

# CSE 332: Data Structures & Parallelism

## Lecture 6: Recurrences

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(Worst Case)  $O(\log_2 n)$

# Recursive Binary Search

binarySearch(2st 1, 15)

5	8	13	42	75	79	88	90	95	99
0	1	2	3	4	5	6	7	8	9

```
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);
```

```
    }
```

```
}
```

$$T(n) = 9 + T\left(\frac{n}{2}\right)$$

4  
2  
2

# 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

$$T(n) = 9 + T\left(\frac{n}{2}\right)$$

- 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?

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

$T(n)$  = “cost” of running the entire algorithm on an array of length  $n$

# Let's Solve the Recurrence!

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

$$T(1) = 1$$

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

$$1 + T\left(\frac{n}{4}\right)$$

$$1 + T\left(\frac{n}{8}\right)$$

...

1

$$\frac{n}{2^i} = 1$$

$$\log_2(2^i) = \log_2(n)$$

$$i = \log_2 n$$

$$T(n) = \sum_{i=1}^{\log_2 n} 1 = \log_2 n$$

$$\rightarrow \frac{n}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \dots$$

Substitute until  $T(1)$

So  $\log_2 n$  steps

Base case

$$\frac{n}{2^i} = 1$$

$i$  = number of substitutions needed until get to base case

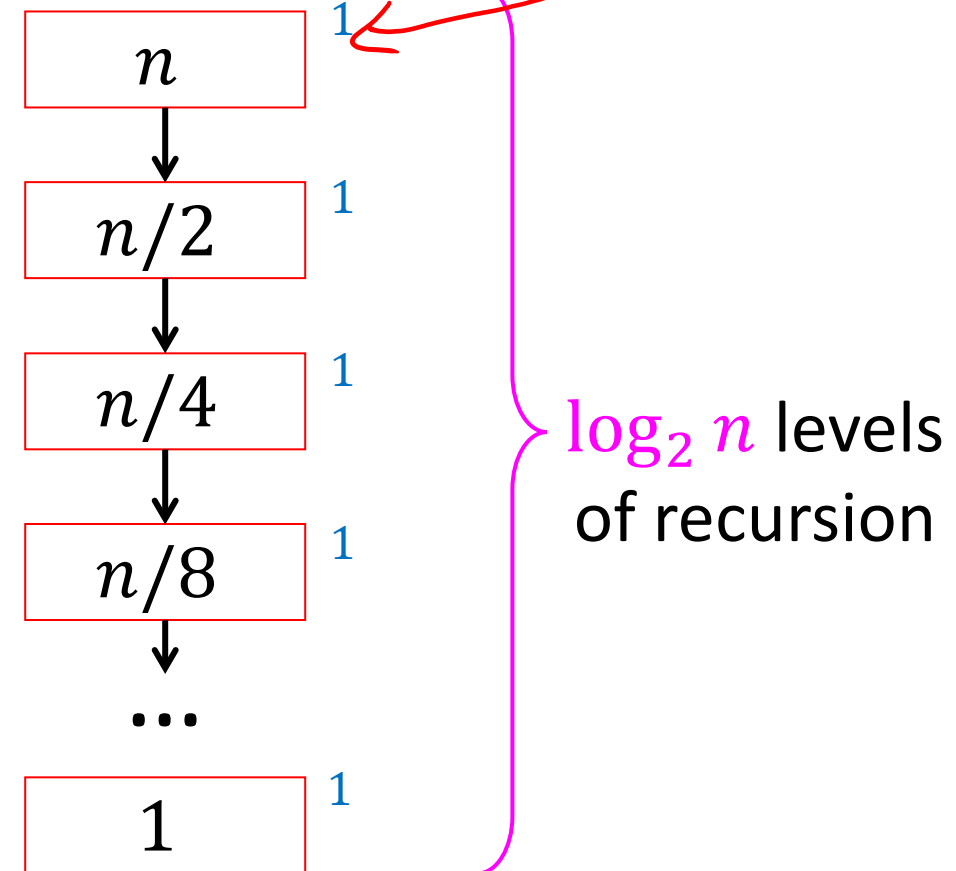
$$T(n) \in \Theta(\log n)$$

# Make our process “prettier”

- 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”

