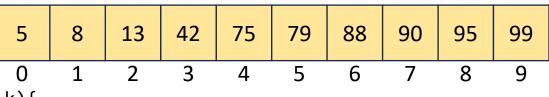
# CSE 332: Data Structures & Parallelism Lecture 6: Recurrences

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### **Recursive Binary Search**



```
public static boolean binarySearch(List<Integer> lst, int k){
        return binarySearch(lst, k, 0, lst.size());
private static boolean binarySearch(List<Integer> lst, int k, int start, int end){
    if(start == end)
        return false;
    int mid = start + (end-start)/2;
    if(lst.get(mid) == k){
        return true;
    } else if(lst.get(mid) > k){
        return binarySearch(lst, k, start, mid);
    } else{
        return binarySearch(lst, k, mid+1, end);
    }
}
```

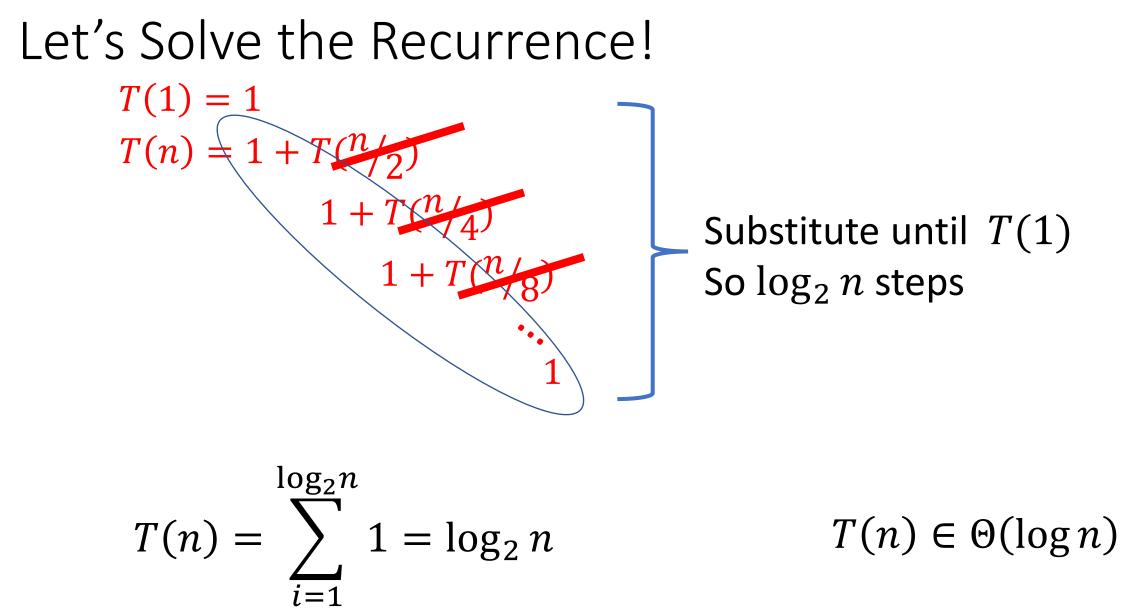
## Analysis of Recursive Algorithms

- Overall structure of recursion:
  - Do some non-recursive "work"
  - Do one or more recursive calls on some portion of your input
  - Do some more non-recursive "work"
  - Repeat until you reach a base case
- Running time:  $T(n) = T(p_1) + T(p_2) + \dots + T(p_x) + f(n)$ 
  - The time it takes to run the algorithm on an input of size *n* is:
  - The sum of how long it takes to run the same algorithm on each smaller input
  - Plus the total amount of non-recursive work done at that step
- Usually:
  - $T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n)$ 
    - Called "divide and conquer"
  - T(n) = T(n-c) + f(n)
  - Called "chip and conquer"

