



Middlebury

CSCI 200: Math Foundations of Computing

Spring 2026

Lecture 4W: Recurrence Relations II

Goals for today:

- Identify divide-and-conquer recurrence relations,
- Use the Tree Theorem to analyze the complexity of divide-and-conquer recurrences.

```
1 def binary_search(lst, val, lo, hi):  
2     if lo > hi:  
3         return -1  
4     m = (lo + hi) // 2  
5     if lst[m] == val:  
6         return m  
7     elif lst[m] > val:  
8         return binary_search(lst, val, lo, m - 1)  
9     else:  
10        return binary_search(lst, val, m + 1, hi)
```

Similar to Lab 3
sqrt 2
power 3

Let's analyze the work done by Merge-Sort.

```
1 def merge_sort(a: list) -> list:
2     n = len(a)
3     if n <= 1: # base case
4         return a
5
6     # divide stage: sort two sublists
7     m = n // 2
8     a1 = merge_sort(a[:m]) # sort left sublist
9     a2 = merge_sort(a[m:]) # sort right sublist
10
11    # conquer stage: merge the sorted sublists
12    merged = []
13    i = j = 0
14    while i < len(a1) and j < len(a2):
15        if a1[i] < a2[j]:
16            merged.append(a1[i])
17            i += 1
18        else:
19            merged.append(a2[j])
20            j += 1
21
22    while i < len(a1):
23        merged.append(a1[i])
24        i += 1
25
26    while j < len(a2):
27        merged.append(a2[j])
28        j += 1
29    return merged
```

