

DAA Assignment - 1

Ques:- What do you understand by Asymptotic notations. Define different Asymptotic notation with example.

Ans:- Asymptotic Notation is used to describe the running time of an algorithm - how much time an algorithm take with a given input, n .

Different types of notations:-① Big O Notation

$$f(n) \leq O(g(n))$$

Big O Notation refers an upper bounds on the growth rate of a function.

Example:- If an algorithm has time complexity of $O(n)$, it means that the running time of an algorithm grows linearly with the size of input.

(2) Omega Notation (n):-

$$f(n) \geq g(n)$$

Omega notation represents a lower bound on growth rate of a function.

Example:- If an algorithm have time complexity of $\Theta(n^2)$ or $\Omega(n^2)$. It means its running time grows at least quadratically with size of input.

(3) Theta Notation (Θ):-

$$f(n) = \Theta(g(n))$$

Theta Notation represents a lower bound and upper bound ^{both} on the growth rate of function.

Example:- If an algorithm have time complexity of $\Theta(n)$, it means running time grows linearly with size of input.

$$f(n) \leq g(n)$$

small o notation represents an upper bound that is not tight. It indicates that a function grows strictly slower than another function.

Ex If an algorithm has time complexity of $o(n^2)$ it means that running time grows strictly slower than n^2 .

5 Small omega (ω)

$$f(n) > \omega(g(n))$$

Small omega notation represents a lower bound that is not tight. It indicates that function grows, strictly faster than another function

Example if an algorithm has a time complexity of $\omega(n)$, it means that the running time grows strictly faster than linear.

Q2: What should be time complexity of -

for $\{ i \leftarrow n \}$

S

$i \leftarrow i^2;$

}

The loop continues until 'i' exceeds or equal to 'n'. Therefore the no. of iterations can be expressed as:-

$$2^k \geq n$$

log both sides

$$k \geq \log_2(n)$$

So, loop runs $O(\log(n))$.

Q3: $T(n) = \begin{cases} 3 + (n-1) & \text{if } n > 0 \\ 1 & \text{otherwise} \end{cases}$

1) Base case $\rightarrow T(0) = 1.$

2) Recursive case \rightarrow for $n > 0$, $T(n) = 3T(n-1)$

Expanding $T(n)$:-

$$T(n) = 3T(n-1)$$

$$\rightarrow 3^2T(n-2) = 3^3T(n-3) = \dots = 3^nT(0)$$

Substitute $\rightarrow T(0) = 1.$

$$T(n) = 3^n$$

So, solⁿ to xclusive selection is

$$T(n) = 3^n.$$

So, solⁿ to recursive selection is $T(n) = 3^n$.

Time complexity in in recursive selection is exponential in term of n .

Q4: $T(n) = \begin{cases} 2T(n-1) - 1 & \text{if } n > 0, \\ 1 & \text{otherwise} \end{cases}$

1) Base Case: $T(0) = 1$.

2) Recursive Case:

for $n > 0$, the recursive selection $T(n) =$
 $2T(n-1) - 1$

Now, let's expand $T(n)$

$$T(n) = 2T(n-1) - 1 = 2(2T(n-2) - 1) - 1$$

$$= 2^2 T(n-2) - 2 =$$

$$2^3 T(n-3) - 3 = \dots = 2^n T(0) - (2^{n-1} + 2^{n-2} + 2^{n-3} + \dots + 1)$$

The sum $2^{n-1} + 2^{n-2} + \dots + 1$ is a geometric series and its sum is $2^n - 1$.

Substitute the base case :-

$$T(n) = 2^n - (2^n - 1) = 1$$

$$T(n) = 1$$

$$O(1)$$

Q5: What should be time complexity of :-

```
int i=1, s=1;
while (s <= n) {
    i++;
    s = s + i;
    printf("#");
}
```

$$\text{After 1st iteration} = s = 1 + 2 = 3$$

$$\text{After 2nd iteration} = s = 3 + 3 = 6$$

$$\text{After 3rd iteration} = s = 6 + 4 = 10$$

$$\text{After } k\text{th iteration} = 1 + 2 + 3 + 4 + 5 + \dots + k$$

$$= \frac{k(k+1)}{2}$$

So, finding value of k

$$\frac{k(k+1)}{2} > n$$

$$k^2 + k > n$$

$$\frac{k^2 + k}{2} > 2n$$

$$k > \sqrt{n}$$

Time complexity is $O(K)$

$$= O(\sqrt{n})$$

Q6: Time complexity of :-

```
void function (int n) {
    int i, count = 0;
    for (i = 1; i * i <= n; i++)
        count++;
}
```

Loop have condition $i * i \rightarrow i = i \leq n$ so,

loop will continue until it become greater than root n.

Count \rightarrow take constant time.

$$O(\sqrt{n})$$

Q7: Time complexity of :-

```
void function (int n) {
    int i, j, k, count = 0;
    for (i = n/2; i <= n; i++)
        for (j = 1; j <= n; j = j + 2)
```

for ($k > 1$; $k \leq n$; $k = k \cdot 2$)

wanted;

}

Outer loop runs $\rightarrow n/2$ times.

Middle j loop runs $\rightarrow O(\log n)$ times because
 j doubles in iteration.

Inner loop also runs $\rightarrow O(\log_2 n)$ times
 because K doubles in each iteration.

$$\begin{aligned} \text{Total no. of iterations} &= O\left(\frac{n}{2} \cdot \log n \cdot \log_2 n\right) \\ &= O(n \log^2 n) \end{aligned}$$

Time complexity $= O(n \log^2 n)$.

