

# Divide and Conquer

# Divide and Conquer

- Break problem into several subproblems that are similar to the original problem but smaller in size
- Solve the subproblems recursively
- Combine these solutions to create a solution to the original problem
- Most of the time written recursively

# Decrease and Conquer

- Similar to divide and conquer
- Instead of partitioning a problem into multiple subproblems of smaller size, we use some technique to reduce our problem into a single problem that is smaller than the original.
- Solution of smaller size problem is solution of original problem

# Binary Search

- **Input:** An array of elements,  $A$  and a search element  $S$
- **Output:** Index of  $S$  in  $A$  if present and  $-1$  otherwise
- **Assumption:** All elements in  $A$  are positive
- **Pre-requisite:** Elements in  $A$  are in sorted order

# Illustration

A

1	2	3	4	5	6
21	34	37	41	45	50

S

37

# Illustration

A

1	2	3	4	5	6
21	34	37	41	45	50

S

50

# Illustration

A

1	2	3	4	5	6
21	34	37	41	45	50

S

41

# Illustration

A

1	2	3	4	5	6
21	34	37	41	45	50

S

43



# Algorithm

```
int Binary_Search(A,n,Key):
```

```
    l = 1; h = n
```

```
    while (l<=h):
```

```
        mid = (l+h)/2
```

```
        if (key==A[mid])
```

```
            return mid
```

```
        if (key<A[mid])
```

```
            h = mid-1
```

```
        else
```

```
            l = mid+1
```

```
    return 0
```

# Proof of Correctness

Strong Induction is used to prove

**Base case:** Prove that the proposition holds for  $n = 0$ , i.e., prove that  $P(0)$  is true.

**induction hypothesis:** Assume that  $P(n)$  holds for all  $n$  between 0 and  $k$

**Inductive step:** Prove that  $P(k+1)$  is true.

Conclude by strong induction that  $P(n)$  holds for all  $n \geq 0$ .

# Proof of Correctness

**Base case:** When  $n = 0$ ,  $l = 1$  and  $h = 0$ , loop of the algorithm do not get executed even once and 0 will be returned

Correct for base case

# Proof of Correctness

**Induction hypothesis:** When the size of the array,  $n \leq k$ , the algorithm will return correct answer

**Inductive step:** Prove that when the size of the array,  $n = k + 1$ , the algorithm will return correct answer

# Proof of Correctness

**Case 1:** When  $S = A[\text{mid}]$ , algorithm will return mid

**Case 2:** When  $S < A[\text{mid}]$ , since the array is sorted,  $S$  will be present in the subarray from index 1 to  $\text{mid}-1$ , as per induction hypothesis the algorithm will return correct answer

**Case 3:** When  $S > A[\text{mid}]$ , symmetrical case of case 2

# Induction or Substitution Method

$$T(n) = 1 \quad n=1$$

$$T(n/2) + 1 \quad n>1$$

$$T(n) = T(n/2) + 1 \quad \text{-----}(1)$$

$$T(n/2) = T(n/2^2) + 1 \quad \text{-----}(2)$$

Applying (2) in (1)

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

# Induction or Substitution Method

$$T(n/2^2) = T(n/2^3) + 1 \quad \text{-----}(3)$$

Applying (2) in (1)

$$T(n) = T(n/2^3) + 3$$

•  
•  
•

$$T(n) = T(n/2^k) + k$$

At the kth iteration, there will be only one element  $\Rightarrow n = 1$

$$n/2^k = 1$$

# Induction or Substitution Method

$$n = 2^k \text{ and } k = \log n$$

$$T(n) = 1 + \log n$$

$$O(\log n) \text{ or } \theta(\log n)$$



# Recurrence Tree Method

# Master's Theorem for Divide and Conquer

Consider a recurrence relation of the form:

