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BTech ISE 5th Sem
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Q

Q1 →

ans. Asymptotic notations are used to write fastest and slowest possible running time for an algorithm. These are also referred to as 'best case' and 'worst case' respectively.

Three types of asymptotic notation to represent the growth of any algo, are →

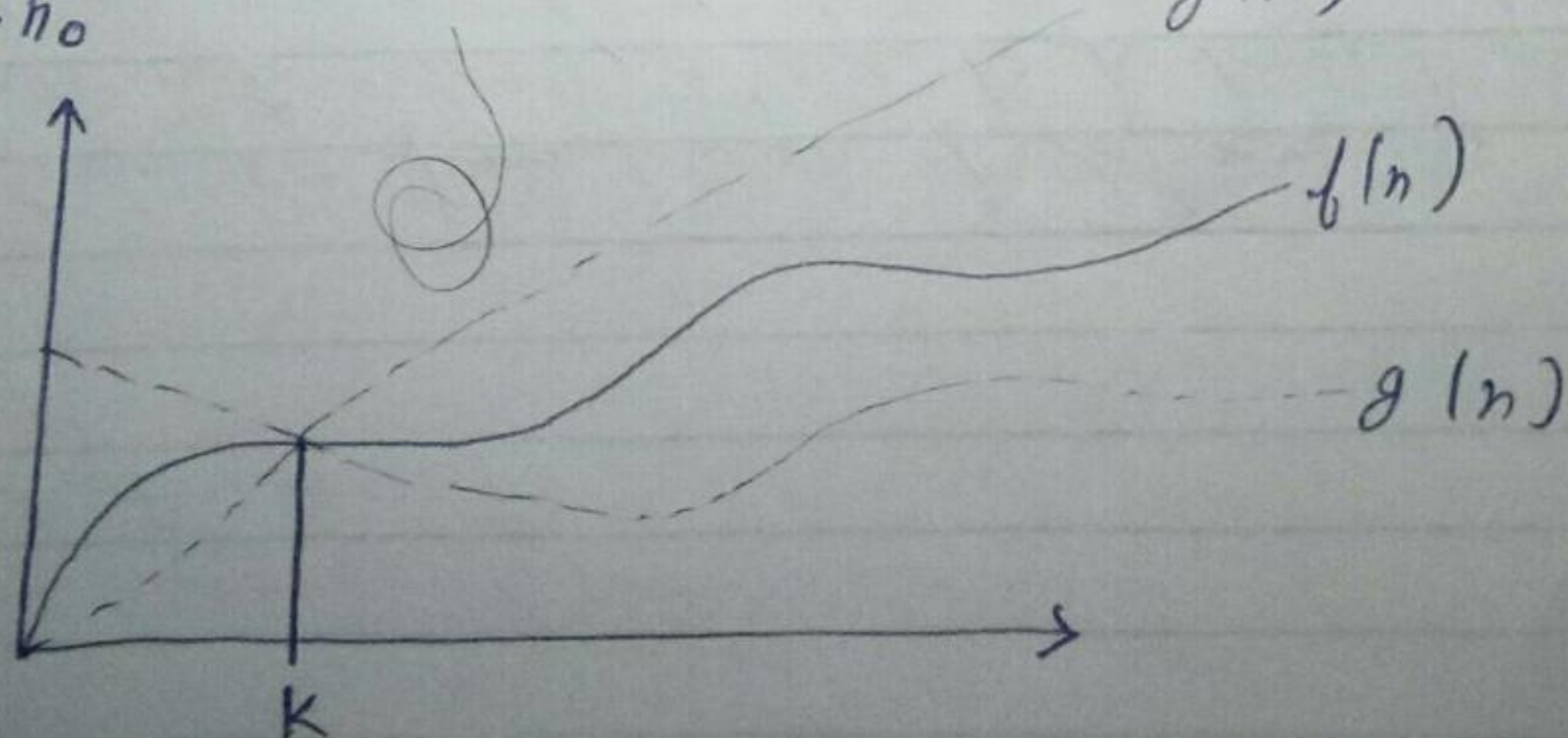
- (i) → Big Theta (Θ)
- (ii) → Big -oh (O)
- (iii) → Big Omega (Ω)

(i) - The time complexity represented by the Big Θ notation is like the average value within which the actual time of execution of the algo will be

