
Divide and Conquer Approach

Merge Sort, Quick Sort, Binary Search

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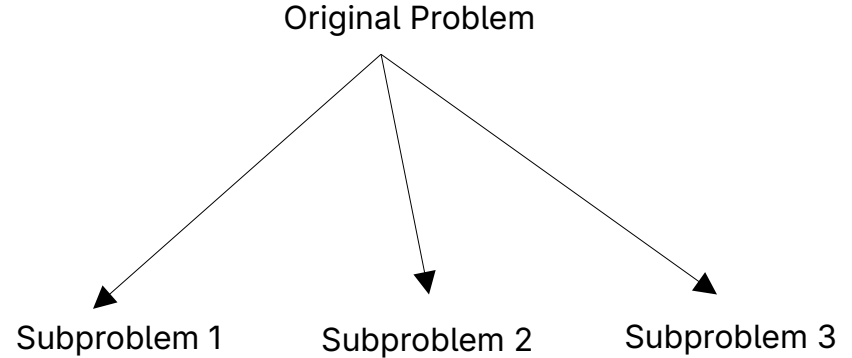
Recursion

- Recursion is a method of solving a problem where the solution depends on solutions to smaller instances of the same problem.
- In programming, this means a function calls itself.
- Example: Calculating a Factorial – $n! = n(n-1)! = n(n-1)(n-2)!$ and so on.

Divide and Conquer Approach

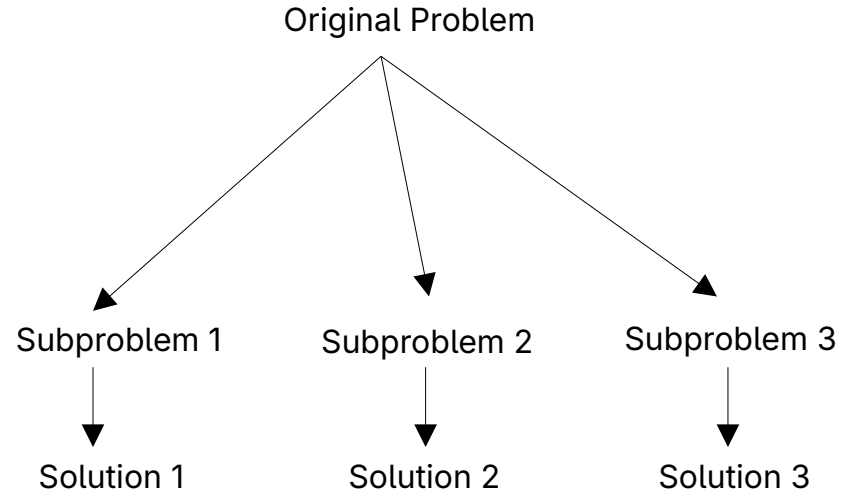
The divide and conquer paradigm involves three steps at each level of the recursion:

- **Divide** the problem into a number of subproblems that are smaller instances of the same problem.



Divide and Conquer Approach

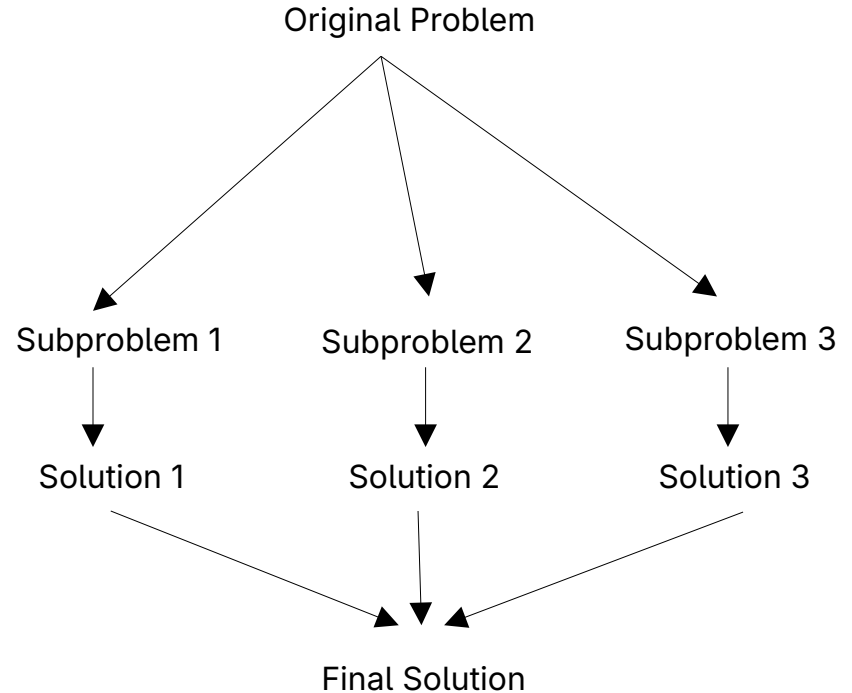
- **Conquer** the subproblems by solving them recursively. If the subproblem sizes are small enough, however, just solve the subproblems in a straightforward manner.



