

CSI 2103: Data Structures

Recursion (Ch 4)

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Aims

- Various operations operate **repeatedly** on data structures
 - Loops
- Discuss **recursion**
- Examples
- Time complexity

Recursion

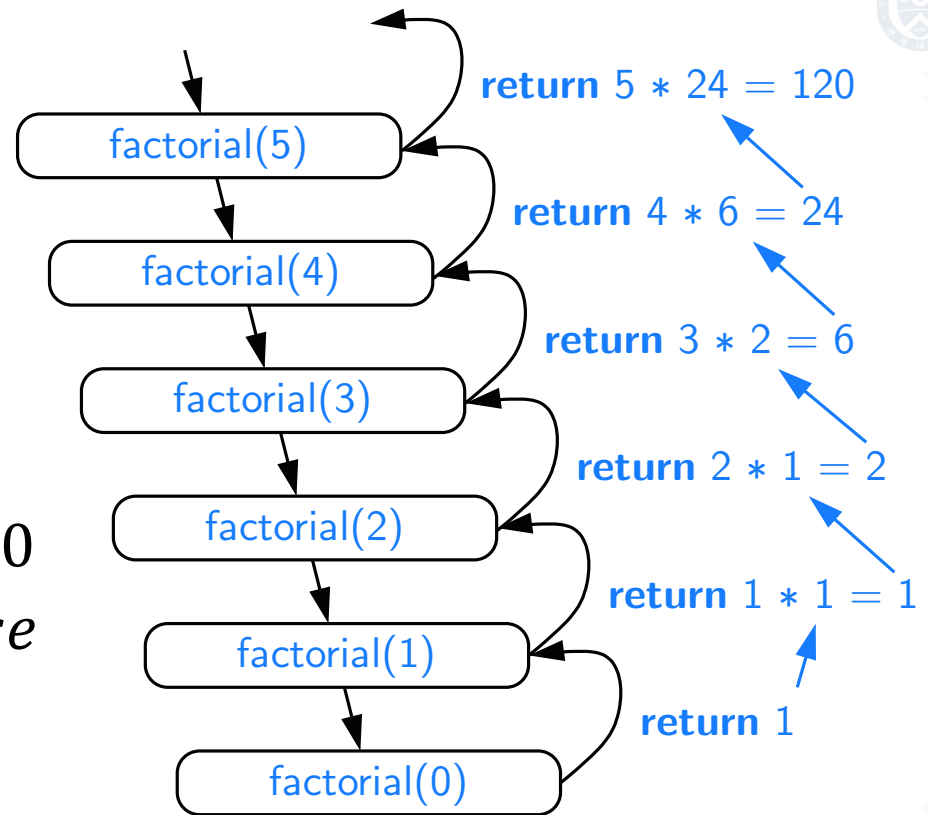
- A recursive method which calls itself recursively
 - **Base case**: when to stop the recursion
 - **Recursive calls**: Calls itself to reach the base case
- Factorial $f(n) = n!$:

$$f(n) = \begin{cases} 1 & \text{if } n = 0 \\ n \cdot f(n - 1) & \text{else} \end{cases}$$

```
1 def factorial(n):
2     if n == 0:
3         return 1
4     else:
5         return n * factorial(n-1)
```

Factorial

$$f(n) = \begin{cases} 1 & \text{if } n = 0 \\ n \cdot f(n - 1) & \text{else} \end{cases}$$



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Recursion

- A recursive method which calls itself recursively
- Factorial $f(n) = n!$:

$$f(n) = \begin{cases} 1 & \text{if } n = 0 \\ n \cdot f(n - 1) & \text{else} \end{cases}$$

- Fibonacci sequence:

$$f(n) = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ f(n - 1) + f(n - 2) & \text{else} \end{cases}$$

