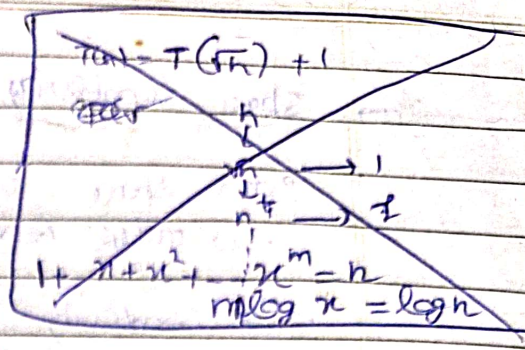


⇒ Analysis of Algorithms

An algo. Depends on
 (i) Input Size
 (ii) Input order



So, we use Asymptotic Analysis

It evaluates the performance (time & space) of an algo. in terms of input size. How does the time taken increase by input size increase.

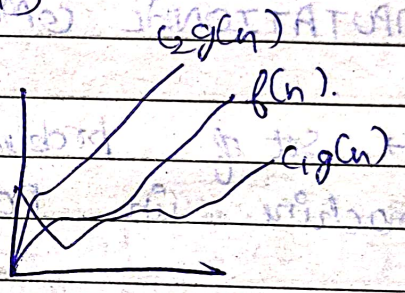
Three cases

- (i) Best case → Lower Bound
- (ii) Worst case → Upper Bound
- (iii) Average case → All possible inputs & calculate computing time divided by sum of total no. of inputs.

$$= \frac{\sum_{i=1}^{n+1} \theta(i)}{(n+1)} \quad \text{all possible inputs}$$

Three Notations:

- (i) θ Notation Both
- (ii) O Notation $g(n)$
- (iii) Ω Notation $c, g(n)$
- (iv) little o strict upper bound
- (v) little ω strict lower bound



→ Amortized Analysis:

Used for algorithms where an occasional operation is very slow, but most of other operations are faster.

e.g. Dynamic Table (or Arrays)

Amortized cost for n operations = $\frac{(1+1+\dots+1n) + (1+2+4+\dots)}{n}$

$$= \frac{n + 2n}{n} < 3$$

→ Space Complexity & Auxiliary Space :

Extra space or temporary space used by an algorithm, is called Auxiliary space.

Space complexity is total space taken by the algorithm w.r.t. input size. It includes both auxiliary space & space used by input. e.g. Insertion sort have $O(n)$ space complexity but take $O(1)$ auxiliary space.

→ Pseudo-polynomial :

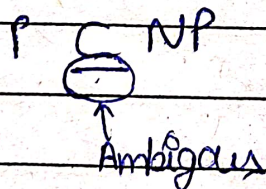
An algorithm whose worst case time complexity depends on numeric value of input (not no. of inputs) is called Pseudo-polynomial algorithm.

e.g. algorithm depending on max. value of inputs.

⇒ COMPUTATIONAL COMPLEXITY :

P → Set of problems solvable by deterministic Turing machine in Polynomial time.

NP → Set of decision problems solvable by Non-Deterministic Turing Machine in Polynomial time.



