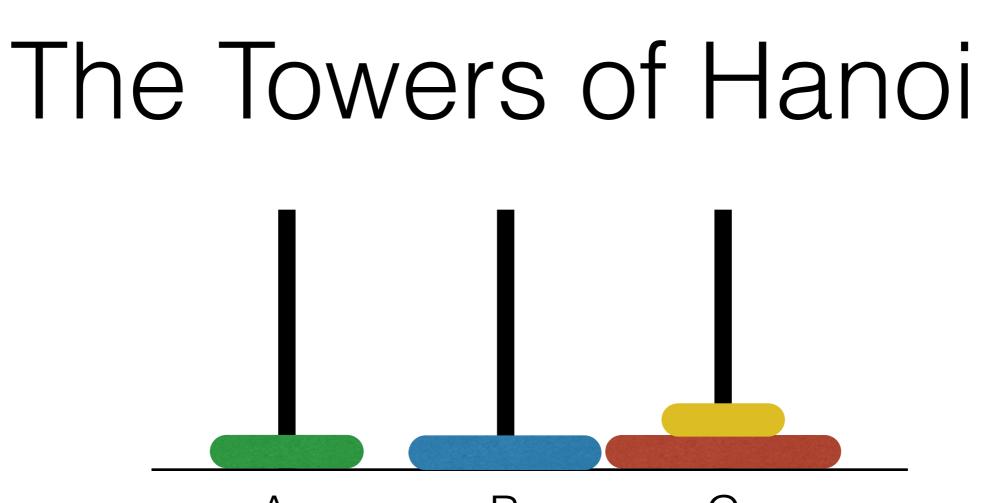


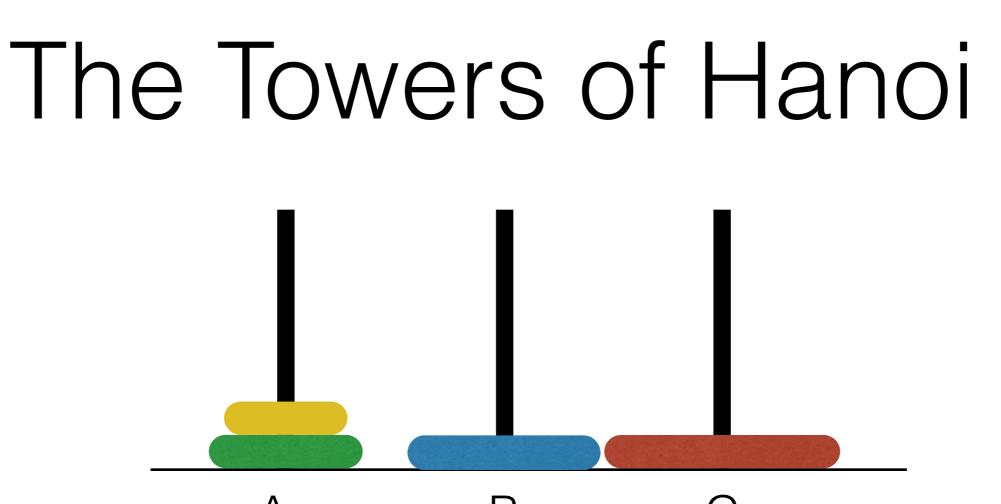


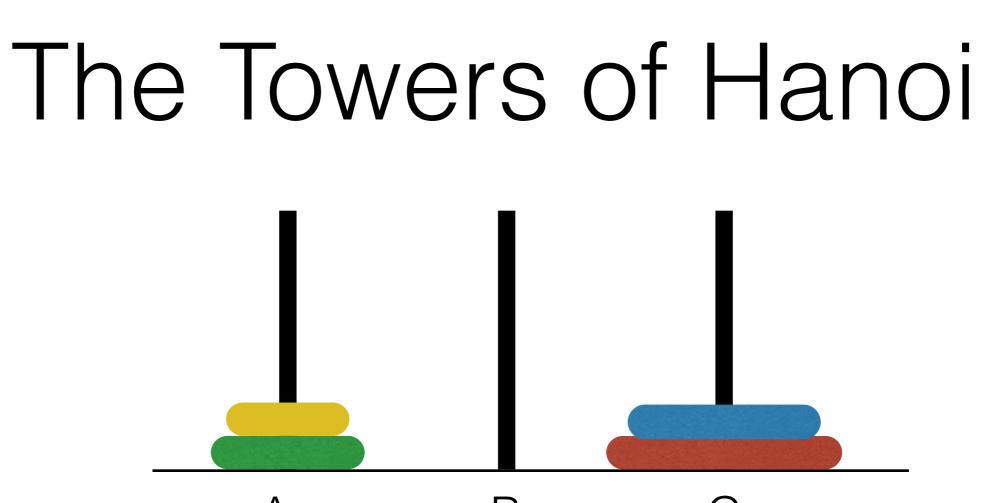


