$n^{0.1}$  $(\log n)^{10}$ (iv)  $n2^n$ **Solution:** The definition for the relations between f(n) and g(n) are as follows :  $f(n) = O(g(n)) \iff \exists c, n_o > 0 \text{ s.t. } \forall n \geq n_o, f(n) \leq c \cdot g(n)$  $f(n) = \Omega(g(n)) \iff \exists c, n_o > 0 \text{ s.t. } \forall n \geq n_o, f(n) \geq c \cdot g(n)$  $f(n) = \Theta(g(n)) \iff \exists \ c_1, c_2, n_o > 0 \ \mathrm{s.t.} \ orall \ n \geq n_o, \ \ c_1 \cdot g(n) \leq f(n) \leq c_2 \cdot g(n)$ We will use the above definitions for deciding the relation between f(n) and g(n). (i) f(n) = n - 100, g(n) = n - 200We see that  $\forall n \in \mathbb{R}, n-200 \le n-100 \implies f(n) = \Omega(g(n))$ We see that for the inequality,  $n-100 \le c(n-200)$  to be satisfied,  $(c-1)n-200c+100\geq 0$ . Let c = 2  $\implies n-300\geq 0 \implies n\geq 300$ For  $n_o = 300, c = 2, f(n) \le c \cdot g(n) \implies f(n) = O(g(n))$ Thus, for  $n_0 = 300$ ,  $c_1 = 1$ ,  $c_2 = 2$ ,  $c_1 \cdot g(n) \le f(n) \le c_2 \cdot g(n)$ The conditions for  $\Theta$  is satisfied. Thus  $ig|f=\Theta(g)ig|$ (ii)  $f(n) = \log 2n, g(n) = \log 3n$ We see that  $\forall n > 0$ ,  $\log 2n \leq \log 3n \implies f(n) = O(g(n))$ We see that for the inequality,  $\log 2n > c \log 3n$  to be satisfied  $(1-c)\log n + \log 2 - c\log 3 \ge 0$ Let  $c=0.5 \implies 0.5 \log n + \log \frac{2}{\sqrt{3}} \geq 0 \implies n \geq \frac{3}{4}$ For  $n_o = rac{3}{4}, c = 0.5, f(n) \geq c \cdot g(n) \implies f(n) = \Omega(g(n))$ Thus, for  $n_o=rac{3}{4}, c_1=0.5, c_2=1$  ,  $c_1\cdot g(n)\leq f(n)\leq c_2\cdot g(n)$ The conditions for  $\Theta$  is satisfied. Thus  $ig|f=\Theta(g)ig|$ (iii)  $f(n) = n^{0.1}, g(n) = (\log n)^{10}$ Let  $n_o = 2^{1000}, c = 1$ , we see that  $f(n) \geq c \cdot g(n) \ orall \ n \geq n_o$ Also,  $\lim_{n\to\infty}\frac{n^{0.1}}{(\log n)^{10}}=\infty$ , which implies that the value of  $n^{0.1}$  is greater than  $(\log n)^{10}$  asymptotically. Thus the conditions for  $\Omega$  is satisfied. Also, there is no  $n_o,c$  to satisfy the condition for O. Thus  $|f=\Omega(g)|$ (iv)  $f(n) = n2^n, g(n) = 3^n$ Let  $n_o=3, c=1$ , we see that  $f(n) \leq c \cdot g(n) \ \forall \ n \geq n_o$ Also,  $\lim_{n \to \infty} n \left( \frac{2}{3} \right)^n = 0$ , which implies the value of  $n2^n$  is greater than  $3^n$  asymptotically. Thus the conditions for O is satisfied. Also, there is no  $n_o,c$  to satisfy the conditions for  $\Omega$ . Thus f=O(g)Q3. Describe an efficient algorithm for finding the ten largest elements in a sequence of size n. What is the running time of your algorithm? **Solution** : O(n) **Time Complexity** We will use the following algorithm for finding the k largest elements in an array. For k = 10, we will get the solution for the above question. Let the array arr be of length n(>=k) We first find the (n-k)th largest element in the array, which is found using the Random QuickSelect algorithm as follows: We first select a pivot in the given array randomly which is between two given indices I and r • We push elements which are less than the *pivot* element to the left of it and the elements which are greater than the *pivot* element to the right of it. After the above operation, the pivot element will be in the correct position of the array. Thus, we can do the following based on the • If *pivot* position is equal to (n-k), we return that element. • If pivot position is greater than (n-k), we recursively search in the left part of the array (I,pivot-1) • If *pivot* position is less than (n-k), we recursively search in the right part of the array (*pivot+1,r*) • The time complexity of the above algorithm is O(n) on average. The proof is given as follows: Assuming we split the array into half on average (due to the random pivot select), we see that the size of the array we need to compute on reduces by half in each step. The maximum number of steps we take is  $\log n$  where n is the size of the array. • Thus the time taken would be as follows : (Note :  $2^{\log n} = n$ )  $T(n) = (n + rac{n}{2} + rac{n}{4} + rac{n}{8} \ldots)_{\log n ext{ times}} = n \left(rac{1 - \left(rac{1}{2^{\log n}}