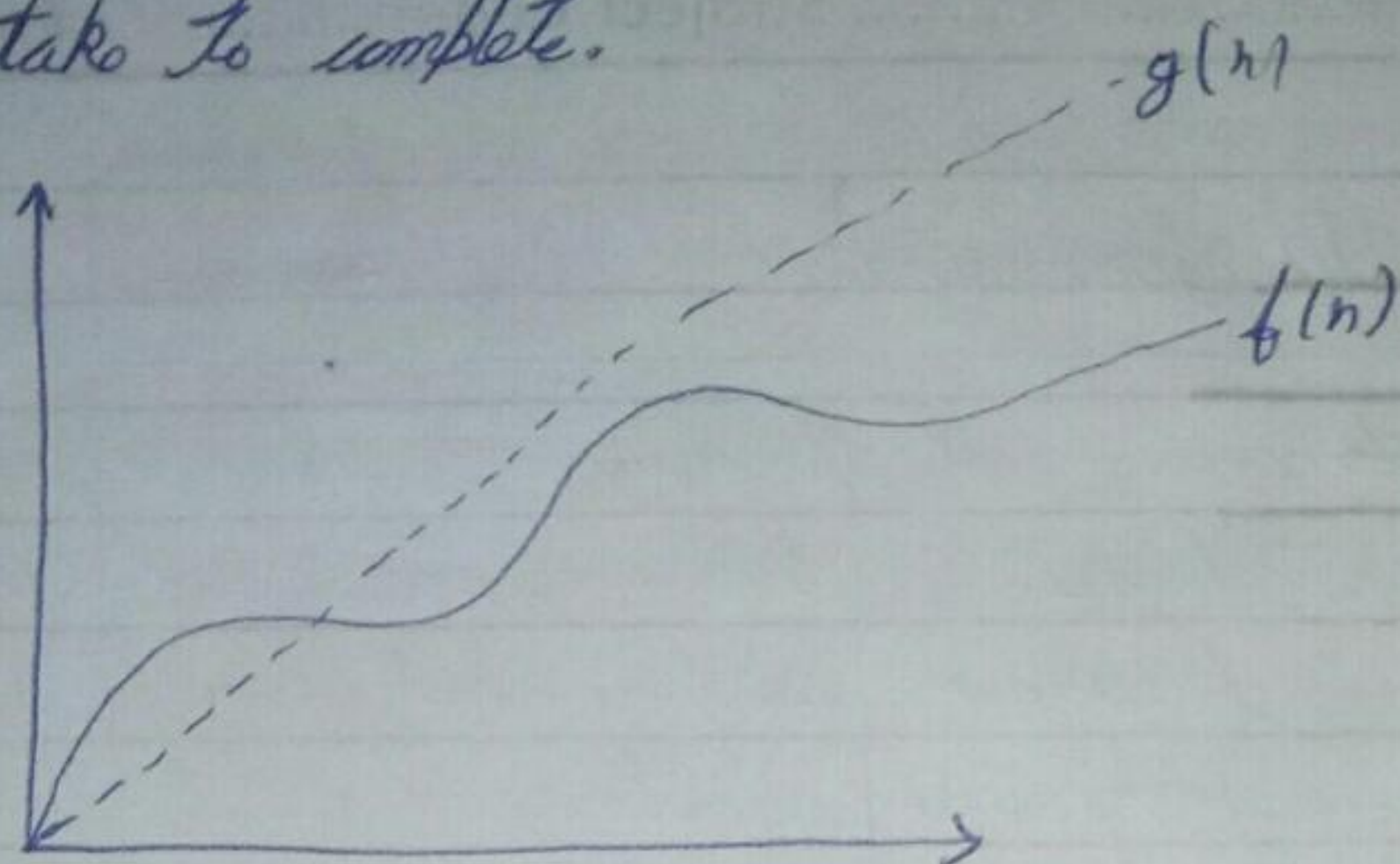
eg. $3n^2 + 5n$

we use the Big Θ notation to represent this, then the time complexity would be $O(n^2)$ ignoring the constant coefficient and remaining insignificant part, which is $5n$.