#### How Efficient Is It?

- $T(n) = 1 + T\left(\left[\frac{n}{2}\right]\right)$
- Base case: T(1) = 1

T(n) = "cost" of running the entire algorithm on an array of length n

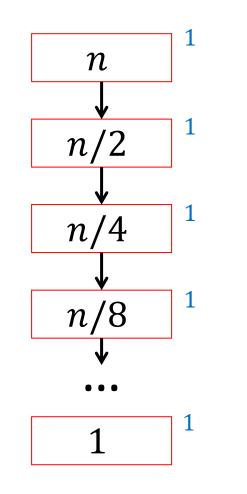


## Make our process "prettier"

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

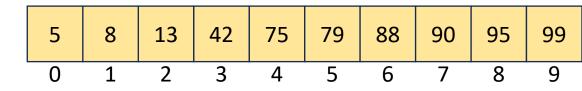
- Draw a picture of the recursion
- Identify the work done per stack frame
- Add up all the work!
  - Sum is the answer!
  - In this case  $\Theta(\log_2 n)$

## The "Tree Method"



 $\log_2 n$  levels of recursion

#### Recursive Linear Search

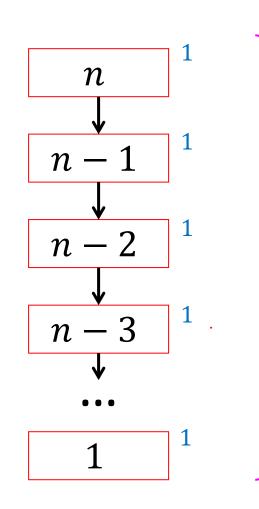


```
public static boolean linearSearch(List<Integer> lst, int k){
        return linearSearch(lst, k, 0, lst.size());
    }
private static boolean linearSearch(List<Integer> lst, int k, int start, int end){
    if(start == end){
        return false;
    } else if(lst.get(start) == k){
        return true;
    } else{
        return linearSearch(lst, k, start+1, end);
    }
}
```

## Make our method "prettier"

- Identify the work done per stack frame
- Add up all the work!

#### Running time: $\Theta(n)$



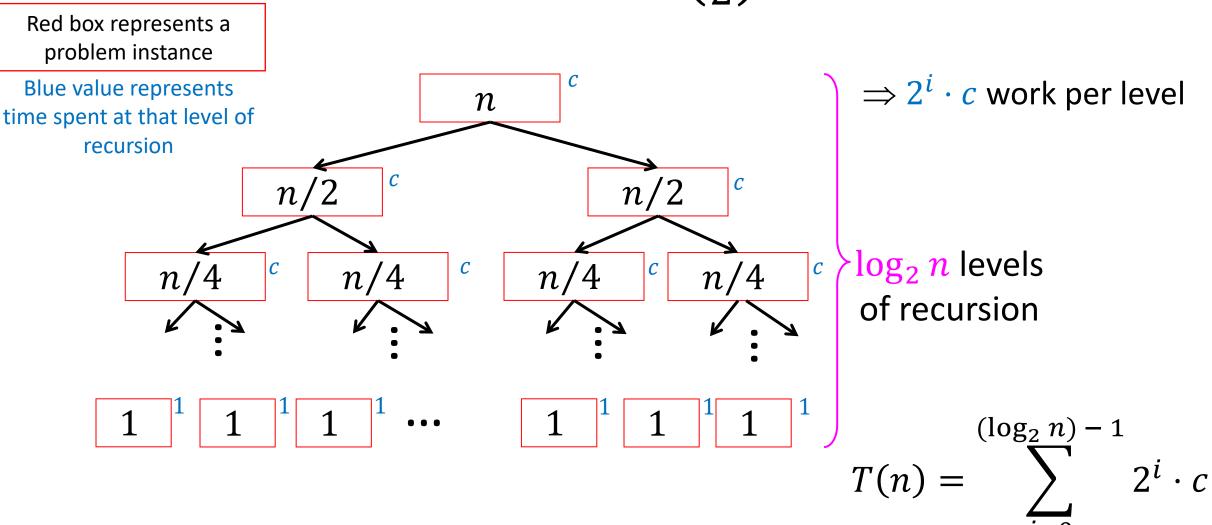
T(n) = T(n-1) + 1

*n* levels of recursion

### **Recursive List Summation**

```
public int sum(int[] list){
    return sum_helper(list, 0, list.size);
}
private int sum_helper(int[] list, int low, int high){
    if (low == high){ return 0; }
    if (low == high-1){ return list[low]; }
    int middle = (high+low)/2;
    return sum_helper(list, low, middle) + sum_helper(list, middle, high);
}
```

