

Lecture 12: Problem Paradigms & Complexity Classes

Contents to cover:

- ~~Divide and conquer~~
- ~~Important: Stimulation & Randomization~~
- ~~Monte Carlo method for calculating π~~
- ~~Approximation~~
- Optimization
- Generate-and-Test (L11)
- Understanding the NP problem } \rightarrow Not assessable
- Algorithms classifications

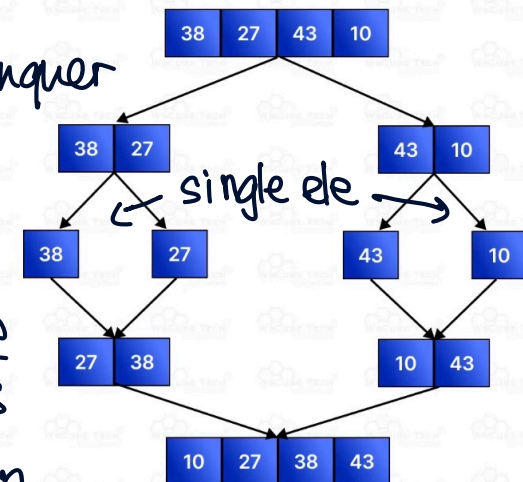
Merge Sort follows the divide-and-conquer paradigm (pattern):

Divide: split the array into halves at midpoint

Conquer: Recursively sort each half until you reach single elements

Merge: Merge the sorted halves in sorted order

Merge Sort Algorithm

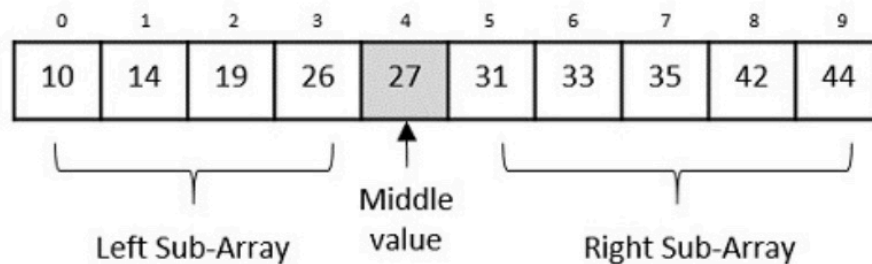


Has consistent, guaranteed time complexity of $O(n \log n)$
 $O(n \log n) = O(n)$ operations * $O(\log n)$ levels

Binary Search (\neq Binary Search Tree - BST) works on sorted arrays

- How Binary search works:

- check the middle element of the sorted array (ASC)
- if
 - target == middle element \rightarrow return the index
 - target < middle element \rightarrow return the left half
 - target > middle element \rightarrow return the right half
- recurse until found or search space is empty
- Time complexity:
 - $O(1)$ - Best case
 - $O(\log n)$ - Average & Worst case



Quick sort follows the divide-and-conquer paradigm

Pivot (Last element)

Partition: Arrange the array s.t. smaller elements come before it; larger elements come after it

Recursively sort: Apply quicksort to the subarrays on both sides of the pivot

Combine

Time complexity:

Best & Average case: $O(n \log n) = O(\log n) \text{ level} \times O(n) \text{ elements}$

Worst case: $O(n^2)$ - the pivot is always the smallest or largest element in the array

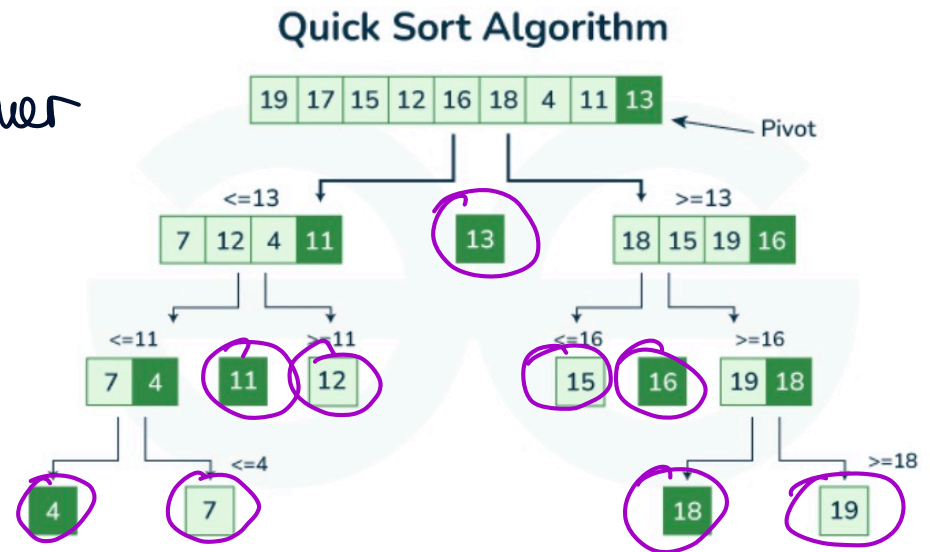
\Rightarrow Unbalanced partition: 1 side as n element, 1 side has 0 element

Longest Common prefix (LCP)