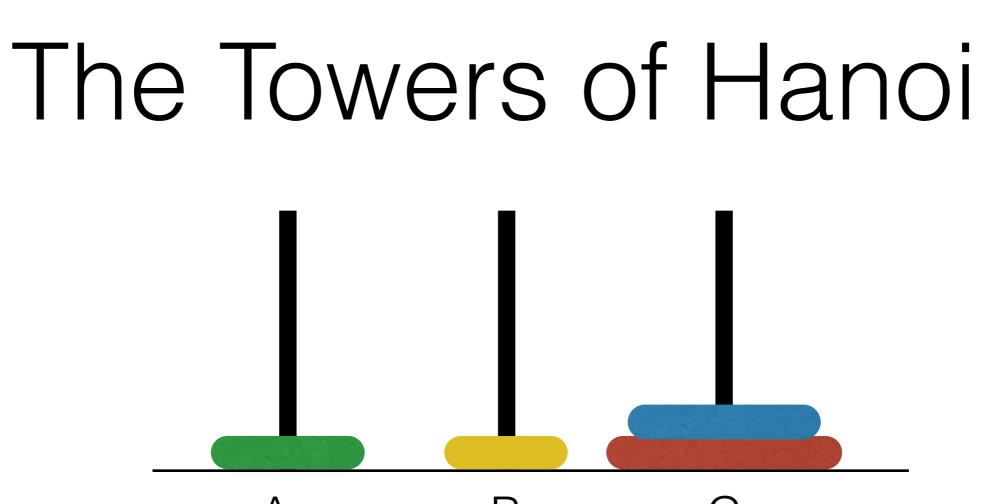


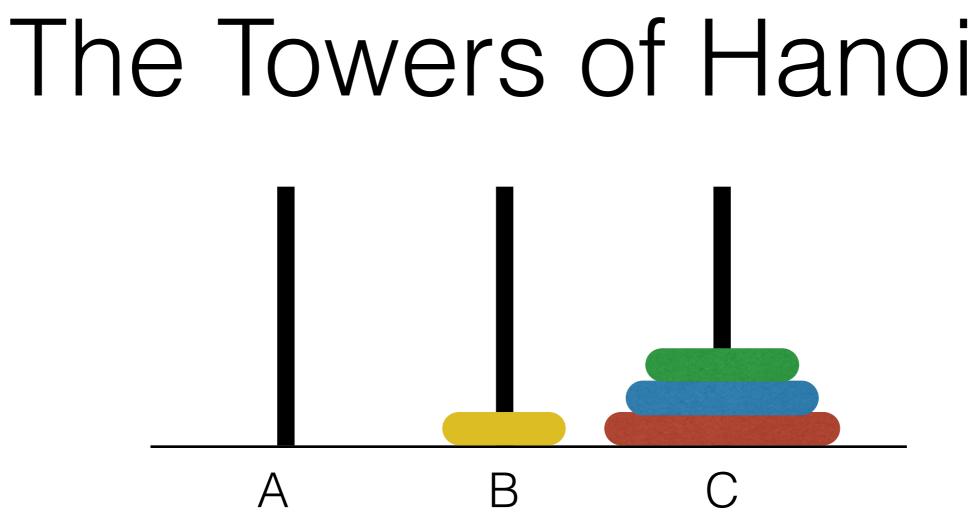