Divide and Conquer Approach

- **Combine** the solutions to the subproblems into the solution for the original problem.

Examples: Merge sort, Quick sort, Binary search.

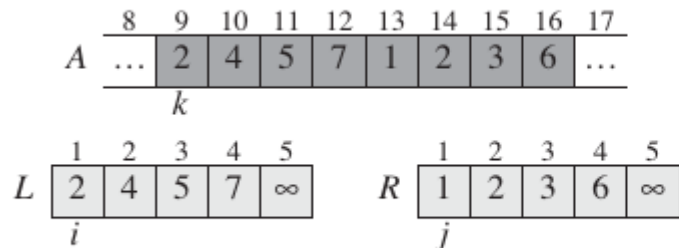


Merge Sort

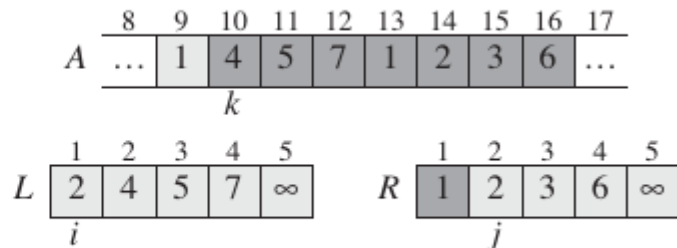
The merge sort algorithm closely follows the divide-and-conquer paradigm. Intuitively, it operates as follows.

- **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.

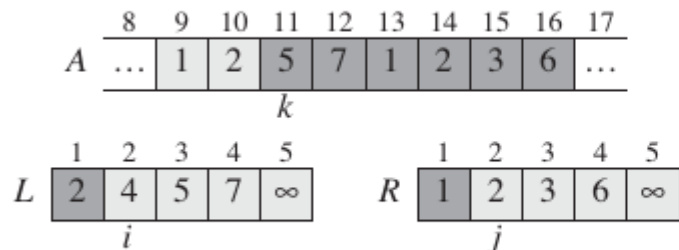
Merge Sort Simulation



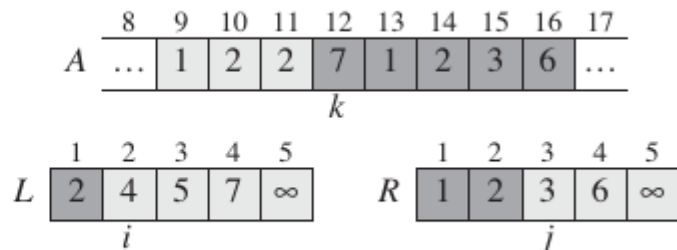
(a)



(b)

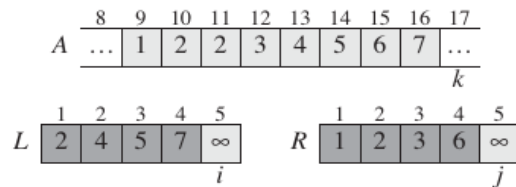
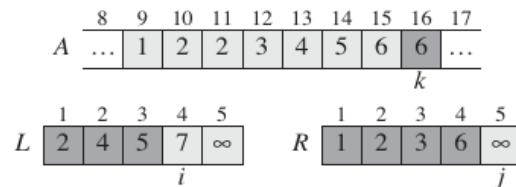
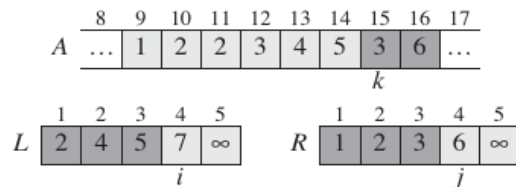
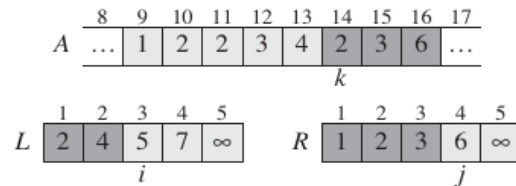
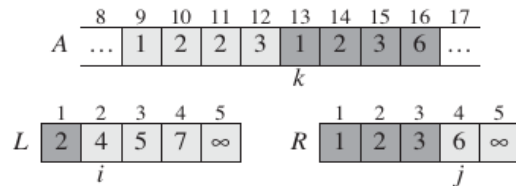


(c)



(d)

Merge Sort Simulation



Algorithms and Complexities

MERGE-SORT(A, p, r)

```
1  if  $p < r$ 
2       $q = \lfloor (p + r)/2 \rfloor$ 
3      MERGE-SORT( $A, p, q$ )
4      MERGE-SORT( $A, q + 1, r$ )
5      MERGE( $A, p, q, r$ )
```

- **Time Complexity:** $O(n \log n)$ in all cases. MERGE procedure takes time $\Theta(n)$ at each of the $\log n$ levels of recursion.
- **Space Complexity:** $O(n)$ because it requires an auxiliary array of size proportional to the input array size, n , to hold the merged elements during the sorting process.