# Tree Method: $T(n) = 2T\left(\frac{n}{2}\right) + c$



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#### **Recursive List Summation**

$$T(n) = \sum_{i=0}^{(\log_2 n) - 1} 2^i \cdot c$$

$$= c \cdot \sum_{i=0}^{(\log_2 n) - 1} 2^i$$

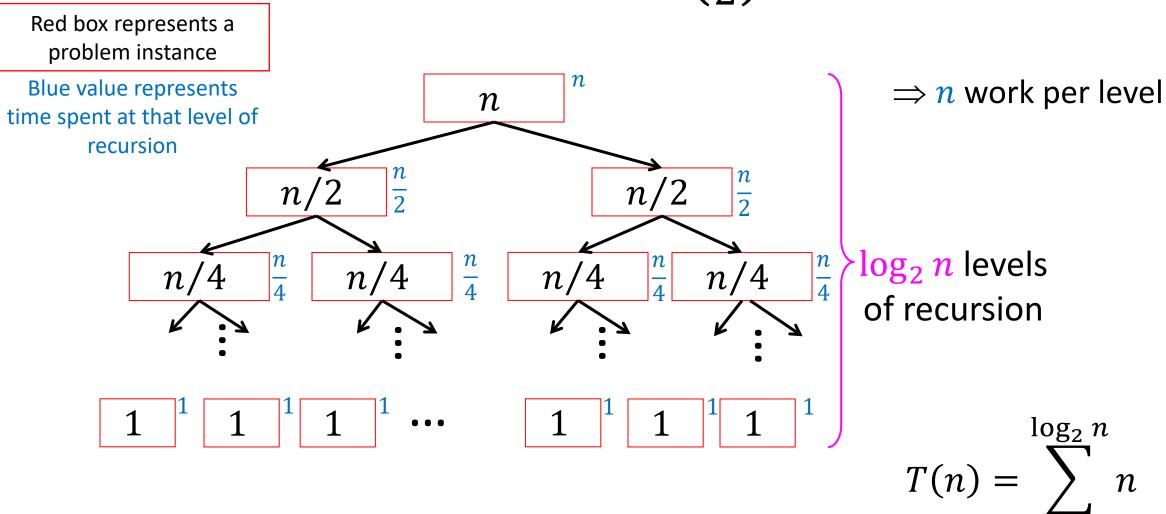
A "useful" Math Identity  
(see link on exercises page)
$$\sum_{i=0}^{n-1} x^{i} = \frac{1-x^{n}}{1-x}$$

$$= c \left( \frac{1 - 2^{\log_2 n}}{1 - 2} \right)$$

#### Let's do some more!

- For each, assume the base case is n = 1 and T(1) = 1
- $T(n) = 2T\left(\frac{n}{2}\right) + n$ •  $T(n) = 2T\left(\frac{n}{2}\right) + n^2$ •  $T(n) = 2T\left(\frac{n}{8}\right) + 1$

