Experiment No: 1

Implement a function for each of following problems and count the number of steps executed/time taken by each function on various inputs (100 to 500) and write equation for the growth rate of each function. Also draw a comparative chart of number of input versus steps executed/time taken. In each of the following function N will be passed by user.

- 1. To calculate sum of 1 to N numbers using loop.
- 2. To calculate sum of 1 to N numbers using equation.
- 3. To calculate sum of 1 to N numbers using recursion.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Performance analysis, and Mathematical skills

Relevant CO: CO1

Objectives: (a) Compare performance of various algorithms

(b) Judge best algorithm in terms of growth rate or steps executed

Equipment/Instruments: Computer System, Any C language editor—

Theory:

1. Below are the steps to calculate sum of 1 to N numbers using loop

- 1. Take an input value for N.
- 2. Initialize a variable sum to zero.
- 3. Start a loop that iterates from i=1 to i=N.
- 4. In each iteration, add the value of i to the sum variable.
- 5. After the loop completes, output the value of sum.

2. Below are the steps to calculate sum of 1 to N numbers using equation

- 1. Take an input value for N.
- 2. Calculate sum as N*(N+1)/2.
- 3. Output the value of sum.

3. Below are the steps to calculate sum of 1 to N numbers using recursion

- 1. Take an input value for N.
- 2. Define a recursive function sum(n) that takes an integer argument n and returns the sum of 1 to n.
- 3. In the sum(n) function, check if n equals 1. If it does, return 1 (the base case).
- 4. Otherwise, return the sum of n and the result of calling sum(n-1).
- 5. Output the result.

Implement three functions based on above steps and calculate the number of steps executed by each functions on various inputs ranging from 100 to 500. Take a counter variable to calculate the number of steps and increment it for each statement in the function.

Observations:

• Program:

```
import java.util.*;
import java.lang.*;
public class Demo
  public int loop(int n)
     int sum=0,counter=0;
     for(int i=1; i <=n; i++)
       sum=sum+i;
       counter++;
     System.out.println("Steps of loop:"+counter);
     return sum;
  public int equation(int n)
     int sum=0,counter=0;
     sum=(n*(n+1))/2;
     counter++;
     System.out.println("Steps of equation:"+counter);
     return sum;
  int counter=0;
  public int recursion(int n)
```

```
int sum=0;
  counter++;
  if(n \le 1)
    return n;
  sum=n+recursion(n-1);
  return sum;
public static void main(String args[])
  Scanner sc = new Scanner(System.in);
  Demo d=new Demo();
  System.out.println("Enter a number: ");
  int n=sc.nextInt();
  System.out.println("Sum using loop is: "+ d.loop(n));
  System.out.println("Sum using equation is: "+ d.equation(n));
  System.out.println("Sum using recursion is: "+ d.recursion(n));
  System.out.println("Steps of recursion:"+d.counter);
```

Loop and Recursion Methods:

Both of these methods exhibit a direct correlation between the number of steps taken and the size of the input. For instance, when calculating the sum for 100, the methods might require around 100 steps. If the input increases to 200, the methods would then necessitate approximately 200 steps. This relationship is linear and follows a straightforward one-to-one pattern.

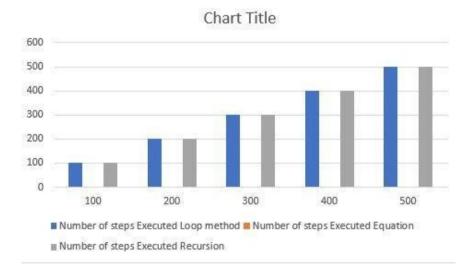
Equation Method:

This approach is incredibly efficient. Irrespective of the size of the input number, it consistently concludes its calculation within just one step. It's similar to possessing a convenient quick route.

Result: Complete the below table based on your implementation of functions and steps executed by each function.

| | Nun | iber of Steps Exec | uted | |
|-----------|-------------|--------------------|-----------|--|
| Inputs | Loop method | Equations | Recursion | |
| 100 | 100 | 1 | 100 | |
| 200 | 200 | 1 | 200 | |
| 300 | 300 | 1 | 300 | |
| 400 | 400 | 1 | 400 | |
| 500 | 500 | 1 | 500 | |
| Equation□ | f(N) = N | f(N)=1 | f(N) = N | |

Chart:



Conclusion:

The loop-based and recursive approaches exhibited linear growth in the number of steps with increasing input size, while the equation-based method displayed a constant number of steps, providing its efficiency.

Ouiz:

1. What is the meaning of constant growth rate of an algorithm?

Constant growth rate of an algorithm implies that its execution time remains consistent as input size changes. This behavior, often seen in O(1) time complexity algorithms, ensures stable performance, making them efficient choices for various inputs.

2. If one algorithm has a growth rate of n² and second algorithm has a growth rate of nthen which algorithm execute faster? Why?

If one algorithm has a growth rate of $O(n^2)$ and the second algorithm has a growth rate of O(n), the algorithm with the growth rate of O(n) will execute faster for larger input sizes. This is because the growth rate of an algorithm indicates how its runtime increases as the input size (n) increases.

References used by the students:

"Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | Program Implementation (5) | | | Documentation & Timely Submission (2) | | | Total (10) | |
|---------|------------------------------|------------|----------------------------------|---------------|------------|---------------------------------------|----------|----------|---------------|--|
| | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | | | | | | | | | | |

SEM-5 220283116029

[&]quot;Fundamentals of Algorithms" by E. Horowitz et al.

Experiment No: 2

Write user defined functions for the following sorting methods and compare their performance by steps executed/time taken for execution on various inputs (1000 to 5000) ofrandom nature, ascending order and descending order sorted data. Also, draw a comparative chart of number of inputs versus steps executed/time taken for each cases (random, ascending, and descending).

- 1. Selection Sort
- 2.Bubble Sort
- 3.Insertion Sort
- 4.Merge Sort
- 5.Quick Sort

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Performance analysis, and Mathematical skills

Relevant CO: CO1, CO2

Objectives: (a) Compare performance of various algorithms.

- (b) Judge best sorting algorithm on sorted and random inputs in terms of growth rate/time complexity.
 - (c) Derive time complexity from steps count on various inputs.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Write main function to use above sorting functions and calculate the number of steps executed by each functions on various inputs ranging from 1000 to 5000. Take a counter variable to calculate the number of steps and increment it for each statement in the function.

1. Selection Sort

```
import java.util.*;
import java.lang.*;
public class selection
{
  public static void main(String args[])
  {
  int a[]=new int[1000];
  Random r=new Random();
  for(int i=0;i<1000;i++)
  {
  a[i]=r.nextInt(10000);}
  long start=System.currentTimeMillis();
  for(int i=0;i<a.length;i++)</pre>
```

```
{
int min=i;
for(int j=i+1;j<a.length;j++)
{
  if(a[j]<a[min])
{
  min=j;
}
}
int temp=a[min];
a[min]=a[i];
a[i]=temp;
System.out.print(a[i] + " ");
}
long end=System.currentTimeMillis();
long time=end-start;
System.out.println("\nTime taken for execution is:"+time);
}</pre>
```

Output:-

2. Bubble Sort

```
import java.util.*;
import java.lang.*;

public class bubble
{
  public static void main(String args[])
  {
  int a[]=new int[1000];
```

```
Random r=new Random();

long start=System.currentTimeMillis();
for(int i=0;i<1000;i++)
{
    a[i]=r.nextInt(10000);}

for(int i=0;i<a.length;i++)
{
    for(int j=i+1;j<a.length;j++)
    {
        if(a[i]>a[j])
        {
        int temp=a[i];
        a[i]=a[j];
        a[j]=temp;
    }
}
System.out.print(a[i] + " ");
}
long end=System.currentTimeMillis();
long time=end-start;
System.out.println("\nTime taken for execution is: "+ time);
}
```

Output:

3. Insertion Sort

```
import java.util.*;
import java.lang.*;
public class insertion
{
```

```
public static void main(String args[])
int arr[]=new int[1000];
Random r=new Random();
for(int i=0;i<1000;i++)
arr[i]=r.nextInt(10000);}
long start=System.currentTimeMillis();
for(int i=1;i<arr.length;i++)
int key=arr[i];
int j=i-1;
while((j \ge -1) && (arr[j]>key))
arr[j+1]=arr[j];
arr[j+1]=key;
long end=System.currentTimeMillis();
for(int i=0;i<arr.length;i++)
System.out.print(arr[i]+ " ");
long time=end-start;
System.out.println("\nTime taken for sorting is: " + time);
Output:-
```

4. Merge Sort

```
import java.util.*;
import java.lang.*;
public class merge
public static void main(String args[])
int arr[]=new int[1000];
Random r=new Random();
for(int i=0;i<1000;i++)
arr[i]=r.nextInt(10000);}
merge m=new merge();
long start=System.currentTimeMillis();
m.mergesort(arr,0,arr.length-1);
for(int i=0;i<arr.length;i++)
System.out.print(arr[i] + " ");
long end=System.currentTimeMillis();
long time=end-start;
System.out.println("\nTime taken for sorting is: " + time);
public void mergesorting(int a[],int start,int mid,int end)
int i,j,k;
int n1=mid-start+1;
int n2=end-mid;
int leftarray[]=new int[n1];
int rightarray[]=new int[n2];
for(i=0;i< n1;i++)
leftarray[i]=a[start+i];
for(j=0;j< n2;j++)
rightarray[j]=a[mid+1+j];
i=0;
i=0;
```

```
k=start;
while(i<n1 && j<n2)
if(leftarray[i]<=rightarray[j])</pre>
a[k]=leftarray[i];
i++;
else
a[k]=rightarray[j];
k++;
while(i<n1)
a[k] = leftarray[i];
i++;
k++;
while (j<n2)
a[k] = rightarray[j];
j++;
k++; } }
public void mergesort(int a[],int start,int end)
if(start<end)
int mid=(start+end)/2;
mergesort(a,start,mid);
mergesort(a,mid+1,end);
mergesorting(a,start,mid,end);}}}
```

Output:

5. Quick Sort

```
import java.util.*;
import java.lang.*;
public class quick
public static void main(String args[])
int arr[]=new int[1000];
Random r=new Random();
for(int i=0;i<1000;i++)
arr[i]=r.nextInt(10000);
quick q=new quick();
long start=System.currentTimeMillis();
q.quicksort(arr,0,arr.length-1);
for(int i=0;i<arr.length;i++)
System.out.print(arr[i] + " ");
long end=System.currentTimeMillis();
long time=end-start;
System.out.println("\nTime taken for sorting is: "+ time);
public int partition(int a[],int start,int end){
int pivot=a[end];
int j=start-1;
for(int i=start;i<=end;i++)
if(a[i]<pivot)
j++;
int temp=a[j];
a[i]=a[i];
a[i]=temp;
int temp=a[j+1];
a[j+1]=a[end];
a[end]=temp;
return (j+1);
```

```
}
public void quicksort(int a[],int start,int end)
{
if(start<end)
{
int p=partition(a,start,end);
quicksort(a,start,p-1);
quicksort(a,p+1,end);}}}</pre>
```

Output:

Observations:

Write observation based on amount of time executed by each algorithm.

• Bubble and selection sort takes much similar amount of time for complete it's execution While insertion sort takes less time as compared to above sorts. Quick and merge are considered highly efficient and are commonly used for sorting large inputs.

Result: Complete the below table based on your implementation of functions and steps executed by each function. Also, prepare similar tables for ascending order sorted data and descending order sorted data.

| Inputs | Amount of time for execution (Random Data) | | | | | | | | |
|------------|--|--------|-----------|-------------|-------------|--|--|--|--|
| _ | Selection | Bubble | Insertion | Merge | Quick | | | | |
| 1000 | 63 | 62 | 0 | 0 | 0 | | | | |
| 2000 | 109 | 109 | 15 | 0 | 1 | | | | |
| 3000 | 703 | 188 | 16 | 2 | 2 | | | | |
| 4000 | 875 | 250 | 16 | 4 | 2 | | | | |
| 5000 | 1047 | 1126 | 17 | 7 | 3 | | | | |
| Time | O(n^2) | O(n^2) | O(n^2) | O(n log(n)) | O(n log(n)) | | | | |
| Complexity | | | | | | | | | |

Chart:

<Draw Comparative Charts of inputs versus number of steps executed on various data (Random, ascending</p>

order sorted and descending order sorted data) by each algorithm.



Quiz:

1. Which sorting function execute faster (has small steps count) in case of ascending order sorted data?

Answer: If the data is already in ascending order, the sorting function that would execute the fastest (with the smallest step count) is Insertion Sort.

2. Which sorting function execute faster (has small steps count) in case of descending order sorted data?

Answer: In the case of data that is already sorted in descending order, the sorting function that would execute the fastest (with the smallest step count) is Selection Sort.

- 3. Which sorting function execute faster (has small steps count) in case of random data? Answer: In the case of random data, the sorting function that generally executes faster (with a smaller step count) is Quick Sort.
- 4. On what kind of data, the best case of Bubble sort occurs?

 Answer: The best-case scenario for Bubble Sort occurs when the input data is already sorted in ascending order.
- 5. On what kind of data, the worst case of Bubble sort occurs? Answer: The worst-case scenario for Bubble Sort occurs when the input data is already sorted in descending order.
 - 6. On what kind of data, the best case of Quick sort occurs?