$\Theta(f(n)) = (g(n))$ if and only if $g(n) = O(f(n))$ and $g(n) = \Omega(f(n))$ for all $n > n_0$



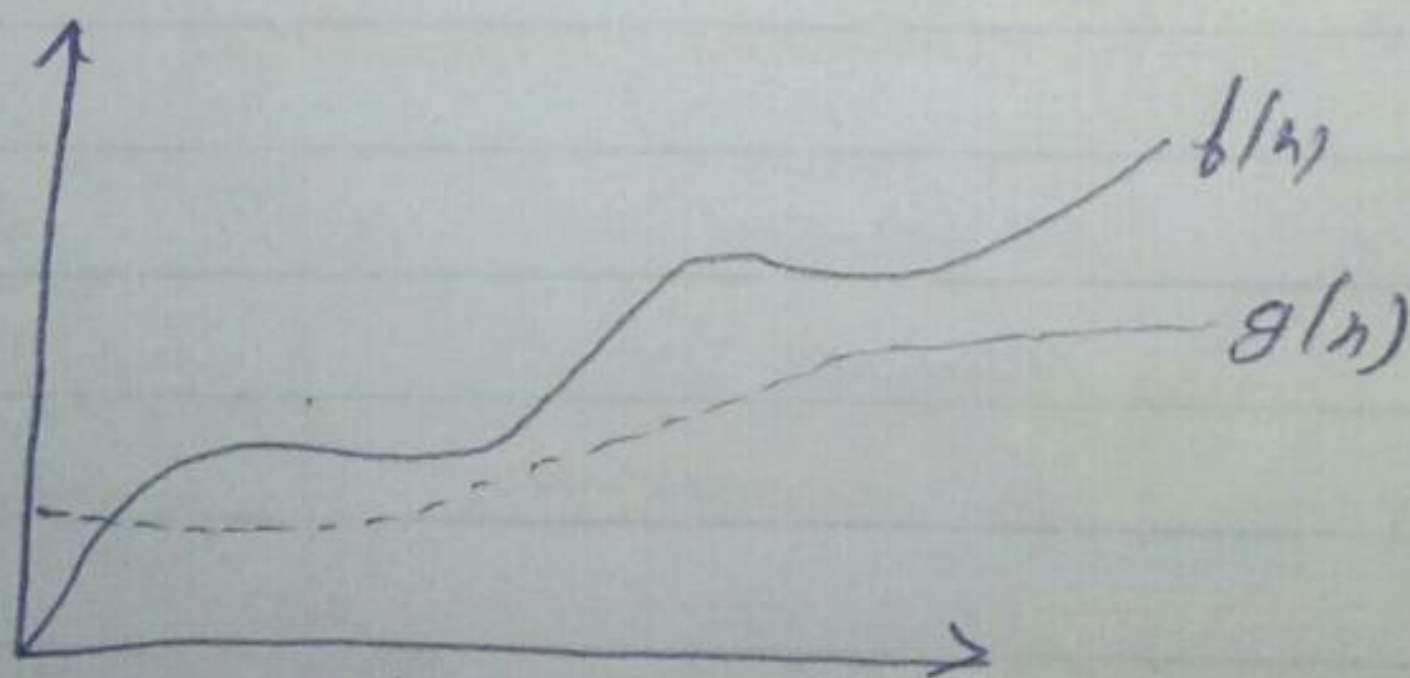
- (i) Big Oh Notation (O)
- It is the formal way to express upper bound of all algorithm running time. It measures the worst case time complexity or the longest amount of time all algo can possibly take to complete.



e.g. $O(f(n)) = \{g(n); \text{there exist } c > 0 \text{ and } n_0 \text{ such that } f(n) \leq c \cdot g(n) \text{ for all } n > n_0\}$

(ii) Omega notation (Ω)

→ it is the formal way to express the lower bound of an algo.'s running time. It measures the best case time an algo can possibly take to complete.



$\Omega(f(n)) = \{g(n); \text{there exist } c > 0 \text{ and } n_0 \text{ such that } g(n) < c \cdot f(n) \text{ for all } n > n_0\}$

2.

Sol. for $(i=1; i \leq n; i=i*2)$

1, 2, 3, 4, 3, ...

$$T(n) = O(\log_2 n)$$

3.

Sol. $T(n) = 3T(n-1) - \text{①}$

put $n=n-2$ in eqn ① we get

$$T(n-1) = 3T(n-2) - \text{②}$$

put value of $T(n-1)$ in eqn ①

$$T(n) = 3\{3T(n-2)\}$$

$$T(n) = 3^2 T(n-2) - \text{③}$$

put value of $n=n-2$ in eqn ① we get

$$T(n-1) = 3T(n-3) - \text{④}$$

put the value

② \Rightarrow for $(i=1 \text{ to } n) \{i=i*2\}$

series will be

1, 2, 4, 8, 16, 32, 64, ... n

k term

so using ap $a_n = a + r^{k-1}$

$$a_n = n, a = 1, r = 2$$

$$n = 1 \cdot 2^{k-1}$$

$$n = \frac{2^k}{2} \Rightarrow 2n = 2^k$$

taking log base 2 on both side

$$\log_2 2n = \log_2 2^k$$

$$\log_2 2 + \log_2 n = k \log_2 2$$

$$1 + \log_2 n = k$$

$$k = T = O(\log_2 n) = \log n$$

3.
sol.

