

# Design and Analysis of Algorithms

## Session -33

# BASIC CONCEPTS

The computing times of algorithms fall into two groups.

- **Group1**– Consists of problems whose solutions are bounded by the **polynomial of small degree**. Example – Binary search  $O(\log n)$  , sorting  $O(n \log n)$ , matrix multiplication  $O(n^{2.81})$
- **Group2** – Contains problems whose best known algorithms are non polynomial.

Example –Traveling salesperson problem  $O(n^2 2^n)$ , knapsack problem  $O(2^{n/2})$  etc.

There are two classes of non polynomial time problems

**1. NP-Complete:** Have the property that it can be solved in polynomial time if all other NP-Complete problems can be solved in polynomial time.

**2. NP-Hard-** If it can be solved in polynomial time then all NP-Complete can be solved in polynomial time.

**“All NP-Complete problems are NP-Hard but not all NP Hard problems are not NP-Complete”**

# Deterministic and Non-Deterministic Algorithms

Algorithms with the property that the result of every operation is **uniquely** defined are termed deterministic.

- Such algorithms agree with the way programs are executed on a computer.
- When the outcome is not uniquely defined but is limited to a specific set of possibilities, we call it non deterministic algorithm

To specify such algorithms in SPARKS, we introduce three statements

- choice(S) : arbitrarily chooses one of the elements of the set S.
- failure : Signals an unsuccessful completion.
- Success : Signals a successful completion.

# EXAMPLE OF A DETERMINISTIC ALGORITHM

```
Algorithm Lsearch (A, n, x) {  
    for i:= 1 to n do {  
        if(a[i]=x) then  
            return i;  
    }  
    return 0;  
}
```

# EXAMPLE OF A NON DETERMINISTIC ALGORITHM

Algorithm Search(A, n, x) {

    j:= choice(1,n);

    if(a[j]=x) then {

        return j;

        success();

    }

    return 0;

    failure();

}

# DEFINITIONS

- **Decision problem**

Any problem whose answer is yes or no

- **Decision algorithm**

An algorithm for a decision problem is termed as a decision algorithm

- **Optimization problem**

Any problem that involves the identification of an optimal (either minimum or maximum) values of a given cost function is known as an optimization problem.

- **Optimization algorithm**

An optimization algorithm is used to solve an optimization problem.

- **P** is the set of all decision problems solvable by a deterministic algorithm in polynomial time.

### Sample Problems in P :

Fractional Knapsack, MST , Sorting.

- **NP** is the set of all decision problems solvable by a nondeterministic algorithm in polynomial time.

### Sample Problems in NP :

Fractional Knapsack, MST , Sorting and Hamiltonian Cycle (Traveling Salesman), Graph Coloring



# SATISFIABILITY

- Let  $x_1, x_2, x_3, \dots, x_n$  denotes Boolean variables.
- Let  $\neg X_i$  denotes the negation of  $x_i$ .
- A literal is either a variable or its negation.
- A formula in the propositional calculus is an expression that can be constructed using literals and the operators  $\wedge$ (AND) and  $\vee$ (OR).

- A formula is in Conjunctive Normal Form (CNF) iff it is represented as  $\bigwedge C_i$ , where the  $C_i$  are clauses each represented as  $\bigvee l_{ij}$ .
- It is in Disjunctive Normal Form (DNF) iff it is represented as  $\bigvee C_i$  and each clause is represented as  $\bigwedge l_{ij}$
- The satisfiability problem is to determine if a formula is true for some assignment of truth values to the variables
- CNF-Satisfiability is the satisfiability problem for CNF formulas

# ALGORITHM FOR SATISFIABILITY

Algorithm EVAL(E, n)

{

for i := 1 to n do

    xi := choice(true, false);

if E(x1,...,xn)

    then success();

else failure();

}

## Reducibility

Let  $L_1$  and  $L_2$  be problems.  $L_1$  reduces to  $L_2$  ( $L_1 \alpha L_2$ ) if and only if there is a deterministic polynomial time algorithm to solve  $L_1$  that solves  $L_2$  in polynomial time.

➤ If  $L_1 \alpha L_2$  and  $L_2 \alpha L_3$  then  $L_1 \alpha L_3$ .

## COOK's Theorem

Satisfiability is in P if and only if  $P = NP$ .

