

Chandpur Science And Technology University



Lab Report

Course Code : CSE 2201

Course Title : Algorithm Design and Analysis

Experiment no : 01

Experiment Name: Introduction to Algorithm Design & Complexity Analysis

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Experiment # 1: Linear Search & Step Analysis

Objective:

To implement the Linear Search algorithm in C, analyze its performance by counting the number of steps taken to find an element in arrays of different sizes ($n = 10, 100, 1000$), and compare theoretical and practical results for worst-case scenarios.

Algorithm:

- 1) Start from the first element of the array.
- 2) Compare the current element with the key:
 - If the element matches the key, return its index (position).
- 3) If the element does not match, move to the next element.
- 4) Repeat steps 2 and 3 until:
 - The key is found, or
 - The end of the array is reached.
- 5) If the key is not found after scanning the entire array, return -1 to indicate "not found".

Theoretical Solution of given problem :

- Best Case: Key is at the first index \rightarrow Steps = 1
- Average Case: Key is in the middle \rightarrow Steps = $n / 2$
- Worst Case: Key is at the last index or not found \rightarrow Steps = n

Expected Worst-Case Step Counts:

Array Size (n)	Expected Steps (Worst Case)
10	10
100	100
1000	1000

Practical Work:

a. Pseudocode:

```
FUNCTION LinearSearch(array, size, key):  
    SET steps = 0  
    FOR i FROM 0 TO size-1 DO:  
        steps = steps + 1  
        IF array[i] == key THEN:
```

RETURN (i, steps)
END FOR
RETURN (-1, steps)

FUNCTION TestSearch(size, key):
 CREATE array of given size
 FILL array with sorted values from 1 to size
 CALL LinearSearch(array, size, key)
 PRINT "Steps taken"

Source Code in C :

```
#include <stdio.h>

int linearSearch(int arr[], int n, int key, int* steps) {
    for (int i = 0; i < n; i++) {
        (*steps)++;
        if (arr[i] == key)
            return i;
    }
    return -1;
}
```

```
void testSearch(int n, int key) {
    int arr[n];
    int steps = 0;
```

```
    for (int i = 0; i < n; i++) {
        arr[i] = i + 1;
    }
```

```
    int index = linearSearch(arr, n, key, &steps);
```

```
    printf("Array size: %d, Search key: %d\n", n, key);
    if (index != -1)
        printf("Key found at index %d\n", index);
}
```

```
else
    printf("Key not found.\n");

printf("Steps taken: %d\n\n", steps);
}
// worst case
int main() {
    testlinearsearch(10, 10);
    testlinearsearch(100, 100);
    testlinearsearch(1000, 1000);

    return 0;
}
```

Output:

Array size: 10, Search key: 10

Key found at index 9

Steps taken: 10

Array size: 100, Search key: 100

Key found at index 99

Steps taken: 100

Array size: 1000, Search key: 1000

Key found at index 999

Steps taken: 1000

Observations:

- Step Count Grows Linearly

As expected, the number of steps taken is directly proportional to the array size. For every additional element, one more comparison is needed in the worst case.

- Experimental vs. Theoretical Results

The experimental results matched the theoretical analysis exactly:

- For $n = 10 \rightarrow \text{Steps} = 10$

- For $n = 100 \rightarrow \text{Steps} = 100$

- For $n = 1000 \rightarrow \text{Steps} = 1000$

- Predictable Behavior

Linear Search behaves consistently — the steps are predictable and increase at a steady rate as the size of the array increases.

Challenges:

- Handling Large Input Sizes

When testing with large arrays (e.g., $n = 1000$ or more), memory management becomes crucial, especially in environments with limited stack size. Using variable-length arrays can lead to stack overflow in some compilers.

- Worst-Case Identification

Ensuring that the search key is placed at the worst-case position (i.e., last index) during testing is important for accurate step analysis. Any mistake here can give misleading performance data

Conclusion :

In this project, we successfully implemented the Linear Search algorithm in C and analyzed its performance across different input sizes ($n = 10, 100, 1000$). By counting the number of steps (comparisons) taken during the search, we validated the theoretical time complexity of Linear Search.