Binary Search

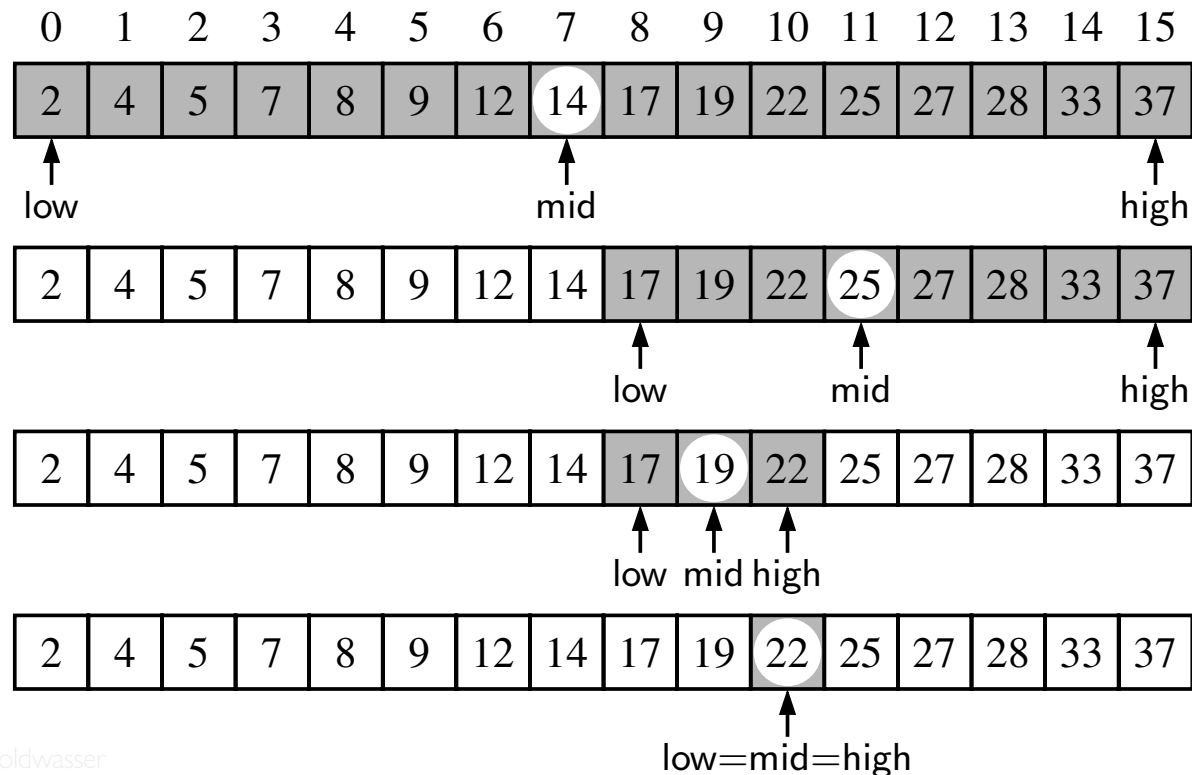
- Search for an integer in a sorted list
 - If unsorted, we need the sequential search $O(n)$
- When the sequence is **sorted** and **indexable**
 - $i < j \Rightarrow A[i] \leq A[j]$
 - Which also means for any index i ,
 - All values at indices $0, \dots, i - 1 \leq$ the value at i
 - All values at indices $i + 1, \dots, n - 1 \geq$ the value at i

| | | | | | | | | | | | | | | | |
|---|---|---|---|---|---|----|----|----|----|----|----|----|----|----|----|
| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 |
| 2 | 4 | 5 | 7 | 8 | 9 | 12 | 14 | 17 | 19 | 22 | 25 | 27 | 28 | 33 | 37 |

Binary Search

$$\text{mid} = \lfloor (\text{low} + \text{high}) / 2 \rfloor$$

- Three cases:
 - If $\text{target} = \text{data}[\text{mid}]$, then we have found the target
 - If $\text{target} < \text{data}[\text{mid}]$, then we recur on the first half (low to mid-1)
 - If $\text{target} > \text{data}[\text{mid}]$, then we recur on the second half (mid+1 to high)