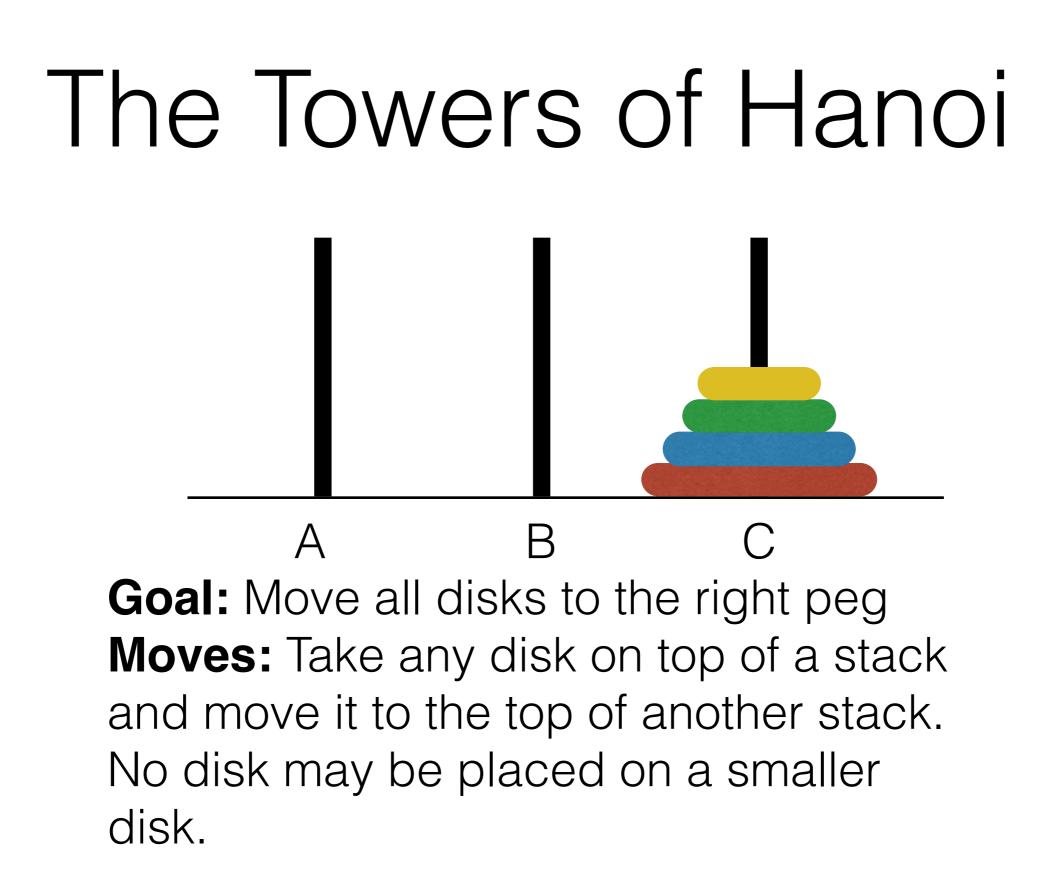


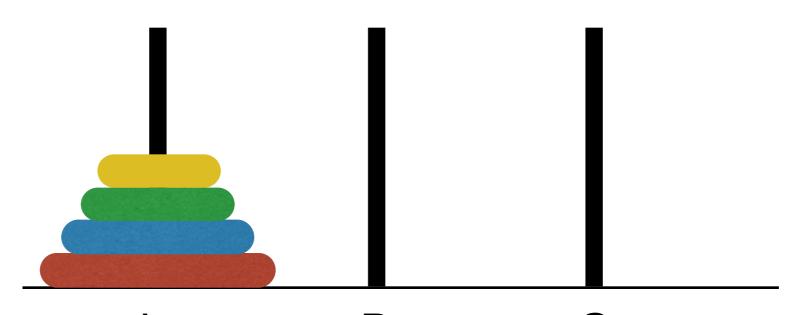