Key takeaways:

- Linear Search has a time complexity of $O(n)$, meaning its performance degrades linearly with the size of the array.
- The number of steps in the worst case is equal to the size of the array.
- While it is simple and easy to implement, Linear Search is not suitable for large datasets due to its inefficiency.
- More optimized algorithms like Binary Search are recommended when the data is sorted and speed is essential.

Experiment # 2: Binary Search & Step Analysis:

Objective:

To implement the Binary Search algorithm in C, count the number of steps (comparisons) taken for different input sizes ($n = 10, 100, 1000$), and compare the practical step count with the theoretical analysis for worst-case scenarios.

Algorithm:

Binary Search is an efficient search algorithm that works on sorted arrays. It repeatedly divides the search interval in half. If the key is less than the middle element, it searches in the left half; otherwise, it searches in the right half.

Steps:

Start with two pointers: low and high.

While $\text{low} \leq \text{high}$:

Calculate $\text{mid} = (\text{low} + \text{high}) / 2$.

If $\text{key} == \text{arr}[\text{mid}]$, return mid.

If $\text{key} < \text{arr}[\text{mid}]$, search the left half.

If $\text{key} > \text{arr}[\text{mid}]$, search the right half.

If the key is not found, return -1.

Theoretical Solution of the Problem:

Time complexity of Binary Search:

1) Best Case: $O(1)$ (when the key is at the middle)

2) Average & Worst Case: $O(\log_2 n)$

Theoretical Step Counts (Worst Case):

Array Size	(n)	$\log_2(n)$	(approx) Steps	(Worst Case)
------------	-------	-------------	----------------	--------------

10		≈ 4	4	
----	--	-------------	---	--

100		≈ 7	7	
-----	--	-------------	---	--

Array Size (n) $\log_2(n)$ (approx) Steps (Worst Case)

1000 \approx 10 10

Practical Work:

a. Pseudocode:

WHILE low \leq high DO

 INCREMENT steps_ref by 1

 SET mid \leftarrow (low + high) / 2

 IF array[mid] = key THEN

 RETURN mid // Key found

ELSE IF array[mid] < key THEN

SET low \leftarrow mid + 1 // Search right half

ELSE

SET high \leftarrow mid - 1 // Search left half

END WHILE

RETURN -1 // Key not found

b. Source Code in C:

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

```
int binarySearch(int arr[], int n, int key, int* steps) {  
    int low = 0, high = n - 1;
```

```
    while (low <= high) {  
        (*steps)++;  
        int mid = (low + high) / 2;
```

```
        if (arr[mid] == key)  
            return mid;  
        else if (arr[mid] < key)  
            low = mid + 1;  
        else  
            high = mid - 1;  
    }
```

```
    return -1;  
}
```

```
void testbinarysearch(int n, int key) {  
    int arr[n];  
    int steps = 0;
```

```
    for (int i = 0; i < n; i++) {  
        arr[i] = i + 1;  
    }
```

```
    int index = binarySearch(arr, n, key, &steps);
```

```
    printf("Array size: %d, Search key: %d\n", n, key);  
    if (index != -1)  
        printf("Key found index %d\n", index);  
    else  
        printf("Key not found.\n");
```

```
    printf("Steps taken: %d\n\n", steps);  
}
```

```
// worst case  
int main() {  
    testbinarysearch(10, 10);  
    testbinarysearch(100, 100);  
    testbinarysearch(1000, 1000);
```

```
    return 0;
```

```
}
```

Output:

Key found at index 9

Steps taken: 4

Array size: 100, Search key: 100

Key found at index 99

Steps taken: 7

Array size: 1000, Search key: 1000

Key found at index 999

Observations:

The number of steps grows logarithmically with the size of the array.

Experimental results confirm the theoretical estimates:

For $n = 10 \rightarrow \text{Steps} \approx 4$

For $n = 100 \rightarrow \text{Steps} \approx 7$

For $n = 1000 \rightarrow \text{Steps} \approx 10$

Binary Search is extremely efficient compared to Linear Search for large sorted arrays.