Binary Search

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```
1 def binary_search(data, target, low, high):
2     """Return True if target is found in indicated portion of a Python list.
3
4     The search only considers the portion from data[low] to data[high] inclusive.
5     """
6     if low > high:
7         return False                                # interval is empty; no match
8     else:
9         mid = (low + high) // 2
10        if target == data[mid]:                      # found a match
11            return True
12        elif target < data[mid]:
13            # recur on the portion left of the middle
14            return binary_search(data, target, low, mid - 1)
15        else:
16            # recur on the portion right of the middle
17            return binary_search(data, target, mid + 1, high)
```


Binary Search

$$\text{mid} = \lfloor (\text{low} + \text{high}) / 2 \rfloor$$

- Why this complicated way of searching?
 - Because it's super fast!
- Each recursive call divides the search space in half

$$(\text{mid} - 1) - \text{low} + 1 = \left\lfloor \frac{\text{low} + \text{high}}{2} \right\rfloor - \text{low} \leq \frac{\text{high} - \text{low} + 1}{2}$$

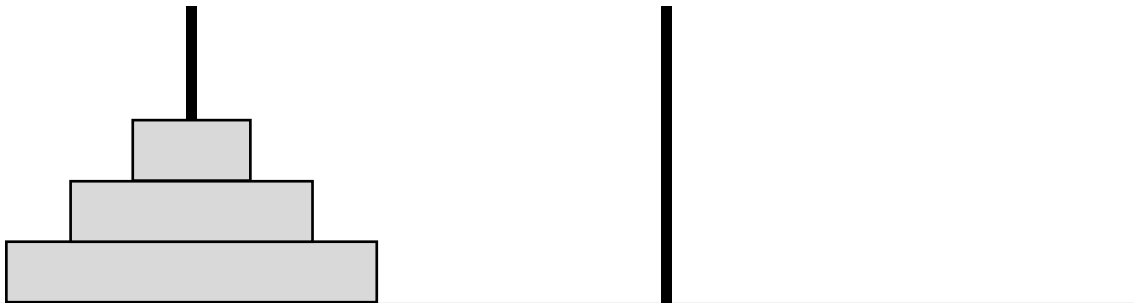
or

$$\text{high} - (\text{mid} + 1) + 1 = \text{high} - \left\lfloor \frac{\text{low} + \text{high}}{2} \right\rfloor \leq \frac{\text{high} - \text{low} + 1}{2}$$

- The maximum number of recursive calls is the smallest integer r such that $\frac{n}{2^r} < 1$ so $r = \lfloor \log n \rfloor + 1$
- Time complexity of binary search: $O(\log n)$ (base 2)
 - If $n = 1,000,000,000$, then $\log n = 30$

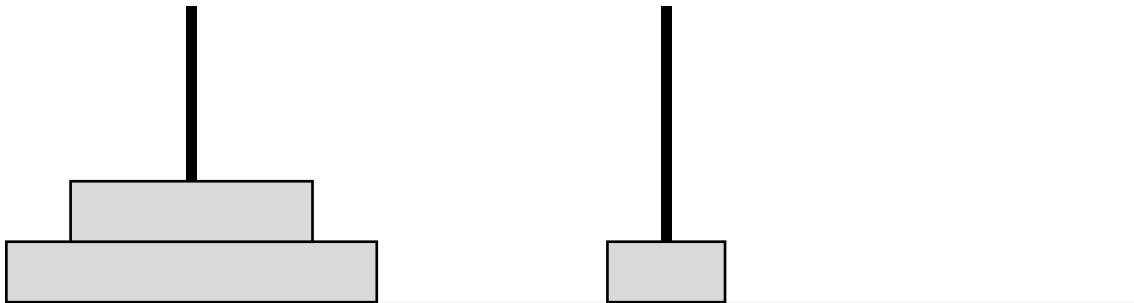
Tower of Hanoi

- n discs, 3 rods
 - Move one disc at a time to move all n discs to the second rod
 - No disc can be placed on a smaller disc



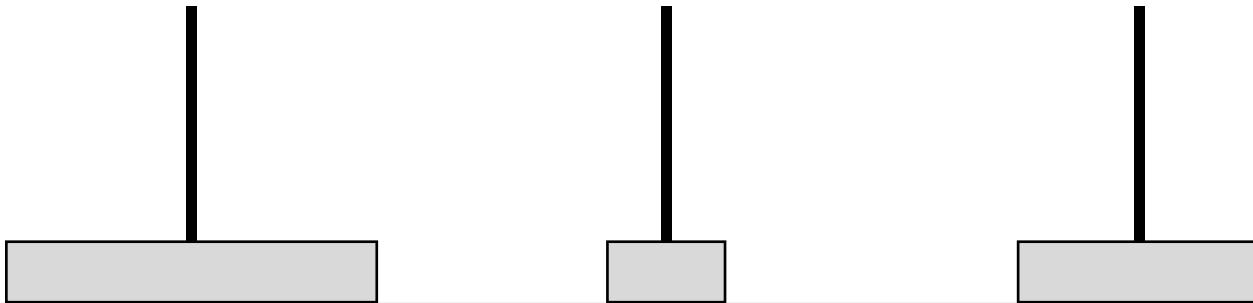
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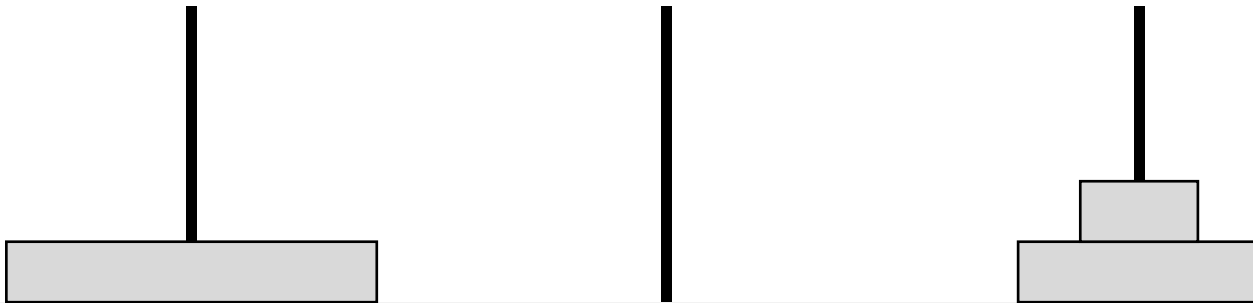