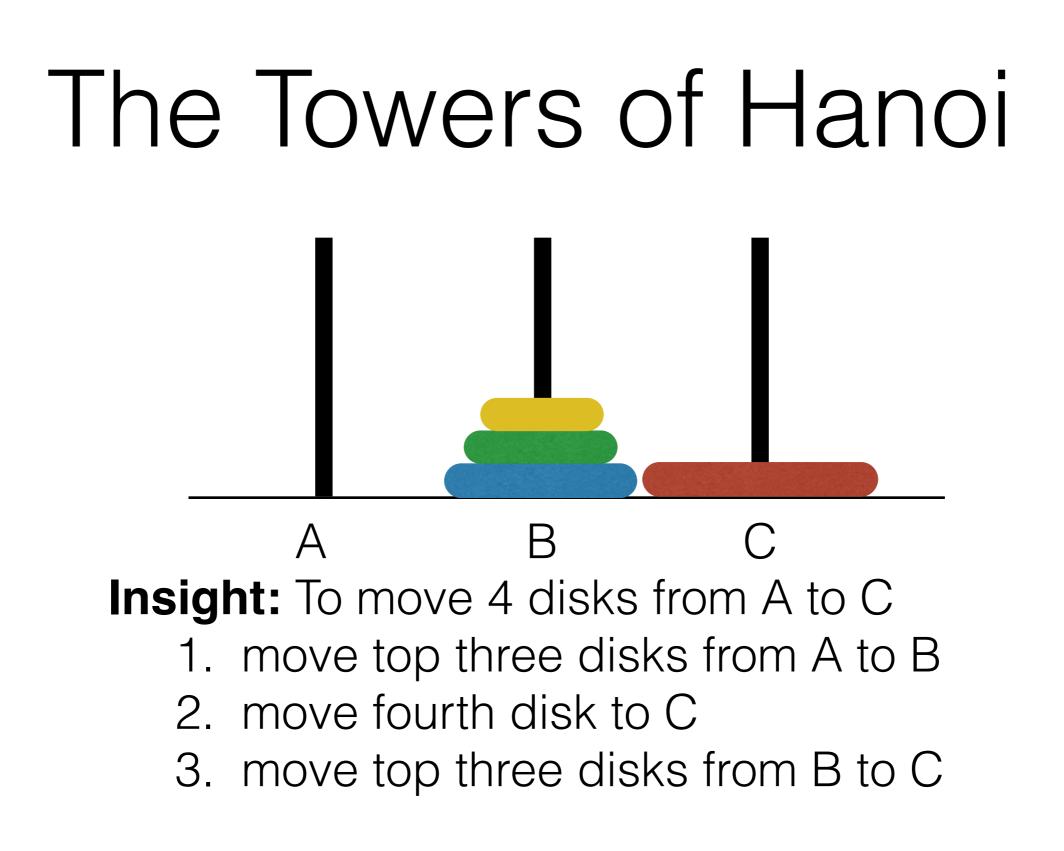


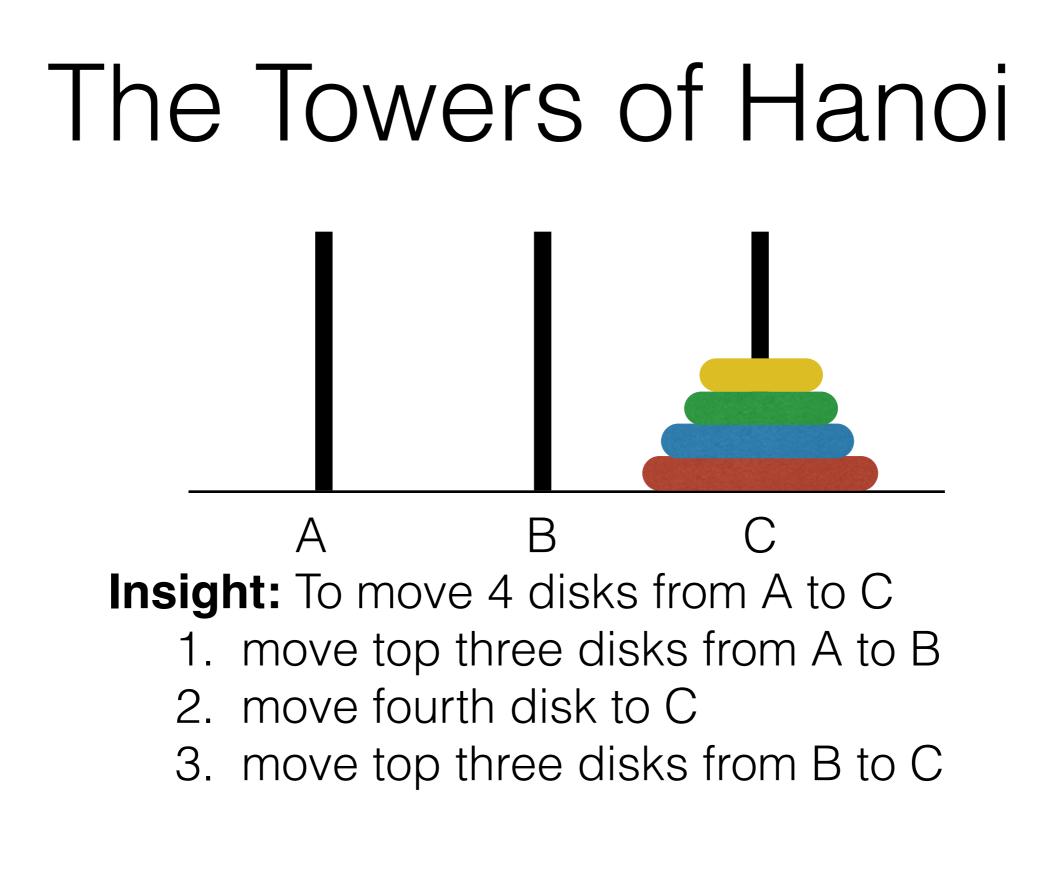
A B C Insight: To move 4 disks from A to C