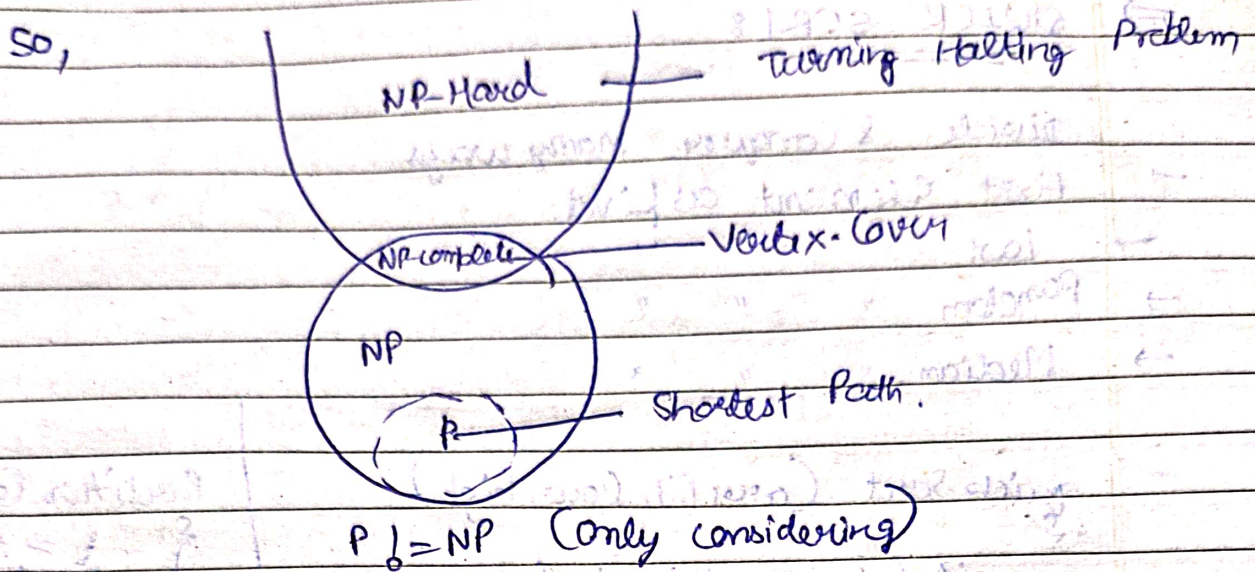
"Lucky" Algorithm which always makes a right choice.

NP Complete, A decision problem L is NP-Complete if :

→ L is in NP.

→ Every problem in NP is reducible to L in polynomial time.

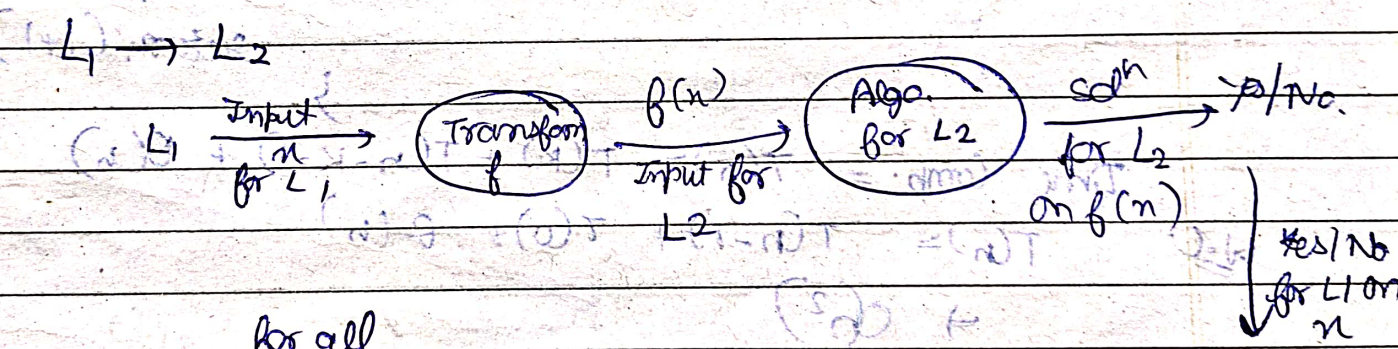
NP-Hard



Notes Decision problems vs Optimization Problems:

- Note 1: Decision \approx Optimization.
- Note 2: Decision is easier than Optimization.

REDUCTION:



for all $x \in L_1$, $f(x) \in L_2$ and vice versa.
 NP-Complete Set

Reverse also apply.

1st NP-Complete problem was SAT (Boolean Satisfiability Problem)

$$\Rightarrow C_N = N+1 + 2 \sum_{k=1}^{N-1} C_{k-1}$$

$$N C_N = N(N+1) + 2 \sum_{k=1}^N C_{k-1} \quad \text{--- (1)}$$

$$(N-1) C_{N-1} = (N-1)N + 2 \sum_{k=1}^{N-1} C_{k-1} \quad \text{--- (2)}$$

Subtracting (2) from (1)