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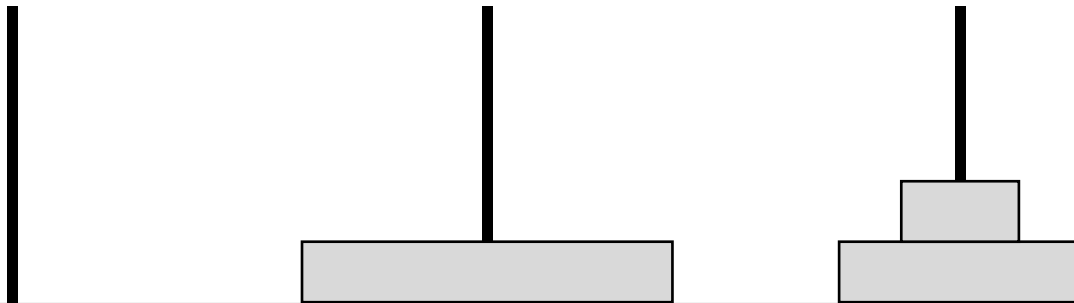
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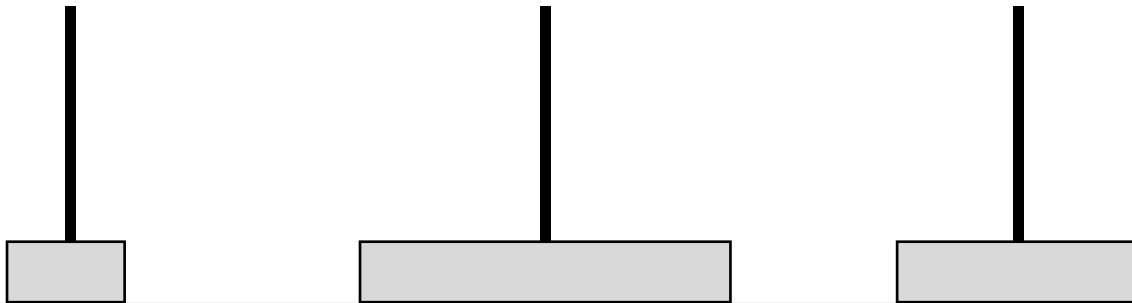
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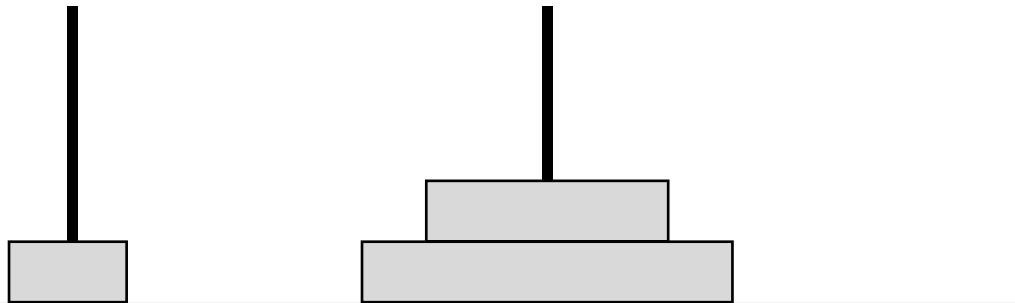
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- n discs, 3 rods
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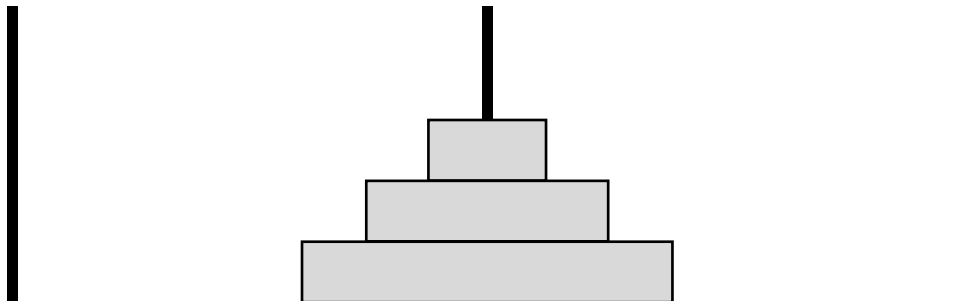
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Tower of Hanoi

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 - Move one disc at a time to move all n discs to the second rod
 - No disc can be placed on a smaller disc



Tower of Hanoi

- Move 3 disks from rod 1 to rod 2
 - Base case?
 - To move n disks from 1 to 2, we need to move $n-1$ disks from 1 to 3, move the bottom-most disk from 1 to 2, then move the $n-1$ disks from 3 to 2
- Time complexity? $O(2^n)$