MERGE(A, p, q, r)

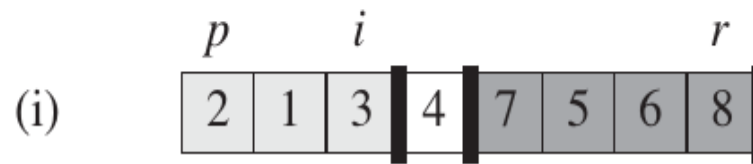
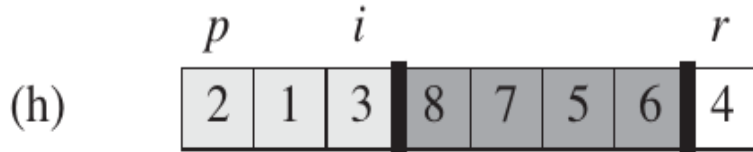
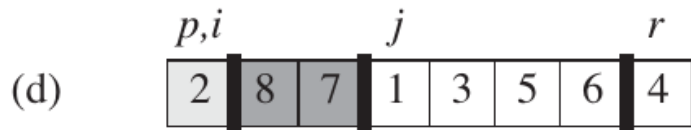
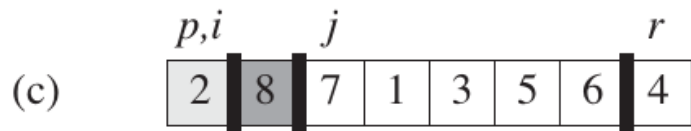
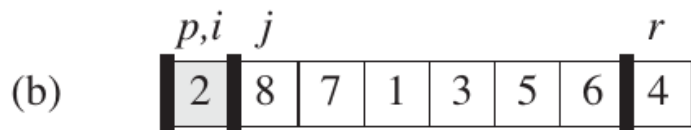
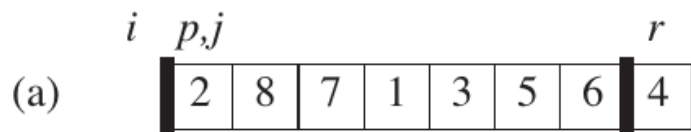
```
1   $n_1 = q - p + 1$ 
2   $n_2 = r - q$ 
3  let  $L[1..n_1 + 1]$  and  $R[1..n_2 + 1]$  be new arrays
4  for  $i = 1$  to  $n_1$ 
5       $L[i] = A[p + i - 1]$ 
6  for  $j = 1$  to  $n_2$ 
7       $R[j] = A[q + j]$ 
8   $L[n_1 + 1] = \infty$ 
9   $R[n_2 + 1] = \infty$ 
10  $i = 1$ 
11  $j = 1$ 
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          $j = j + 1$ 
```

Quick Sort

Here is the three-step divide-and-conquer process for sorting a typical subarray $A[p \dots r]$:

- **Divide:** Partition (rearrange) the array $A[p \dots r]$ into two (possibly empty) subarrays $A[p \dots q-1]$ and $A[q+1 \dots r]$ such that each element of $A[p \dots q-1]$ is less than or equal to $A[q]$, which is, in turn, less than or equal to each element of $A[q+1 \dots r]$. Compute the index q as part of this partitioning procedure.
- **Conquer:** Sort the two subarrays $A[p \dots q-1]$ and $A[q+1 \dots r]$ by recursive calls to quicksort.
- **Combine:** Because the subarrays are already sorted, no work is needed to combine them. The entire array $A[p \dots r]$ is now sorted.

The Partition Operation



Algorithms and Complexities

QUICKSORT(A, p, r)

```
1  if  $p < r$ 
2       $q = \text{PARTITION}(A, p, r)$ 
3      QUICKSORT( $A, p, q - 1$ )
4      QUICKSORT( $A, q + 1, r$ )
```

PARTITION(A, p, r)

```
1   $x = A[r]$ 
2   $i = p - 1$ 
3  for  $j = p$  to  $r - 1$ 
4      if  $A[j] \leq x$ 
5           $i = i + 1$ 
6          exchange  $A[i]$  with  $A[j]$ 
7  exchange  $A[i + 1]$  with  $A[r]$ 
8  return  $i + 1$ 
```

To sort an entire array A , the initial call is QUICKSORT($A, 1, A.\text{length}$).

- **Time Complexity:** $O(n^2)$ in the worst case when the pivot selection consistently results in highly unbalanced partitions (e.g., always picking the smallest or largest element). Time complexity is $O(n \log n)$ in other cases.
- **Space Complexity:** Typically $O(\log n)$, because it uses a recursive call stack whose maximum depth is proportional to the level of partitions.

Reference

- ***Introduction to Algorithms, Third Edition***
 - ***Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, Clifford Stein***