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



# Recursive Linear Search

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

5	8	13	42	75	79	88	90	95	99
0	1	2	3	4	5	6	7	8	9

```
public static boolean linearSearch(List<Integer> lst, int k){
    return linearSearch(lst, k, 0, lst.size());
}

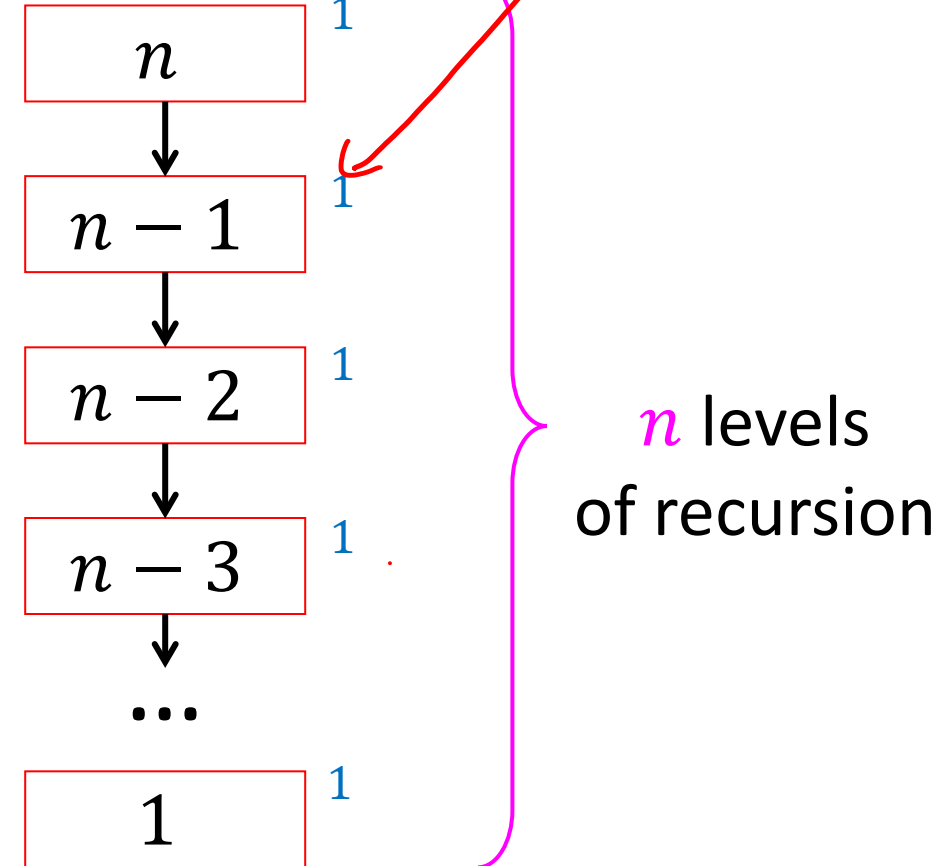
private static boolean linearSearch(List<Integer> lst, int k, int start, int end){
    if(start == end){
        return false;
    } else if(lst.get(start) == k){
        return true;
    } else{
        return linearSearch(lst, k, start+1, end);
    }
}
```

# Make our method “prettier”

- Draw a picture of the recursion
- Identify the work done per stack frame
- Add up all the work!