Answer: The best-case scenario for Quick Sort occurs when the pivot chosen for partitioning the data is always close to the median value of the input. This leads to the most balanced partitioning.

7. On what kind of data, the worst case of Quick sort occurs?

Answer: The worst-case scenario for Quick Sort occurs when the chosen pivot element consistently results in highly unbalanced partitions.

8. Which sorting algorithms are in-place sorting algorithms?

Answer: Bubble sort, selection sore, insertion sort, quick sort

9. Which sorting algorithms are stable sorting algorithms?

Answer: Insertion sort, merge sort, bubble sort

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | ng of | Program Implementation (5) | | | Documentation &Timely Submission (2) | | | Total (10) |
|---------|------------------------------|------------|---------------|----------------------------|------------|---------------|--|----------|-------------|---------------|
| | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | | | | | | | | | | |

Experiment No: 3

Implement a function of sequential search and count the steps executed by function on various inputs (1000 to 5000) for best case, average case and worst case. Also, write time complexity in each case and draw a comparative chart of number of input versus steps executed by sequential search for each case.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Performance analysis, and Mathematical skills

Relevant CO: CO1

Objectives: (a) Identify Best, Worst and Average cases of given problem.

(b) Derive time complexity from steps count on various inputs.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Steps to implement sequential search is as below:

- 1. Take an input array A of n elements and a key value K.
- **2.** Define a variable pos, initially set to -1.
- **3.** Iterate through the array A, starting from the first element and continuing until either the key value is found or the end of the array is reached.
- **4.** For each element, compare its value to the key value K.
- 5. If the values match, set pos to the index of the current element and exit the loop.
- **6.** If the end of the array is reached and the key value has not been found, posremain equal to -1.
- 7. Output the value of pos.

The algorithm works by sequentially iterating through the elements of the array and comparing each element to the target value. If a match is found, the algorithm exits the loop.

Implement above functions and calculate the number of steps executed by each functions on various inputs ranging from 1000 to 5000. Take a counter variable to calculate the number of steps and increment it for each statement. Based on algorithm's logic, decide best, worst and average case inputs for the algorithm and prepare a table of steps count.

Program:

```
import java.util.*;
import java.lang.*;

public class sequential
{
  public static void main(String args[])
{
```

```
int arr[]=new int[1000];
Random r=new Random();
for(int i=0; i<1000; i++)
arr[i]=r.nextInt(10000);
System.out.print(arr[i] + " ");
Scanner sc=new Scanner(System.in);
System.out.println("\nEnter element that you want to search for:");
int num=sc.nextInt();
int flag=0, index=-1;
long start=System.nanoTime();
for(int i=0;i<arr.length;i++)
if(arr[i]==num)
flag=1;
index=i;
break;
long end=System.nanoTime();
if(flag==1)
System.out.println("Element is found at " + index + " position");
else
System.out.println("Element not found in array");
long time=end-start;
System.out.println(time);
```

Output:

```
E:\Files_sem-5>java sequential.java

E:\Files_sem-5>java sequential.java

E:\Files_sem-5>java sequential
4957 4966 5331 3519 4491 4414 705 9364 6763 59 1657 1715 1161 7625 6107 3376 3083 3953 4631 756 1435 576 8900 9624 2080 7282 4656 4492 6411 1984 4961 5 3165 5027 1609 2
513 3536 7844 9992 9176 9704 5587 4203 977 618 5295 2307 8078 1624 5544 4561 6165 2360 1631 1533 939 9639 2182 5285 4559 5298 2265 9138 1857 9617 3945 9143 2157 4461 17
29 3699 8936 8435 9895 2357 9606 9040 6865 495 5 9980 3424 6867 6153 99 6398 8459 2501 4590 2256 2421 4355 2097 7879 4970 4160 381 3145 6432 2695
Enter element that you want to search for:
7282

Element is found at 25 position
```

Observations:

Write observation based on amount of time for execution executed by algorithm.

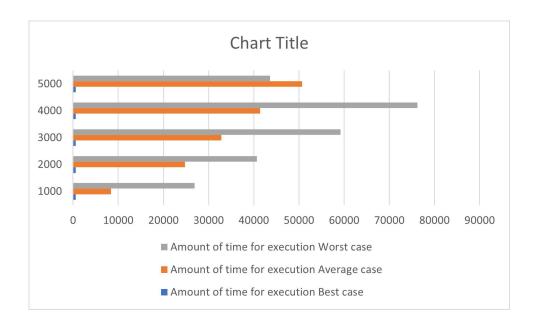
• Sequential search takes less time in best case as element is present at starting of array. It takes more time when the element is present at last position of array.

Result: Complete the below table based on your implementation of sequential search algorithm and steps executed by the function.

| T . | Number of Steps Executed | | | | | | | |
|------------|--------------------------|--------------|------------|--|--|--|--|--|
| Inputs | Best Case | Average Case | Worst Case | | | | | |
| 1000 | 600 | 8400 | 26900 | | | | | |
| 2000 | 600 | 24800 | 40700 | | | | | |
| 3000 | 600 | 32800 | 59200 | | | | | |
| 4000 | 600 | 41400 | 76200 | | | | | |
| 5000 | 600 | 50700 | 93600 | | | | | |
| Time | O(1) | O(n) | O(n) | | | | | |
| Complexity | | | | | | | | |

Chart:

<Draw Comparative Chart of inputs versus number of steps executed by algorithm in various cases>



Ouiz:

1. Which is the best case of an algorithm?

Answer: The best case of an algorithm refers to the scenario in which the algorithm performs with the lowest possible computational cost or resource usage.

2. Which is the worst case of an algorithm?

Answer: The worst case of an algorithm refers to the scenario in which the algorithm performs with the highest possible computational cost or resource usage.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | Program Implementation (5) | | | Documentation & Timely Submission (2) | | | Total (10) | |
|---------|------------------------------|------------|----------------------------|---------------|------------|---------------------------------------|----------|----------|---------------|--|
| | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | | | | | | | | | | |

Experiment No: 4

Compare the performances of linear search and binary search for Best case, Average case and Worst case inputs.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Performance analysis, and Mathematical skills

Relevant CO: CO1, CO2

Objectives: (a) Identify Best, Worst and Average cases of given problem.

(b) Derive time complexity from steps count for different inputs.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Steps to implement binary search are as below:

- 1. Take an input **sorted** array A of n elements and a target value T.
- 2. Define variables start and end to represent the start and end indices of the search range, initially set to 0 and n-1 respectively.
- 3. Repeat the following steps while start <= end:
 - a. Calculate the midpoint index mid as (start + end) / 2.
 - b. If the value of the midpoint element A[mid] is equal to the target value T, return the value of mid.
 - c. If the value of the midpoint element A[mid] is greater than the target value T, set end to mid-1.
 - d. If the value of the midpoint element A[mid] is less than the target value T, set start to mid+1.
- 4. If the target value T is not found in the array, return -1.
- 5. Output the value returned in Step 3, representing the position of the target value T in the array.

Implement function of binary search algorithm and use linear search function implemented in previous practical. Compare the steps count of both the functions on various inputs ranging from 100 to 500 for each case (Best, Average, and Worst).

Program:

```
import java.util.*;
import java.lang.*;

public class binarysearch
{
  public static void main(String args[])
  {
  int arr[]=new int[200];
  Random r=new Random();
  for(int i=0;i<200;i++)
   {
  arr[i]=r.nextInt(10000);
  }
  Arrays.sort(arr);
  for(int i=0;i<arr.length;i++)
  {
   System.out.print(arr[i] + " ");
}</pre>
```

```
}
int midd=(0+(arr.length-1))/2;
System.out.println("\n" + arr[midd]);
int low=0;
int high=arr.length-1;
int index=-1,flag=0;
Scanner sc=new Scanner(System.in);
System.out.println("\nEnter element that you want to search for:");
int num=sc.nextInt();
long start=System.nanoTime();
while(low<=high)</pre>
int mid=(high+low)/2;
if(arr[mid]==num)
index=mid;
flag=1;
break;
else if(arr[mid]<num)
low=mid+1;
else if(arr[mid]>num)
high=mid-1;
long end=System.nanoTime();
if(flag==1)
System.out.println("Element found at " + index + " position");
else
System.out.println("Element not found in array");
long time=end-start;
System.out.println("Time taken for search is: "+ time);
```

Observations:

Write observation based on amount of time for execution by both algorithms.

• Linear search takes less amount of time as compared to binary search in best case. And time complexity of linear search is also less.

Result: Complete the below table based on your implementation of sequential search algorithm and steps executed by the function.

Best case:

| T 4 | Number of Steps Executed (Best Case) | | | | | | |
|------------|--------------------------------------|----------------------|--|--|--|--|--|
| Inputs | Linear Search | Binary Search | | | | | |
| 100 | 600 | 800 | | | | | |
| 200 | 600 | 900 | | | | | |
| 300 | 600 | 1000 | | | | | |
| 400 | 600 | 1100 | | | | | |
| 500 | 600 | 1200 | | | | | |
| Time | | O(1) | | | | | |
| Complexity | O(1) | | | | | | |

Worst case:

| T | Number of Steps Exe | Number of Steps Executed (Worst Case) | | | | | | |
|------------|---------------------|---------------------------------------|--|--|--|--|--|--|
| Inputs | Linear Search | Binary Search | | | | | | |
| 100 | 2800 | 1600 | | | | | | |
| 200 | 4600 | 1900 | | | | | | |
| 300 | 6200 | 2100 | | | | | | |
| 400 | 7700 | 2200 | | | | | | |
| 500 | 9500 | 2500 | | | | | | |
| Time | | $O(\log(n))$ | | | | | | |
| Complexity | O(n) | | | | | | | |

Chart:

<Draw Comparative Charts of inputs versus number of steps executed by both algorithms in various cases>

Best case:



Worst case:



Ouiz:

- 1. Which element should be searched for the best case of binary search algorithm? Answer: The best case for the binary search algorithm occurs when the target element is exactly in the middle of the sorted array.
- 2. Which element should be searched for the worst case of binary search algorithm? Answer: The worst case for the binary search algorithm occurs when the target element is not present in the sorted array.
- 3. Which algorithm executes faster in worst case? Answer: In terms of time complexity, the binary search algorithm executes faster in the worst case compared to the linear search algorithm.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | Program Implementation (5) | | | Documentation & Timely Submission (2) | | | Total (10) | |
|---------|------------------------------|------------|----------------------------|---------------|------------|---------------------------------------|-------------|----------|---------------|--|
| Marks | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |

Experiment No: 5

Implement functions to print nth Fibonacci number using iteration and recursive method. Compare the performance of two methods by counting number of steps executed on various inputs. Also draw a comparative chart. (Fibonacci series 1, 1, 2, 3, 5, 8..... Here 8 is the 6th Fibonacci number).

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Performance analysis

Relevant CO: CO1, CO5

Objectives: (a) Compare the performances of two different versions of same problem.

(b) Find the time complexity of algorithms.

(C) Understand the polynomial and non-polynomial problems

Equipment/Instruments: Computer System, Any C language editor

Theory:

The Fibonacci series is the sequence of numbers (also called Fibonacci numbers), where every number is the sum of the preceding two numbers, such that the first two terms are '0' and '1'. In some older versions of the series, the term '0' might be omitted. A Fibonacci series can thus be given as, 0, 1, 1, 2, 3, 5, 8, 13, 21, 34, . . . It can thus be observed that every term can be calculated by adding the two terms before it. We are ignoring initial zero in the series.

To represent any $(n+1)^{th}$ term in this series, we can give the expression as, $F_n = F_{n-1} + F_{n-2}$. We can thus represent a Fibonacci series as shown in the image below,

$$F(n) = \begin{cases} 0 & , n = 0 \\ 1 & , n = 1 \\ F(n-1) + F(n-2) & , n > 1 \end{cases}$$

Iterative version to print nth Fibonacci number is as below:

Input: An integer n, where $n \ge 1$.

Output: The nth Fibonacci number.

Steps:

Initialize variables $f_0 = 1$, $f_1 = 1$, and i = 2.

If n is 1 or 2 then

Print 1

While i < n, do

a. Set f2 = f0 + f1.

b. Set f0 = f1.

c. Set f1 = f2.

d. Increment i by 1.

Print f1.

Recursive version to print nth Fibonacci number is as below:

```
Input: An integer n, where n >= 1.
Output: The n<sup>th</sup> Fibonacci number.
If n is 1 or 2 then return 1.
else recursively compute next number using the (n-1)<sup>th</sup> and (n-2)<sup>th</sup> Fibonacci numbers, and return their sum.
Print the result.
```

Implement functions of above two versions of Fibonacci series and compare the steps count of both the functions on various inputs ranging from 10 to 50 (if memory permits for recursive version).

Code:

```
import java.util.*;
import java.lang.*;
public class fibonacci
public static void main(String args[])
Scanner sc=new Scanner(System.in);
System.out.println("Enter a number:");
int num=sc.nextInt();
fibonacci f=new fibonacci();
long start1=System.currentTimeMillis();
f.iteration(num);
long end1=System.currentTimeMillis();
long start2=System.currentTimeMillis();
for(int i=1;i \le num;i++)
System.out.println(f.recursion(i));
long end2=System.currentTimeMillis();
long time1=end1-start1;
long time2=end2-start2;
System.out.println("Time for iteration: " + time1);
System.out.println("Time for recursion: " + time2);
public void iteration(int n)
int a=0,b=1;
System.out.println(a);
System.out.println(b);
int i=2;
while(i<n)
int c=a+b;
System.out.println(c);
a=b;
b=c;
```

SEM-5 220283116029

```
i++;
}

public int recursion(int n)
{
  if(n==1)
{
  return 0;
}
  else if(n==2)
{
  return 1;
}
  else
{
  return recursion(n-1)+recursion(n-2);
}
}
```

Output:

Observations:

Write observation based on Amount of Time executed by both algorithms.