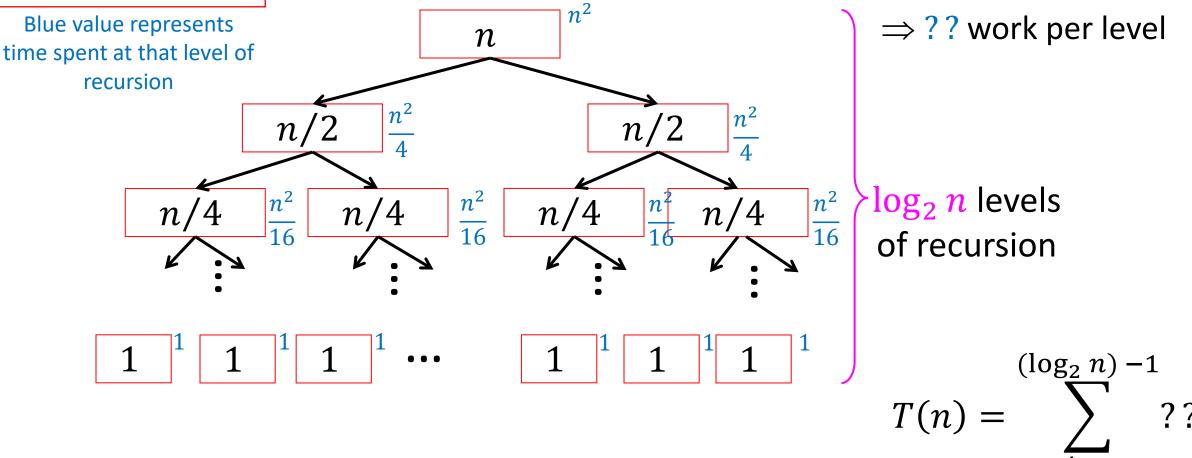
# Tree Method: $T(n) = 2T\left(\frac{n}{2}\right) + n$



# Tree Method: $T(n) = 2T\left(\frac{n}{2}\right) + n^2$

Red box represents a problem instance Blue value represents

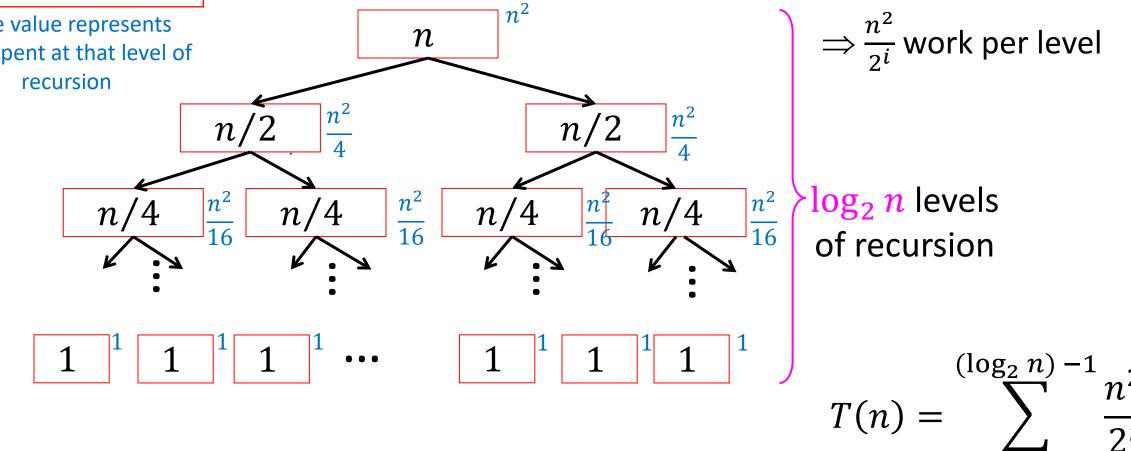
recursion



# Tree Method: $T(n) = 2T\left(\frac{n}{2}\right) + n^2$

Red box represents a problem instance

Blue value represents time spent at that level of recursion



i=0

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$$T(n) = \sum_{i=0}^{(\log_2 n) - 1} \frac{n^2}{2^i}$$

$$= n^2 \cdot \sum_{i=0}^{(\log_2 n) - 1} \left(\frac{1}{2}\right)^i$$

<u>A "useful" Math Identity</u> (see link on <u>exercises page</u> )
$\sum_{i=0}^{n-1} x^{i} = \frac{1-x^{n}}{1-x}$