Running time:  $\Theta(n)$

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





# Recursive List Summation

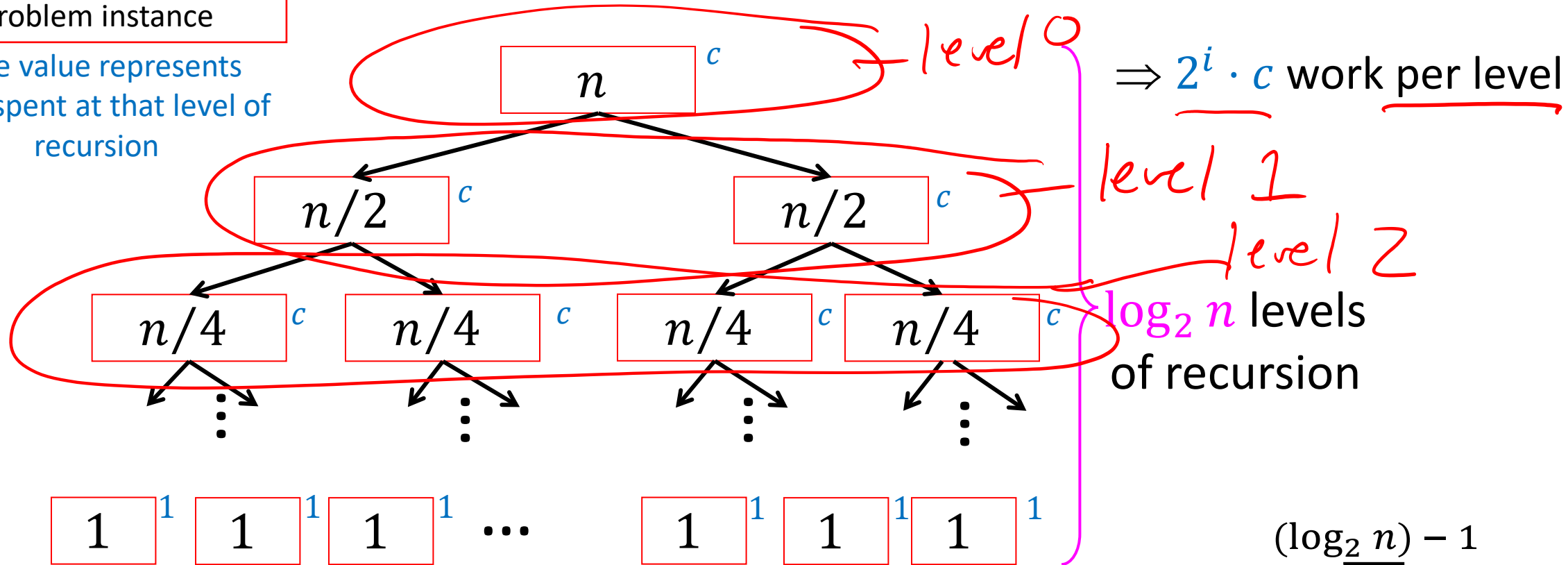
$$T(n) = c + 2 \cdot T\left(\frac{n}{2}\right)$$