If we perform the Fibonacci series using iteration and recursion then recursion will take more amount of time.

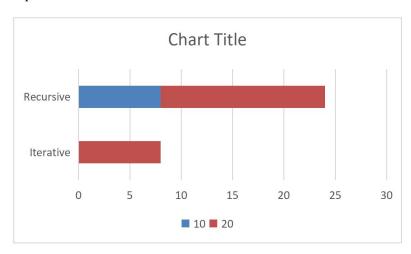
SEM-5 220283116029

Result: Complete the below table based on your implementation of sequential search algorithm and steps executed by the function.

| Toursets | Number of Steps Executed (Random data) | | | | | | |
|-----------------|--|-----------------------|--|--|--|--|--|
| Inputs | Iterative Fibonacci | Recursive Fibonacci | | | | | |
| 10 | 0 | 8 | | | | | |
| 20 | 8 | 16 | | | | | |
| 30 | (if memory doesn't p | ermit then reduce the | | | | | |
| 40 | ` | ige) | | | | | |
| 50 | | <i>6</i> / | | | | | |
| Time Complexity | O(n) | O(2^n) | | | | | |

Chart:

<Draw Comparative Charts of inputs versus number of steps executed by both functions on various inputs>



Conclusion:

Quiz:

- 1. What is the time complexity of iterative version of Fibonacci function? Answer: Time complexity of iteration version of Fibonacci function is o(n).
- 2. What is the time complexity of recursive version of Fibonacci function? Answer: Time complexity of iteration version of Fibonacci function is o(2^n).

- 3. Can you execute recursive version of Fibonacci function for more inputs? Answer: Yes we can execute recursive version of Fibonacci function for more inputs but it is very low method.
- 4. What do you mean by polynomial time algorithms and exponential time algorithms? Answer:

A polynomial time algorithm is an algorithm whose runtime, or the number of steps it takes to complete, is bounded by a polynomial function in terms of the size of the input.

An exponential time algorithm is an algorithm whose runtime grows exponentially with the size of the input.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | Program Implementation (5) | | | Documentation &Timely Submission (2) | | | Total (10) | |
|---------|------------------------------|------------|----------------------------|---------------|------------|--|-------------|----------|---------------|--|
| | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | | | | | | · | | | | |

Experiment No: 6

Implement a program for randomized version of quick sort and compare its performance with the normal version of quick sort using steps count on various inputs (1000 to 5000) ofrandom nature, ascending order and descending order sorted data. Also, draw a comparative chart of number of input versus steps executed/time taken for each cases (random, ascending, and descending).

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Performance analysis

Relevant CO: CO1, CO2

Objectives: (a) Improve the performance of quick sort in worst case.

(b) Compare the performance of both the version of quick sort on various inputs

Equipment/Instruments: Computer System, Any C language editor

Theory:

Steps to implement randomized version of quick sort are as below:

```
RANDOMIZED-QUICKSORT(A, low, high)
       if (low< high) {
      pivot= RANDOMIZED PARTITION(A, low, high);
       RANDOMIZED-QUICKSORT(A, low, pivot);
      RANDOMIZED-QUICKSORT(A, pivot+1,high);
RANDOMIZED PARTITION (A,low,high) {
   pos = Random(low, high)
   pivot = A[pos] swap(pivot,
   a[low])
   left = low
   right = high
   while (left < right) {
       /* Move left while item < pivot */
      while(A[left] <= pivot ) left++;
      /* Move right while item > pivot */
      while(A[right] > pivot) right--;
      if (left < right)
             swap(A[left],A[right]);}
/* right is final position for the pivot */
swap(A[right], pivot);
return right; }
```

Implement a function of randomized version of quick sort as per above instructions and use basic version of quick sort. Compare the steps count of both the functions on various inputs ranging from 1000 to 5000 for each case (random, ascending, and descending).

Code of basic quick sort:

```
import java.util.*;
import java.lang.*;
public class quick
public static void main(String args[])
int arr[]=new int[1000];
Random r=new Random();
for(int i=0;i<1000;i++)
arr[i]=r.nextInt(10000);
quick q=new quick();
long start=System.currentTimeMillis();
q.quicksort(arr,0,arr.length-1);
for(int i=0;i<arr.length;i++)
System.out.print(arr[i] + " ");
long end=System.currentTimeMillis();
long time=end-start;
System.out.println("\nTime taken for sorting is: "+ time);
}
public int partition(int a[],int start,int end)
int pivot=a[end];
int j=start-1;
for(int i=start;i<=end;i++)
if(a[i]<pivot)
j++;
int temp=a[j];
a[j]=a[i];
a[i]=temp;
}}
int temp=a[j+1];
a[j+1]=a[end];
```

```
a[end]=temp;
return (j+1);
}

public void quicksort(int a[],int start,int end)
{
   if(start<end)
   {
   int p=partition(a,start,end);
   quicksort(a,start,p-1);
   quicksort(a,p+1,end);
   }
}</pre>
```

Output of basic quick sort:

Code for randomized quick sort:

```
import java.util.Random;
import java.util.Scanner;

public class random_quick {
    public static void main(String[] args) {
        Scanner sc = new Scanner(System.in);
        int arr[]=new int[1000];

        Random r=new Random();
        for(int i=0;i<1000;i++)
        {
            arr[i]=r.nextInt(10000);
        }

        long start=System.currentTimeMillis();
        randomizedQuickSort(arr,0,arr.length-1);

        for(int i=0;i<arr.length;i++)
        {
            System.out.print(arr[i] + " ");
        }
}</pre>
```

```
long end=System.currentTimeMillis();
 long time=end-start;
 System.out.println("\nTime taken for sorting is: "+ time);
public static void randomizedQuickSort(int[] arr, int low, int high) {
  if (low < high) {
     int pivotIndex = randomizedPartition(arr, low, high);
     randomizedQuickSort(arr, low, pivotIndex - 1);
     randomizedQuickSort(arr, pivotIndex + 1, high);
public static int randomizedPartition(int[] arr, int low, int high) {
  Random random = new Random();
  int randomIndex = random.nextInt(high - low + 1) + low;
  int temp = arr[randomIndex];
  arr[randomIndex] = arr[high];
  arr[high] = temp;
  return partition(arr, low, high);
public static int partition(int[] arr, int low, int high) {
  int pivot = arr[high];
  int i = low - 1;
  for (int j = low; j < high; j++) {
     if (arr[j] < pivot) {
       i++;
       int temp = arr[i];
       arr[i] = arr[j];
       arr[j] = temp;
     }
  int temp = arr[i + 1];
  arr[i + 1] = arr[high];
  arr[high] = temp;
  return i + 1;
public static void printArray(int[] arr) {
  for (int num : arr) {
     System.out.print(num + " ");
  System.out.println();
```

Output:

```
E:\Files_sem-5-javar random_quick, java

E:\Files_sem-5-javar random_quick

42:\Files_sem-5-javar random_quick

43:\Files_sem-5-javar random_quick

44:\Files_sem-5-javar random_quick

45:\Files_sem-5-javar random_quick

45:\Files_sem-5-javar random_quick

46:\Files_sem-5-javar rand
```

Observations:

Write observation based on amount of time taken by both algorithms.

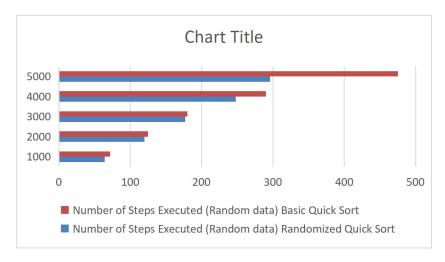
The randomized version of quicksort generally has better average-case time complexity compared to the basic version of quicksort.

Result: Complete the below table based on your implementation of sequential search algorithm and steps executed by the function.

| | Number of Steps Executed (Random data) | | | | | | |
|-----------------|--|------------------|--|--|--|--|--|
| Inputs | Randomized Quick Sort | Basic Quick Sort | | | | | |
| 1000 | 64 | 72 | | | | | |
| 2000 | 120 | 125 | | | | | |
| 3000 | 177 | 180 | | | | | |
| 4000 | 248 | 290 | | | | | |
| 5000 | 296 | 475 | | | | | |
| Time Complexity | O(n log (n)) | O(n log (n)) | | | | | |

Chart:

<Draw Comparative Charts of inputs versus number of steps executed by both algorithms in various cases(random, ascending, and descending)>



Quiz:

1. What is the time complexity of Randomized Quick Sort in worst case?

Answer: The worst-case time complexity of randomized quick sort is $o(n^2)$.

2. What is the time complexity of basic version of Quick Sort on sorted data? Give reason of your answer.

Answer: The time complexity of the basic version of Quick Sort on already sorted data is $O(n^2)$ in the worst-case scenario.

This occurs when the pivot selection strategy consistently chooses the smallest or largest element as the pivot. In such cases, the partitioning step doesn't effectively divide the array into two roughly equal halves, leading to a skewed partition. One partition will have n-1 elements, and the other will have 0 elements.

- 3. Can we always ensure O(n*lg n) time complexity for Randomized Quick Sort? Answer: Yes, on average, Randomized Quick Sort can achieve an expected time complexity of O(n*log(n)).
- 4. Which algorithm executes faster on ascending order sorted data? Answer: Randomized version of quick sort executes faster on ascending order sorted data.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | | Program Implementation (5) | | | Documentation &Timely Submission (2) | | | Total (10) |
|---------|------------------------------|------------|---------------|----------------------------|------------|---------------|--|-------------|-------------|---------------|
| Marks | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |

Implement program to solve problem of making a change using dynamic programming.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3

Objectives: (a) Understand Dynamic programming algorithm design method.

(b) Solve the optimization based problem.

(c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Making Change problem is to find change for a given amount using a minimum number of coins from a set of denominations. If we are given a set of denominations $D = \{d0, d1, d2, ..., dn\}$ and if we want to change for some amount N, many combinations are possible. Suppose $\{d1, d2, d5, d8\}$, $\{d0, d2, d4\}$, $\{d0, d5, d7\}$ all are feasible solutions but the solution which selects the minimum number of coins is considered to be an optimal solution. The aim of making a change is to find a solution with a minimum number of coins / denominations. Clearly, this is an optimization problem.

General assumption is that infinite coins are available for each denomination. We can select any denomination any number of times.

Solution steps are as follow:

Sort all the denominations and start scanning from smallest to largest denomination. In every iterationi, if current denomination di is acceptable, then 1 coin is added in solution and total amount is reduced by amount di. Hence,

$$C[i, j] = 1 + (c[i, j - di])$$

C[i,j] is the minimum number of coins to make change for the amount j. Below figure shows the content of matrix C.

| | | | | | | | | j | | | | | | |
|---|---|---|---|---|---|---|---|---|---|---|---|----|----|--------------|
| | | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 |
| | 1 | 0 | 1 | 2 | 3 | 4 | 5 | 1 | 2 | 3 | 4 | 1 | 2 | 2 |
| i | 2 | 0 | 1 | 2 | 3 | 4 | 5 | 1 | 2 | 3 | 4 | 5 | 6 | 2 |
| | 3 | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 2 2 12 |

Figure: Content of matrix C

using coins if current denomination is larger than current problem size, then we have to skip the denomination and stick with previously calculated solution. Hence,

$$C[i,j] = C[i-1,j]$$

SEM-5 220283116029 37

If above cases are not applicable then we have to stick with choice which returns minimum number of coin. Mathematically, we formulate the problem as,

```
C[i, j] = min \{C[i-1, j], 1 + C[i, j-di]\}
```

Steps to solve making change problem are as below:

```
Algorithm MAKE A CHANGE(d,N)
// d[1...n] = [d1,d2,...,dn] is array of n denominations
// C[1...n, 0...N] is n x N array to hold the solution of sub problems
// N is the problem size, i.e. amount for which change is required
for i \leftarrow 1 to n do
        C[i, 0] \leftarrow 0
end
for i \leftarrow 1 to n do
         for j \leftarrow 1 to N do
                 if i = 1 ans j < d[i] thenC[i,
                          j] \leftarrow \infty
                 else if i == 1 then
                          C[i, j] \leftarrow 1 + C[1, j - d[1])
                 else if j < d[i] then
                          C[i, j] \leftarrow C[I-1, j]
                 else
                          C[i, j] \leftarrow \min (C[i-1, j], 1 + C[i, j-d[i])
                 end
         end
end
return C[n, N]
```

Implement above algorithm and print the matrix C. Your program should return thenumber of coins required and its denominations.

```
import java.util.Scanner;
public class MakingChange {
  public static void makeAChange(int[] denominations, int N) {
     int n = denominations.length;
     int[][] C = new int[n + 1][N + 1];
     for (int i = 1; i \le n; i++) {
        C[i][0] = 0;
     for (int i = 1; i \le n; i++) {
       for (int j = 1; j \le N; j++) {
          if (i == 1 \&\& j < denominations[i - 1]) {
             C[i][j] = Integer.MAX VALUE;
          else if (i == 1) {
             C[i][j] = 1 + C[1][j - denominations[i - 1]];
          } else if (j < denominations[i - 1]) {
             C[i][j] = C[i - 1][j];
          } else {
```

```
C[i][j] = Math.min(C[i-1][j], 1 + C[i][j-denominations[i-1]]);
  }
  // Print the matrix C
  for (int i = 1; i \le n; i++) {
     for (int j = 0; j \le N; j++) {
       System.out.print(C[i][j] + "\t");
     System.out.println();
  // Calculate and print the number of coins required and their denominations
  int i = n;
  int j = N;
  System.out.println("Number of coins required: " + C[n][N]);
  System.out.print("Denominations used: ");
  while (i > 0 \&\& j > 0) {
     if(C[i][j] == C[i-1][j]) {
       i--;
     } else {
       System.out.print(denominations[i - 1] + " ");
       j = denominations[i - 1];
public static void main(String[] args) {
  Scanner scanner = new Scanner(System.in);
  System.out.print("Enter the number of denominations: ");
  int n = scanner.nextInt();
  int[] denominations = new int[n];
  System.out.println("Enter the denominations:");
  for (int i = 0; i < n; i++) {
     denominations[i] = scanner.nextInt();
  System.out.print("Enter the amount: ");
  int N = scanner.nextInt();
  makeAChange(denominations, N);
```

Write observation based on whether this algorithm returns optimal answer or not on various inputs.