$$N C_N = 2N + 2 C_{N-1} + N C_{N-1} - C_{N-1}$$

$$N C_N = (N+1) C_{N-1} + 2N$$

$$\frac{C_N}{(N+1)} = \frac{C_{N-1}}{N} + \frac{2}{(N+1)}$$

$$= \frac{C_{N-2}}{N-1} + \frac{2}{N} + \frac{2}{N+1}$$

$$\frac{C_N}{N+1} = \frac{C_2}{3} + \sum_{k=3}^N \frac{2}{k+1}$$

$$\frac{C_N}{(N+1)} \approx \int_3^N \frac{2}{k+1} \cdot dk$$

$$\boxed{C_N \approx (N+1) \ln N}$$

3-WAY QUICK SORT / DUTCH NATIONAL FLAG

while (mid <= high)

if (a[mid] < pivot) swap (&a[low++], &a[mid++]);
 else if (a[mid] == pivot) mid++;
 else swap (&a[mid], &a[high--]);

return {low-1, mid, high}

→ Merge sort for linked list:

→ Merge sort:

```
mergeSort(0, n-1) {  
    if (low < high) {  
        int mid = (l+h)/2  
        mergeSort(l, mid)  
        mergeSort(mid+1, h)  
        merge(l, mid, h)  
    }  
}
```

```
merge {  
    arr1 = l, mid  
    arr2 = mid+1, high  
    mergeThem  
}
```

n log n for all



Insertion Sort:



```

for (i=0; i<n; i++) {
    int num = a[i];
    int j = i-1;
    while (j >= 0 && a[j] > num) {
        swap(a[j], a[j+1]);
        j--;
    }
}

```

$O(n^2)$ → Worst
 $O(n)$ → Best
 $O(n^2)$ → Average



Selection Sort:

```

Select min.
for (i=0; i<n; i++) {
    int mini = a[i];
    for (j=i+1; j<n; j++) {
        if (a[j] < mini)
            mini = a[j];
    }
    swap(mini_ind, a[i]);
}

```

$O(n^2)$ → Worst
 $O(n)$ → Best
 $O(n^2)$ → Average



Bubble Sort:

```

for (i=0; i<n; i++) {
    for (j=1; j<n; j++) {
        if (a[j-1] > a[j])
            swap
    }
}

```

$O(n^2)$ → Worst
 $O(n)$ → Best
 $O(n^2)$ → Average



Heap DS:

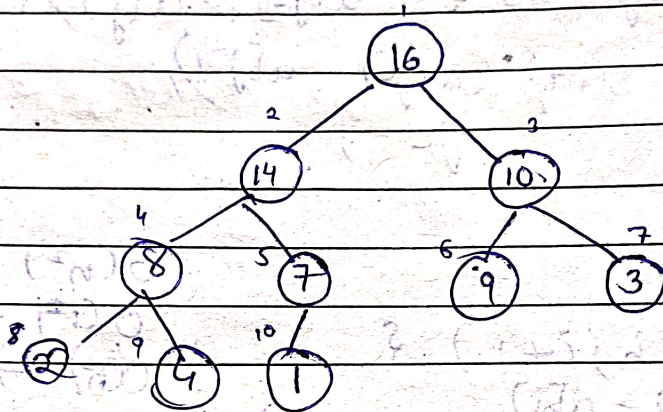
Tree-based DS in which tree is complete binary tree.

Max-Heap Min-Heap

An array visualised ^{heap} complete binary tree.

e.g.

1	2	3	4	5	6	7	8	9	10
16	14	10	8	7	9	3	2	4	1



root \rightarrow first element ($i=1$)

Parent (i) = $i/2$

Left (i) = $2i$

Right (i) = $2i+1$

~~For min-Heap~~

```

void heapify(int i) {
    int l = left(i);
    int r = right(i);
    int small = i;
    if (l < n && a[l] > a[i])
        small = l;
    if (r < n && a[r] > a[i])
        small = r;
    if (small != i) {
        swap(a[i], a[small]);
        heapify(small);
    }
}

```

Other operations like insert, delete can be done by a while loop only.

