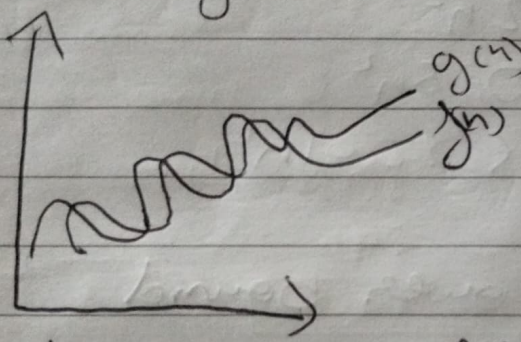
Assignment-1
DAA

⇒ These notations are used to tell the complexity of an algorithm

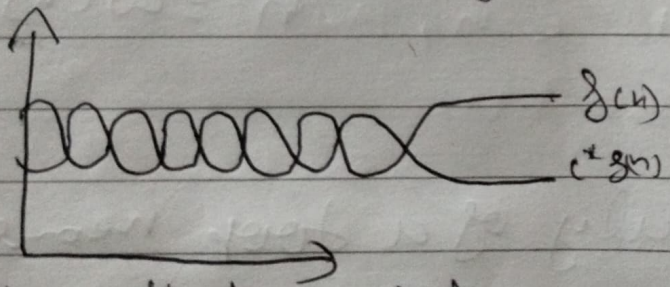
-> It describes the algorithm efficiency & performance in a meaningful way.

1) big oh notation - The function $f(n) = O(g(n))$ if & only if there exists a const $(+ve)$ & $f(n) \leq c\, g(n)$ for all $n, n \geq k$



$f(n) = O(g(n))$
if
$f(n) \leq c\, g(n)$
$\forall\ n \geq n_0$
so, constant $c > 0$

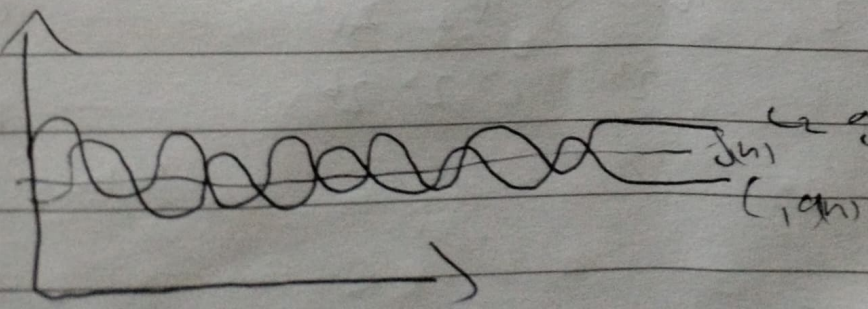
2) big omega notation - The function $f(n) = \Omega(g(n))$ if there exist a +ve constant c & k such that $f(n) \geq c^* g(n)$ for $n, n \geq k$



$f(n) = c\, g(n)$
$f(n) \geq c \cdot g(n)$
$\forall\ n \geq n_0$ & $c > 0$

3) big theta notation

Similarly,

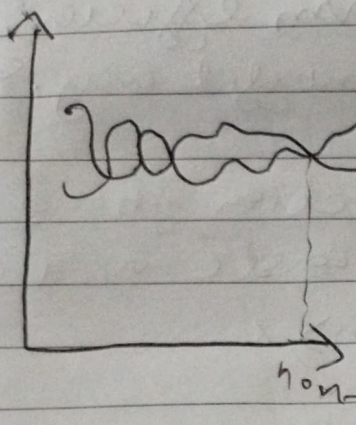


$f(n) = \Theta(g(n))$
if $c_1 g(n) \leq f(n) \leq c_2$
$\forall\ n \geq \max(n_1, n_2$

4) Small o notation

$f(n) = o(g(n))$: $g(n)$ is upper bound of $f(n)$ if & only if

$$f(n) << g(n)$$



$\forall\ n > n_0$ and for all constraints

$c > 0$

$n = o(n^2)$

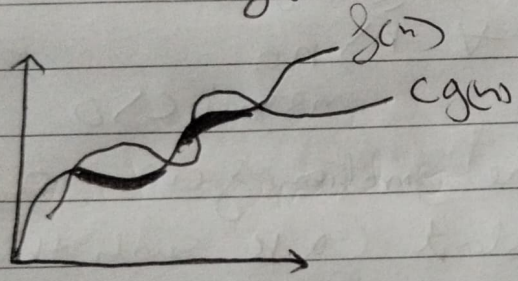
$n < 1n^2$

$2n^2$

$0.5n^2$

$n < 0.001n^2\ n_0$

5) Small omega



lower bound

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

$f(n) > c.g(n)$

$\forall\ n > n_0$ & $\forall c > 0$

$n^2 = \omega(n)$

(q2) for (i = 1 ; i < n ; i = i * 2)

