

Keynotes and Preliminary Knowledge from Algorithm Design and Applications

Chapter 8.1: Merge-Sort

Merge-Sort Overview:

- Divide-and-Conquer Paradigm:

1. Divide: Split the input data into two disjoint subsets.
2. Recur: Recursively solve the subproblems.
3. Conquer: Merge the solutions of the subproblems to solve the original problem.

Steps in Merge-Sort:

1. Divide: If the sequence has one or zero elements, it's already sorted. If not, split it into two sequences S_1 and S_2 .
2. Recur: Recursively sort S_1 and S_2 .
3. Conquer: Merge S_1 and S_2 into a single sorted sequence.

Merge-Sort Tree:

- Each node represents a recursive call.
- The height of the tree is $\log n$, where n is the number of elements.
- The merge operation takes $O(n)$ time.

Analysis:

- The merge step involves sequentially accessing data, which is efficient for large datasets.
- Merge-sort has a time complexity of $O(n \log n)$.

Pseudocode for Merge:

Algorithm merge(S_1 , S_2 , S):

Input: Two arrays, S_1 and S_2 , of size n_1 and n_2 , respectively, sorted in non-decreasing order, and an empty array, S , of size at least $n_1 + n_2$

Output: S , containing the elements from S_1 and S_2 in sorted order

$i \leftarrow 1$

$j \leftarrow 1$

while $i \leq n_1$ and $j \leq n_2$ do

 if $S_1[i] \leq S_2[j]$ then

$S[i + j - 1] \leftarrow S_1[i]$

$i \leftarrow i + 1$

 else

$S[i + j - 1] \leftarrow S_2[j]$

$j \leftarrow j + 1$

while $i \leq n_1$ do

$S[i + j - 1] \leftarrow S_1[i]$

$i \leftarrow i + 1$

while $j \leq n_2$ do

$S[i + j - 1] \leftarrow S_2[j]$

$j \leftarrow j + 1$

Chapter 8.2: Quick-Sort

Quick-Sort Overview:

- Steps in Quick-Sort:

1. Divide: Choose a pivot and partition the sequence into L , E , and G .
2. Recur: Recursively sort L and G .

3. Conquer: Concatenate L, E, and G back into S.

In-Place Quick-Sort:

- Utilizes the input array for storing subarrays.
- Partitions the array using a pivot, and recursively sorts subarrays.

Visualization:

- Uses a binary recursion tree where each node represents a recursive call and its pivot.

Chapter 11.1: Recurrences and the Master Theorem

Master Theorem for Divide-and-Conquer Recurrences:

- Used to solve recurrence relations of the form $T(n) = aT(n/b) + f(n)$.
- Cases:
 1. If $f(n)$ is $O(n^{(\log_b a - \epsilon)})$, then $T(n)$ is $\Theta(n^{\log_b a})$.
 2. If $f(n)$ is $\Theta(n^{\log_b a} \log^k n)$, then $T(n)$ is $\Theta(n^{\log_b a} \log^{k+1} n)$.
 3. If $f(n)$ is $\Omega(n^{(\log_b a + \epsilon)})$ and $af(n/b) \leq \delta f(n)$, then $T(n)$ is $\Theta(f(n))$.

Examples:

- $T(n) = 4T(n/2) + n$ resolves to $\Theta(n^2)$.
- $T(n) = 2T(n/2) + n \log n$ resolves to $\Theta(n \log^2 n)$.

Chapter 11.2: Integer Multiplication

Problem:

- Multiplying large integers efficiently, which has applications in data security.

Divide-and-Conquer Algorithm:

- Split integers into halves and recursively compute products.
- Time complexity can be reduced to $O(n^{1.585})$ using the divide-and-conquer approach.

Theorem:

- Multiplying two n -bit integers can be done in $O(n^{(\log_2 3)})$.

Chapter 11.3: Matrix Multiplication

Problem:

- Efficiently multiplying two $n \times n$ matrices.

Strassen's Algorithm:

- Uses 7 recursive multiplications instead of 8.
- Running time: $O(n^{2.808})$.

Further Improvements:

- More advanced algorithms can achieve running times as low as $O(n^{2.376})$.

Chapter 11.4: The Maxima-Set Problem

Problem:

- Finding a set of maximal points from a set of points in a plane.

Divide-and-Conquer Algorithm:

- Sort points lexicographically.
- Recursively find maxima sets on the left and right.
- Merge these sets by removing dominated points.

Time Complexity:

- $O(n \log n)$.