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

$$T(n) = 3T(n-1) \quad T(0) = 2$$

using substitution

$$T(n) = 3T(n-1) \rightarrow T(n-1) = 3T(n-2)$$

$$T(n) = 3(3T(n-2)) \rightarrow T(n-2) = 3T(n-3)$$

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

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

for k terms

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

$$\text{let } T(n-k) = T(0)$$

$$n = k$$

Using $n = k$

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

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

4 \Rightarrow

sol. $T(n) = 2T(n-1) - 1 \quad T(0) = 1$

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

$$T(n) = 2(2T(n-2) - 1) - 1 \rightarrow T(n-2) = 2T(n-3) - 1$$

$$T(n) = 2[2(2T(n-3) - 1) - 1] - 1$$

$$T(n) = 2[4T(n-3) - 2] - 1 = 8T(n-3) - 7$$

for k^{th} term

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

$$\text{let } T(n-k) = T(0)$$

$$n = k$$

using $k = n$

$$T(n) = 2^n (T(0)) - (2^n - 1)$$

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

$$T(n) = 1$$

5 →

sol →

$i=1, S=1;$
while ($S \leq n$)

{ $i++; S=S+i;$ }

Here no. of steps

1
2
3
4
5

i
1
2
3
4
5

S
1
3
6
10
15

so order of growth is not const. so general term can be taken as $\frac{k(k+1)}{2}$ let at n no. of terms is k so $n = \frac{k^2+k}{2} \Rightarrow 2n = k^2+k$

ignoring lower order terms

$$n = k^2$$

$$k = \sqrt{n}$$

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

6 → for ($i=1; i*i \leq n; i++$)
count++

so i is moving from 1 to \sqrt{n} with linear growth so

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

7 → for ($i=n/2; i \leq n; i++$)

for ($j=1; j \leq n; j=j*2$)

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

count++

k and j loop are moving from 1 to n with exponential growth so T(

for these two loop will be $O(\log n)$

and i loop going through $n/2$ to n so with constant argument

$$\text{so } T(n) = O(n/2) = O(n)$$

so overall complexity of the fⁿ will be $T(n) = O(n \cdot \log n \cdot \log n)$

$$8 \rightarrow T(n) = T(n-3) + n^2$$

$$T(1) = 1$$

$$T(n) = T(n-3) + n^2 \rightarrow T(n-6) + (n-3)^2$$

$$T(n) = T(n-6) + (n-3)^2 + n^2 \rightarrow T(n-9) + (n-6)^2$$

$$T(n) = T(n-9) + (n-6)^2 + (n-3)^2 + n^2 \rightarrow T(n-12) + (n-9)^2$$

$$T(n) = T(n-12) + (n-9)^2 + (n-6)^2 + (n-3)^2 + n^2 \text{ so for } k^{\text{th}} \text{ term}$$

$$T(n) = T(n-k) + (n-(k-3))^2 + (n-(k-6))^2 + (n-(k-9))^2 + (n-(k+2))^2 + \dots + (n-(k-k))^2$$

$$\text{let } T(n-k) = T(1)$$

$$n = 1 + k = 1 + (n-1)$$

$$T(n) = T(n-(n-1)) + [n-(n-1-3)]^2 + [n-(n-1-6)]^2 + [n-(n-1-9)]^2 + \dots + n^2$$

$$T(n) = T(1) + 4^2 + 7^2 + 10^2 + \dots + n^2$$

$$T(n) = 1 + 4^2 + 7^2 + 10^2 + \dots \rightarrow n^2$$

$$\Rightarrow n = (3k-2)^2$$

$$n = 9k^2 + 4 - 12k$$

$$n = k^2$$

$$k = \sqrt{n}$$

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

9 → for $i=1$ to n

for $j=i; j \leq n; j=j+i$

print("*")

steps

i 1 2 3 4 ... n

j n n/2 n/3 n/4 ... n/n=1

so total complexity is

$$T(n) = n + n/2 + n/3 + n/4 + n/5 + \dots + n/n$$

$$\Rightarrow n[1 + 1/2 + 1/3 + 1/4 + 1/5 + \dots + 1/n]$$

$$\Rightarrow n \int_1^n \frac{1}{x} dx$$

$$\Rightarrow n[\log x]_1^n = n[\log n - \log 1] = n \log n$$

10 → if $c > 1$, then the exponential c^n for outgrows any term, so the ans is n^k is $O(c^n)$