Average: $O(\log n)$ due to recursion stack

Worst: $O(m \times n)$ m -string; n -shortest string

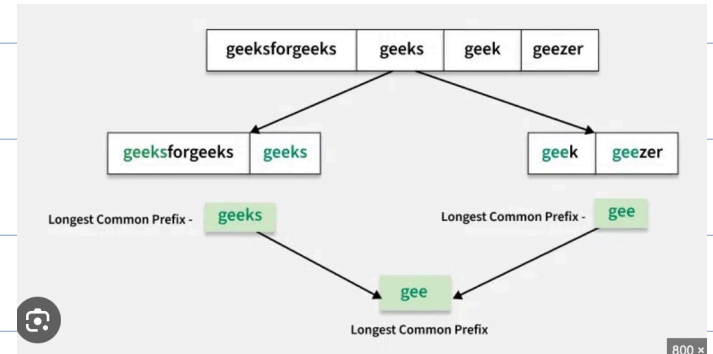
Divide: Split the array of strings into 2 halves



Conquer: Recursively find the longest common prefix for each half

Combine: Compare the 2 prefixes char by char to find the common prefix

- Vital in bioinformatics for DNA and protein sequence analysis



n^{th} Fibonacci #: Time complexity $O(2^n)$ – Not factorial time

```
#include <stdio.h>
```

```
// Function to calculate the nth Fibonacci number using recursion
```

```
int nthFibonacci(int n){
```

```
    // Base case: if n is 0 or 1, return n
```

```
    if (n <= 1){
```

```
        return n;
```

```
    }
```

```
    // Recursive case: sum of the two preceding Fibonacci numbers
```

```
    return nthFibonacci(n - 1) + nthFibonacci(n - 2);
```

```
}
```

Note: Factorial time complexity will occur if you were generating all possible permutations. For Fib, it just makes overlapping calls

Fibo problem using Dynamic Programming (DP) - solve complex problem by breaking them into simpler subproblems and storing the result to avoid redundancy

2 main approaches

Memorization (Top down): uses recursion with a cache
original problem \rightarrow Base case

Tabulation (Bottom up): uses iteration to build solutions from smallest subproblems upwards

Common DP problems

Fibonacci

Knapsack ?

LC subsequences

Matrix chain

Fibo - Tabulation/ Bottom-Up

```
#include <stdio.h>
```

```
// Function to calculate the nth Fibonacci number  
// using iteration
```

```
int nthFibonacci(int n) {
```

```
    // Handle the edge cases
```

```
    if (n <= 1) return n;
```

```
    // Create an array to store Fibonacci numbers
```

```
    int dp[n + 1];
```

```
    // Initialize the first two Fibonacci numbers
```

```
    dp[0] = 0;
```

```
    dp[1] = 1;
```

```
    // Fill the array iteratively
```

```
    for (int i = 2; i <= n; ++i) {  
        dp[i] = dp[i - 1] + dp[i - 2];  
    }
```

```
    // Return the nth Fibonacci number
```

```
    return dp[n];
```

```
}
```


Uses random number generation to solve problems that's difficult/expensive to solve analytically

Monte Carlo π estimation:

Mathematical foundation:

- Consider a unit circle inscribed in a square
- $\frac{\text{Circle area}}{\text{Square area}} = \frac{\pi}{4}$
- Random points falling inside circle vs. total point: $\frac{\pi}{4}$

Algorithm (Pseudocode)

```
c
for (many iterations) {
  x = random number between -1 and 1
  y = random number between -1 and 1

  if (x2 + y2 <= 1) {
    points_inside_circle++
  }
  total_points++
}

estimated_pi = 4 * (points_inside_circle / total_points)
```

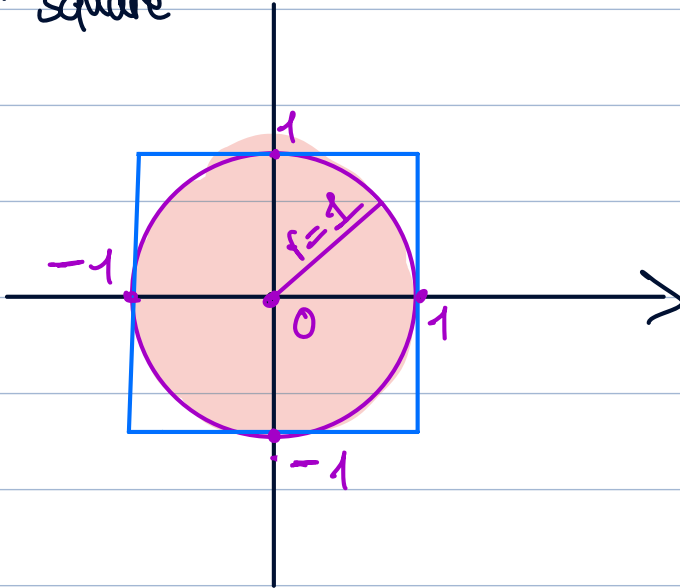
Convergence: As you increase the number of random samples, your estimate approaches the true value of π 2 3.

②

Stimulation & Randomisation

$$A_{\text{circle}} = \pi r^2 = \pi 1^2 = \pi$$

$$A_{\text{square}} = 2^2 = 4$$



$$\therefore \frac{\text{circle area}}{\text{square area}} = \frac{\pi}{4}$$

$$\therefore \pi = 4 \times \frac{\text{circle area}}{\text{square area}}$$

Monte Carlo vs. Las Vegas Algorithm

Monte Carlo	Las Vegas
<ul style="list-style-type: none">· Can run forever· No guarantee of correct answer· Provide approximate / probabilistic results· Used when approximate ans is OK, statistical sampling, simulation problems	<ul style="list-style-type: none">· Always terminate w/ finite runtime· Always correct when terminate· Use randomness to improve average-case performance. Eg:<ul style="list-style-type: none">+ Randomised qsort+ Randomised BST (to maintain balance)+ ... primality testing (certain variants)

③ **Approximation** solve a simpler version of the problem that approximates the original with a known and converging error
→ Numerical analysis