Challenges:

Ensuring that the array is sorted before applying Binary Search is mandatory.

Correctly handling integer division when calculating the mid index.

In systems where array sizes are very large, memory handling and stack limits might need attention.

Conclusion:

Binary Search significantly reduces the number of comparisons needed to find an element in a sorted array. Its logarithmic time complexity makes it far more efficient than Linear Search for large datasets. The implementation successfully matched theoretical expectations, confirming that Binary Search is optimal for sorted data.

Let me know if you'd like this formatted into a Word or PDF report — or need a comparison table between linear and binary search included!

Experiment # 3: Bubble Sort – Complexity Analysis:

Objective :

To implement the Bubble Sort algorithm in C, analyze the number of steps (comparisons and swaps) for different input sizes, and compare practical results with the theoretical time complexity.

Algorithm:

Bubble Sort is a simple comparison-based sorting algorithm. It repeatedly compares adjacent elements and swaps them if they are in the wrong order. This process continues until the array is fully sorted.

Steps:

Repeat $(n - 1)$ times:

For $i = 0$ to $n - j - 1$:

If $\text{arr}[i] > \text{arr}[i + 1]$, swap them.

The largest unsorted element "bubbles" to the end of the array in each pass.

Theoretical Solution of the Problem:

Time Complexity of Bubble Sort:

Case	Comparisons	Time Complexity
Best Case	$n - 1$	$O(n)$ (if optimized with a flag)
Worst Case	$n(n - 1)/2$	$O(n^2)$
Average	$n(n - 1)/2$	$O(n^2)$

Practical Work:

a. Pseudocode :

1) Set steps $\leftarrow 0$

- 2) For $i \leftarrow 0$ to $n - 2$ do a. Set $\text{swapped} \leftarrow \text{false}$ b. For $j \leftarrow 0$ to $n - i - 2$ do i. $\text{steps} \leftarrow \text{steps} + 1$ ii. If $\text{arr}[j] > \text{arr}[j + 1]$ then A. Swap $\text{arr}[j]$ and $\text{arr}[j + 1]$ B. Set $\text{swapped} \leftarrow \text{true}$ c. If $\text{swapped} = \text{false}$ then i. Break // array is already sorted
- 3) Return steps

b. Source Code in C:

```
void bubbleSort(int arr[], int n, int* steps)
{
    for (int i = 0; i < n - 1; i++)
    {
        int swapped = 0;
        for (int j = 0; j < n - i - 1; j++) {
            (*steps)++;
```

```
            if (arr[j] > arr[j + 1]) {

                int temp = arr[j];
                arr[j] = arr[j + 1];
                arr[j + 1] = temp;
                swapped = 1;
            }
        }
    }
}
```

```
if (swapped == 0)
    break;
}
}
```

```
void testbubblesort(int n)
{
    int arr[n];
    int steps = 0;
    for (int i = 0; i < n; i++)
    {
        arr[i] = n - i;
    }
}
```

```
bubbleSort(arr, n, &steps);
```

```
printf("Array size: %d\n", n);  
printf("Steps : %d\n\n", steps);  
}
```

```
int main()  
{ testbubblesort(10);  
  testbubblesort(100);  
  testbubblesort(1000);  
  return 0;  
}
```

Output:

Array size: 10

Steps (comparisons): 45

Array size: 100

Steps (comparisons): 4950

Array size: 1000

Steps (comparisons): 499500

Observations:

Array Size (n)	Expected Comparisons	Actual Comparisons
10	45	45
100	4950	4950
1000	499500	499500

Challenges:

Bubble Sort is inefficient for large datasets.

Without optimization (e.g., break early if no swaps), even already sorted arrays go through full iterations.

Manually counting steps and verifying swap behavior for small datasets is straightforward, but harder with large n .

Conclusion:

Bubble Sort is a fundamental sorting algorithm used to demonstrate basic sorting principles. While easy to implement and understand, it is inefficient for large inputs. The practical results closely match the theoretical complexity of $O(n^2)$. For performance-critical applications, advanced algorithms like Merge Sort or Quick Sort are preferred.