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

Red box represents a problem instance

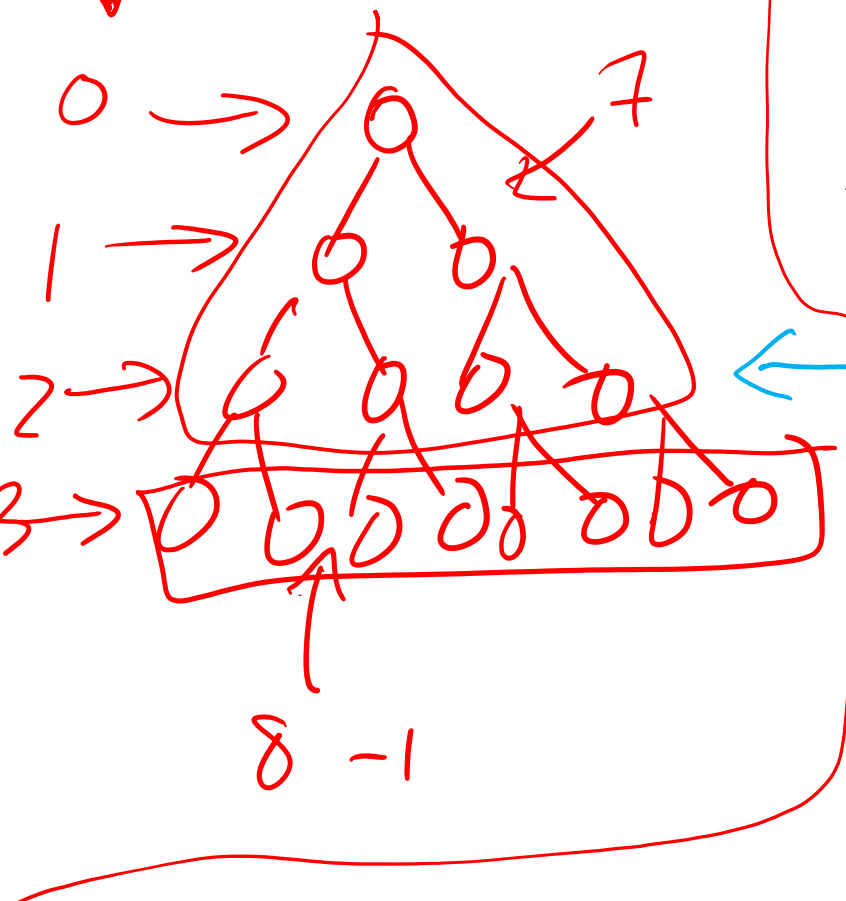
Blue value represents time spent at that level of recursion



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

Remember? (see lecture 4 slides)  
 Number of nodes in a perfect binary tree of height  $h$ ?

# Recursive List Summation



$$\sum_{i=0}^h 2^i = 2^{h+1} - 1$$

#nodes of that level

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

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

or use the more general formula

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

A "useful" Math Identity  
 (see link on [exercises page](#))