$$T(n) = aT(n/b) + \theta(n^k \log^p n)$$

1) If  $a > b^k$ , then  $T(n) = \theta(n^{\log_b a})$

2) If  $a = b^k$ , then

a. If  $p > -1$ , then  $T(n) = \theta(n^{\log_b a} \log^{p+1} n)$

b. If  $p = -1$ , then  $T(n) = \theta(n^{\log_b a} \log \log n)$

c. If  $p < -1$  then  $T(n) = \theta(n^{\log_b a})$

# Master's Theorem for Divide and Conquer

3) If  $a < b^k$ ,

a. If  $p \geq 0$ , then  $T(n) = \theta(n^k \log^p n)$

b. If  $p < 0$ , then  $T(n) = \theta(n^k)$

# Master's Theorem for Binary Search

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

$$a = 1, b = 2, k = 0, p = 0$$

Case 2 of master theorem  $a = b^k$  and  $p > -1$

$$\theta(n^0 \log^1 n) = \theta(\log n)$$

# Master's Theorem for Binary Search

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

$$a = 2, b = 2, k = 1, p = 0$$

Case 2 of master theorem  $a = b^k$  and  $p > -1$

$$\theta(n^1 \log^1 n) = \theta(n \log n)$$

# Master's Theorem for Binary Search

$$T(n) = 4T(n/2) + n^2$$

$$a = 4, b = 2, k = 2, p = 0$$

Case 2 of master theorem  $a = b^k$  and  $p > -1$

$$\theta(n^2 \log^1 n) = \theta(n^2 \log n)$$

# Master's Theorem for Binary Search

$$T(n) = 4T(n/2) + n^2 \log n$$

$$a = 4, b = 2, k = 2, p = 1$$

Case 2 of master theorem  $a = b^k$  and  $p > -1$

$$\theta(n^2 \log^2 n)$$

# Master's Theorem for Binary Search

$$T(n) = 4T(n/2) + n^2 \log^2 n$$

$$a = 4, b = 2, k = 2, p = 2$$

Case 2 of master theorem  $a = b^k$  and  $p > -1$

$$\theta(n^2 \log^3 n)$$



# Master's Theorem for Binary Search

$$T(n) = 8T(n/2) + n^3$$

$$a = 8, b = 2, k = 3, p = 0$$

Case 2 of master theorem  $a = b^k$  and  $p > -1$

$$\theta(n^3 \log n)$$

# Merge Sort

Closely follows the divide–and–conquer paradigm

**Divide:** Divide the  $n$ –element sequence to be sorted into two subsequences of  $n/2$  elements each

**Conquer:** Sort the two subsequences recursively using merge sort

**Combine:** Merge the two sorted subsequences to produce the sorted answer

# Merge Sort

- Recursion stops when sequence to be sorted has a length 1
- Since every sequence of length 1 is already in sorted order
- Key operation of merge sort algorithm is merging of two sorted sequences in “combine” step
- Let us use an Auxiliary procedure MERGE to do the combine operation

# Merge Sort

- MERGE (A, p, q, r), where A is an array and p, q, and r are indices into the array such that  $p \leq q < r$
- Procedure assumes that the subarrays A[p..q] and A[q+1..r] are in sorted order
- It merges them to form a single sorted subarray that replaces the current subarray A[p..r]

# Merge in Merge Sort

- MERGE procedure takes time  $\theta(n)$ , where  $n = r - p + 1$  is the total number of elements being merged
- Shall be illustrated with two piles of cards face up on a table
- Each pile is sorted, with the smallest cards on top
- We wish to merge the two piles into a single sorted output pile, which is to be face down on the table

# Merge in Merge Sort

## Basic Step

- (i) Choose the smaller of the two cards on top of the face-up piles, removing it from its pile (which exposes a new top card),
  - (ii) Place this card face down onto the output pile
- repeat this step until one input pile is empty then take the remaining input pile and place it face down onto the output pile

# Merge in Merge Sort

- Computationally, each basic step takes constant time, since we are comparing just the two top cards
- Since we perform at most  $n$  basic steps, merging takes  $\theta(n)$  time
- We shall make the algorithm more clear by adding a sentinel card which has a very large value