Heap Sort

Heapify (make heap)
for each N
heapify & extract

→ Time Complexity of heap build:

For a height h with n nodes in Binary tree

$$\text{at most} = \left\lceil \frac{n}{2^{h+1}} \right\rceil$$

$$\text{So, } T(n) = \sum_{h=0}^{\lg n} \left\lceil \frac{n}{2^{h+1}} \right\rceil \cdot O(h)$$

$$= O\left(n * \sum_{h=0}^{\lg n} \frac{h}{2^h}\right)$$

$$= O\left(n * \sum_{h=0}^{\infty} \frac{h}{2^h}\right)$$

$$= O(n * 2)$$

$$= O(n)$$

$$\sum_{h=0}^{\infty} x^h = \frac{1}{1-x}, \quad |x| < 1$$

$$\sum_{h=0}^{\infty} h x^{h-1} = \frac{1}{(1-x)^2}$$

$$\sum_{h=0}^{\infty} h x^h = \frac{x}{(1-x)^2}$$

replacing x by $\frac{1}{2}$ & n by h

$$\sum_{h=0}^{\infty} h \frac{1}{2^h} = \frac{\frac{1}{2}}{\left(\frac{1}{2}\right)^2} = \frac{1}{\frac{1}{2}} = 2$$

⇒ Linked List: Before studying any of it, let's study pointers.

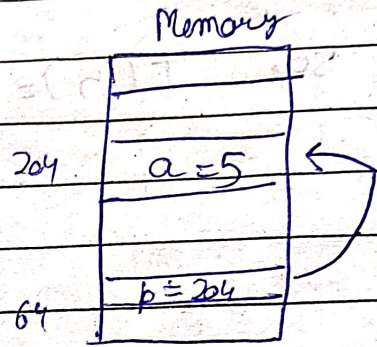
POINTERS (size of pointer depends on architecture)

8 byte → 64 bit
2 byte → 16 bit

⇒ Variables that store address of other variables.

```
int a;
int *p;
char c; char *pc;
```

```
p = &a
*p = 8
a // 8.
```



⇒ Pointer Arithmetic

```
int *p = &a // 2002
```

```
(p+1) // 2006
```

Now, it gets "dangerous"

(p-9) Also possible

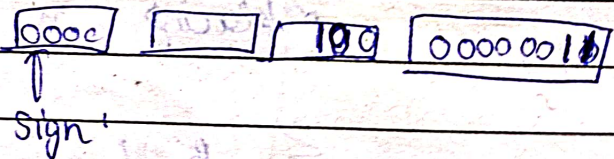
```
(p-1) 1998
```

→ Why we need strongly typed pointers?

Ans Due to ^{dc} referencing

Now, let's see a mind blowing trick:

```
int a = 1027
int *p = &a;
```



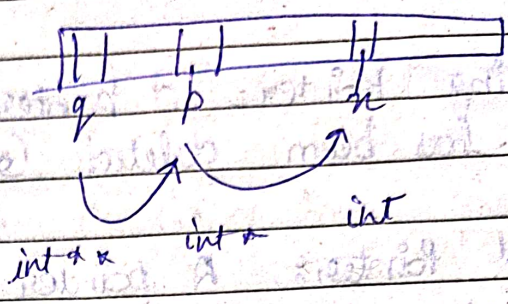
```
p // 2002
*p // 1027
```

```
char *p0 = (char*) p;
```

```
p0 // 2002
*p0 // 3
```

⇒ Pointer to Pointer:

```
int n = 5;
int * p = &n;
int ** q;
q = &p;
```



⇒ Pointer & Arrays:

```
A = &A[0]
Address - &A[i] or (A+i)
Value - A[i] or *(A+i)
```

⇒ Pointer & String (Char array)

```
char * s = "Abcd";
char * s = (char *) "Abc";
(++)
```

Array as Function

```
int A[] ⇔ int * A
```

⇒ 2-D array in function

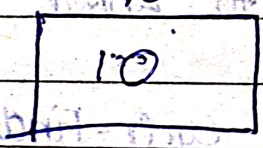
```
(int * a)
*(a + i * m) + h
```

⇒ Pointer & Multi-dimensional Arrays

$$B[i][j] = *(B[i] + j) = **(&(B+i) + j)$$

⇒ Reference Variables in C++ (Alias)

```
int n = 10;
int &y = n
```



Reference Variable
(Internal pointer)

- * Must be initialized with variable.
- * Alternative name for existing variable

⇒ Types of pointers

- (i) Dangling Pointer: A pointer pointing to a memory location that has been deleted (or freed)
- (ii) Void pointer: A pointer pointing to some data location in storage, which doesn't have any specific type.
- (iii) Null Pointer: A pointer pointing to nothing
- (iv) Wild Pointer: A pointer which has not been initialized to anything (not even NULL).