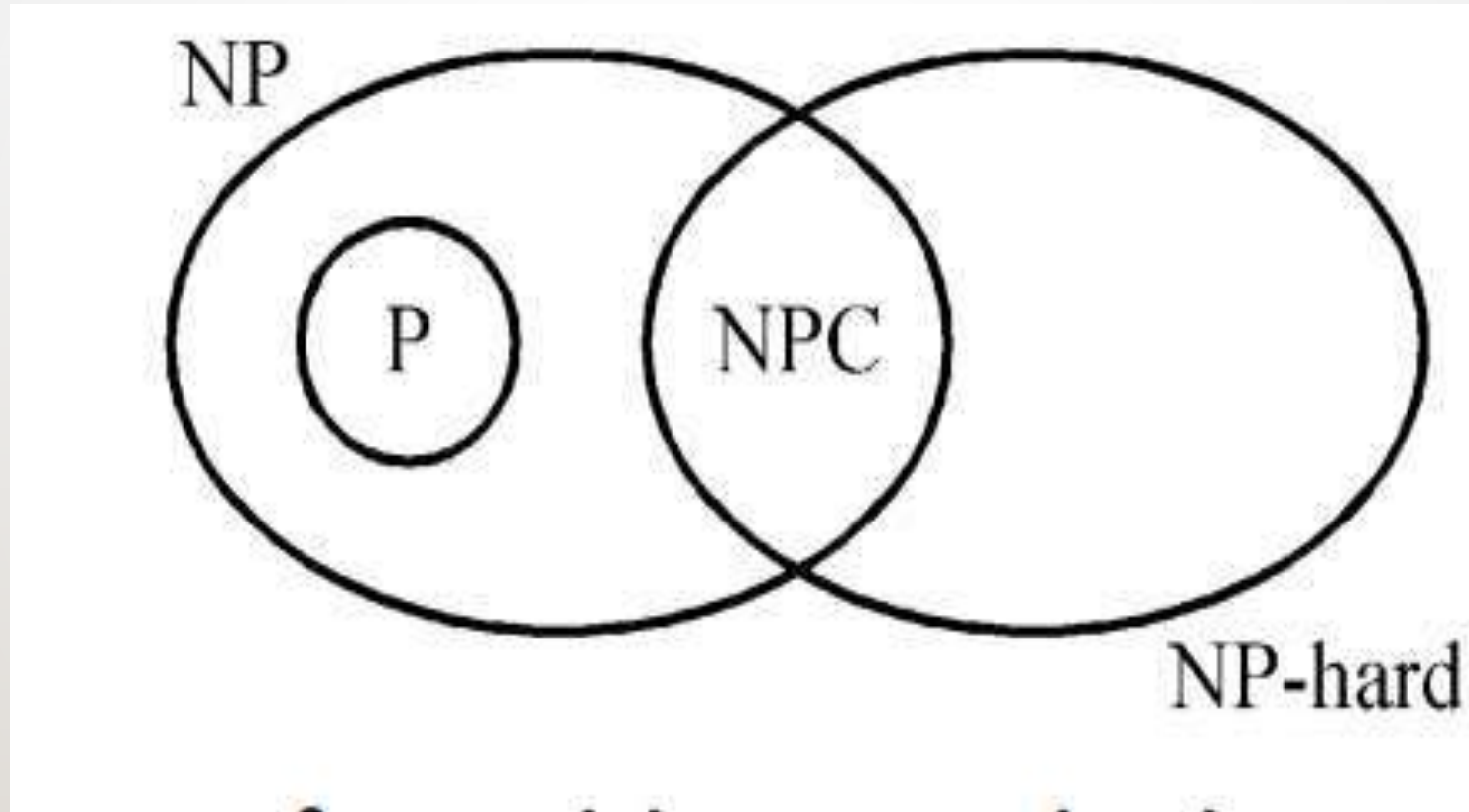
### NP-Hard

A problem  $L$  is NP-hard if any only if satisfiability reduces to  $L$ .

### NP-complete

A problem  $L$  is NP-complete if and only if  $L$  is NP-hard and  $L \in NP$ .

# RELATIONSHIP BETWEEN P, NP, NP-HARD AND NP-COMPLETE



# COOK'S THEOREM

- Theorem states that satisfiability is in P if and only if  $P = NP$ .
- To prove this, we show how to obtain from any polynomial time nondeterministic decision algorithm  $A$  and input  $I$  a formula  $Q(A, I)$  such that  $Q$  is satisfiable if  $A$  has a successful termination with input  $I$ .
- If the length of  $I$  is  $n$  and the time complexity of  $A$  is  $p(n)$  for some polynomial  $p()$ , then the length of  $Q$  is  $O(p^3(n) \log n) = O(p^4(n))$ .

- The time needed to construct  $Q$  is also  $O(p^3(n)\log n)$ .
- A deterministic algorithm  $Z$  to determine the outcome of  $A$  on any input  $I$  can be easily obtained.
- Algorithm  $Z$  simply computes  $Q$  and then uses a deterministic algorithm for the satisfiability problem to determine whether  $Q$  is satisfiable.
- If  $O(q(m))$  is the time needed to determine whether a formula of length  $m$  is satisfiable, then the complexity of  $Z$  is  $O(p^3(n) \log n + q(p^3(n) \log n))$ .

- If satisfiability is in P, then  $q(m)$  is a polynomial function of  $m$  and the complexity of  $Z$  becomes  $O(r(n))$  for some polynomial  $r()$ . Hence, if satisfiability is in P, then for every nondeterministic algorithm  $A$  in NP we can obtain a deterministic  $Z$  in P.
- So, the above construction shows that if satisfiability is in P, then  $P = NP$ .



# Difference Between NP-Hard and NP-Complete

NP-Hard	NP-Complete
NP-Hard problems(say X) can be solved if and only if there is a NP-Complete problem(say Y) that can be reducible into X in polynomial time.	NP-Complete problems can be solved by a non-deterministic Algorithm/Turing Machine in polynomial time.
To solve this problem, it do not have to be in NP .	To solve this problem, it must be both NP and NP-hard problems.
Do not have to be a Decision problem.	It is exclusively a Decision problem.
<b>Example:</b> Halting problem, Vertex cover problem, etc.	<b>Example:</b> Determine whether a graph has a Hamiltonian cycle, Determine whether a Boolean formula is satisfiable or not, Circuit-satisfiability problem, etc.

**Example :** Halting problem is NP-hard decision problem, but it is not NP-complete.

## **Halting problem is NP-hard**

To show that Halting problem is NP-hard, we show that

**satisfiability  $\alpha$  halting problem.**

For this let us construct an algorithm A

whose input is a propositional formula X.

- Suppose X has n variables.
- Algorithm A tries out all  $2^n$  possible truth assignments and verifies if X is satisfiable.

Halting problem is un-decidable.

- Hence there exists no algorithm to solve this problem.
- So, it is not in NP. Therefore, it is not NP-complete.

## Questions:

1. With suitable example, explain nondeterministic algorithm.
2. Explain terminology used in Satisfiability Problem.
3. Explain Cook's theorem.
4. Differentiate NP-Hard with NP-Completeness.

# THANK YOU