ight)}{1 - rac{1}{2}}
ight) = 2n \left(1 - rac{1}{n}
ight) = 2n - 2 = O(n)$ • The average time complexity is O(n) assuming we split the array equally on average due to the random pivot selection. But the worst-case time complexity is  $O(n^2)$  which happens when we repeatedly select the least or greatest element of the subarray as pivot. • After finding the (n-k)th largest element in the array, we again iterate through the array and take only elements which are greater than or equal to the (n-k)th array. This gives us the required k greatest numbers in the array. This operation is O(n) time complexity. Thus the overall time complexity of the algorithm is O(n) + O(n) = O(n) time complexity on average. In [1]: Python code to get the 10 largest elements from an array # Importing random library to randomly select pivot from random import randint Function to sort the array around a random pivot Input -> arr - Array, l - Left Index, r - Right Index Output -> Array which is sorted around pivot and the pivot index def randPartition(arr, l, r): # Selecting the random pivot m = randint(l, r)# Swap the pivot to the end for easier sorting arr[r], arr[m] = arr[m], arr[r]# Variable to store current swap position # Pivot element value x = arr[r]# Loop to sort the array around the pivot for j in range(l, r): # If element is smaller than pivot, **if** arr[j] <= x: # Swap element to ith position arr[i], arr[j] = arr[j], arr[i] # Increment i i += 1 # Swap the pivot to the correct position arr[r], arr[i] = arr[i], arr[r] # Return the pivot position return i Function to find the kth largest element Input -> arr - Array, l - Left Index, r - Right Index, k Output -> k th largest element def kthLargestElement(arr, l, r, k): # Check if k is less than size of array **if** k > 0 **and**  $k \le r - l + 1$ : # Apply randomPartition and get the pivot position p = randPartition(arr, l, r) # Compare pivot position to k and act accordingly **if** p - l == k - 1: return arr[p] **elif** p - l < k - 1: return kthLargestElement(arr, p + 1, r, k - p + l - 1) return kthLargestElement(arr, l, p - 1, k) Function to return k largest element Input -> arr - Array, k Output -> k largest elements def kLargestElements(arr, k): # Get the (n-k)th largest element  $k_l = kthLargestElement(arr, 0, len(arr) - 1, len(arr) - k + 1)$ # Variable to store answer ans = [] for i in arr: # Store values in answer which are >= (n-k)th element **if** i >= k\_l: ans.append(i) return ans # Checking the function arr = [3, 2, 1, 4, 5, 9, 8, 7, 10, 6, 11, 12, 15, 14, 13]print("10 largest elements :", kLargestElements(arr, 10)) 10 largest elements : [6, 7, 8, 10, 13, 11, 12, 15, 14, 9] Q4. Use the divide and conquer integer multiplication algorithm to multiply the two binary integers 10011011 and 10111010. Solution :  $O(n^{1.6})$  Time Complexity We will be using the Karatsuba divide and conquer algorithm for multiplication. • Given two binary numbers x and y to be multiplied, we will write x as  $2^{(n/2)}a+b$  and y as  $2^{(n/2)}c+d$ • Now we can write  $x \cdot y$  as  $2^n(a \cdot c) + 2^n(a \cdot c) + 2^{(n/2)}(a \cdot d + b \cdot c) + b \cdot d$ • Now we recursively call the function for the products  $a \cdot c$ ,  $a \cdot d$ ,  $(a+b) \cdot (c+d)$ • The term  $a \cdot d + b \cdot c$  can be calculated as  $(a+b) \cdot (c+d) - a \cdot c - b \cdot d$ • The stopping condition is when we have 1 digit products, where we just directly return the result The time complexity is given as follows: We do  $3^k$  steps at each level, and the last level is of size  $\log n$ . Thus the final time complexity is  $3^{\log n} = n^{\log 3} \approx O(n^{1.6})$ Function to multiply two binary numbers Input -> Two binary numbers x,y Output -> Binary number which is the product of x and y def binary\_multiply(x, y): # Get the length of the array