```
def Hanoi(n, loc_from, loc_to):  
    if n <= 0:  
        return  
    loc_aux = 6 - loc_from - loc_to  
    Hanoi(n-1, loc_from, loc_aux)  
    print("Move a disk from rod " + str(loc_from) + " to rod " + str(loc_to))  
    Hanoi(n-1, loc_aux, loc_to)
```

```
Hanoi(3, 1, 2)
```

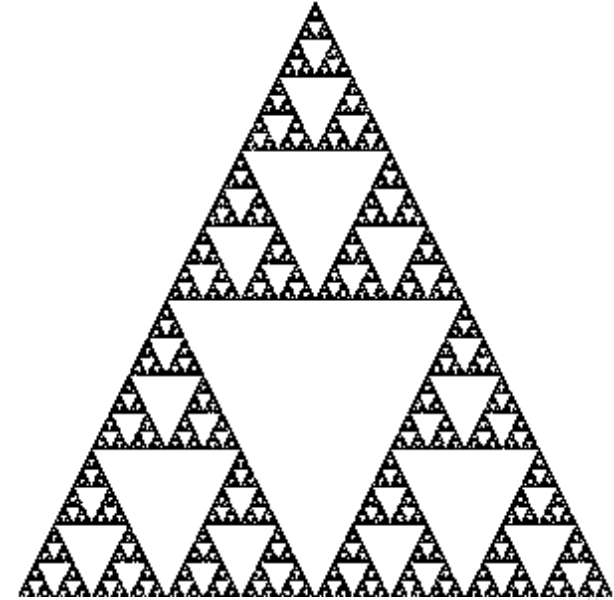
```
Move a disk from rod 1 to rod 2  
Move a disk from rod 1 to rod 3  
Move a disk from rod 2 to rod 3  
Move a disk from rod 1 to rod 2  
Move a disk from rod 3 to rod 1  
Move a disk from rod 3 to rod 2  
Move a disk from rod 1 to rod 2
```

Tower of Hanoi

- Wikipedia illustration:
https://upload.wikimedia.org/wikipedia/commons/2/20/Tower_of_Hanoi_recursion_SMIL.svg
- Game: <https://www.mathsisfun.com/games/towerofhanoi.html>

Fractal

- Endlessly recursive pattern
 - ex: Sierpinski triangle
 - triangles inside triangles inside triangles...



Linear Recursion

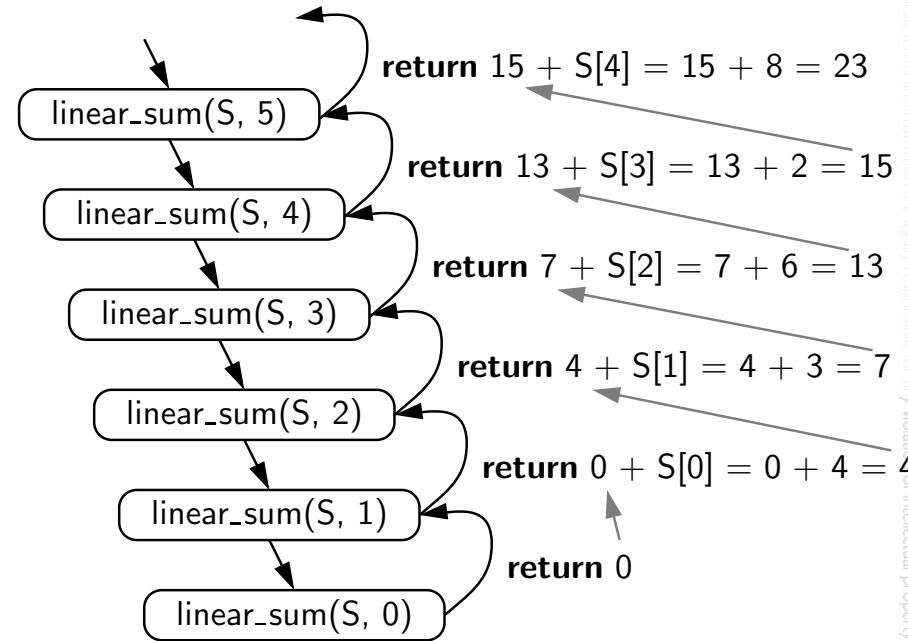
- Sum all elements in an array

| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 |
|---|---|---|---|---|---|---|---|---|---|----|----|----|----|----|----|
| 4 | 3 | 6 | 2 | 8 | 9 | 3 | 2 | 8 | 5 | 1 | 7 | 2 | 8 | 3 | 7 |

```

1 def linear_sum(S, n):
2     """ Return the sum of the first n numbers of sequence S. """
3     if n == 0:
4         return 0
5     else:
6         return linear_sum(S, n-1) + S[n-1]

```



Binary Recursion

- Two recursive calls at a time