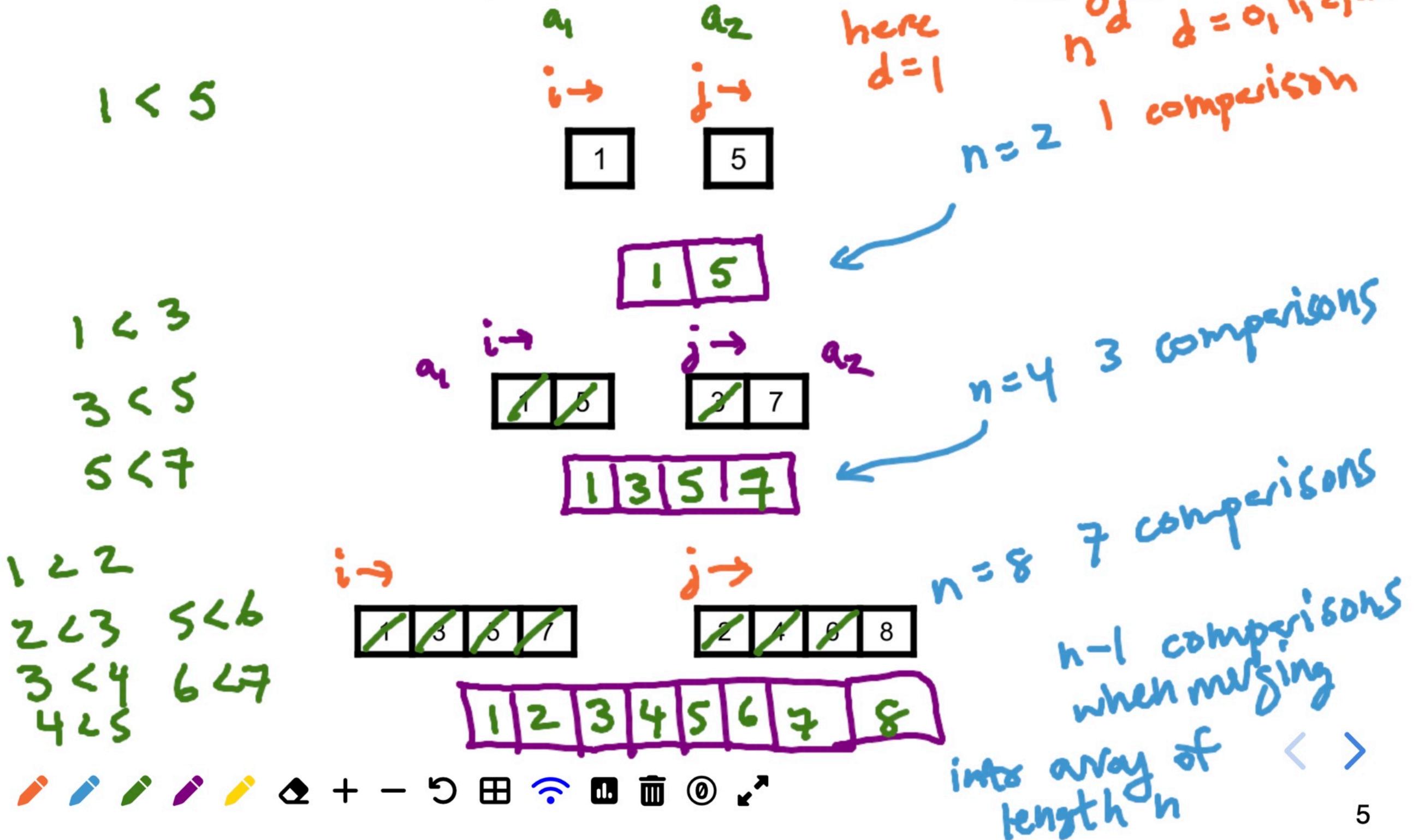
} 2 subproblems

← this what we will count

Merge-Sort in action (from Wikipedia)

6 5 3 1 8 7 2 4

How many times do you need to ask "which leading element is smallest?" when merging these two subarrays?



Adding up all the work done (Merge-Sort).

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

2 subproblems, breaking problem in half.
 $\sum_{i=0}^{\infty} r^i = \frac{1-r^{D+1}}{1-r}$

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

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

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

$$T(n) = (n - 1) + 2\left(\left(\frac{n}{2} - 1\right) + 2T\left(\frac{n}{4}\right)\right)$$

when $n=1$
 $T(1) = 0$

$$T(n) = (n-1) + (n-2) + 2^2 T\left(\frac{n}{4}\right)$$

$$T(n) = (n-1) + (n-2) + 2^2 \left[\frac{n}{4} - 1 + 2T\left(\frac{n}{8}\right) \right]$$

$$T(n) = (n-1) + (n-2) + (n-4) + 2^3 T\left(\frac{n}{8}\right)$$

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

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

$$= n \log_2 n + (n-1)$$

$$O(n \log n)$$

$\frac{n}{2^{\log_2 n - 1}} = 1 \rightarrow 2^{\log_2 n - 1} = n$
 (upper bound of sum)