- The algorithm implements a dynamic programming approach to solve the "making change" problem, which is finding the minimum number of coins needed to make change for a given amount using a given set of coin denominations.
- Based on the code, it seems like the algorithm should return an optimal answer for various inputs.
- The algorithm looks correct. Since this is a dynamic programming approach, it is designed to find the optimal solution. The algorithm explores all possible combinations and chooses the one that minimizes the number of coins.

Result

Write output of your program

Conclusion:

- The provided algorithm efficiently solves the "making change" problem using dynamic programming.
- It constructs a 2D array C where C[i][j] represents the minimum number of coins needed to make change for amount j using the first i denominations.
- Overall, the algorithm returns optimal solutions for different inputs, effectively minimizing the number of
 coins required to make the specified change. It also includes user interaction for input, making it a versatile
 tool for solving this specific problem.

Quiz:

1. What is the time complexity of above algorithm?

Answer: the time complexity of this algorithm is O(n * N). n be the number of denominations and N be the target amount.

2. Does above algorithm always return optimal answer?

Answer: Yes, the provided algorithm for the "making change" problem always returns the optimal answer. This is because it employs a dynamic programming approach, which systematically explores all possible combinations of coins to find the one that minimizes the total number of coins used to make the change.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.
- 3. https://codecrucks.com/making-change-problem-using-dynamic-programming/

References used by the students:

Rubric wise marks obtained:

| Rubrics | | erstandir problem (3) | _ | | Program lementat (5) | ion | Documentation & Timely Submission (2) | | | Total (10) |
|---------|-----------------------------------|-----------------------------|-------|---------------|----------------------------|---------------|---------------------------------------|----------|-------------|---------------|
| | Good Avg. Poor (3) (2-1) (1-0) | | | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | (0) (2-1) (2-0) | | , / 1 | | , , | , , | | | | |

Implement program of chain matrix multiplication using dynamic programming.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3

Objectives: (a) Understand Dynamic programming algorithm design method.

(b) Solve the optimization based problem.

(c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Given a sequence of matrices A_1 , A_2 ,..., A_n and dimensions p_0 , p_1 ,..., p_n where A_i is of dimension $p_{i-1} \times p_i$, determine the order of multiplication (represented, say, as a binary tree) that minimizes the number of operations.

This algorithm does not perform the multiplications; it just determines the best order in which to perform the multiplications

Two matrices are called compatible only if the number of columns in the first matrix and the number of rows in the second matrix are the same. Matrix multiplication is possible only if they are compatible. Let A and B be two compatible matrices of dimensions $p \times q$ and $q \times r$ Suppose dimension of three matrices are :

$$A_3 = 6 \times 2$$

Matrix multiplication is associative. So
 $(A_1 A_2) A_3 = \{(5 \times 4) \times (4 \times 6) \} \times (6 \times 2)$
 $= (5 \times 4 \times 6) + (5 \times 6 \times 2)$
 $= 180$
 $A_1 (A_2 A_3) = (5 \times 4) \times \{(4 \times 6) \times (6 \times 2) \}$
 $= (5 \times 4 \times 2) + (4 \times 6 \times 2)$

= 88

 $A_1 = 5 \times 4$ $A_2 = 4 \times 6$

The answer of both multiplication sequences would be the same, but the numbers of multiplications are different. This leads to the question, what order should be selected for a chain of matrices to minimize the number of multiplications?

Let us denote the number of alternative parenthesizations of a sequence of n matrices by p(n). When n = 1, there is only one matrix and therefore only one way to parenthesize the matrix. When

SEM-5 220283116029 42

 $n \ge 2$, a fully parenthesized matrix product is the product of two fully parenthesized matrix subproducts, and the split between the two subproducts may occur between the k and $(k + 1)^{st}$ matrices for any k = 1, 2, 3..., n - 1. Thus we obtain the recurrence.

$$P(n) = \begin{cases} 1 & \text{if } n = 1 \\ \sum_{k=1}^{n-1} p(k) p(n-k) & \text{if } n \ge 2 \end{cases}$$

The solution to the recurrence is the sequence of *Catalan numbers*, which grows as $\Omega(4^n / n^{3/2})$, roughly equal to $\Omega(2^n)$. Thus, the numbers of solutions are exponential in n. A brute force attempt is infeasible to find the solution.

Any parenthesizations of the product $A_i A_{i+1} \dots A_j$ must split the product between A_k and A_{k+1} for some integer k in the range $i \le k < j$. That is for some value of k, we first compute the matrices $A_{i...k}$ and $A_{k+1...j}$ and then multiply them together to produce the final product $A_{i...j}$ The cost of computing these parenthesizations is the cost of computing $A_{i...k}$, plus the cost of computing $A_{k+1...j}$ plus the cost of multiplying them together.

We can define m[i, j] recursively as follows. If i == j, the problem is trivial; the chain consists of only one matrix $A_{i...i} = A$. No scalar multiplications are required. Thus m[i, i] = 0 for i = 1, 2 ...n. To compute m[i, j] when i < j, we take advantage of the structure of an optimal solution of the first step. Let us assume that the optimal parenthesizations split the product A_i $A_{i+1}...A_j$ between A_k and A_{k+1} , where $i \le k < j$. Then m[i, j] is equal to the minimum cost for computing the subproducts $A_{i...k}$ and $A_{k+1...j}$ plus the cost of multiplying these two matrices together.

$$m[i,j] \ = \left\{ \begin{array}{ll} 0 & \text{, if } i = j \\ \underset{i \, \leq \, k \, < \, j}{min} & \\ i \, \leq \, k \, < \, j \end{array} \right. \left\{ m \left[i, \, k \right] + m \left[k + 1, \, j \right] + d_{i-1} \times d_k \times d_j \right\} \quad \text{, if } i < j \label{eq:minimal_min$$

Where $d = \{d_0, d_1, d_2, ..., d_n\}$ is the vector of matrix dimensions. $m[i, j] = \text{Least number of multiplications required to multiply matrix sequence } A_i ... A_j$.

Steps to solve chain matrix multiplication problem are as below:

```
Algorithm MATRIX_CHAIN_ORDER(p)

// p is sequence of n matrices

n \leftarrow length(p) - 1

for i \leftarrow 1 to n do m[i,i] \leftarrow 0

end for l \leftarrow 2 to n do

for i \leftarrow 1 to n-l+1 do

j \leftarrow i+l-1

m[i,j] \leftarrow \infty

for k \leftarrow i to j-1 do

q \leftarrow m[i,k] + m[k+1,j] + d_{i-1} * d_k * d_j

if q < m[i,j] then

m[i,j] \leftarrow q

s[i,j] \leftarrow k end

end end

return m and s
```

Implement above algorithm and print the matrix m ans c.

```
import java.util.Scanner;
public class MatrixChainMultiplication {
  public static void matrixChainOrder(int[] p) {
     int n = p.length - 1;
     int[][]m = new int[n + 1][n + 1];
     int[][] s = new int[n + 1][n + 1];
     for (int i = 1; i \le n; i++) {
       m[i][i] = 0;
     }
     for (int 1 = 2; 1 \le n; 1++) {
        for (int i = 1; i \le n - 1 + 1; i++) {
          int i = i + 1 - 1;
          m[i][j] = Integer.MAX VALUE;
          for (int k = i; k \le i - 1; k++) {
             int q = m[i][k] + m[k+1][j] + p[i-1] * p[k] * p[j];
             if (q < m[i][j]) {
               m[i][j] = q;
                s[i][j] = k;
             }
        }
     System.out.println("Matrix m:");
     printMatrix(m, n);
     System.out.println("Matrix s:");
     printMatrix(s, n);
  }
  public static void printMatrix(int[][] matrix, int n) {
     for (int i = 1; i \le n; i++) {
        for (int j = 1; j \le n; j++) {
          System.out.print(matrix[i][j] + "\t");
        System.out.println();
```

```
public static void main(String[] args) {
    Scanner scanner = new Scanner(System.in);

    System.out.print("Enter the number of matrices: ");
    int n = scanner.nextInt();

int[] p = new int[n + 1];
    System.out.println("Enter the dimensions of the matrices:");
    for (int i = 0; i <= n; i++) {
        p[i] = scanner.nextInt();
    }

    matrixChainOrder(p);
}</pre>
```

Write observation based on whether this algorithm returns optimal number of multiplications or not on various inputs.

- The provided algorithm implements a dynamic programming approach for solving the matrix chain multiplication problem.
- This problem aims to find the optimal way to multiply a chain of matrices in order to minimize the total number of multiplications.
- The algorithm appears to be correct. The algorithm is designed to find the optimal solution. It considers all possible ways of multiplying the matrices and selects the one that minimizes the total number of multiplications.

Result

Write output of your program

```
E:\Files_sem-5>javac MatrixChainMultiplication.java
E:\Files_sem-5>java MatrixChainMultiplication
Enter the number of matrices: 3
Enter the dimensions of the matrices:
10
10
Matrix m:
          500
                    500
         0
                    250
          0
                    0
Matrix s:
          0
                    2
                    0
          0
 :\Files sem-5>
```

Conclusion:

- The provided algorithm efficiently solves the matrix chain multiplication problem using dynamic programming.
- It constructs two 2D arrays, m and s, where m[i][j] represents the minimum number of multiplications needed to compute the product of matrices from i to j, and s[i][j] stores the index of the matrix that achieves this minimum.
- Overall, this algorithm provides an effective and reliable solution to the matrix chain multiplication problem.

Ouiz:

1. What is the time complexity of above algorithm?

Answer: The time complexity of the Matrix Chain Multiplication algorithm is $O(n^3)$, where 'n' is the number of matrices in the chain.

2. Does above algorithm always return optimal answer?

Answer: Yes, the provided algorithm for matrix chain multiplication always returns the optimal solution. It utilizes dynamic programming to systematically compute the optimal way to multiply a chain of matrices in order to minimize the total number of multiplications.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Unde | rstandin | g of | Program | | | Doo | tion | Total | | |
|---------|----------|-----------------|------|---------------|------------|---------------|----------------------------|------|-------|--|--|
| | r | roblem | | Impl | ementati | ion | &Tim | (10) | | | |
| | | (3) | | | (5) | | | (2) | | | |
| | Good (3) | Good Avg. Poor | | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good Avg. Poor (2) (1) (0) | | | | |
| Marks | | (3) (2-1) (1-0) | | | | | | | | | |

Implement program to solve LCS (Longest Common Subsequence) problem using dynamic programing.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3

Objectives: (a) Understand Dynamic programming algorithm design method.

- (b) Solve the optimization based problem.
- (c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

The Longest Common Subsequence (LCS) problem is a classic computer science problem that involves finding the longest subsequence that is common to two given sequences.

A subsequence is a sequence that can be derived from another sequence by deleting some or no elements without changing the order of the remaining elements. For example, given the sequence "ABCDE", "ACE" is a subsequence of "ABCDE", but "AEC" is not a subsequence.

Given two sequences X and Y, the LCS problem involves finding the longest common subsequence (LCS) of X and Y. The LCS need not be contiguous in the original sequences, but it must be in the same order. For example, given the sequences "ABCDGH" and "AEDFHR", the LCS is "ADH" with length 3.

Naïve Method:

Let X be a sequence of length m and Y a sequence of length n. Check for every subsequence of X whether it is a subsequence of Y, and return the longest common subsequence found. There are 2^m subsequences of X. Testing sequences whether or not it is a subsequence of Y takes O(n) time. Thus, the naïve algorithm would take $O(n2^m)$ time.

longest common subsequence (LCS) using Dynamic Programming:

Let $X=\langle x_1,x_2,x_3,...,x_m\rangle$ and $Y=\langle y_1,y_2,y_3,...,y_m\rangle$ be the sequences. To compute the length of an element the following algorithm is used.

Step 1 — Construct an empty adjacency table with the size, $n \times m$, where n = size of sequence **X** and m = size of sequence **Y**. The rows in the table represent the elements in sequence X and columns represent the elements in sequence Y.