Beware of Wild Pointer.

⇒ Dynamic Memory Allocation:

Allocated on Heap just like global or static variables.

```
int *p = new int
```

```
int *p = new int [10];
```

```
delete p;
```

```
delete [] p;
```

Back to Linked List:

(i) Floyd's Cycle-Finding Algorithm

Move one pointer by one then by two, if they are same at a moment return true.

* Try everything via two pointers, three pointers
done
you are pro!

- Detect loop & Remove (detect loop + get length)
- Reverse a linked list
- Swap two nodes

→ DMA

~~C~~ int *p = (int *) malloc (size of ^{required} ~~array~~)
returns void pointer

(int *) calloc (No. of blocks, ^{byte} size)

contiguous allocation

(int *) realloc (p, new size);
↓
address

free (p);

~~C++~~

new int;

new int [10];

delete p;

delete [] p;

⇒ No. of binary trees with n nodes

Unlabeled = n^{th} catalan number

$$= \sum_{i=1}^n C(n-i) C(i-1) = \text{No. of BSTs}$$

labeled = n^{th} catalan number $\times (n)!$

* Also n^{th} catalan number = $\frac{(2n)!}{(n+1)! n!}$

(Inorder/Preorder/Postorder)

(Level order)

DFS vs BFS in binary tree

	Extra space	Extra space
Height of BT	$O(h)$	$O(w)$
	$O(h)$	$O(n)$
		width of BT

	Normal	Iterative
Worst	$O(h)$	$O(n)$

Preorder	Root	L	R
Inorder	L	Root	R
Postorder	L	R	Root

Iterative

right left

Print, ~~right~~, ~~left~~

If tmp → right push, and go to leftmost

push tmp → right & tmp.

⇒ Binary Search Tree :

* Inorder traversals produce sorted output
* We can create BST from preorder, postorder or inorder

Das

Range Search

Nearest Neighbours

Insert

Delete

⇒ Delete in BST

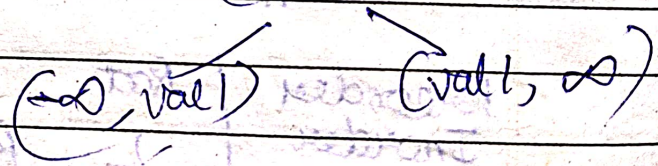
→ If leaf delete

→ If one child, assume it was never there

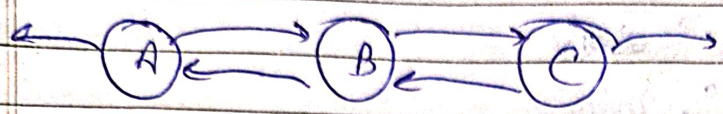
→ If two child, swap with its inorder successor and delete current one

⇒ Check if B.T. is BST :

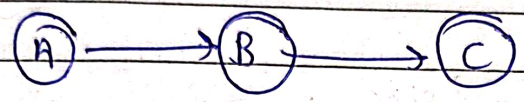
Given each node a range $(-\infty, \infty)$



⇒ Doubly linked list



with single pointer



A next contains XOR of NULL & address of B
B next contains " " address of A & " " C

⇒ Hashing:

Array, LL, Heaps, BSTs are not giving best search insert & delete.

Hash gives (or) Direct access table

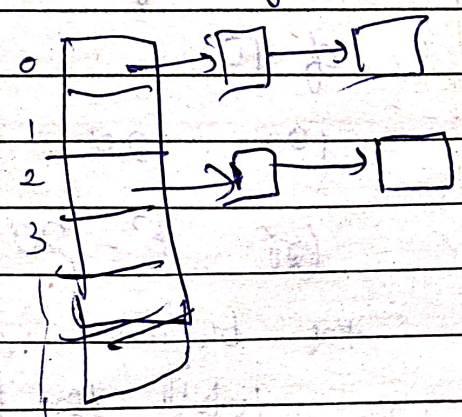
↓ Requires Huge Size

Converts into smaller index (using hash funcⁿ)

- ↳ efficiently computable
- ↳ uniformly distribute keys.