11 → int $j=1; j=0;$

while $(i < n)$

{ $i=i+j;$

$j=j+i;$

}

i = 0 1 2 3 6 10 15 ...

j = 1 2 3 4 5 6 ...

so i will go on till n and general formula for kth term is

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

$$\text{so } T(n) = O(\sqrt{n})$$

12.

Sol.

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

$$\text{let } T(n-1) = T(n-2)$$

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

$$T(n) = 2T(n-1) + C \rightarrow T(n-1) = 2T(n-2) + C$$

$$T(n) = 2[2T(n-2) + C] + C = 4T(n-2) + 3C \rightarrow T(n-2) = 2T(n-3) + C$$

$$T(n) = 4[2T(n-3) + C] + 3C = 8T(n-3) + 7C$$

so on k^{th} term will be

$$T(n) = 2^k T(n-k) + (2^k - 1)C$$

and $T(n-k)$ for $k =$ but term n will be 1 for $n=1$

$$2^n \cdot 1 + (2^n - 1)C = T(n)$$

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

neglecting the constants

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

and max depth for recursion of n is based on n so space complexity is $O(n)$

13.

Sol. $n(\log n)$

```
for (int i=0; i<n; i++)
```

```
for (int j=0; j<n; j*=2)
```

```
print ("* ");
```

n^3

```
for (int i=0; i<n; i++)
```

```
for (int j=0; j<n; j++)
```

```
for (int k=0; k<n; k++)
```

```
print ("* ");
```

$$T(n) = T(n/2) + T(n/2) + Cn^2$$

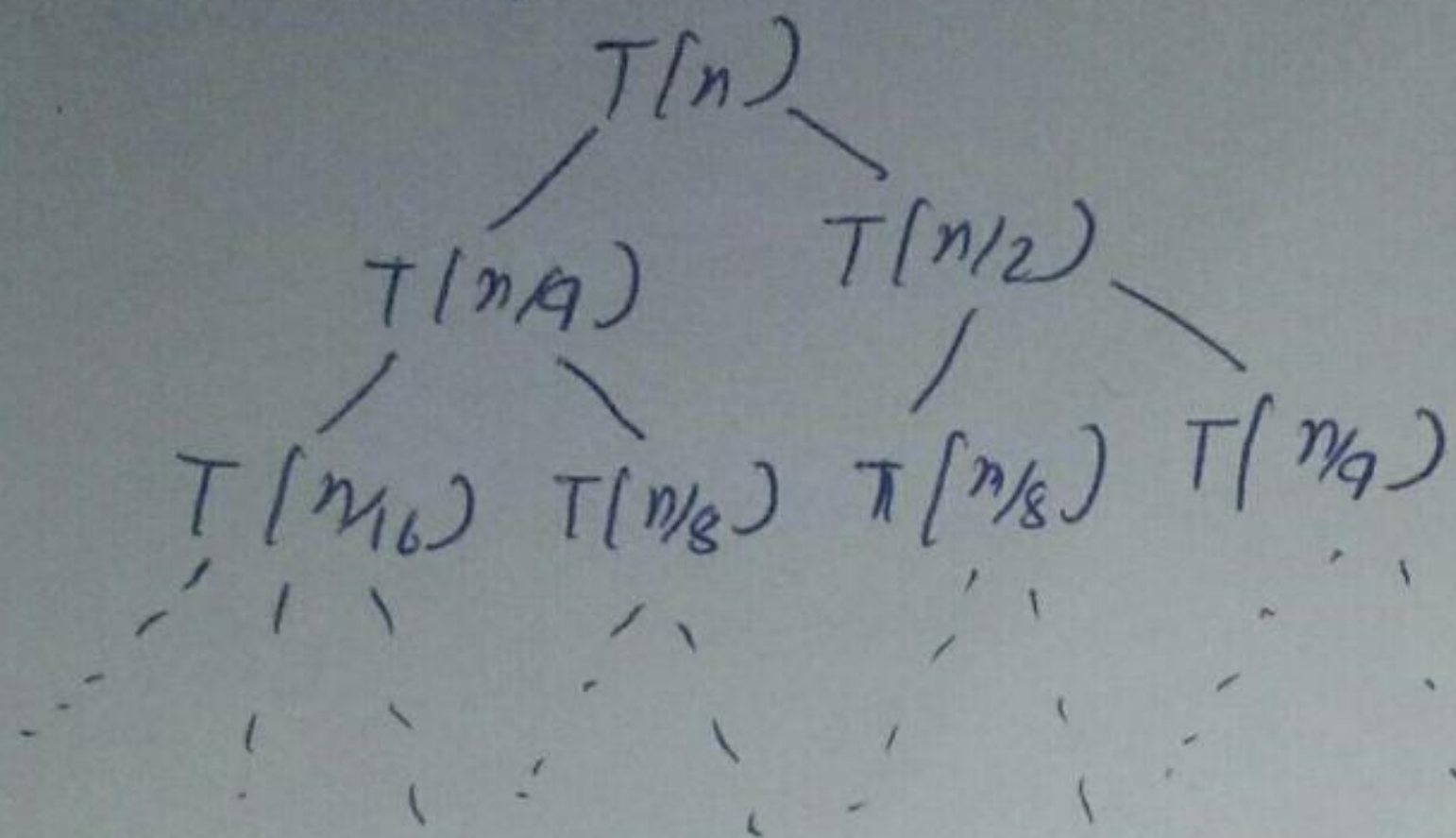
$$\log(\log n)$$

```
for (int i=0; i<n; i=i+1)
```

```
print ("* ");
```