$$\sum_{i=0}^{n-1} x^i = \frac{1-x^n}{1-x}$$

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

$$= c \cdot (n - 1) \rightarrow \Theta(n)$$

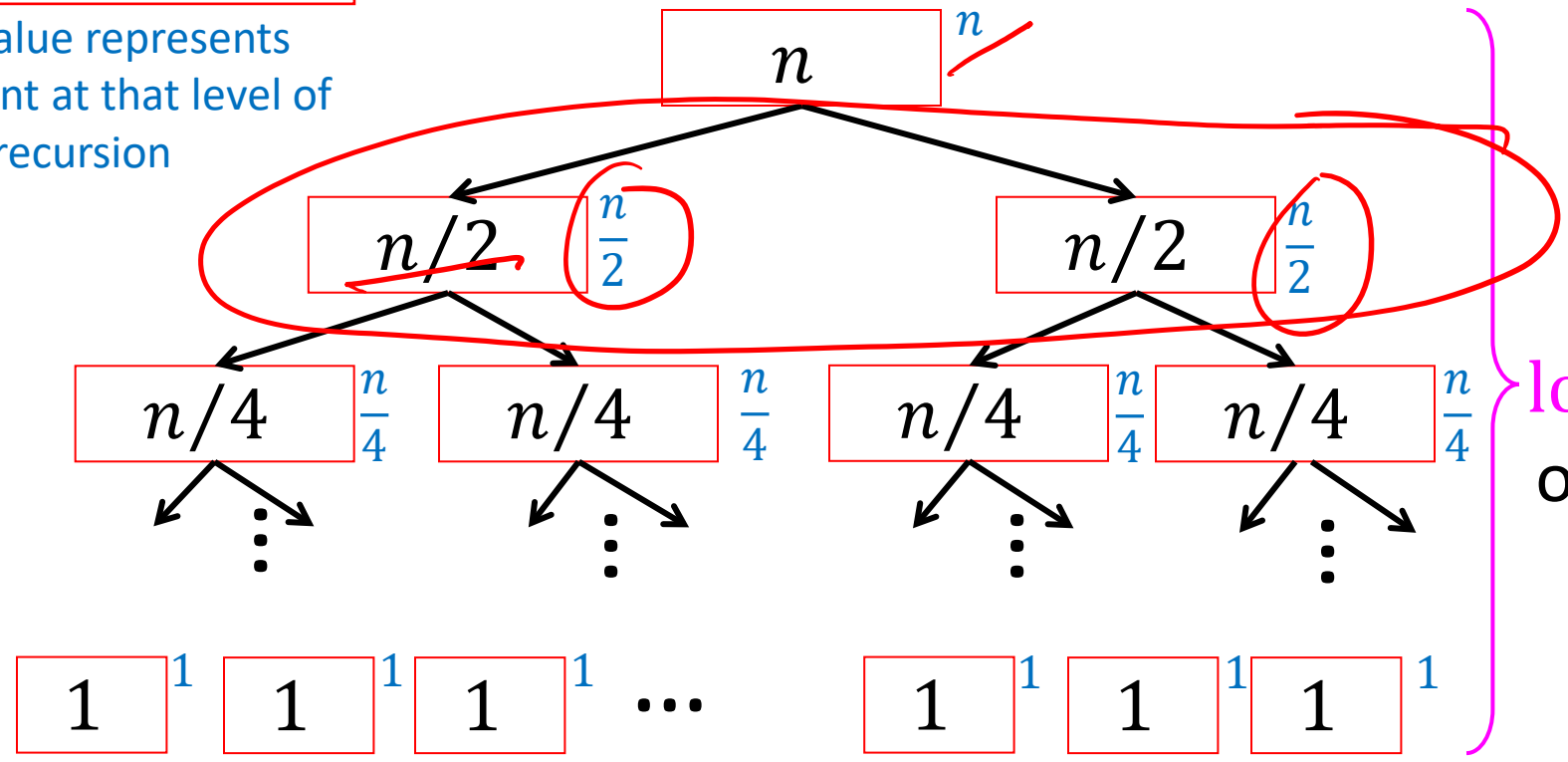
# Let's do some more!

- 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: $T(n) = 2T\left(\frac{n}{2}\right) + \underline{n}$

