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**Theory of Computation II: COM 5108**  
*Lecture II*

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## 1 P, NP, coNP and Complete Problems

We have already defined the basic complexity classes **P** and **NP** using polynomial time bounded deterministic and nondeterministic Turing machines. In this section we start with a different definition of the class **NP**. We show that these two definitions coincide. Then following Cook and Levin we establish the existence of an important **NP-complete** problem, the *satisfiability* problem of propositional logic.

Many of us have experience that solving a mathematical problem is more difficult than verifying the correctness of a given solution. It seems, it is also manifested in computing (no one knows for sure). There are decision problems for which there is no known *polynomial time* algorithm (Turing machine solving the problem within the number of steps bounded by some polynomial over the length of an input), but if an appropriate “*witness*” or “*proof*” of polynomial size is provided for a *positive answer* ( $x \in L$ ), then that can be verified in polynomial time.

We have already defined  $\mathbf{NP} = \bigcup_{k \geq 1} \mathbf{NTIME}(n^k)$ . Following is an alternate and interesting definition using polynomial time ‘proof verifier’.

**Definition 1.** A language  $L \subseteq \{0, 1\}^*$  is in **NP**, if there is a polynomial time bounded Turing Machine  $V$  called a *verifier* and a polynomial  $p(n)$  over  $\mathbb{N}_0$ , such that

$$x \in L \text{ if and only if } \exists w \in \{0, 1\}^{p(|x|)} \text{ such that } V \text{ accepts } \langle x, w \rangle,$$

where  $w$  is called an *witness*, *certificate*, or *proof* of  $x \in L$ . It is natural that the witness cannot be *too long*. Its length must be polynomial bounded. Otherwise reading the proof will take ‘long’ time.

If  $L \in \mathbf{P}$ , then it must be in **NP**, and the verifier is the decider of  $L$  with *null string* as the witness. So  $\mathbf{P} \subseteq \mathbf{NP}$ . The class **coNP** is defined as follows:

$$\mathbf{coNP} = \{L \subseteq \{0, 1\}^* : \Sigma^* \setminus L \in \mathbf{P}\}^1.$$

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<sup>1</sup>The notion of  $\Sigma^* \setminus L$  is a bit relaxed. We treat PRIME as a complement of COMPOSITE, and do not bother about 1.

The following proposition shows the equivalence of two definitions of **NP**.

**Proposition 1.** **NP** as defined above is same as  $\bigcup_{k \geq 1} \mathbf{NTIME}(n^k)$ .

**Proof:** Let  $L$  is decided by an NDTM  $N$  within  $p(n)$  number of steps, where  $p(n)$  is a polynomial. So for each  $x \in L$ , there is a sequence of choices of transitions<sup>2</sup> that leads to an *accept halt*. The length of this sequence cannot be longer than  $p(n)$ .

Given the input  $x$ , the description of  $N$  and the choice sequence, a deterministic Turing machine  $M$  can verify in polynomial time that  $x$  drives  $N$  to an *accept halt*. The length of witness, the nondeterministic choice sequence, is bounded by a polynomial, the description of  $N$  is fixed (it does not depend on  $x$ ). So the language  $L$  is in **NP**.

Let  $L \in \mathbf{NP}$ , so there is a polynomial time verifier  $V$ . Following is the polynomial time NTM.

$N$ : input  $x$

1. Nondeterministically creates a witness string  $w$  of the choices of transitions<sup>3</sup>. The length of  $w$  is bounded by  $p(n)$ , where  $n = |x|$ .
2. Use the verifier  $V$  on  $\langle x, w \rangle$ .
3. Accept if  $V(x, w) = 1$ , else reject.

The running time of  $N$  is bounded by polynomial.

QED.

A few examples of problems in **NP**.

**Example 1.**

1. **COMPOSITE** =  $\{n \in \mathbb{N} : \exists p, q \in \mathbb{N} (p, q > 1 \text{ and } n = p \cdot q)\}$ . The certificate is a pair of factors which cannot be longer than  $c \lceil \log_2 n \rceil$ , where the input is of length  $\lceil \log_2 n \rceil$ .  
As a consequence **PRIME**, the collection of *prime numbers*, is in **coNP**. But now it is known that **PRIME** is in **P**. So **COMPOSITE** is also in **P**.
2. Whether a graph  $G$  has an *independent set* of certain size.  
**INDSET** =  $\{\langle G, k \rangle : \exists S \subseteq V(G), |S| \geq k, \text{ and } \forall u, v \in S, \{u, v\} \notin E(G)\}$ ,  $k \in \mathbb{N}$  specifies the size of the *independent set*. The certificate is a set of vertices.
3. **TSP** =  $\{\langle G = (V, E), d : E \rightarrow \mathbb{N}, k \rangle : \text{there is a travelling salesperson's tour of distance } \leq k\}$ .  $V = \{1, 2, \dots, n\}$  is a set of nodes,  $\binom{n}{2}$  distances  $d_{ij}$  between nodes  $i$  and  $j$ , and  $k$  is a number. Decide whether there is a tour that visits every node exactly once and the total length traversed is at most  $k$ . The certificate is a sequence of such nodes.
4. **GRAPH-ISO** =  $\{\langle G_1, G_2 \rangle : \text{graph } G_1 \text{ and graph } G_2 \text{ are isomorphic}\}$ . Given two adjacency matrices  $E_1$  and  $E_2$  corresponding to  $G_1$  and  $G_2$ , decide whether there is a permutation  $\pi : V_1 \rightarrow V_2$  so that  $E_1$  after reordering is same as  $E_2$ .
5. **FACTORING** =  $\{\langle n, l, u \rangle : n, l, u \in \mathbb{N}, n \text{ has a factor } p, l \leq p \leq u\}$ . The certificate is  $p$ .