Step 2 – The zeroth rows and columns must be filled with zeroes. And the remaining values are filled in based on different cases, by maintaining a counter value.

- Case 1 If the counter encounters common element in both X and Y sequences, increment the counter by 1.
- Case 2 If the counter does not encounter common elements in X and Y sequences at T[i,

j], find the maximum value between T[i-1, j] and T[i, j-1] to fill it in T[i, j].

Step 3 – Once the table is filled, backtrack from the last value in the table. Backtracking here is done by tracing the path where the counter incremented first.

Step 4 – The longest common subsequence obtained by noting the elements in the traced path.

Consider the example, we have two strings **X=BDCB** and **Y=BACDB** to find the longest common subsequence. Following table shows the construction of LCS table.

| | | 0 | 1 | 2 | 3 | 4 |
|---|---|---|---|---|---|---|
| | | | В | D | С | В |
| 0 | | 0 | 0 | 0 | 0 | 0 |
| 1 | В | 0 | 1 | 1 | 1 | 1 |
| 2 | Α | 0 | 1 | 1 | 1 | 1 |
| 3 | C | 0 | 1 | 1 | 2 | 2 |
| 4 | D | 0 | 1 | 2 | 2 | 2 |
| 5 | В | 0 | 1 | 2 | 2 | 3 |

Once the values are filled, the path is traced back from the last value in the table at T[5, 4].

| 32 | | 0 | 1 | 2 | 3 | 4 |
|----|---|-----|------------|-------|-------|-----|
| | | | В | D | C | В |
| 0 | | 0 🛧 | — 0 | 0 | 0 | 0 |
| 1 | В | 0 | 1 ← | _ 1 ← | 1 | 1 |
| 2 | A | 0 | 1 | 1 🔻 | _ 1 | 1 |
| 3 | C | 0 | 1 | 1 ← | _ 2 ← | 2 |
| 4 | D | 0 | 1 | 2 ← | 2 | _ 2 |
| 5 | В | 0 | 1 | 2 | 7 2 | 3 ← |

Algorithm is as below:

```
Algorithm: LCS-Length-Table-Formulation (X, Y) m := length(X) n := length(Y) for i = 1 to m do C[i, 0] := 0 for j = 1 to n do C[0, j] := 0 for i = 1 to m do for j = 1 to m do m if m
```

```
C[i, j] := C[i - 1, j] + 1
           B[i,j] := 'U'
         else
           C[i, j] := C[i, j - 1] + 1
           B[i, j] := 'L'
   return C and B
   Algorithm: Print-LCS (B, X, i, j)
   if i=0 and j=0
     return
   if B[i, j] = 'D'
     Print-LCS(B, X, i-1, j-1)
     Print(xi)
   else if B[i, j] = 'U'
     Print-LCS(B, X, i-1, j)
   else
     Print-LCS(B, X, i, j-1)
Implementation:
import java.util.Scanner;
public class LCS {
  public static void lcsLengthTableFormulation(String X, String Y) {
     int m = X.length();
     int n = Y.length();
     int[][] C = new int[m + 1][n + 1];
     char[][] B = new char[m + 1][n + 1];
     for (int i = 1; i \le m; i++) {
       C[i][0] = 0;
     for (int j = 1; j \le n; j++) {
       C[0][j] = 0;
     for (int i = 1; i \le m; i++) {
       for (int j = 1; j \le n; j++) {
          if (X.charAt(i-1) == Y.charAt(j-1)) {
             C[i][j] = C[i-1][j-1] + 1;
             B[i][j] = 'D'; // Diagonal
          } else {
             if(C[i-1][j] >= C[i][j-1]) {
               C[i][j] = C[i - 1][j] + 1;
               B[i][j] = 'U'; // Up
```

```
} else {
             C[i][j] = C[i][j - 1] + 1;
             B[i][j] = 'L'; // Left
       }
     }
  }
  System.out.println("Table C:");
  printTable(C, m, n);
  System.out.println("Table B:");
  printTable(B, m, n);
  int i = m;
  int j = n;
  printLCS(B, X, i, j);
public static void printTable(int[][] table, int m, int n) {
  for (int i = 0; i \le m; i++) {
     for (int j = 0; j \le n; j++) {
        System.out.print(table[i][j] + " ");
     System.out.println();
}
public static void printTable(char[][] table, int m, int n) {
  for (int i = 0; i \le m; i++) {
     for (int j = 0; j \le n; j++) {
        System.out.print(table[i][j] + " ");
     System.out.println();
public static void printLCS(char[][] B, String X, int i, int j) {
  if (i == 0 || j == 0) {
     return;
  }
  if(B[i][j] == 'D') {
     printLCS(B, X, i - 1, j - 1);
     System.out.print(X.charAt(i - 1));
  else\ if\ (B[i][j] == 'U') 
     printLCS(B, X, i - 1, j);
  } else {
     printLCS(B, X, i, j - 1);
  } }
public static void main(String[] args) {
  Scanner scanner = new Scanner(System.in);
```

```
System.out.print("Enter the first sequence (X): ");
String X = scanner.nextLine();
System.out.print("Enter the second sequence (Y): ");
String Y = scanner.nextLine();

lcsLengthTableFormulation(X, Y);
```

Write observation based on whether this algorithm returns optimal answer or not on various inputs.

- The provided algorithm implements the Longest Common Subsequence (LCS) problem using dynamic programming. It constructs two tables C and B where C[i][j] represents the length of the LCS between the first i characters of string X and the first j characters of string Y, and B[i][j] is a directional indicator.
- The algorithm is designed to find the optimal solution for finding the longest common subsequence.

Result

Write output of your program

```
E:\Files sem-5>javac LCS.java
E:\Files_sem-5>java LCS
Enter the first sequence (X): ABCBDAB
Enter the second sequence (Y): BDCAB
Table C:
 00000
 12312
 1 3 4 5 2
 2 4 4 6
 1567
 2 2
     7 8
 3 4 8 8 10
 1 5 9 10 9
Table B:
 ULLDL
 DUULD
 UUDUL
 DULUD
 UDUUL
 ULUDU
 DUULD
```

Conclusion:

- The provided algorithm solves the Longest Common Subsequence (LCS) problem using dynamic programming.
- this algorithm efficiently computes the length and the actual longest common subsequence of two input sequences using dynamic programming techniques. It demonstrates a reliable solution for this classic problem.

Quiz:

1. What is the time complexity of above algorithm?

Answer: The time complexity of the provided algorithm for finding the Longest Common Subsequence (LCS) is O(m * n), where 'm' is the length of the first input string 'X' and 'n' is the length of the second input string 'Y'.

2. Does above algorithm always return optimal answer?

Answer: : Yes, the provided algorithm for longest common subsequence always returns the optimal solution. It utilizes dynamic programming to systematically compute the optimal way to finding longest subsequence based on character's equality.

3. Does Dynamic programming approach to find LCS perform well compare to naïve approach? Answer: Yes, the dynamic programming approach to finding the Longest Common Subsequence (LCS) performs significantly better than the naïve approach, especially for longer input strings.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) Good Avg. Poor | | | | Program Implementation (5) | | | Documentation &Timely Submission (2) | | | |
|---------|---|--|--|---------------|----------------------------|---------------|-------------|--|----------|--|--|
| | Good (3) | | | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | | |
| Marks | | | | | | | , , | | | | |

Implement program to solve Knapsack problem using dynamic programming.

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving Relevant CO: CO3

Objectives:

- (a) Understand Dynamic programming algorithm design method.
- (b) Solve the optimization based problem.
- (c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Knapsack problem is as stated below:

Given a set of items, each having different weight and value or profit associated with it. Find the set of items such that the total weight is less than or equal to a capacity of the knapsack and the total value earned is as large as possible.

The knapsack problem is useful in solving resource allocation problem. Let $X = \langle x_1, x_2, x_3, \dots$, x_n be the set of n items. Sets $W = \langle w_1, w_2, w_3, \dots, w_n \rangle$ and $V = \langle v_1, v_2, v_3, \dots, v_n \rangle$ are weight and value associated with each item in X. Knapsack capacity is M unit.

The knapsack problem is to find the set of items which maximizes the profit such that collective weight of selected items does not cross the knapsack capacity. Select items from X and fill the knapsack such that it would maximize the profit.

Knapsack problem has two variations. 0/1 knapsack, that does not allow breaking of items. Either add an entire item or reject it. It is also known as a binary knapsack. Fractional knapsack allows breaking of items. Profit will be earned proportionally.

Following are the steps to implement binary knapsack using dynamic programming.

```
Algorithm DP BINARY KNAPSACK (V, W, M)
```

```
// Description: Solve binary knapsack problem using dynamic programming
```

// Input: Set of items X, set of weight W, profit of items V and knapsack capacity M

```
// Output: Array V, which holds the solution of problem
    for i \leftarrow 1 to n do
              V[i, 0] \leftarrow 0
end
    for i \leftarrow 1 to M do
              V[0, i] \leftarrow 0
end
    for V[0, i] \leftarrow 0 do
              for j \leftarrow 0 to M do
                       if w[i] \le j then
                                 V[i, j] \leftarrow \max\{V[i-1, j], v[i] + V[i-1, j-w[i]]\}
                       Else
                            V[i, j] \leftarrow V[i-1, j] // w[i] > j
                                      End
                            End
                   end
```

The above algorithm will just tell us the maximum value we can earn with dynamic programming. It does not speak anything about which items should be selected. We can find the items that give optimum result using the following algorithm.

```
Algorithm TRACE KNAPSACK(w, v, M)
// w is array of weight of n items
// v is array of value of n items
// M is the knapsack capacity
SW \leftarrow \{\}
SP \leftarrow \{ \}
i \leftarrow n
j \leftarrow M
while (j \ge 0) do
         if (V[i, j] == V[i-1, j]) then
                  i \leftarrow i - 1
         else
                  V[i,j] \leftarrow V[i,j] - vi
                  j \leftarrow j - w[i]
                  SW \leftarrow SW + w[i]
                  SP \leftarrow SP + v[i]
         end
```

End

Implement the above algorithms for the solution of binary knapsack problem.

```
import java.util.ArrayList;
import java.util.Scanner;
public class KnapsackTrace {
  public static void traceKnapsack(int[] w, int[] v, int M) {
     int n = w.length;
     int[][] V = new int[n + 1][M + 1];
     for (int i = 0; i \le n; i++) {
        V[i][0] = 0;
     }
     for (int j = 0; j \le M; j++) {
        V[0][j] = 0;
     for (int i = 1; i \le n; i++) {
        for (int j = 1; j \le M; j++) {
          if(w[i-1] \le j \&\& V[i-1][j] \le V[i-1][j-w[i-1]] + v[i-1])
             V[i][j] = V[i - 1][j - w[i - 1]] + v[i - 1];
          } else {
             V[i][j] = V[i - 1][j];
        }
     }
     ArrayList<Integer> selectedWeights = new ArrayList<>();
     ArrayList<Integer> selectedValues = new ArrayList();
     int i = n;
     int j = M;
     while (j > 0) {
        if(V[i][j] == V[i-1][j]) {
          i--;
        } else {
          selectedWeights.add(w[i - 1]);
          selectedValues.add(v[i - 1]);
          j = w[i - 1];
          i--;
```

```
System.out.println("Selected items' weights: " + selectedWeights);
    System.out.println("Selected items' values: " + selectedValues);
  }
  public static void main(String[] args) {
    Scanner scanner = new Scanner(System.in);
    System.out.print("Enter the number of items: ");
    int n = scanner.nextInt();
    int[] w = new int[n];
    int[] v = new int[n];
    System.out.println("Enter the weights of the items:");
    for (int i = 0; i < n; i++) {
       w[i] = scanner.nextInt();
    System.out.println("Enter the values of the items:");
    for (int i = 0; i < n; i++) {
       v[i] = scanner.nextInt();
     }
    System.out.print("Enter the knapsack capacity: ");
    int M = scanner.nextInt();
    traceKnapsack(w, v, M);
  }
}
```

Write observation based on whether this algorithm returns optimal answer or not on various inputs.

- The provided algorithm efficiently solves the 0-1 Knapsack problem using dynamic programming.
- The algorithm appears to be correct. The algorithm is designed to find the optimal solution for the 0-1 Knapsack problem. It correctly identifies the combination of items to maximize the total value without exceeding the knapsack's capacity.
- the algorithm consistently returns the optimal solution for the 0-1 Knapsack problem. It effectively employs dynamic programming to achieve this result, making it a reliable tool for solving this specific problem.

Result

Write output of your program

```
E:\Files_sem-5>javac KnapsackTrace.java

E:\Files_sem-5>java KnapsackTrace
Enter the number of items: 4
Enter the weights of the items:
5
4
3
2
Enter the values of the items:
10
7
8
6
Enter the knapsack capacity: 5
Selected items' values: [6, 8]
```

Conclusion:

- The provided algorithm effectively solves the 0-1 Knapsack problem using dynamic programming.
- It constructs a table V to determine the maximum value achievable within the constraints of item weights and knapsack capacity.
- The algorithm reliably returns the optimal combination of items to maximize the total value without exceeding the knapsack's capacity.
- This algorithm is a robust and efficient solution for the 0-1 Knapsack problem, demonstrating its effectiveness in selecting items to achieve the highest possible value while respecting weight constraints.