# Illustration of Merge in Merge Sort

	8	9	10	11	12	13	14	15	16	17
A	...	2	4	5	7	1	2	3	6	...
	$k$									
L	1	2	3	4	5					
	2	4	5	7	$\infty$					
	$i$									
R	1	2	3	4	5					
	1	2	3	6	$\infty$					
	$j$									

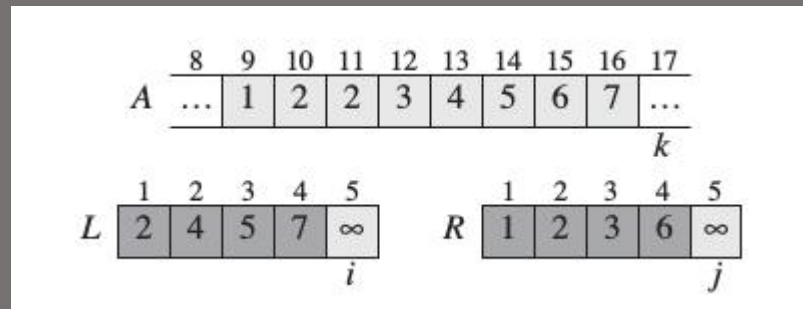
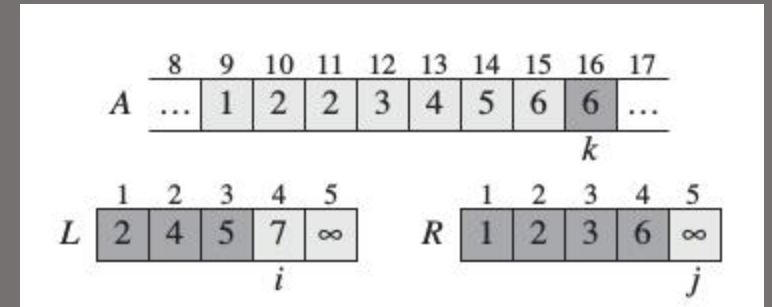
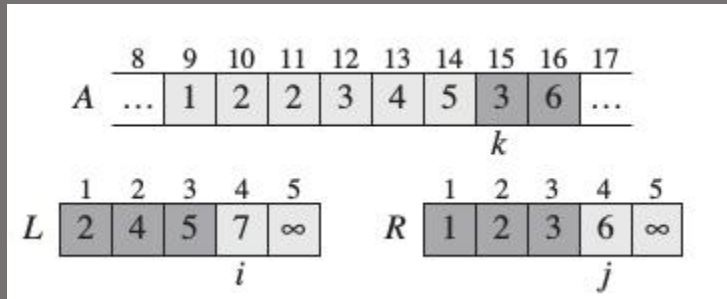
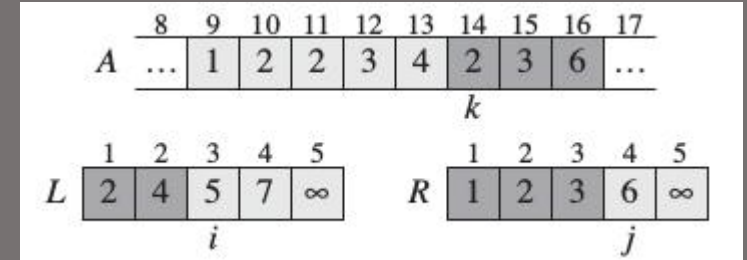
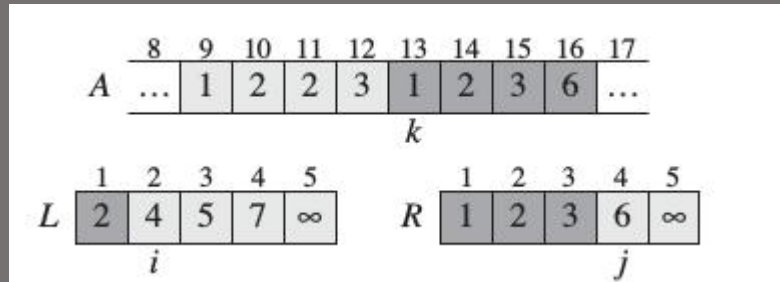
	8	9	10	11	12	13	14	15	16	17
A	...	1	2	5	7	1	2	3	6	...
	$k$									
L	1	2	3	4	5					
	2	4	5	7	$\infty$					
	$i$									
R	1	2	3	4	5					
	1	2	3	6	$\infty$					
	$j$									