Red box represents a problem instance

Blue value represents time spent at that level of recursion



$\Rightarrow \underline{n}$  work per level

$\log_2 n$  levels of recursion



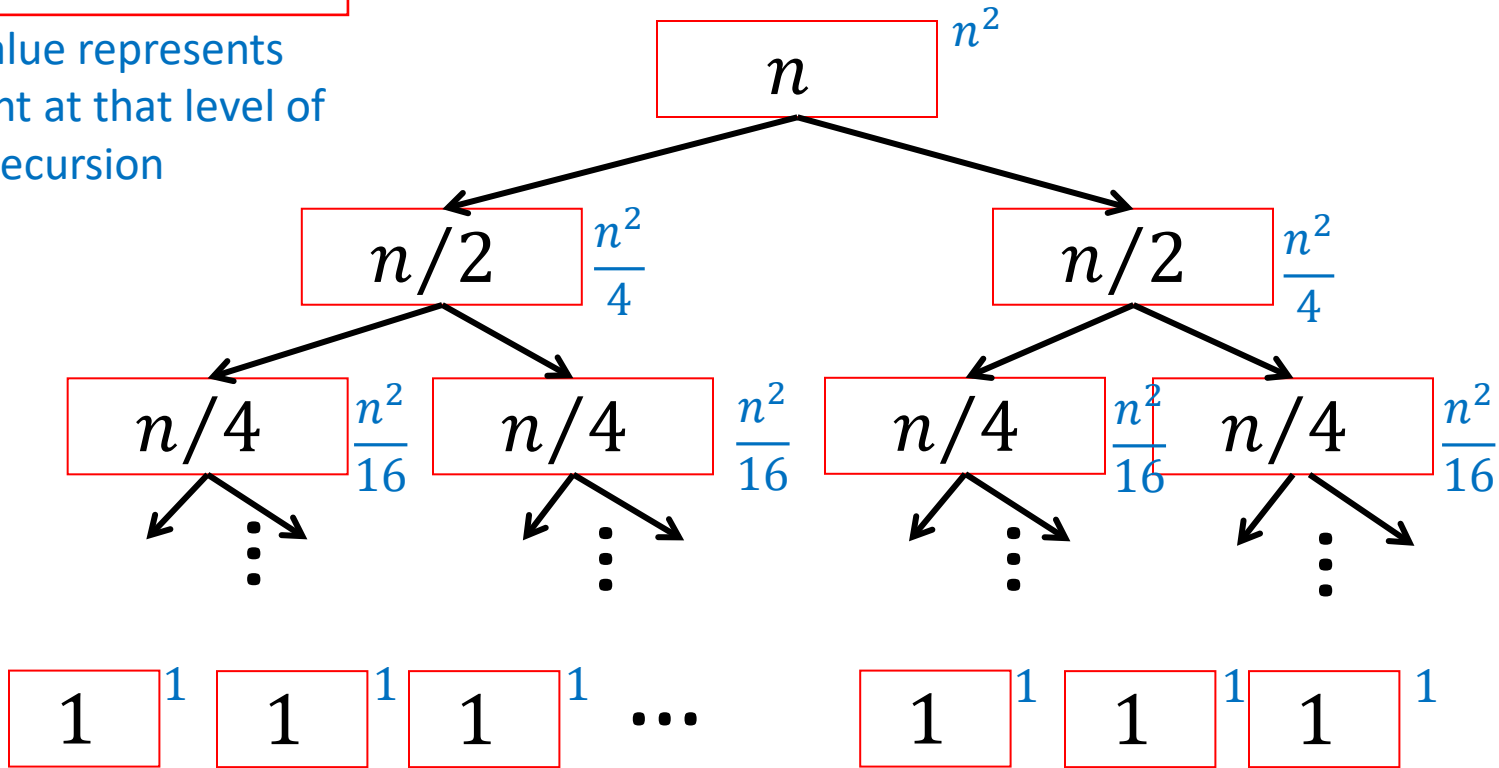
$\Theta(n \log n)$

$$T(n) = \sum_{i=1}^{\log_2 n} n$$

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

Red box represents a problem instance

Blue value represents time spent at that level of recursion



⇒ ?? work per level

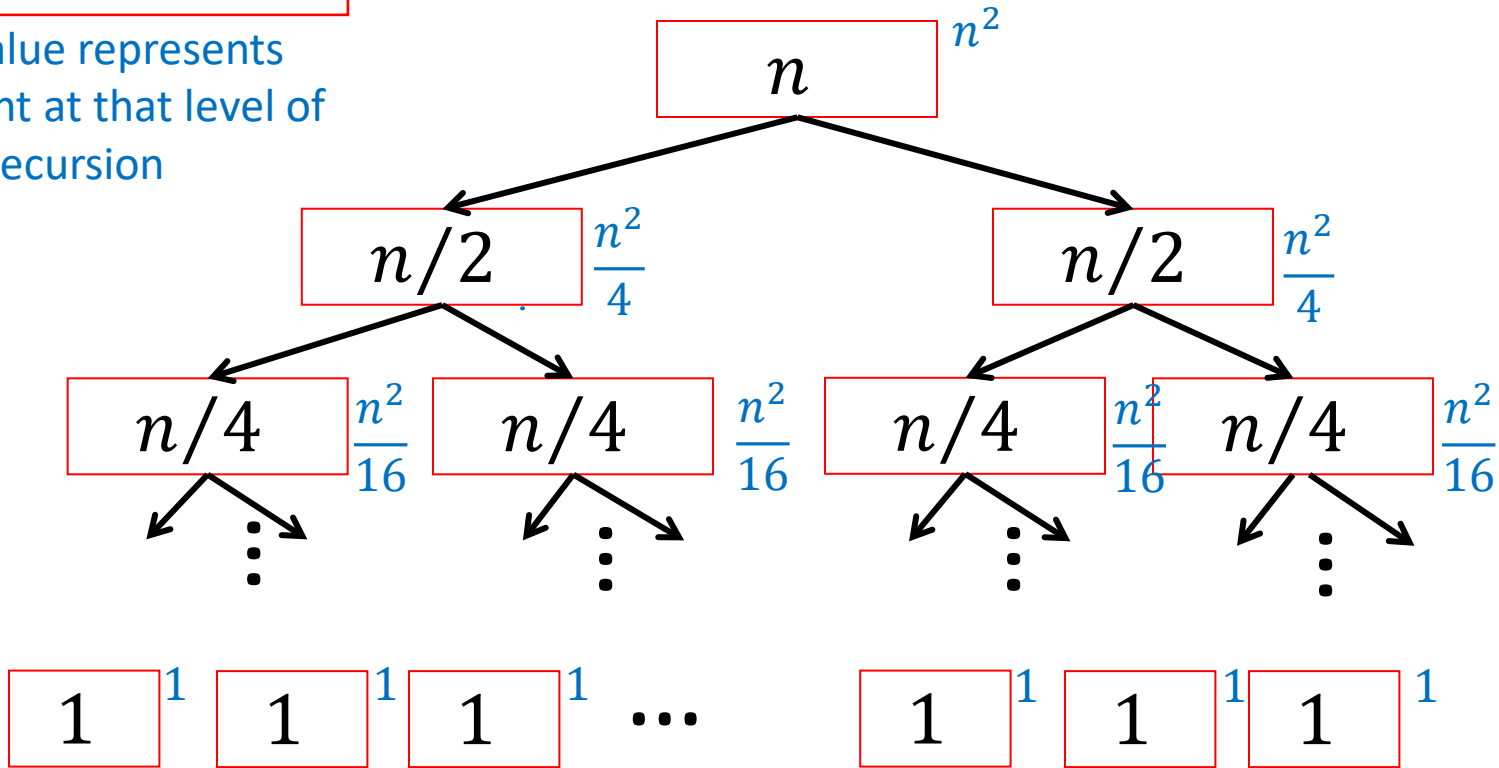
$\log_2 n$  levels of recursion

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

# 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



$\Rightarrow \frac{n^2}{2^i}$  work per level

$\log_2 n$  levels of recursion

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

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

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

A “useful” Math Identity  
(see link on [exercises page](#))

$$\sum_{i=0}^{n-1} x^i = \frac{1-x^n}{1-x}$$



$$\begin{aligned}
T(n) &= \sum_{i=0}^{(\log_2 n) - 1} \frac{n^2}{2^i} \\
&= n^2 \cdot \sum_{i=0}^{(\log_2 n) - 1} \left(\frac{1}{2}\right)^i \\
&= n^2 \cdot \left(\frac{\frac{1}{n} - 1}{\frac{1}{2} - 1}\right) = \underline{\Theta(n^2)}
\end{aligned}$$