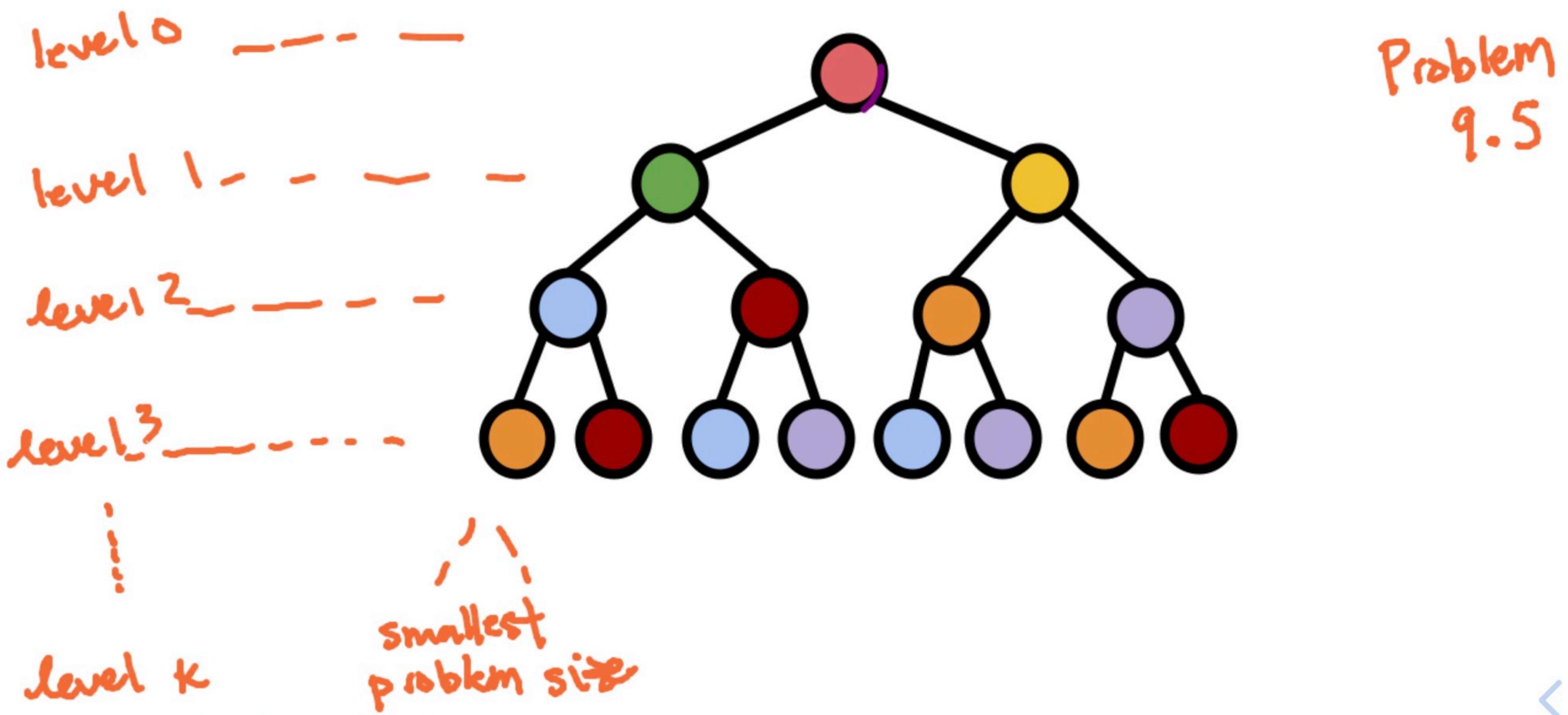
$i = \log_2 n - 1$

$D = \log_2 n$

Adding up all the work done (general divide-and-conquer recurrences).

Questions we need to answer:

1. How many times is the original length- n array broken up until we get to subarrays of length 1?
2. How much *total* work (operations) are done during the recursive step (i.e. to *merge*)?



The Tree Theorem for divide & conquer recurrences.

$$\text{General recurrence: } f(n) = a \cdot f\left(\frac{n}{b}\right) + c \cdot n^d$$

subproblems

shrink factor
on problem
size

work done
outside
recursive
calls

e.g. merge sort

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

$$a = 2$$

$$b = 2$$

$$d = 1$$

$$f(n) = \begin{cases} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

problem 9.5

← merge-sort

$$T(n) = O(n^1 \log n) = O(n \log n)$$



Use the Tree Theorem to determine the complexity of binary search. Discuss and then vote.

```
1 def binary_search(lst, val, lo, hi):
2     if lo > hi:
3         return -1
4     m = (lo + hi) // 2
5     if lst[m] == val:
6         return m
7     elif lst[m] > val:
8         return binary_search(lst, val, lo, m - 1)
9     else:
10        return binary_search(lst, val, m + 1, hi)
```

- A. $O(n)$
- B. $O(n \log n)$
- C. $O(\log n)$
- D. $O(1)$

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

$$f(n) = \begin{cases} O(n^d) & \text{if } a < b^d \\ O(n^d \log n) & \text{if } a = b^d \\ O(n^{\log_b a}) & \text{if } a > b^d \end{cases}$$

$$b = 2$$
$$d = 0$$

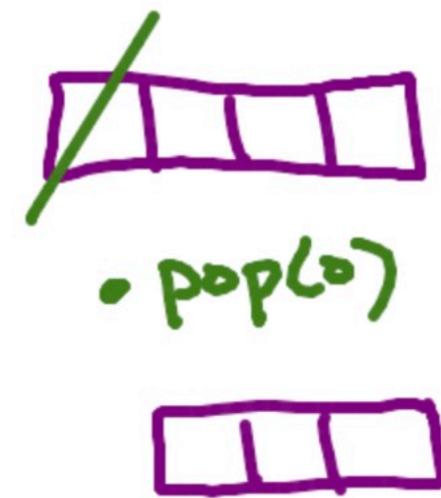
$$1 = 2^0$$
$$O(n^0 \log n) = O(\log n)$$



What about this implementation of **Merge-Sort**? Look up the complexity of **.pop(0)**. Discuss and then vote.

```
1 def merge_sort(a: list) -> list:
2     n = len(a)
3     if n <= 1: # base case
4         return a
5
6     # divide stage: sort two sublists
7     m = n // 2
8     a1 = merge_sort(a[:m]) # sort left sublist
9     a2 = merge_sort(a[m:]) # sort right sublist
10
11    # conquer stage: merge the sorted sublists
12    merged = []
13    while len(a1) > 0 and len(a2) > 0:
14        if a1[0] < a2[0]:
15            merged.append(a1.pop(0))
16        else:
17            merged.append(a2.pop(0))
18
19    while len(a1) > 0:
20        merged.append(a1.pop(0))
21
22    while len(a2) > 0:
23        merged.append(a2.pop(0))
24    return merged
```

$O(n^2)$



- A. $O(n^2)$
- B. $O(n \log n)$
- C. $O(n^2 \log n)$
- D. $O(\log(n^2))$

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

$d ? 2$

$$a < b^d$$
$$2 < 2^2$$
$$O(n^2)$$