	8	9	10	11	12	13	14	15	16	17
A	...	1	4	5	7	1	2	3	6	...
	$k$									
L	1	2	3	4	5					
	2	4	5	7	$\infty$					
	$i$									
R	1	2	3	4	5					
	1	2	3	6	$\infty$					
	$j$									

	8	9	10	11	12	13	14	15	16	17
A	...	1	2	2	7	1	2	3	6	...
	$k$									
L	1	2	3	4	5					
	2	4	5	7	$\infty$					
	$i$									
R	1	2	3	4	5					
	1	2	3	6	$\infty$					
	$j$									



# Illustration of Merge in Merge Sort



# Merge in Merge Sort

**MERGE**( $A, p, q, r$ )

```
1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  let  $L[1 \dots n_1 + 1]$  and  $R[1 \dots n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
12 for  $k = p$  to  $r$ 
13     if  $L[i] \leq R[j]$ 
14          $A[k] = L[i]$ 
15          $i = i + 1$ 
16     else  $A[k] = R[j]$ 
17          $j = j + 1$ 
```

# Proof of Correctness of Merge Operation

Lines 10 – 17, of the algorithm, perform the  $r - p + 1$  basic steps by maintaining the following loop invariants:

- (i) At the start of each iteration of the for loop of lines 12 – 17, the subarray  $A[p..k-1]$  contains  $k - p$  smallest elements of  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$ , in sorted order.
- (ii) Moreover,  $L[i]$  and  $R[j]$  are the smallest elements of their arrays that have not been copied back into  $A$ .

## To Show

- Loop invariant holds prior to first iteration of for loop of lines 12 – 17
- Each iteration of the loop maintains the invariant
- Invariant provides a useful property to show correctness when the loop terminates

# Initialization

- Prior to the first iteration of the loop, we have  $k = p$ , so that the subarray  $A[p..k-1]$  is empty
- This empty subarray contains the  $k - p = 0$  smallest elements of  $L$  and  $R$ , and since  $i = j = 1$ , both  $L[i]$  and  $R[j]$  are the smallest elements of their arrays that have not been copied back into  $A$ .

# Maintenance of Loop Invariant

- Let  $L[i] \leq R[j]$ , then  $L[i]$  is the smallest element not yet copied back into  $A$
- Because  $A[p..k-1]$  contains the  $k - p$  smallest elements, after line 14 copies  $L[i]$  into  $A[k]$ , the subarray  $A[p .. k]$  will contain the  $k - p + 1$  smallest elements

# Termination of Loop Invariant

- At termination,  $k = r + 1$ . By the loop invariant, the subarray  $A[p..k-1]$ , which is  $A[p..r]$ , contains the  $k - p = r - p + 1$  smallest elements of  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$ , in sorted order
- The arrays  $L$  and  $R$  together contain  $n_1 + n_2 + 2 = r - p + 3$  elements

# Termination of Loop Invariant

- All but the two largest have been copied back into A, and these two largest elements are the sentinels.