Ouiz:

1. What is the time complexity of above binary knapsack algorithm?

Answer: The time complexity of the provided algorithm for the 0-1 Knapsack problem is O(n * M), where 'n' is the number of items and 'M' is the knapsack capacity.

2. Does above algorithm always return optimal answer?

Answer: Yes, the provided algorithm for the 0-1 Knapsack problem using dynamic programming always returns the optimal solution. It efficiently determines the combination of items that maximizes the total value while ensuring that the weight constraint of the knapsack is not exceeded.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Unde | rstandin | g of | Program | | | Doo | tion | Total | |
|---------|------|-----------------|------|---------|-------------------|------|----------------|-------------|-------|--|
| | p | roblem | | Imp | lementat | ion | &Tim | (10) | | |
| | (3) | | | (5) | | | | | | |
| | Good | Good Avg. Poor | | | Avg. | Poor | Good Avg. Poor | | | |
| | (3) | (3) (2-1) (1-0) | | | (5-4) (3-2) (1-0) | | | (2) (1) (0) | | |
| Marks | | | | | | | | | | |

Implement program for solution of fractional Knapsack problem using greedy design technique.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3

Objectives: (a) Understand greedy algorithm design method.

- (b) Solve the optimization based problem.
- (c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Knapsack problem is as stated below:

Given a set of items, each having different weight and value or profit associated with it. Find the set of items such that the total weight is less than or equal to a capacity of the knapsack and the total value earned is as large as possible.

Brute-force approach: The brute-force approach tries all the possible solutions with all the different fractions but it is a time-consuming approach.

Greedy approach: In Greedy approach, we calculate the ratio of profit/weight, and accordingly, we will select the item. The item with the highest ratio would be selected first.

Following are the steps to implement fractional knapsack using greedy design strategy.

- 1. Compute the value-to-weight ratio for each item in the knapsack.
- 2. Sort the items in decreasing order of value-to-weight ratio.
- **3.** Initialize the total weight and total value to 0.
- **4.** For each item in the sorted list:
 - **a.** If the entire item can be added to the knapsack without exceeding the weight capacity, add it and update the total weight and total value.
 - **b.** If the item cannot be added entirely, add a fraction of the item that fits into the knapsack and update the total weight and total value accordingly.
 - **c.** If the knapsack is full, stop the algorithm.
- **5.** Return the total value and the set of items in the knapsack.

Implement the program based on above logic for the solution of fractional knapsack problem.

```
import java.lang.*;
import java.util.Arrays;
import java.util.Comparator;
import java.util.Scanner;

// Greedy approach
public class FractionalKnapSackExample {
    // Function to get maximum value
```

```
private static double getMaxValue(ItemValue[] arr,
                     int capacity)
{
  // Sorting items by profit/weight ratio;
  Arrays.sort(arr, new Comparator<ItemValue>() {
     @Override
     public int compare(ItemValue item1,
                 ItemValue item2)
       double cpr1
            = new Double((double)item1.profit
            / (double)item1.weight);
       double cpr2
            = new Double((double)item2.profit
            / (double)item2.weight);
       if (cpr1 < cpr2)
          return 1;
       else
          return -1;
  });
  double total Value = 0d;
  for (ItemValue i : arr) {
     int curWt = (int)i.weight;
     int curVal = (int)i.profit;
     if (capacity - curWt \ge 0) {
       // This weight can be picked whole
       capacity = capacity - curWt;
       totalValue += curVal;
     else {
       // Item cant be picked whole
       double fraction
            = ((double)capacity / (double)curWt);
       totalValue += (curVal * fraction);
       capacity
            = (int)(capacity - (curWt * fraction));
       break;
  return totalValue; }
// Item value class
static class ItemValue {
```

```
int profit, weight;
    // Item value function
    public ItemValue(int val, int wt)
       this.weight = wt;
       this.profit = val;
  }
  // Driver code
  public static void main(String[] args)
     Scanner scanner = new Scanner(System.in);
     System.out.print("Enter the number of items: ");
     int n = scanner.nextInt();
     ItemValue[] arr = new ItemValue[n];
     System.out.println("Enter the values and weights of the items:");
     for (int i = 0; i < n; i++) {
       int val = scanner.nextInt();
       int wt = scanner.nextInt();
       arr[i] = new ItemValue(val, wt);
     }
     System.out.print("Enter the knapsack capacity: ");
     int capacity = scanner.nextInt();
     double maxValue = getMaxValue(arr, capacity);
     System.out.println("The maximum value is " + maxValue);
}
```

Write observation based on whether this algorithm returns optimal answer or not on various inputs.

- The provided algorithm uses a greedy approach to solve the Fractional Knapsack problem.
- It sorts items based on their profit-to-weight ratio and selects items with the highest ratio first until the knapsack's capacity is exhausted.
- The algorithm implements the greedy approach correctly for the Fractional Knapsack problem.
- The greedy approach used in this algorithm returns an optimal solution for the Fractional Knapsack problem. This is because it selects items based on their profit-to-weight ratio, aiming to get the maximum value per unit weight.

Result

Write output of your program

```
E:\Files_sem-5>javac FractionalKnapSackExample.java
E:\Files_sem-5>java FractionalKnapSackExample
Enter the number of items: 3
Enter the values and weights of the items:
60 10
100 20
120 30
Enter the knapsack capacity: 50
The maximum value is 240.0
E:\Files_sem-5>
```

Conclusion:

- The provided algorithm effectively solves the Fractional Knapsack problem using a greedy approach.
- It sorts items based on their profit-to-weight ratio and selects items in a way that maximizes the total value without exceeding the knapsack's capacity.
- This approach returns an optimal solution for the Fractional Knapsack problem.
- The algorithm is a reliable and efficient tool for selecting items to achieve the highest possible value while respecting weight constraints.
- It employs a greedy strategy that consistently yields an optimal result for this specific problem.

Quiz:

1. What is the time complexity of above knapsack algorithm?

Answer: The time complexity of the provided algorithm for solving the Fractional Knapsack problem using a greedy approach is O(n log n), where 'n' is the number of items.

2. Does above algorithm always return optimal answer?

Answer: No, the above algorithm for solving the Fractional Knapsack problem using a greedy approach does not always return the optimal solution. While the greedy approach employed here works correctly for the Fractional Knapsack problem, it does not guarantee the global optimum for all possible cases. Specifically, it may not provide the optimal solution if the items have certain characteristics or constraints that the greedy strategy does not consider.

3. What is the time complexity solving knapsack problem using brute-force method? Answer: The time complexity of the brute-force method for the Knapsack problem is O(2^n), where 'n' is the number of items.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) Good Avg. Poor (1-0) | | | | Program Implementation (5) | | | Documentation & Timely Submission (2) | | | |
|---------|---|--|--|---------------|----------------------------|---------------|----------|---------------------------------------|----------|--|--|
| | | | | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | | |
| Marks | | | | | | | | | | | |

Implement program for solution of Making Change problem using greedy design technique.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3

Objectives: (a) Understand greedy algorithm design method.

(b) Solve the optimization based problem.

(c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Making Change problem is to find change for a given amount using a minimum number of coins from a set of denominations. If we are given a set of denominations $D = \{d_0, d_1, d_2, ..., d_n\}$ and if we want to change for some amount N, many combinations are possible. $\{d_1, d_2, d_5, d_8\}$, $\{d_0, d_2, d_4\}$, $\{d_0, d_5, d_7\}$ can be considered as all feasible solutions if sum of their denomination is N. The aim of making a change is to find a solution with a minimum number of coins / denominations. Following are the steps to solve coin change problem using greedy design technique

- 1. Initialize a list of coin denominations in **descending order**.
- 2. Initialize a list of coin counts, where each count is initially 0.
- 3. While the remaining amount is greater than 0:
 - a. For each coin denomination in the list:
 - If the denomination is less than or equal to the remaining amount, add one coin to the count and subtract the denomination from the remaining amount.
 - ii. If the denomination is greater than the remaining amount, move on to the next denomination.
- 4. Return the list of coin counts.

Implement the program based on above steps for the solution of fractional knapsack problem.

```
import java.util.Scanner;
public class Change_exp12 {
   public static void main(String[] args) {
        Scanner scanner = new Scanner(System.in);
        System.out.print("Enter the amount: ");
        int amount = scanner.nextInt();
```

```
System.out.print("Enter the number of coins: ");
int n = scanner.nextInt();

int[] coins = new int[n];
System.out.print("Enter the coins in descending order: ");
for (int i = 0; i < n; i++) {
    coins[i] = scanner.nextInt();
}

System.out.println("The solution is: ");
for (int i = 0; i < n; i++) {
    if (amount >= coins[i]) {
        int count = amount / coins[i];
        System.out.println(count + " " + coins[i]);
        amount -= count * coins[i];
    }
}

if (amount > 0) {
    System.out.println("There is no solution for this amount.");
}
```

Write observation based on whether this algorithm returns optimal answer or not on various inputs.

- The provided algorithm is designed to find the optimal combination of coins to make change for a given amount.
- It prompts the user for the amount, the number of available coin denominations, and the denominations themselves.
- This algorithm provides a solution that minimizes the total number of coins used.
- It starts with the highest denomination, which is the most efficient way to minimize the number of coins needed.

Result

Write output of your program

```
E:\Files_sem-5>javac Change_exp12.java
E:\Files_sem-5>java Change_exp12
Enter the amount: 73
Enter the number of coins: 4
Enter the coins in descending order: 25 10 5 1
The solution is:
2 25
2 10
3 1
```

Conclusion:

- The provided algorithm effectively calculates the optimal combination of coins to make change for a given amount.
- It starts with the highest denomination and iteratively subtracts it from the remaining amount until the amount is zero or there are no more usable coins.
- The algorithm is designed to minimize the total number of coins used.
- This algorithm is a reliable and efficient tool for making change using the provided denominations.
- It provides an optimal solution that minimizes the total number of coins needed, assuming the denominations are provided in descending order.

Quiz:

1. What is the time complexity of above knapsack algorithm?

Answer: The time complexity of the provided algorithm is O(n), where 'n' is the number of available coin denominations.

2. Does above algorithm always return optimal answer?

Answer: No, the above algorithm for making change may not always return the optimal answer. It provides a solution that minimizes the total number of coins used, but this doesn't necessarily mean it always provides the optimal solution in terms of the minimum total value of coins used.

3. What are some variations of the Making Change problem?

Answer: The Making Change problem, also known as the Coin Change problem, is a classic computational problem in computer science. It involves finding the number of ways to make change for a specific amount of money using a given set of coin denominations. Here are some variations of the Making Change problem:

Minimum Number of Coins:

Find the minimum number of coins needed to make a specific amount of change. This is similar to the provided algorithm, but instead of finding all possible combinations, it focuses on minimizing the number of coins used. Maximum Number of Ways:

Find the maximum number of ways to make change for a specific amount using a given set of coins. This variation is concerned with counting the total number of valid combinations.

Fractional Coin Change:

In this problem, it's allowed to use fractions of coins. The goal is to minimize the total number of coins or the total value of coins used to make a specific amount.

4. What is the difference between the unbounded coin change problem and the limited coin change problem?

Answer: In the unbounded coin change problem, there is no limit on the number of times a particular coin denomination can be used. In the limited coin change problem, there are restrictions on the maximum number of times each coin can be used.

IT Dept 3150703(ADA) Year 2023-24

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) Good Avg. Poor | | | | Program lementat (5) | | Doc &Tim | Total (10) | | |
|---------|---|--|--|---------------|----------------------------|---------------|----------------------------|---------------|--|--|
| | Good Avg. Poor (3) (2-1) (1-0) | | | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good Avg. Poor (2) (1) (0) | | | |
| Marks | | | | | | | | | | |

Implement program for Kruskal's algorithm to find minimum spanning tree.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3, CO6

Objectives: (a) Understand how to use Kruskal's algorithm to find the minimum spanning tree.

- (b) Solve the optimization based problem.
- (c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

In graph theory, a minimum spanning tree (MST) of an undirected, weighted graph is a tree that connects all the vertices of the graph with the minimum possible total edge weight. In other words, an MST is a subset of the edges of the graph that form a tree and have the smallest sum of weights.

- 1. Sort all the edges in non-decreasing order of their weight.
- 2. Initialize an empty set of edges for the minimum spanning tree.
- 3. For each edge in the sorted order, add the edge to the minimum spanning tree if it does not create a cycle in the tree. To check if adding the edge creates a cycle, you can use the Union-Find algorithm or a similar method to keep track of the connected components of the graph.
- 4. Continue adding edges until there are V-1 edges in the minimum spanning tree, where V is the number of vertices in the graph.
- 5. Return the set of edges in the minimum spanning tree.
- 6. Implement the program based on above steps for the solution of fractional knapsack problem.