```

1  def binary_sum(S, start, stop):
2      """ Return the sum of the numbers in implicit slice S[start:stop]. """
3      if start >= stop:                                # zero elements in slice
4          return 0
5      elif start == stop-1:                            # one element in slice
6          return S[start]
7      else:                                            # two or more elements in slice
8          mid = (start + stop) // 2
9          return binary_sum(S, start, mid) + binary_sum(S, mid, stop)

```

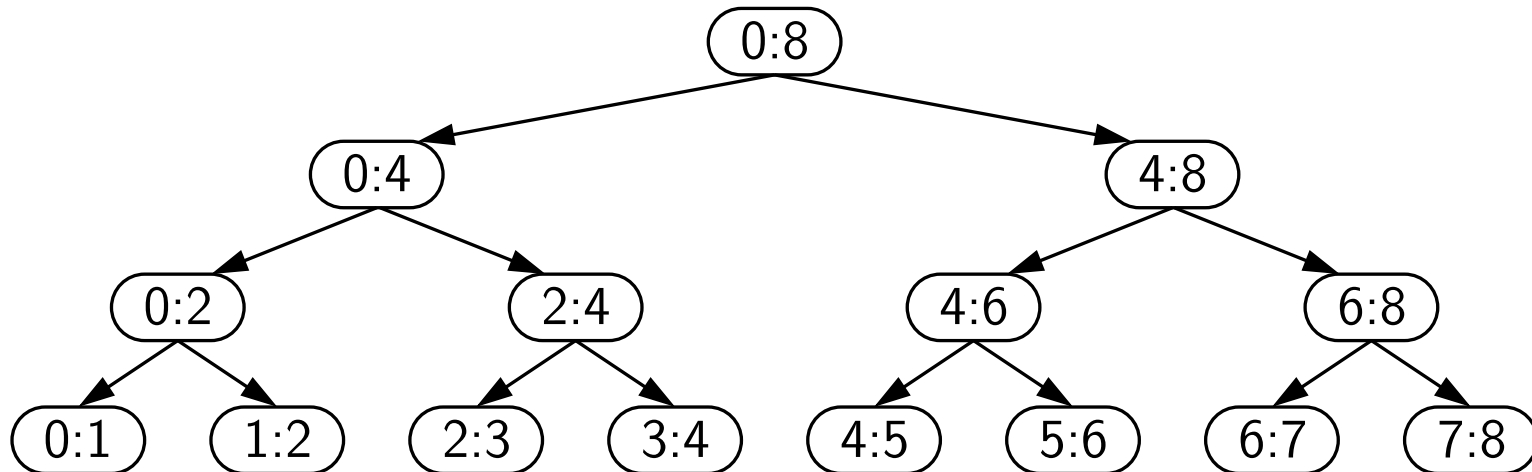


Figure 4.13: Recursion trace for the execution of `binary_sum(0, 8)`.

Bad Recursion?

- Fibonacci sequence:

$$f(n) = \begin{cases} 0 & \text{if } n = 0 \\ 1 & \text{if } n = 1 \\ f(n-1) + f(n-2) & \text{else} \end{cases}$$

- Technically correct, but terribly inefficient
 - Each recursion makes 2 calls
 - Similar to the Tower of Hanoi problem

```
1  def bad_fibonacci(n):
2      """ Return the nth Fibonacci number. """
3      if n <= 1:
4          return n
5      else:
6          return bad_fibonacci(n-2) + bad_fibonacci(n-1)
```

Bad Recursion!

- c_n = number of calls performed in `bad_fibonacci(n)`
- When $n = n + 2$, the number of calls more than doubles
- $c_n > 2^{n/2}$

$$c_0 = 1$$

$$c_1 = 1$$

$$c_2 = 1 + c_0 + c_1 = 1 + 1 + 1 = 3$$

$$c_3 = 1 + c_1 + c_2 = 1 + 1 + 3 = 5$$

$$c_4 = 1 + c_2 + c_3 = 1 + 3 + 5 = 9$$

$$c_5 = 1 + c_3 + c_4 = 1 + 5 + 9 = 15$$

$$c_6 = 1 + c_4 + c_5 = 1 + 9 + 15 = 25$$

$$c_7 = 1 + c_5 + c_6 = 1 + 15 + 25 = 41$$

$$c_8 = 1 + c_6 + c_7 = 1 + 25 + 41 = 67$$

Good Recursion

- Why was `bad_fibonacci` so inefficient?
 - Each iteration requires 2 recursive calls: F_{n-1} and F_{n-2}
 - But computing F_{n-1} makes its own call of F_{n-2}
 - Can we reuse the information?
- Have the function to return two values: F_n and F_{n-1}
 - Thus, the next recursion has both F_{n-1} and F_{n-2} without recomputing them
 - Turns in to a **linear recursion**