# Time Complexity of Merge Operation

- Each of lines 1 – 3 and 8 – 11 takes constant time
- for loops of lines 4 – 7 takes  $\theta(n_1 + n_2) = \theta(n)$  time
- There are  $n$  iterations of the for loop of lines 12 – 17, each of which takes constant time
- Here  $n = r - p + 1$

# Forming Recurrence Relation

- When we have  $n > 1$  elements running time is as follows:
- **Divide:** Just computes the middle of the subarray, which takes constant time. Thus,  $D(n) = \theta(1)$
- **Conquer:** We recursively solve two subproblems, each of size  $n/2$ , which contributes  $2T(n/2)$  to the running time.
- **Combine:** MERGE procedure on an  $n$ -element subarray takes time  $\theta(n)$ , and so  $C(n) = \theta(n)$

# Recurrence Relation

- $T(n) = \theta(1)$  if  $n = 1$   
 $= 2T(n/2) + \theta(n)$  if  $n > 1$

# Substitution Method

$$T(n) = 2T(n/2) + n \text{-----(1)}$$

$$T(n/2) = 2 * T(n/2^2) + n/2 \text{-----(2)}$$

Substiute (2) in (1)

$$T(n) = 2 * (2 * T(n/2^2) + n/2) + n$$

$$= 2^2 * T(n/2^2) + 2*n$$

.....

$$T(n) = 2^k * T(n/2^k) + k*n$$

# Substitution Method

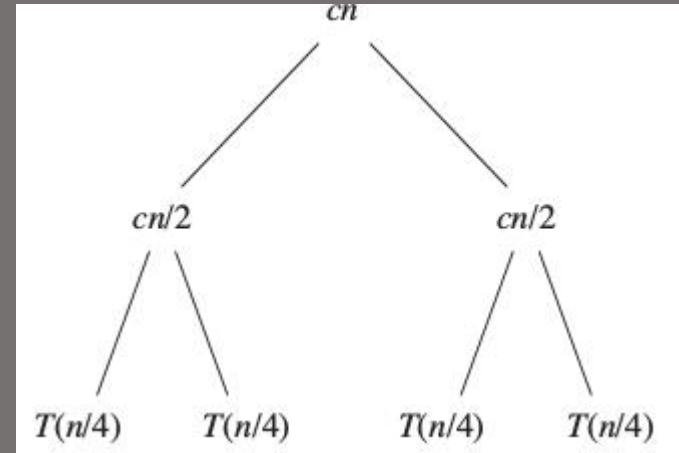
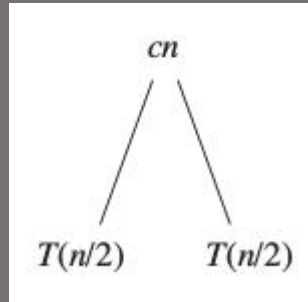
- In the  $k^{\text{th}}$  iteration, the value of  $n$  becomes 1
- therefore  $n/2^k = 1$
- $n = 2^k$
- $k = \log n$
- $T(n) = 2^k * T(n/2^k) + k*n$  becomes
- $T(n) = n + n \log n$
- $T(n) = \theta (n \log n) = O(n \log n)$