> $n_x = len(x)$  $n_y = len(y)$

**if** n == 1:

else:

**while** n % 2:

x = "0" + x y = "0" + y n += 1

# Calculating a,b,c,d

# Calculating a+b and c+d

ac = binary\_multiply(a, c)

bd = binary\_multiply(b, d)

 $(int(ac, 2) \ll n)$ 

+ int(bd, 2)

# Checking the function

Multiplication using D&C

array that runs in  $O(\log n)$  time.

In [3]:

**Solution** :  $O(\log n)$  Time Complexity

Input -> Unimodal Array arr

def max\_unimodal(arr):

n = len(arr)

while l < r:

else:

**if** l == r:

Largest element = 5

Implement this in Python.

Solution :  $O(2^n)$  Time Complexity

We shift the n-1 disks to the middle peg
 We shift the largest disk to the last peg

Python code to solve Tower of Hanoi

Function to draw the current position of disks

Output -> Print the position of disks in pegs

# Variable holding number of disks used

# Loop to print either a disk or a bar (|)
for j in range(num\_disks - 1, -1, -1):

if len(disks[i]) < j + 1:
 print("|", end="\t")</pre>

print(disks[i][j], end="\t")

Function to solve Tower of Hanoi for given number of disks

Input -> disks - Array containing the disks in each pegs

start\_peg - Index pointing to start peg
end\_peg - Index pointing to end peg
temp\_peg - Peg used to shift n-1 disks

def solve\_hanoi(disks, n, start\_peg, end\_peg, temp\_peg):

# If only one disk remains we push it to end peg

# Else, we first shift n-1 disks to temp peg

Function to setup the Tower of Hanoi and solve it

Output -> Solves tower of hanoi and displays each step

# Creating the array of disks to represent pegs
disks = [[i for i in range(n, 0, -1)], [], []]

num\_disks = int(input("Enter the number of disks :"))

# Using deepcopy so that num\_disks is not affected in recursion

print("Starting Solve for n=", n, "\n")

# Solving the Tower of Hanoi
solve\_hanoi(disks, n, 0, 2, 1)

disks[end\_peg].append(disks[start\_peg].pop(-1))

solve\_hanoi(disks, n - 1, start\_peg, temp\_peg, end\_peg)

# We then shift the n-1 disks from temp to the end peg solve\_hanoi(disks, n - 1, temp\_peg, end\_peg, start\_peg)

# We then shift the remaining disk to the end peg
disks[end\_peg].append(disks[start\_peg].pop(-1))

Input -> disks - Array containing the disks in each pegs

from copy import deepcopy

def draw(disks):

global num\_disks

print()
print("\n")

**if** n == 1:

draw(disks)

0.00

1 2

2

3

3

1

def hanoi(n):

draw(disks)

# Checking the code

hanoi(deepcopy(num\_disks))

Enter the number of disks :3

Starting Solve for n= 3

1

2

1

2

3

2

1 2 3

draw(disks)
return

Input -> n - Number of disks

for i in range(3):

n - Number of disks

Output -> Print the position of disks in pegs

3. We then shift back the n-1 disks to the last peg

We follow the steps as follows:

In [10]:

if l == r - 1:

mid = (l + r) // 2

r = mid - 1

l = mid + 1

return arr[l]

print("Largest element =", a)

return arr[mid]

l = 0r = n - 1

# Define l and r

Output -> Maximum number in array

# Get the length of the array

# Loop to apply the above algorithm

return max(arr[l], arr[r])

if arr[mid] > arr[mid - 1] and arr[mid] > arr[mid + 1]:

if arr[mid] < arr[mid - 1] and arr[mid] > arr[mid + 1]:

Q6. Towers of Hanoi: Given a game board with three pegs and a set of disks of different diameter all stacked from smallest to largest on the leftmost peg, moves all of the disks to the rightmost peg following these two rules. First, only one disk may be moved at a time. Second, a larger diameter disk may never be placed on a smaller disk. Any number of disks can be used.