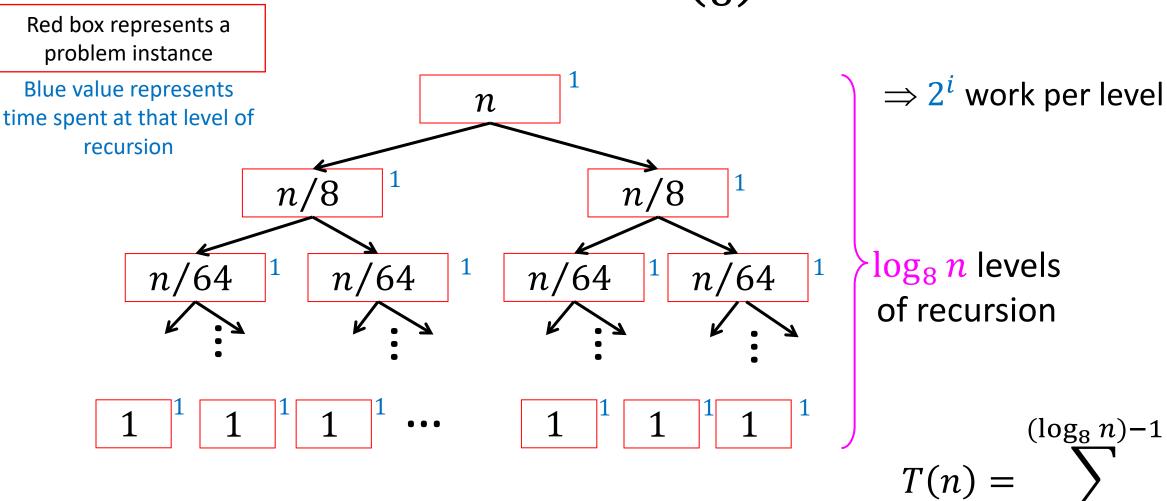
$$T(n) = \sum_{i=0}^{(\log_2 n) - 1} \frac{n^2}{2^i}$$

$$= n^2 \cdot \sum_{i=0}^{(\log_2 n) - 1} \left(\frac{1}{2}\right)^i$$

A "useful" Math Identity  
(see link on exercises page)  
$$\sum_{i=0}^{n-1} x^{i} = \frac{1-x^{n}}{1-x}$$

$$= n^2 \cdot \left(\frac{\frac{1}{n} - 1}{\frac{1}{2} - 1}\right) = \Theta(n^2)$$

# Tree Method: $T(n) = 2T\left(\frac{n}{8}\right) + 1$



 $i=0_{18}$ 

$$T(n) = \sum_{i=0}^{(\log_8 n) - 1} 2^i$$
$$= \left(\frac{1 - 2^{\log_8 n}}{1 - 2}\right)$$
$$= 2^{\log_8 n} - 1$$
$$= n^{\log_8 2} = n^{\frac{1}{3}}$$

A "useful" Math Identity  
(see link on exercises page)  
$$\sum_{i=0}^{n-1} x^{i} = \frac{1-x^{n}}{1-x}$$
$$a^{\log_{b} c} = c^{\log_{b} a}$$

### What matters, recursively

- For  $T(n) = aT\left(\frac{n}{b}\right) + f(n)$ 
  - The following are important for asymptotic behavior:
    - The value of *a*
    - The value of *b*
    - Asymptotic behavior of f(n)
  - The following are not important for asymptotic behavior:
    - Constants and non-dominant terms in f(n)
    - The base case

### Really common recurrences

Should know how to solve recurrences but also recognize some really common ones:

T(n) = O(1) + T(n/2) T(n) = O(1) + 2T(n/2)
T(n) = O(1) + T(n-1) T(n) = O(n) + T(n-1) T(n) = O(1) + 2T(n-1)
T(n) = O(n) + T(n/2) T(n) = O(n) + 2T(n/2)

logarithmic	<i>O</i> (log n)
linear	<i>O</i> (n)
linear	<i>O</i> (n)
quadratic	<i>O</i> (n²)
exponential	<i>O</i> (2 <sup>n</sup> )
linear	<i>O</i> (n)
Ioglinear	<i>O</i> (n <b>log</b> n)