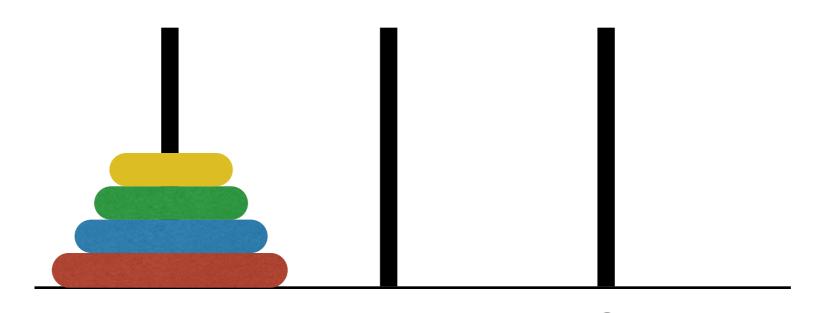
- 1. move top three disks from A to B
- 2. move fourth disk to C
- 3. move top three disks from B to C

A B C Insight: To move 4 disks from A to C 1. move top three disks from A to B 2. move fourth disk to C

3. move top three disks from B to C





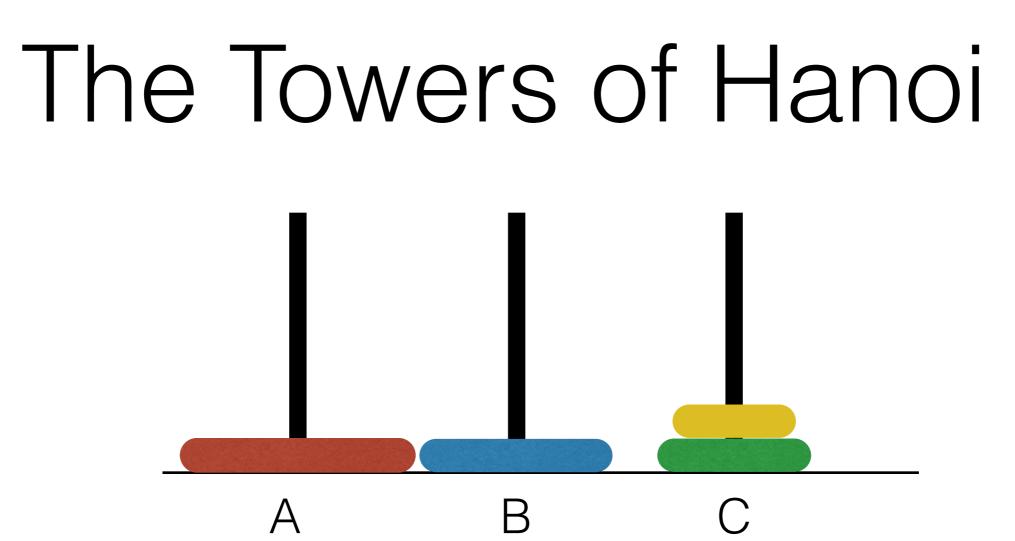


A B C Insight: To move 3 disks from A to B

- 1. move top two disks from A to C
- 2. move third disk to B
- 3. move top two disks from C to B