This is a recursive method where for solving for n-1, we call the solution for n-2 and so on. This makes the solution  $O(2^n)$ 

 $a = max\_unimodal([1, 2, 3, 4, 5, 2, 1, 0, -1, -2, -3, -4])$ 

2. Until  $l \ll r$ , we calculate mid = (l+r)//2

D. If l==r, we return the element at I

E. If l=r-1, we return max of the element at I or r

Function to find maximum number in an unimodal array

 $a_b = bin(int(a, 2) + int(b, 2))[2:]$  $c_d = bin(int(c, 2) + int(d, 2))[2:]$ 

 $ab\_cd = binary\_multiply(a\_b, c\_d)$ 

# Recursively calculating ac, (a+b)(c+d), bd

# Calculating the product x\*y using ac,ad,bc,bd

+ ((int(ab\_cd, 2) - int(ac, 2) - int(bd, 2)) << n // 2)

print("General Multiplication Result\t:", bin(x\_int \* y\_int)[2:])
print("Multiplication using D&C\t:", (binary\_multiply(x, y)))

: 111000010011110

We will use binary search method to find the maximum element of a unimodal array as follows :

B. If the element at mid is lesser than the right element, then we shift l = mid+1 C. If the element at mid is greater than the right element, then we shift r = mid-1

A. If the element at mid is greater than both its nearby elements, then the element at m is maximum

This algorithm runs in  $O(\log n)$  time, since the array needed to be searched is cut into half at every step.

1. We define two pointers l = 0, r = n-1 where n = length of the array

Q5. You are given a unimodal array of n distinct elements, meaning that its entries are in increasing order up until its maximum elements, after which its elements are in decreasing order. Give an algorithm to compute the maximum element of a unimodal

General Multiplication Result : 111000010011110

a = x[: n // 2]b = x[n // 2 :]

c = y[: n // 2]d = y[n // 2 :]

ans = bin(

)[2:] return ans

x = "10011011" y = "10111010"

 $x_{int} = int(x, 2)$  $y_{int} = int(y, 2)$ 

 $n = max(n_x, n_y)$ 

 $x = "0" * (n - n_x) + x$  $y = "0" * (n - n_y) + y$ 

# To make the length of the arrays same

**if** x == "1" **and** y == "1":

return "1"

return "0"

# If the elements are of size 1, we directly return the result

# Making the length of operands even to make implementation easier

EE4371 - Assignment 2

Om Shri Prasath, EE17B113

 $\boxed{2^{10} < 2^{\log(n)} < 3n + 100\log(n) < 4n < n\log(n) < 4n\log(n) + 2n < n^2 + 10n < n^3 < 2^n}$ 

Q2. In each of the following situations, indicate whether f=O(g), or  $f=\Omega(g)$ , or both (in which case  $f=\Theta(g)$  ). Justify your

f(n)

(ii)  $\log 2n$ 

(i) n-100 n-200

g(n)

 $\log 3n$ 

Q1. Order the following functions by asymptotic growth rate.

1.  $a \cdot f(n) < b \cdot f(n)$  for a < b where a, b are constants

3. Only the higher order terms needs to be considered

Also  $2^{\log n} = n$  since the  $\log$  here is of base 2

Using the above rules, we get:

2. constant < logarithmic < polynomial < exponential is the general order

The order of asymptotic growth rate for general functions are given by following thumb rules:

•  $4n\log(n) + 2n$ 

•  $3n + 100 \log(n)$ 

2<sup>10</sup>
2<sup>log(n)</sup>

• 4n•  $2^n$ 

•  $n^3$ 

Solution:

answer.

•  $n^2 + 10n$ 

•  $n \log(n)$