19+

2nd- $T(n) = T(n/4) + T(n/2) + n^2$



$$T(n) = [n^2 + 5n^2/16 + 75n^2/256 + \dots]$$

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

15+ for (int i=1; i<=n; i++)
 for (int j=1; j<=n; j+=1)
 O(1);

i = 1 2 3 4 ... n
 j = n n/2 n/3 n/4 ... n/3

time complexity

$$T(n) = n + n/2 + n/3 + n/4 + \dots + n/n$$

$$T(n) = n \left[1 + 1/2 + 1/3 + 1/4 + \dots + 1/n \right] = n \int_1^n \frac{1}{x} dx = n [\log n - \log 1] = n \log n$$

16- for (int i=2; i<=n; i=floor(i/2))
 O(1)

for any n increasing with exponential rate the T become
 $T = O(\log(\log n))$

18- a+ $100 < \log \log n < \log n < n \log n < \text{root}(n) < n < n^2 < 2^n < 2^{2^n} < 4^n$
 $\log(n!) < n!$

b- $1 < \log(\log n) < \sqrt{\log(n)} < \log n < \log 2n < 2 \log n < n \log n < 2(2^n) < \log n! < n!$

c- $9 < \log_8(n) < \log_2(n) < n \log_6(n) < n \log_2(n) < 5n < 8n^2 < 7n^3 < 8^{2n} < \log(n!) < n!$

19.

```

2.1. Search(a, n)
    low = 0, high = n, key
    while (low < high)
        mid = (low + high) / 2
        if a[mid] == key
            return mid
        else if a[mid] < key
            low = mid + 1
        else high = mid + 1
  
```

20. complexity

$O(n^2)$
 $O(n^2)$
 $O(n^2)$
 $O(n \log n)$
 $O(n \log n)$
 $O(n \log n)$

also
 Bubble sort
 selection "
 Insertion "
 merge "
 Heap "
 Quick "

21. Sorting algo

Bubble

Selection

Insertion

Quick

merge

Heap

Inplace

Yes

Yes

Yes

Yes

no

Yes

Stable

Yes

no

Yes

no

Yes

no

Online

no

no

Yes

no

no

no

23-

```

Search (a, n, Key)
low = 0, high = n
while (low < high)
    mid =  $\frac{low + high}{2}$ 
    if a[mid] is Key
        return mid
    else if a[mid] < Key
        low = mid + 1
    else high = mid - 1

```

Recursion

```

- Search (a, low, high, Key)
    if low < high
        mid =  $\frac{low + high}{2}$ 
        if a[mid] < Key
            Search (a, mid + 1, high, Key)
        else if a[mid] > Key
            Search (a, low, mid - 1, Key)
        else return mid

```

Linear search	$T = O(n)$	$S = \text{constant}$
Binary "	$T = O(\log n)$	$S = 1$

24 - $T(n) = T(n/2) + C$