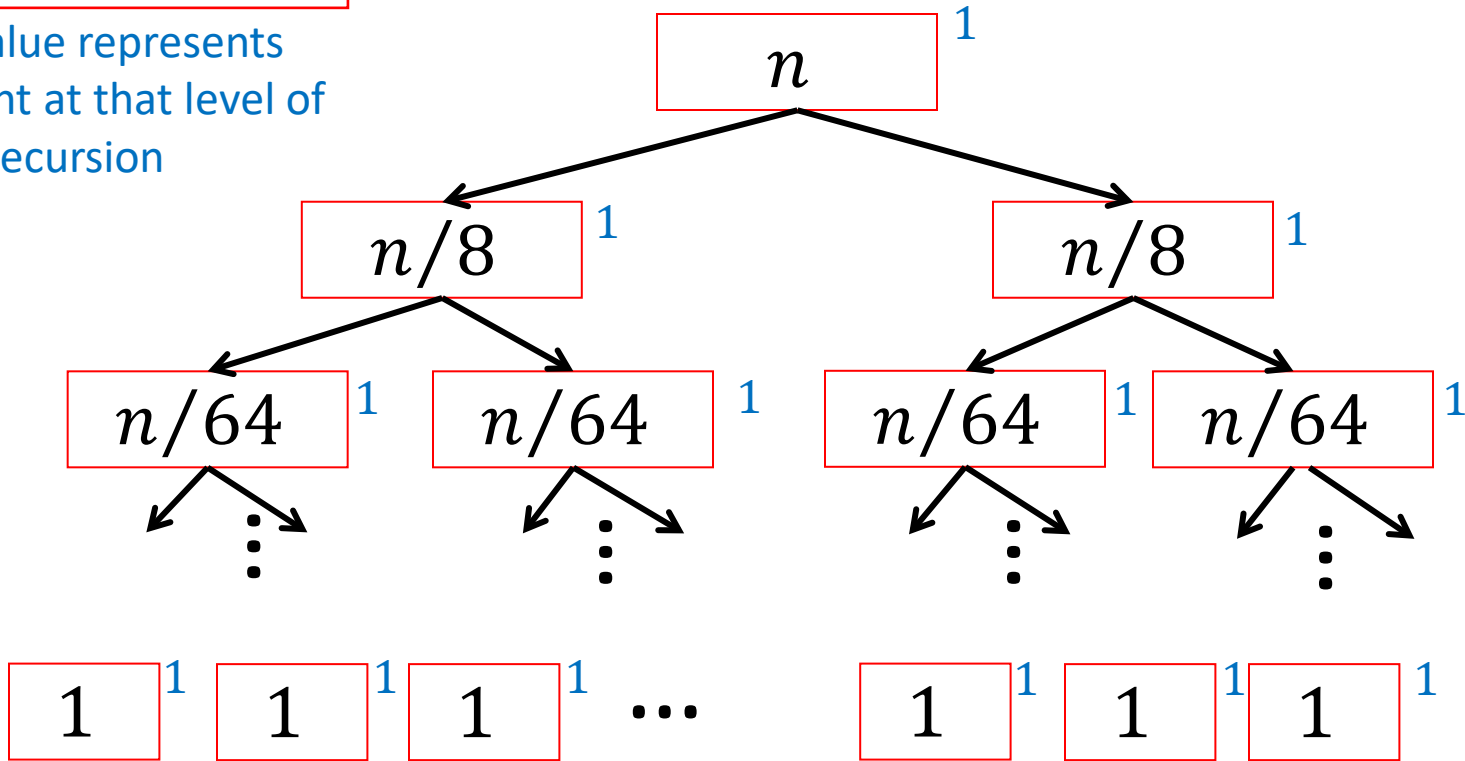
A “useful” Math Identity  
(see link on [exercises page](#))

$$\sum_{i=0}^{n-1} x^i = \frac{1-x^n}{1-x}$$

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

Red box represents a problem instance

Blue value represents time spent at that level of recursion



$\Rightarrow 2^i$  work per level

$\log_8 n$  levels of recursion

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

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

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:

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

logarithmic  $O(\log n)$

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

linear  $O(n)$

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

linear  $O(n)$

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

quadratic  $O(n^2)$

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

exponential  $O(2^n)$

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

linear  $O(n)$

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

loglinear  $O(n \log n)$