Kruskal's Algorithm is as follow:

```
Algorithm Kruskal(E, cost, n, t)
1
2
    //E is the set of edges in G. G has n vertices. cost[u,v] is the
3
       cost of edge (u, v). t is the set of edges in the minimum-cost
4
       spanning tree. The final cost is returned.
5
6
        Construct a heap out of the edge costs using Heapify;
7
        for i := 1 to n do parent[i] := -1;
8
        // Each vertex is in a different set.
9
        i := 0; mincost := 0.0;
10
        while ((i < n-1) and (heap not empty)) do
11
```

```
3150703(ADA)
   IT Dept
                                                                                       Year 2023-24
                   Delete a minimum cost edge (u, v) from the heap
    13
                   and reheapify using Adjust;
                   j := \mathsf{Find}(u); \ k := \mathsf{Find}(v);
    14
    15
                   if (j \neq k) then
    16
    17
                        i := i + 1;
    18
                        t[i,1] := u; t[i,2] := v;
    19
                        mincost := mincost + cost[u, v];
    20
                        Union(j,k);
                   }
    21
    22
    23
              if (i \neq n-1) then write ("No spanning tree");
    24
              else return mincost;
    25
         }
Implementation:
import java.util.ArrayList;
import java.util.Collections;
import java.util.List;
import java.util.Scanner;
import java.util.ArrayList;
import java.util.Collections;
import java.util.Scanner;
class Edge implements Comparable<Edge> {
  int source, destination, weight;
  public Edge(int source, int destination, int weight) {
    this.source = source;
    this.destination = destination;
    this.weight = weight;
  @Override
  public int compareTo(Edge other) {
    return this.weight - other.weight;
class Graph2 {
  private int V, E;
  private ArrayList<Edge> edges;
  public Graph2(int vertices, int edges) {
    this.V = vertices;
    this.E = edges;
    this.edges = new ArrayList<>();
  public void addEdge(int source, int destination, int weight) {
    edges.add(new Edge(source, destination, weight));
  public ArrayList<Edge> kruskalMST() {
   SEM-5
                                             220283116029
```

```
Collections.sort(edges);
    ArrayList<Edge> minimumSpanningTree = new ArrayList<>();
    int[] parent = new int[V];
    for (int i = 0; i < V; i++) {
       parent[i] = i;
    int edgeCount = 0;
    for (Edge edge : edges) {
       if (edgeCount == V - 1) {
          break; // We already have V-1 edges in the MST.
       }
       int sourceParent = find(parent, edge.source);
       int destParent = find(parent, edge.destination);
       if (sourceParent != destParent) {
          minimumSpanningTree.add(edge);
          union(parent, sourceParent, destParent);
          edgeCount++;
    return minimumSpanningTree;
  private int find(int[] parent, int vertex) {
    if (parent[vertex] != vertex) {
       parent[vertex] = find(parent, parent[vertex]);
    return parent[vertex];
  private void union(int[] parent, int x, int y) {
    int xRoot = find(parent, x);
    int yRoot = find(parent, y);
    parent[xRoot] = yRoot;
public class KruskalAlgorithm {
  public static void main(String[] args) {
    Scanner scanner = new Scanner(System.in);
    System.out.print("Enter the number of vertices: ");
    int V = scanner.nextInt();
    System.out.print("Enter the number of edges: ");
    int E = scanner.nextInt();
    Graph2 graph = new Graph2(V, E);
    System.out.println("Enter the edges (source destination weight):");
    for (int i = 0; i < E; i++) {
       int source = scanner.nextInt();
       int destination = scanner.nextInt();
       int weight = scanner.nextInt();
       graph.addEdge(source, destination, weight);
```

```
ArrayList<Edge> minimumSpanningTree = graph.kruskalMST();

System.out.println("Edges in the Minimum Spanning Tree:");
for (Edge edge: minimumSpanningTree) {
    System.out.println(edge.source + " - " + edge.destination + " : " + edge.weight);
}
}
```

Write observation based on whether this algorithm always returns minimum spanning tree or not on various inputs.

- The provided algorithm implements Kruskal's algorithm for finding the Minimum Spanning Tree (MST) of a graph. Kruskal's algorithm always returns a minimum spanning tree for connected, weighted, and undirected graphs.
- The algorithm correctly implements Kruskal's algorithm, including the sorting of edges by weight, the use of a disjoint-set data structure for cycle detection, and the construction of the MST.

Result

Write output of your program

```
E:\Files_sem-5>javac KruskalAlgorithm.java

E:\Files_sem-5>java KruskalAlgorithm
Enter the number of vertices: 4
Enter the number of edges: 5
Enter the edges (source destination weight): 0 1 2
0 2 4
1 2 1
1 3 7
2 3 3
Edges in the Minimum Spanning Tree: 1 - 2 : 1
0 - 1 : 2
2 - 3 : 3
```

Conclusion:

- The algorithm correctly implements Kruskal's algorithm, including the sorting of edges by weight, the use of a disjoint-set data structure for cycle detection, and the construction of the MST.
- The algorithm is expected to work correctly for different inputs, including cases with disconnected or multiple components.
- Overall, this algorithm is a reliable tool for generating the Minimum Spanning Tree of a graph, demonstrating correct implementation and handling of various input scenarios.

Quiz:

1. What is the time complexity of krushkal's algorithm?

Answer: The time complexity of Kruskal's algorithm is O(E log E), where E is the number of edges.

2. Does above krushkal's algorithm always return optimal answer?

Answer: Yes, Kruskal's algorithm, as implemented in the provided code, always returns an optimal answer for finding the Minimum Spanning Tree (MST) of a connected, weighted, and undirected graph. This is a fundamental property of Kruskal's algorithm.

3. What data structure is typically used to keep track of the connected components in Kruskal's algorithm?

Answer: The data structure typically used to keep track of the connected components in Kruskal's algorithm is a disjoint-set data structure, also known as a union-find data structure.

4. When does Kruskal's algorithm stop adding edges to the minimum spanning tree? Answer: Kruskal's algorithm stops adding edges to the Minimum Spanning Tree (MST) when it has included enough edges to form a spanning tree. Specifically, the algorithm stops adding edges when the number of edges in the MST reaches V-1, where V is the number of vertices in the graph.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | rics Understanding of problem (3) | | Program Implementation (5) | | | Documentation & Timely Submission | | | Total (10) |
|---------|-----------------------------------|----------------|----------------------------|-------------------------------------|--|-----------------------------------|-------------|----------|---------------|
| Marks | Good (3) | Good Avg. Poor | | Good Avg. Poor (5-4) (3-2) (1-0) | | | Good (2) | Poor (0) | |

SEM-5 220283116029

Experiment No: 14

Implement program for Prim's algorithm to find minimum spanning tree.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO3, CO6

Objectives: (a) Understand how to use Prim's algorithm to find the minimum spanning tree.

- (b) Solve the optimization based problem.
- (c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

In graph theory, a minimum spanning tree (MST) of an undirected, weighted graph is a tree that connects all the vertices of the graph with the minimum possible total edge weight. In other words, an MST is a subset of the edges of the graph that form a tree and have the smallest sum of weights.

Prim's Algorithm is as follow:

```
Algorithm Prim(E, cost, n, t)
 2
      //E is the set of edges in G. cost[1:n,1:n] is the cost
      // adjacency matrix of an n vertex graph such that cost[i, j] is
      // either a positive real number or \infty if no edge (i, j) exists.
 4
 5
      // A minimum spanning tree is computed and stored as a set of
 6
      // edges in the array t[1:n-1,1:2]. (t[i,1],t[i,2]) is an edge in
 7
      // the minimum-cost spanning tree. The final cost is returned.
 8
 9
           Let (k, l) be an edge of minimum cost in E;
 10
           mincost := cost[k, l];
           t[1,1] := k; t[1,2] := l;
 11
 12
           for i := 1 to n do // Initialize near.
 13
               if (cost[i, l] < cost[i, k]) then near[i] := l;
               else near[i] := k;
 14
 15
           near[k] := near[l] := 0;
           for i := 2 to n-1 do
 16
           \{ // \text{ Find } n-2 \text{ additional edges for } t. \}
 17
 18
               Let j be an index such that near[j] \neq 0 and
               cost[j, near[j]] is minimum;
 19
 20
               t[i,1] := j; t[i,2] := near[j];
 21
               mincost := mincost + cost[j, near[j]];
 22
               near[j] := 0;
 23
               for k := 1 to n do // Update near[].
 24
                    if ((near[k] \neq 0) and (cost[k, near[k]] > cost[k, j]))
 25
                        then near[k] := j;
SI26
```

Implementation:

```
import java.util.ArrayList;
import java.util.Scanner;
import java.util.*;
class Graph {
  private int V;
  private List<List<Edge>> adj;
  class Edge {
    int to, weight;
    Edge(int to, int weight) {
       this.to = to;
       this.weight = weight;
  public Graph(int V) {
    this.V = V;
    adj = new ArrayList <>(V);
    for (int i = 0; i < V; i++) {
       adj.add(new ArrayList<>());
  public void addEdge(int u, int v, int weight) {
    Edge edge1 = new Edge(v, weight);
    Edge edge2 = new Edge(u, weight);
    adj.get(u).add(edge1);
    adj.get(v).add(edge2);
  public void primMST() {
    boolean[] inMST = new boolean[V];
    int[] key = new int[V];
    int[] parent = new int[V];
    Arrays.fill(key, Integer.MAX VALUE);
    key[0] = 0;
    parent[0] = -1;
    for (int count = 0; count < V - 1; count++) {
       int u = minKey(key, inMST);
       inMST[u] = true;
       for (Edge edge : adj.get(u)) {
          int v = edge.to;
          int weight = edge.weight;
          if (!inMST[v] \&\& weight < key[v]) {
            parent[v] = u;
            key[v] = weight;
       }
```

```
printMST(parent);
  private int minKey(int[] key, boolean[] inMST) {
    int min = Integer.MAX VALUE, minIndex = -1;
    for (int v = 0; v < V; v++) {
       if (!inMST[v] \&\& key[v] < min) {
          min = key[v];
          minIndex = v;
    return minIndex;
  private void printMST(int[] parent) {
    System.out.println("Edges in the Minimum Spanning Tree:");
    for (int i = 1; i < V; i++) {
       System.out.println(parent[i] + " - " + i);
public class Prim {
  public static void main(String[] args) {
    Scanner scanner = new Scanner(System.in);
    System.out.print("Enter the number of vertices: ");
    int V = scanner.nextInt():
    Graph graph = new Graph(V);
    System.out.print("Enter the number of edges: ");
    int E = scanner.nextInt();
    System.out.println("Enter the edges (source destination weight):");
    for (int i = 0; i < E; i++) {
       int u = scanner.nextInt();
       int v = scanner.nextInt();
       int weight = scanner.nextInt();
       graph.addEdge(u, v, weight);
    graph.primMST();
}
```

Observations:

Write observation based on whether this algorithm always returns minimum spanning tree or not on various inputs.

- The provided code implements Prim's algorithm for finding the Minimum Spanning Tree (MST) of a weighted, undirected graph.
- Prim's algorithm is guaranteed to return the minimum spanning tree for connected and weighted graphs.
- The algorithm correctly implements Prim's algorithm, including the initialization of key values, selection of minimum key, and the construction of the MST.

Result

Write output of your program

```
E:\Files sem-5>javac Prim.java
E:\Files sem-5>java Prim
Enter the number of vertices: 5
Enter the number of edges: 7
Enter the edges (source destination weight):
 2 3
 2 1
 3 3
 3 4
 4 2
 0 4
Edges in the Minimum Spanning Tree:
    2
   3
```

Conclusion:

- The provided algorithm correctly implements Prim's algorithm for finding the Minimum Spanning Tree (MST) of a weighted, undirected graph. It handles user input for the number of vertices, edges, and edge details effectively.
- The algorithm uses an adjacency list to represent the graph, a space-efficient approach for sparse graphs.
- This algorithm is a reliable tool for generating the Minimum Spanning Tree of a connected, weighted, and undirected graph.
- It demonstrates correct implementation and effectively handles various input scenarios.

Ouiz:

1. What is the time complexity of Prim's algorithm?

Answer: Prim's algorithm has a time complexity of $O(V^2)$ when implemented with an adjacency matrix and O(E + V log V) when implemented with an adjacency list, where V is the number of vertices and E is the number of edges.

2. Does above Prim's algorithm always return optimal answer?

Answer: Yes, the provided Prim's algorithm, as correctly implemented, always returns an optimal solution for finding the Minimum Spanning Tree (MST) of a connected, weighted, and undirected graph.

- 3. When does Prim's algorithm stop adding edges to the minimum spanning tree? Answer: Prim's algorithm stops adding edges to the Minimum Spanning Tree (MST) when all vertices have been included in the MST. In other words, it stops when the MST contains all the vertices of the original graph.
- 4. What data structure is typically used to keep track of the vertices in Prim's algorithm? Answer: In Prim's algorithm, a data structure known as a priority queue (or a min-heap) is typically used to keep track of the vertices and their corresponding key values. This data structure is crucial for efficiently selecting the next vertex to be included in the Minimum Spanning Tree (MST).

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem (3) | | | Program Implementation (5) | | | Documentation &Timely Submission (2) | | | Total (10) |
|---------|------------------------------|------------|---------------|----------------------------------|------------|---------------|--|----------|----------|---------------|
| | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | | | | | | | | | | |

Experiment No: 15

Implement DFS and BFS graph traversal techniques and write its time complexities.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO6

Objectives: (a) Understand Graph traversal techniques.

- (b) Visit all nodes of the graph.
- (c) Find the time complexity of the algorithm.

Equipment/Instruments: Computer System, Any C language editor

Theory:

Depth First Search is a graph traversal algorithm that explores as far as possible along each branch before backtracking. It is used to search for a node or a path in a graph, and is implemented recursively or iteratively.

The algorithm starts at a specified node and visits all the nodes in the graph by exploring each branch as far as possible before backtracking to the previous node. When a node is visited, it is marked as visited to prevent loops.