{

}

; time complexity of a loop means the no of time it was to run

i	1	2	4	8	16	32	----	2^k
	2^1	2^2	2^3	2^4	2^5	2^6	----	n

$i = 1, 2, 4, 8, 16, 32, \ldots, 2^k$ this means

k

i.e $2^k = n$

$k \log_2 2 = \log_2 n \Rightarrow$ $k = \log n$

$\log_2 2 = 1$

$TC = O(\log n)$

(Q3) $T(n) = \begin{cases} 3(T(n-1), & n>0 \\ 1 \end{cases}$

by forward substitution

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

$T(0) = 3(T(-1)) = 0$

$T(1) = 3T(0) = 3$

$T(2) = 3T(1)$

$= 3 * 3 = 9$

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

$3T(2) = 3^1 \cdot 3^2 = 3^3$

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

$TC = O(3^n)$

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

by forward substitution

$T(0) = 1$

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

$= 2^1 - 1$

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

$= 2^2 - 2^1 - 1$

$T(3) = 2T(3-1) - 1$

$= 2^3 - 2^2 - 2^1 - 1$

$2^n - 2^{n-1} - 2^{n-2} - 2^{n-3} \ldots - 2^3 - 2^2$

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

$= 2^n - 2^n + 1 = 1$

$TC = O(1)$

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

The value of i increases for each iteration. The value contained in 's' at the ith iteration is the sum of the first i +ve integers. If k is the total no of iterations taken by any prog.

$1+2+3+\cdots+k$

$[k(k+1)/2] > n$

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

$\therefore T.C = O(\sqrt{n})$

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

$O(n)$. T.C

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

TC = log n * log n
= O(n log n)
TC = O(n log n)

Q8)
```
function (int n)
{
    if (n == 1)
    { return n;
    for (i = 1 to n)
    {
        for (j = 1 to n)
        {
            printf ("*");
        }
    }
    function (n - 3);
}
```

T.C = O(n^2)

Q9)
```
void function (int n)
{
    for (i = 1 to n) {
    for (j = 1; j <= n; j = j + 1)
    { O(n)
        Printf ("*");
    }
}
```

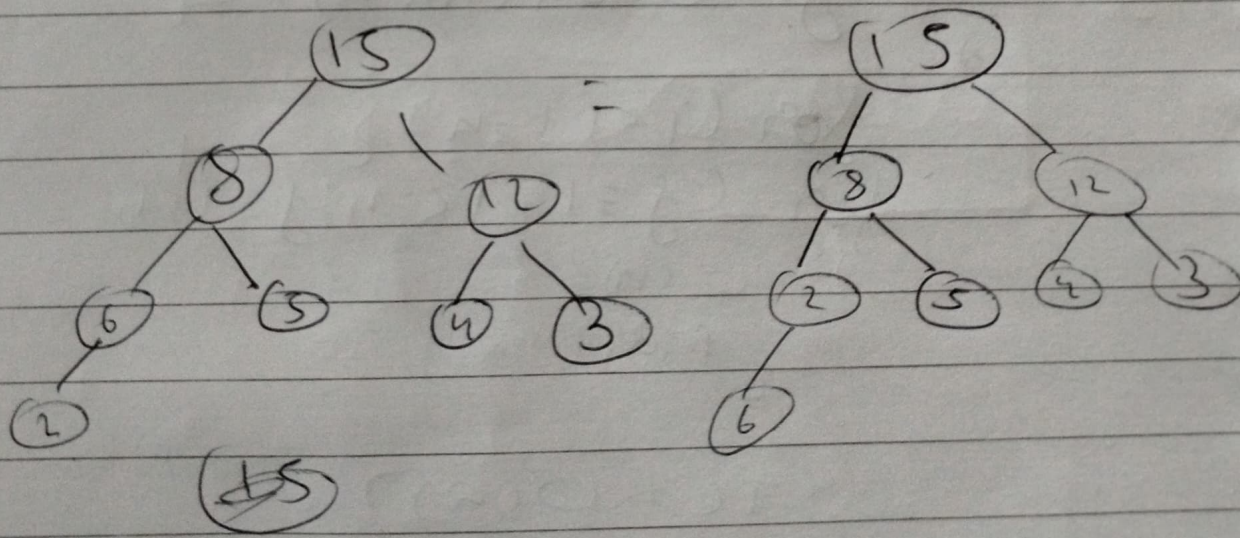
TC = O(n^2)