(Qn:- Time complexity of :-

function (int n) {

if ($n == 1$) return;

for ($i = 1$ to n) {

for ($j = 1$ to n) {

printf(" * ");

}

}

- base case is when $n=1$, then function returns without further recursion.
- outputs outer loop num from $i=1$ to n .
- inner loop num from $j=1$ to n

$O(n^2)$ since, loop inside loop.

The recursive call is made with the parameter ' $n-3$ '. Assuming the initial value of ' n ' is divisible by 3, the recursive call approximately $n/3$ times before reaching the base case where ' n ' becomes 1.

$$T(n) = T(n-3) + O(n^2)$$

], over all time complexity.

Or time complexity of :-

```
void function (int n) {
```

```
    for (i=1 to n) {
```

```
        for (j=1; j<=n; j=j+1)
```

```
            printf(" * ")
```

```
}
```

```
}
```

In inner loop, variable 'j' is incremented by 1 in each iteration. The no. of iterations of the inner loop depends on value of 'i'. Let's denote the no. of iterations of inner loop for specific 'i' as $\frac{n}{i}$

Calculate total no. of iterations:-

$$\frac{n}{1} + \frac{n}{2} + \frac{n}{3} + \dots + \frac{n}{n}$$

Sum of h.p $O(\log n)$.

Time complexity $\rightarrow O(n \log n)$

Qn: for the function n^k and c^n what is the asymptotic relationship b/w these functions?
Assume that $k >= 1$ and $c > 1$ are constant.
find out the value of c & n_0 for which relation holds.

The asymptotic relationship b/w n^k and c^n . depends on values of k and c .

- " If $k > 1$, then n^k grows faster than c^n . In this case, n^k is said to have a polynomial growth rate, c^n has an exponential growth rate. Asymptotic selection $\rightarrow n^k$ is $O(c^n)$

2) If $k=1$, then both n^k and c^n have some growth rate and they are considered to be in the same complexity class. Asymptotic relation $[n^k \text{ is } O(c^n)]$

3) If $0 < k < 1$, then c^n grows faster than n^k . In this case, n^k has a slower growth rate than c^n . Asymptotic Relation $[n^k \text{ is } o(c^n)]$

To find the value of c and n_0 where the relationship holds, you need to consider the limits:

1) for n^k growing faster ($k > 1$):

$$\lim_{n \rightarrow \infty} \frac{n^k}{c^n} = 0$$

In this case any $c > 1$ will work and no can be any positive constant.

2) for c^n growing faster ($0 < k < 1$):

$$\lim_{n \rightarrow \infty} \frac{n^k}{c^n} \rightarrow 0$$

In this case, c must be greater than 1 and no can be any positive constant.

Q11) What will be the time complexity for extraction ()? Given min heap of n nodes
 extraction () :- return the min element for the heap.

```

void heapifydown (int heap[], int n, int i) {
    int smallest = i;
    int left = 2 * i + 1;
    int right = 2 * i + 2;
    if (left < n && heap [left] < heap [smallest])
        smallest = left;
    if (right < n && heap [right] < heap [smallest])
        smallest = right;
    if (smallest != i) {
        swap (&heap [i], &heap [smallest]);
        heapifydown (heap, n, smallest);
    }
}

int extractMin (int heap [], int *n) {
    if (*n == 0)
        return -1;
    swap (&heap [0], &heap [*n - 1]);
    int minElement = heap [*n - 1];
    (*n)--;
    heapifydown (heap, *n, 0);
    return minElement;
}

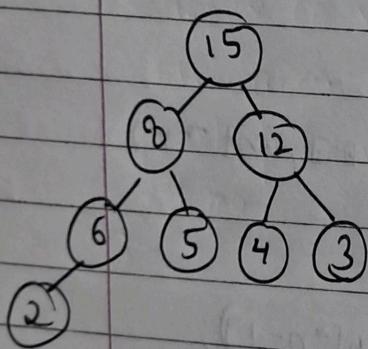
```

```
int main() {  
    int heap[] = {4, 8, 15, 2, 10, 20, 6, 12};  
    int n = sizeof(heap) / sizeof(heap[0]);
```

```
    printf("Min heap before extraction");  
    for (int i = 0; i < n; i++) {  
        printf("%d ", heap[i]);  
    }  
}
```

Time complexity - $O(\log n)$

Q12: Given the following max heap, delete 2
show the heap -



```
#include <iostream>  
#include <vector>
```

```
void swap (int &a, int &b) {  
    int temp = a;  
    a = b;
```

```

b = temp;
}

void heapifyDown (std::vector<int> &heep,
                  int index) {
    int leftchild = 2 * index + 1;
    int rightchild = 2 * index + 2;
    int largest = index;

    if (leftchild < heep.size() && heep[leftchild] >
        heep[largest])
        largest = leftchild;

    if (rightchild < heep.size() && heep[rightchild] >
        heep[largest])
        largest = rightchild;

    if (largest != index) {
        swap (heep[index], heep[largest]);
        heapifyDown (heep, largest);
    }
}

void deleteMax (std::vector<int> &heep) {
    if (heep.empty ()) {
        std::cout << "heep is empty" << std::endl;
        return;
    }
}

```

```
heep[0] = heep.back();
heep.pop_back();
heep.push_back(heep[0]);
}
```

```
void printHeep (const std::vector<int> &sheep) {
    for (int i : heep) {
        std::cout << i << " ";
    }
    std::cout << std::endl;
}
```

```
int main() {
    std::vector<int> maxHeep = {15, 8, 12, 6, 5, 4, 3, 2};
    std::cout << "Max heep before deletion : ";
    printHeep (maxHeep);
    deleteMax (maxHeep);
    std::cout << "Max heep after deletion : ";
    printHeep (maxHeep);
    return 0;
}
```