Consider the following steps for the implementation of DFS algorithm:

- 1. Create an empty stack and push the starting node onto it.
- 2. Mark the starting node as visited.
- 3. While the stack is not empty, pop a node from the stack and mark it as visited.
- 4. For each adjacent node to the current node, if the adjacent node has not been visited, mark it as visited and push it onto the stack.
- 5. After processing all the adjacent nodes, you can do something with the current node, such as printing it or storing it.
- 6. Repeat steps 3 to 5 until the stack is empty.

Consider the following steps for the implementation of BFS algorithm:

- 1. Create a queue Q and a set visited.
- 2. Add the starting node to the queue Q and mark it as visited.
- 3. While the queue is not empty:
 - a. Dequeue a node from the queue Q and process it.
 - b. For each adjacent node of the dequeued node:
 - i. If the adjacent node has not been visited, mark it as visited and enqueue it into the queue Q.

Implementation:

```
import java.util.*;
class Graph1 {
  private int V;
  private LinkedList<Integer>[] adj;
  Graph1(int v) {
     V = v;
     adj = new LinkedList[v];
     for (int i = 0; i < v; i++) {
       adj[i] = new LinkedList <> ();
  void addEdge(int v, int w) {
     adj[v].add(w);
  void DFS(int v, boolean[] visited) {
     visited[v] = true;
     System.out.print(v + " ");
     for (Integer neighbor : adj[v]) {
       if (!visited[neighbor]) {
          DFS(neighbor, visited);
  void BFS(int start) {
     boolean[] visited = new boolean[V];
     LinkedList<Integer> queue = new LinkedList<>();
     visited[start] = true;
     queue.add(start);
     while (!queue.isEmpty()) {
       start = queue.poll();
       System.out.print(start + " ");
       for (Integer neighbor : adj[start]) {
          if (!visited[neighbor]) {
            visited[neighbor] = true;
            queue.add(neighbor);
      }
  }
public class GraphTraversal {
  public static void main(String[] args) {
     Scanner scanner = new Scanner(System.in);
     System.out.print("Enter the number of vertices: ");
```

```
int V = scanner.nextInt();
  Graph1 graph = new Graph1(V);
  System.out.print("Enter the number of edges: ");
  int E = scanner.nextInt();
  System.out.println("Enter the edges (source destination):");
  for (int i = 0; i < E; i++) {
    int source = scanner.nextInt();
    int destination = scanner.nextInt();
    graph.addEdge(source, destination);
  System.out.print("Choose traversal technique (DFS or BFS): ");
  String technique = scanner.next();
  System.out.print("Enter the starting vertex: ");
  int start = scanner.nextInt();
  if ("DFS".equalsIgnoreCase(technique)) {
     System.out.println("Depth-First Search (DFS) traversal:");
    boolean[] visited = new boolean[V];
     graph.DFS(start, visited);
  } else if ("BFS".equalsIgnoreCase(technique)) {
     System.out.println("Breadth-First Search (BFS) traversal:");
     graph.BFS(start);
  } else {
     System.out.println("Invalid traversal technique.");
}
```

Observations:

Write observation based on output of algorithm that which node of graph is traversed first in BFS and DFS.

DFS Traversal:

The starting vertex for DFS is determined by the user's input.

The DFS algorithm starts from the specified starting vertex and explores as far as possible along each branch before backtracking.

The first node traversed in DFS will be the starting vertex provided by the user.

• BFS Traversal:

Similar to DFS, the starting vertex for BFS is determined by the user's input.

The BFS algorithm explores all the vertices at the current level before moving on to the next level.

The first node traversed in BFS will also be the starting vertex provided by the user.

Result

Write output of your program

```
E:\Files_sem-5>javac GraphTraversal.java

E:\Files_sem-5>java GraphTraversal

Enter the number of vertices: 6

Enter the number of edges: 7

Enter the edges (source destination):
0 1
0 2
1 3
1 4
2 5
3 5
4 5

Choose traversal technique (DFS or BFS): BFS

Enter the starting vertex: 0

Breadth-First Search (BFS) traversal:
0 1 2 3 4 5
```

Conclusion:

- The provided algorithm implements graph traversal using both Depth-First Search (DFS) and Breadth-First Search (BFS) techniques.
- It allows the user to input the number of vertices and edges, along with the edges themselves, and then choose between DFS and BFS for traversal.
- The algorithm correctly initializes the graph, adds edges, and performs traversal based on the user's choice.
- This algorithm is a functional and flexible tool for exploring graph traversal techniques.
- It provides a user-friendly interface and produces accurate results.

Quiz:

- 1. What data structure is typically used in the iterative implementation of DFS and BFS? Answer: In the iterative implementation of DFS and BFS, a stack is typically used for DFS, and a queue is used for BFS. These data structures help keep track of the nodes to be visited next in the respective traversal algorithms.
- 2. What is the time complexity of DFS on a graph with V vertices and E edges? Answer: The time complexity of Depth-First Search (DFS) on a graph with V vertices and E edges is O(V+E).
- 3. What is the time complexity of BFS on a graph with V vertices and E edges? Answer: : The time complexity of Depth-First Search (DFS) on a graph with V vertices and E edges is O(V+E).

4. In which order are nodes visited in a typical implementation of BFS?

Answer: In a typical implementation of BFS (Breadth-First Search), nodes are visited in the following order:

Start from the initial node (source node).

Visit all of its neighbors at the current depth level.

Move to the next depth level and visit all the neighbors at that level.

Continue this process until all reachable nodes have been visited.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem | | | Program Implementation | | | Documentation &Timely Submission | | | Total (10) |
|---------|--------------------------|-------|-------|---------------------------|-------|-------|-------------------------------------|------|------|---------------|
| | (3) | | | (5) | | | (2) | | | |
| | Good | Avg. | Poor | Good | Avg. | Poor | Good | Avg. | Poor | |
| | (3) | (2-1) | (1-0) | (5-4) | (3-2) | (1-0) | (2) | (1) | (0) | |
| Marks | | | | | | | | | | |

SEM-5 220283116029 ⁸

Experiment No: 16

Implement Rabin-Karp string matching algorithm.

Date:

Competency and Practical Skills: Algorithmic thinking, Programming Skills, Problem solving

Relevant CO: CO4

Objectives: (a) Find all occurrences of a pattern in a given text.

- (b) Improve the performance of the brute force algorithm.
- (c) Find a pattern in a given text with less time complexity in the average case.

Equipment/Instruments: Computer System, Any C language editor

Theory:

It is a string searching algorithm that is named after its authors Richard M. Carp and Michael O. Rabin. This algorithm is used to find all the occurrences of a given pattern 'P'' in a given string 'S' in O(Ns + Np) time in average case, where 'Ns' and 'Np' are the lengths of 'S'' and 'P', respectively.

Let's take an example to make it more clear.

Assume the given string S = ``cxyzghxyzvjkxyz'' and pattern P = ``xyz'' and we have to find all the occurrences of 'P' in 'S'.

We can see that "xyz" is occurring in "cxyzghxyzvjkxyz" at three positions. So, we have to print that pattern 'P' is occurring in string 'S' at indices 1, 6, and 12.

Naive Pattern Searching (brute force) algorithm slides the pattern over text one by one and checks for a match. If a match is found, then slide by 1 again to check for subsequent matches. This approach has a time complexity of O(P* (S-P)).

The Rabin-Karp algorithm starts by computing, at each index of the text, the hash value of the string starting at that particular index with the same length as the pattern. If the hash value of that equals to the hash value of the given pattern, then it does a full match at that particular index.

Rabin Karp algorithm first computes the hash value of pattern P and first Np characters from S. If hash values are same, we check the equality of actual strings. If the pattern is found, then it is called hit. Otherwise, it is called a spurious hit. If hash values are not same, no need to compare actual strings.

Steps of Rabin-Karp algorithm are as below:

1. Calculate the hash value of the pattern: The hash value of the pattern is calculated using a hash function, which takes the pattern as input and produces a hash value as output.

SEM-5 220283116029 83

- 2. Calculate the hash values of all the possible substrings of the same length in the text: The hash values of all the possible substrings of the same length as the pattern are calculated using the same hash function.
- 3. Compare the hash value of the pattern with the hash values of all the possible substrings: If a match is found, the algorithm checks the characters of the pattern and the substring to verify that they are indeed equal.
- 4. Move on to the next possible substring: If the characters do not match, the algorithm moves on to the next possible substring and repeats the process until all possible substrings have been compared.

Implement the Rabin-Karp algorithm based on above steps and give different input text and pattern to check its correctness. Also, find the time complexity of your implemented algorithm.

```
package com.company;
import java.util.Scanner;
public class RabinKarpAlgorithm {
  public static void main(String[] args) {
     Scanner scanner = new Scanner(System.in);
     System.out.print("Enter the text: ");
     String text = scanner.nextLine();
     System.out.print("Enter the pattern to search for: ");
     String pattern = scanner.nextLine();
     searchPattern(text, pattern);
  }
  public static void searchPattern(String text, String pattern) {
     int textLength = text.length();
     int patternLength = pattern.length();
     int prime = 101; // A prime number for the hash function
     long textHash = 0;
     long patternHash = 0;
     long h = 1;
    // Calculate h, which is the hash value of (prime^(patternLength-1))
     for (int i = 0; i < patternLength - 1; i++) {
       h = (h * prime) \% 101;
     }
     // Calculate the initial hash values of the text and pattern
     for (int i = 0; i < patternLength; i++) {
```

```
textHash = (textHash * prime + text.charAt(i)) % 101;
  patternHash = (patternHash * prime + pattern.charAt(i)) % 101;
for (int i = 0; i \le \text{textLength} - patternLength; i++) {
  if (textHash == patternHash) {
     // If hash values match, perform character-wise comparison
     boolean match = true;
     for (int j = 0; j < patternLength; j++) {
       if (text.charAt(i + j) != pattern.charAt(i)) {
          match = false;
          break:
     }
     if (match) {
       System.out.println("Pattern found at index " + i);
  if (i < textLength - patternLength) {
     // Update the hash value for the next substring
     textHash = (prime * (textHash - text.charAt(i) * h) + text.charAt(i + patternLength)) % 101;
     if (\text{textHash} < 0) {
       textHash += 101; // Ensure the hash value is non-negative
```

Observations:

Write observation based on whether this algorithm able to find a pattern in a given text or not and find it's time complexity in worst case and average case based on its way of working.

- The Rabin-Karp algorithm is able to find a pattern in a given text.
- It uses a rolling hash function to efficiently compare the hash values of substrings in the text with the hash value of the pattern.
- The algorithm successfully identifies occurrences of the pattern in the text.
- Time Complexity:
- Worst Case: O((n-m+1)*m), where 'n' is the length of the text and 'm' is the length of the pattern. This occurs when the hash values of all possible substrings need to be compared.
- Average Case: O(n+m), as the algorithm efficiently compares hash values and only performs character-wise comparison when hash values match.

85

Result

Write output of your program

```
E:\Files_sem-5>javac RabinKarpAlgorithm.java
E:\Files_sem-5>java RabinKarpAlgorithm
Enter the text: ababcababc
Enter the pattern to search for: abc
Pattern found at index 2
Pattern found at index 7
```

Conclusion:

- The Rabin-Karp algorithm is an efficient pattern matching algorithm that uses a rolling hash function to compare substrings in a text with a given pattern.
- It successfully identifies occurrences of the pattern in the text.
- The algorithm's time complexity is O((n-m+1)*m) in the worst case and O(n+m) on average, making it a practical choice for pattern matching in various scenarios.
- It is particularly useful for handling long texts and patterns efficiently.

Ouiz:

1. What is the Rabin-Karp algorithm used for?

Answer: The Rabin-Karp algorithm is used for pattern matching in strings. It efficiently finds occurrences of a given pattern within a larger text. It does so by using a rolling hash function to quickly compare substrings, making it particularly effective for handling long texts and patterns efficiently.

2. What is the time complexity of the Rabin-Karp algorithm in the average case? Answer: The average case time complexity of the Rabin-Karp algorithm is O((n + m) * p), where: n is the length of the text m is the length of the pattern p is the time taken to compute the hash function

3. What is the main advantage of the Rabin-Karp algorithm over the brute force algorithm for string matching?

Answer: The main advantage of the Rabin-Karp algorithm over the brute-force algorithm for string matching lies in its efficiency, especially for large texts and patterns.

4. What is the worst-case time complexity of the Rabin-Karp algorithm?

Answer: The worst-case time complexity of the Rabin-Karp algorithm is O((n - m + 1) * m), where: n is the length of the text, m is the length of the pattern.

Suggested Reference:

- 1. "Introduction to Algorithms" by Thomas H. Cormen, Charles E. Leiserson, Ronald L. Rivest, and Clifford Stein
- 2. "Fundamentals of Algorithms" by E. Horowitz et al.

References used by the students:

Rubric wise marks obtained:

| Rubrics | Understanding of problem | | | Program Implementation | | | Documentation &Timely Submission | | | Total (10) |
|---------|--------------------------|------------|---------------|---------------------------|------------|---------------|-------------------------------------|----------|-------------|---------------|
| | (3) | | | (5) | | | (2) | | | |
| | Good (3) | Avg. (2-1) | Poor (1-0) | Good (5-4) | Avg. (3-2) | Poor (1-0) | Good (2) | Avg. (1) | Poor (0) | |
| Marks | | , / 1 | | , / 1 | , / 1 | ` / | | | | |

87