# Relationship between Time Functions

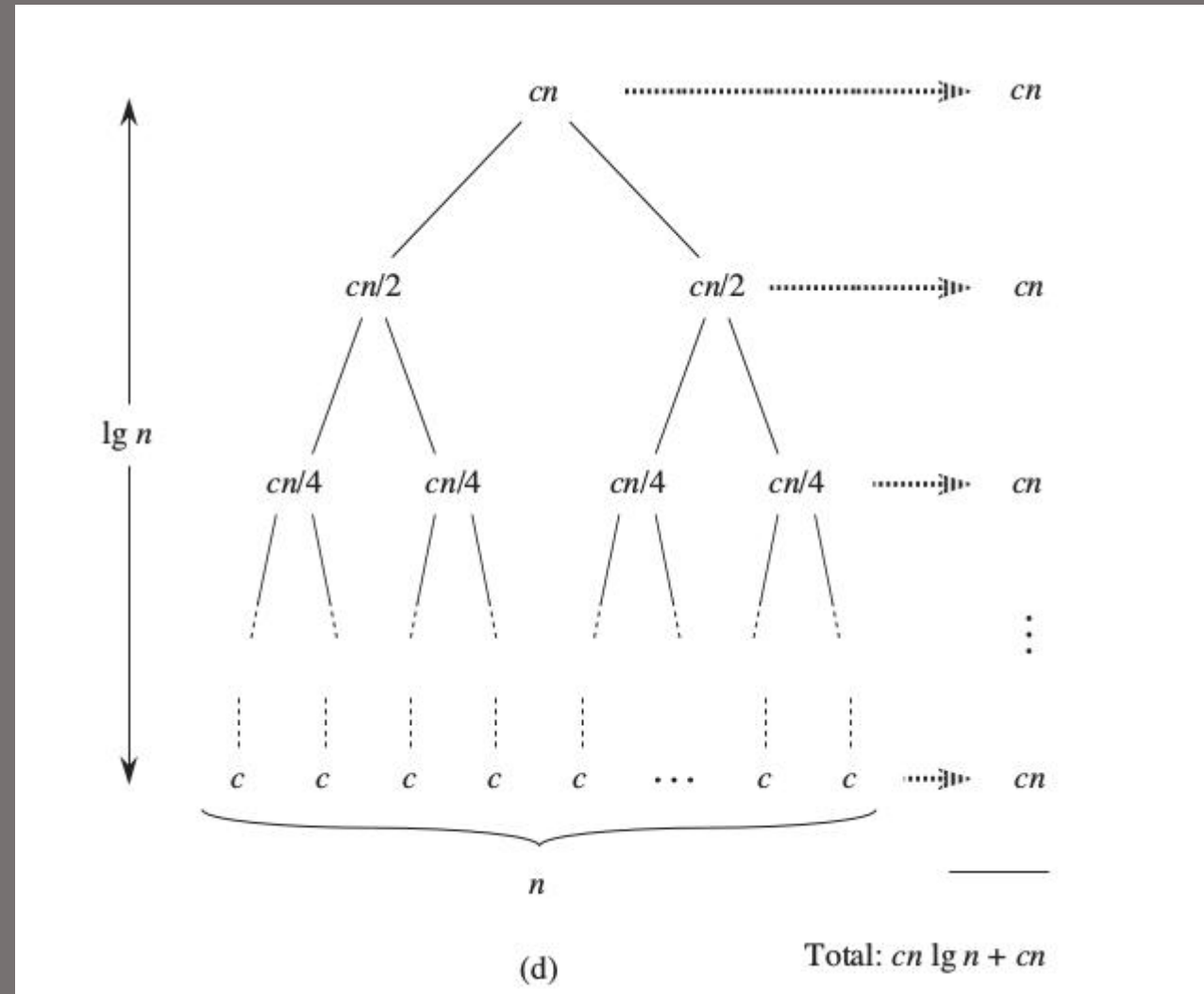
$$1 < \log(n) < \sqrt{n} < n < n \log n < n^2 < n^3 < \dots < 2^n < 3^n \dots n! < \dots < n^n$$

# Recurrence Tree

$T(n)$



# Recurrence Tree





# Master's Theorem for Divide and Conquer

Consider a recurrence relation of the form:

$$T(n) = aT(n/b) + \theta(n^k \log^p n)$$

1) If  $a > b^k$ , then  $T(n) = \theta(n^{\log_b a})$

2) If  $a = b^k$ , then

a. If  $p > -1$ , then  $T(n) = \theta(n^{\log_b a} \log^{p+1} n)$

b. If  $p = -1$ , then  $T(n) = \theta(n^{\log_b a} \log \log n)$

c. If  $p < -1$  then  $T(n) = \theta(n^{\log_b a})$

# Master's Theorem for Divide and Conquer

3) If  $a < b^k$ ,

a. If  $p \geq 0$ , then  $T(n) = \theta(n^k \log^p n)$

b. If  $p < 0$ , then  $T(n) = \theta(n^k)$