```
1  def good_fibonacci(n):
2      """ Return pair of Fibonacci numbers, F(n) and F(n-1). """
3      if n <= 1:
4          return (n, 0)
5      else:
6          (a, b) = good_fibonacci(n-1)
7          return (a+b, a)
```

Good Recursion 2

- Power function: $power(x, n) = x^n$

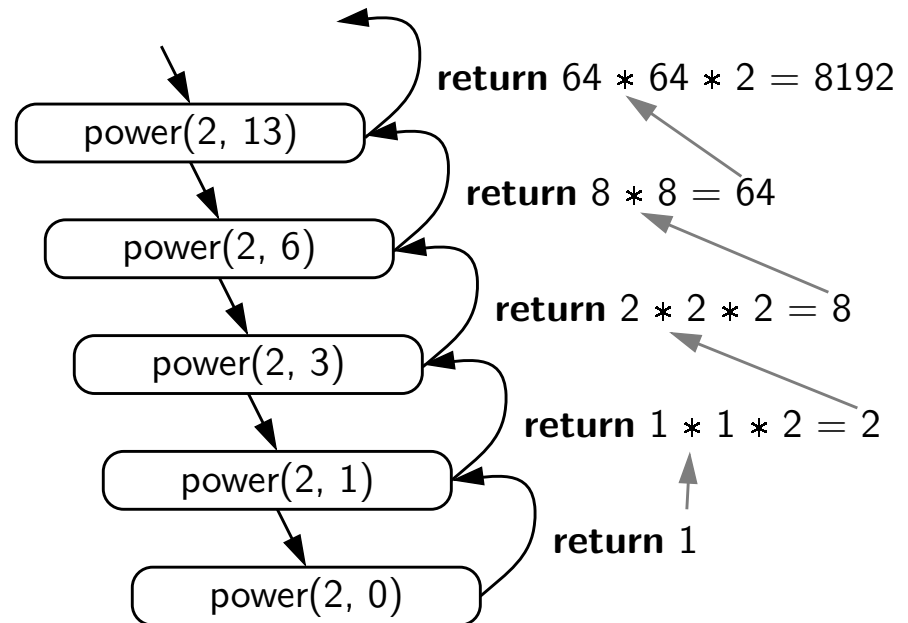
$$power(x, n) = \begin{cases} 1 & \text{if } n = 0 \\ x \cdot power(x, n - 1) & \text{otherwise.} \end{cases}$$

```
1 def power(x, n):
2     """ Compute the value x**n for integer n. """
3     if n == 0:
4         return 1
5     else:
6         return x * power(x, n-1)
```

Good Recursion 2

- (Better) Power function: $power(x, n) = x^n$

$$power(x, n) = \begin{cases} 1 & \text{if } n = 0 \\ x \cdot (power(x, \lfloor \frac{n}{2} \rfloor))^2 & \text{if } n > 0 \text{ is odd} \\ (power(x, \lfloor \frac{n}{2} \rfloor))^2 & \text{if } n > 0 \text{ is even} \end{cases}$$



```

1 def power(x, n):
2     """ Compute the value x**n for integer n. """
3     if n == 0:
4         return 1
5     else:
6         partial = power(x, n // 2)           # rely on truncated division
7         result = partial * partial
8         if n % 2 == 1:                       # if n odd, include extra factor of x
9             result *= x
10    return result

```

Summary

- Often simple and intuitive
- Useful when dealing with “less linearly structured” data
 - e.g., trees, graphs
- In practice: need to be careful as it may not be so efficient
 - Unintendedly large time complexity
 - Good time complexity, but slow in practice due to many costly overhead operations
 - Not so well optimized
- For this course, we will see various recursive algorithms