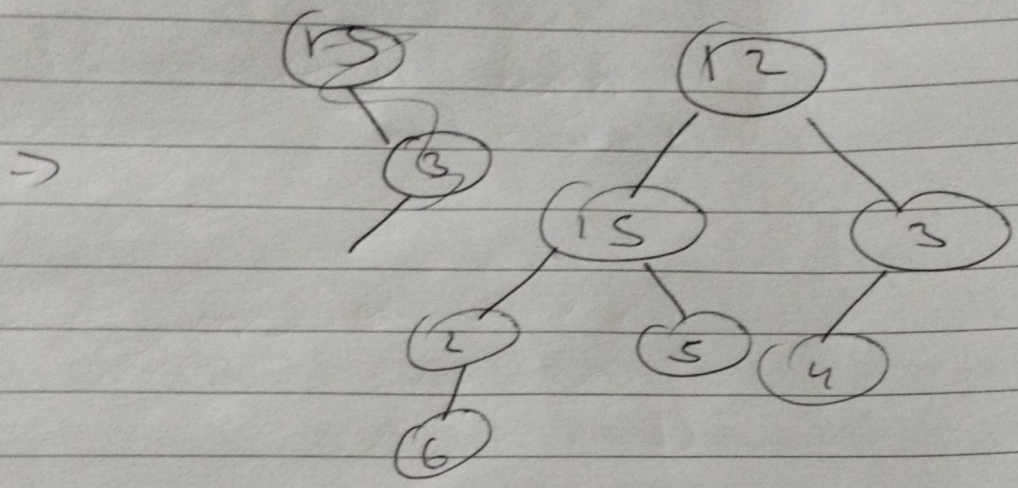
Q10) for the function n^k & c^n, what is asymptotic notations b/w these funcs assume, that $k \geq 1$ & $c > 1$ are const find out the val of c & n which holds n^k is $O(c^n)$

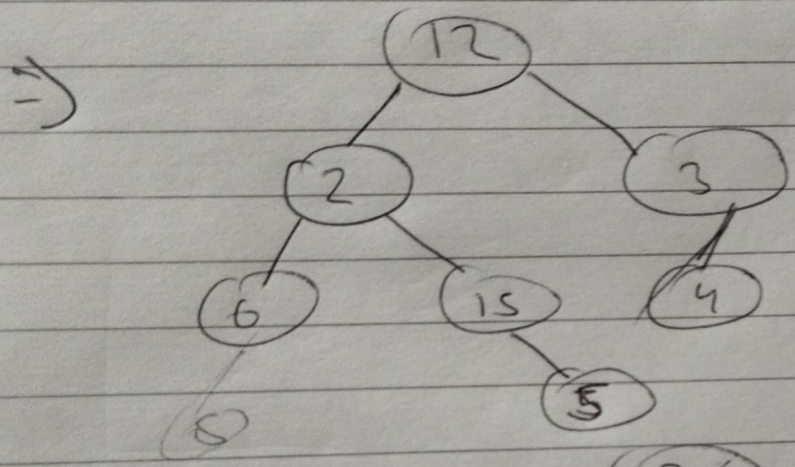
Q11)

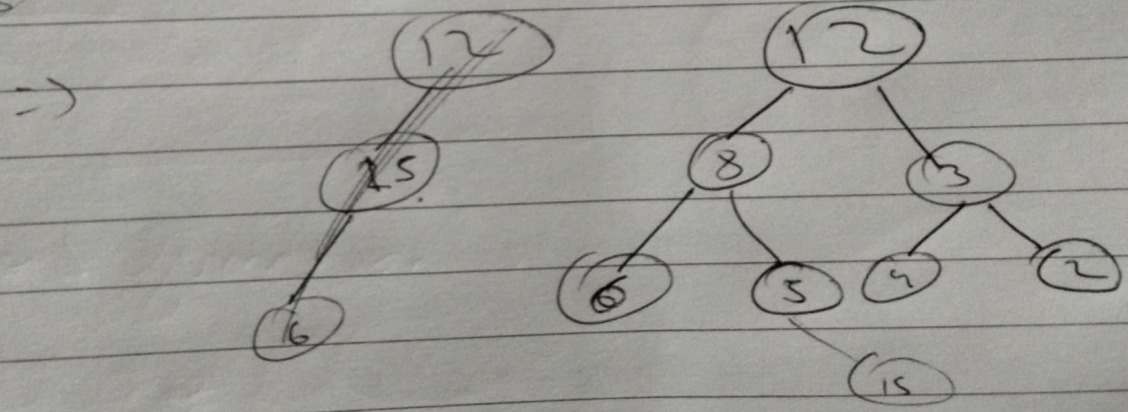
→ The time complexity of 'extractmin()' would depend on the underlying implementation of the data structure. if its a binary heap, the time complexity would be O(1) as the minimum element is always at the top. if its a different data structure the complexity may vary.

Q12) find max heap & del









∴ 15 is not so,