Collisions may occur

- ↳ Use ^{separate} chaining
- ↳ Open addressing



||) separate chaining

1. Linear probing

$$\text{index} = \text{hash} \% \text{tableSize}$$

$$\text{index} = (\text{hash} + 1) \% \text{tableSize}$$

|

2. Quadratic Probing

$$\text{index} = \text{hash} \% \text{tableSize}$$

$$\text{index} = (\text{index} + 1^2) \% \text{tableSize}$$

$$\text{index} = (\text{index} + 2^2) \% \text{tableSize}$$

3. Double Hashing

$$\text{index} = \text{hash} \% \text{tableSize}$$

$$\text{index} = (\text{hash} + 1 * \text{index} * H) \% \text{tableSize}$$

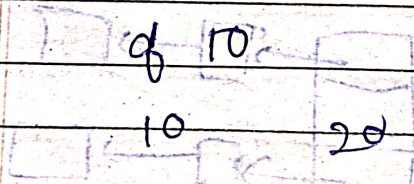
2d

4. Random Probing:

Notes

Prime No. in modulus

→ B/c if data is some pattern based e.g. multiple



for $P = 20$

the buckets will be

10 & 0 only

while for $P = 7$

3, 6, 2, 5, 1

BIT Manipulation:

$n \& (n-1) =$ ~~removes~~ ^{removes} rightmost 1.

i) Power of 2:

$$n \& (n \& (n-1))$$

It needs to print

ii) Count No. of ones:

```
while (n) {  
    n = n & (n-1);  
    cnt++;  
}
```

iii) generate all subsets of A's

```
for (i=0; i < (2^N); i++) {
```

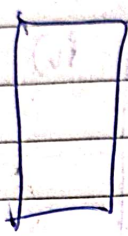
```
    for (j=0; j < N; j++) {
```

```
        if (i & (1 << j))  
            print A[j];  
    }
```

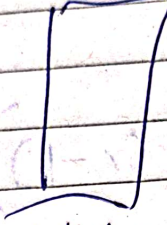
```
    }
```

→ Advance Tree Algos:

I. Reverse Level Order



Queue

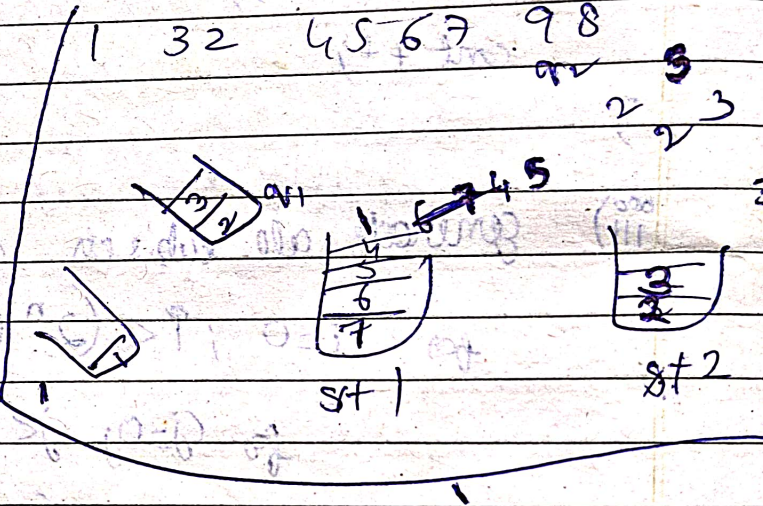
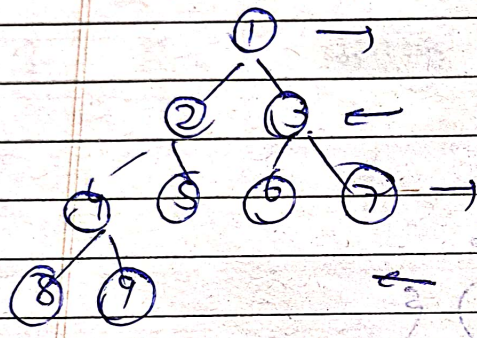


stack

(root right
root left)

(root)

II. Spiral Level order



Queue

III. LCA in BST

if left < root < data & right > root > data
return root.

IV. LCA in BT:

if (curr == n && curr == y)
return curr;
curr == n || curr == y
return |

sd1 = curr -> left
sd2 = curr -> right
if either of 2 is null
return curr