Insight: To move 3 disks from A to B

- 1. move top two disks from A to C
- 2. move third disk to B
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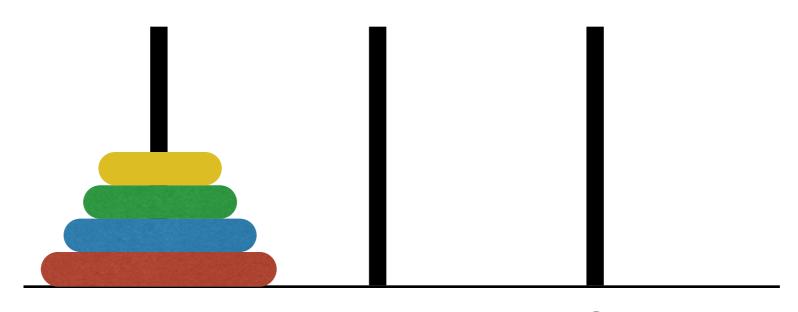


Insight: To move 3 disks from A to B

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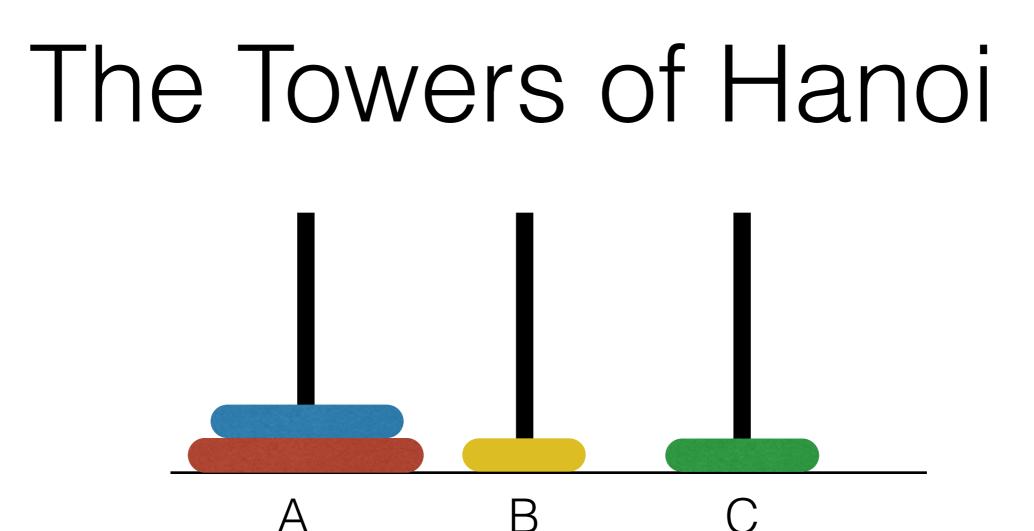


A B C Insight: To move 2 disks from A to C

- 1. move top one disks from A to B
- 2. move third disk to C
- 3. move top one disks from B to C

A B C Insight: To move 2 disks from A to C

- 1. move top one disks from A to B
- 2. move third disk to C
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Insight: To move 2 disks from A to C

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Insight: To move 2 disks from A to C

- 1. move top one disks from A to B
- 2. move third disk to C
- 3. move top one disks from B to C

Algorithm (sketch)

To move *n* disks from A to C 1. move top *n-1* disks from A to B 2. move *n*-th to C 3. move top *n-1* disks from B to C

- A = source peg
- C = target peg
- B = "help" peg (to temporarily store disks)

Peg labels change in each recursive call.

To move *n* disks from A to C

- 1. move top *n*-1 disks from A to B
- 2. move *n*-th to C
- 3. move top *n*-1 disks from B to C

$$T(N) = 2 \cdot T(N-1) + 1$$

$$T(1) = 1$$

Need to solve this recurrence relation!