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<sup>2</sup>In an alternative way we may assume that the degree of nondeterminism is 2, and there two transition functions  $\delta_0$  and  $\delta_1$ .

<sup>3</sup>If  $x \in L$ , there is a sequence that leads to *accept halt*.

6. PRIME is also in NP.

The certificate is more complicated (*Pratt Certificate*). We know that a natural number  $p > 1$  is a prime if and only if  $\mathbb{Z}_p^*$  is a cyclic group of order  $p - 1$  i.e. there is an  $a \in \mathbb{Z}_p^*$ ,  $1 < a < p$ , so that  $a^{p-1} \equiv 1 \pmod{p}$  (*Fermat test*).

And for all prime factors  $q$  of  $p - 1$ ,  $a^{\frac{p-1}{q}} \not\equiv 1 \pmod{p}$  (*Lucas test*).

- (a) So generator  $a$  is part of the certificate. Computation of  $a^{p-1}$  modulo  $p$  can be performed in  $O(l^3)$ , where  $l = \lceil \log p \rceil$ . But it is not sufficient to have  $a$  alone as a certificate as it may fail:  $4^{15-1} \equiv 1 \pmod{15}$ , but 15 is no prime. In this case the second test fails as  $4^2 \equiv 1 \pmod{15}$ .
- (b) So the prime factors of  $p - 1$  is also part of the certificate. But the list of prime factors of  $p - 1$  may be *false*. Each factor also needs the certificate of its primality. As an example, a false certificate of 45 is  $(8; 2, 22)$ , where  $8^{44} \pmod{45} = 1$ . Also  $8^2 \pmod{45} = 19$  and  $8^{22} \pmod{45} = 19$ . This satisfy the second condition also. But if the certificate was correct i.e.  $(8; 2, 11)$ , the second test will fail as  $8^{44/11} \pmod{45} = 8^4 \pmod{45} = 1$ .
- (c) The complete certificate for a prime defined *inductively* is as follows:  
*Basis:*  $C(2) = ()$ , as  $2 - 1 = 1$  has no prime factors.  
*Induction:*  $C(p) = (a; q_1, C(q_1), q_2, C(q_2), \dots, q_k, C(q_k))$ , where  $q_1, \dots, q_k$  are prime factors of  $p - 1$ . The process stops at  $p = 2$ .
- (d) A certificate for 43 is as follows: 2 generates  $\mathbb{Z}_{43}^*$  and the prime factors of  $43 - 1 = 42$  are  $\{2, 3, 7\}$ .

$$\begin{aligned} & (2; (2, C(2)), (3, C(3)), (7, C(7))) \\ &= (2; (2, ()), (3, (2; (2, ()))), (7, (3; (2, C(2)), (3, C(3))))) \\ &= (2; (2, ()), (3, (2; (2, ()))), (7, (3; (2, ()), (3, (2; (2, ())))))) \end{aligned}$$

- (e) It can be proved that the length of the certificate is bounded by  $5(\log p)^2$ , a polynomial over the length of input.
  - i. The bound is true for  $p = 2$  and  $p = 3$ .
  - ii. Let the number of prime factors of  $p - 1$  be  $k < \log_2 p$ , and they are  $q_1 = 2$  ( $p - 1$  is even),  $q_2, \dots, q_k$ .  
The certificate  $C(p) = (a, 2, C(2), q_2, C(q_2), \dots, q_k, C(q_k))$ .  
The length of  $C(p)$  is determined by  $a$  ( $|a| < \log p$ ),  $2k$  separators ( $< 2 \log p$ ), certificate of 2 (of constant length), length of all  $q_i$ 's ( $2 \log p$ ) (Note<sup>4</sup>), and the lengths of  $C(q_i)$ 's.
  - iii. By the induction hypothesis, the length of the certificate for each  $q_i$  is  $5(\log q_i)^2$ .
  - iv. So the total length of the certificate is bounded by

$$|C(p)| \leq 5 \log p + c_1 + 5 \sum_{i=2}^k (\log q_i)^2 < 5(\log p)^2.$$

$$\log(q_2 \cdots q_k) \leq \log \frac{p-1}{2} < \log p - 1 \Rightarrow \log q_2 + \cdots + \log q_k < \log p - 1. \text{ Therefore } \log^2 q_2 + \log^2 q_3 + \cdots + \log^2 q_k < (\log p - 1)^2$$

$$|C(p)| \leq 5 \log p + 5 \log^2 p - 10 \log p + c < 5 \log^2 p.$$

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<sup>4</sup>The number of bits for  $2^n$  is  $\log_2 2^n = n$  and the number of bits for  $n$  2's is  $2n$ . This is the largest possible.

- v. The verification of  $C(p)$  can be performed in  $O(n^4)$  time, where  $n = \log p$ . The computation of modular exponentiation takes  $O(n^3)$  time.

It is not known whether **NP** and **P** are equal. This is considered to be the ‘*Holy Grail*’ of computer science! Most researchers believe that  $\mathbf{P} \neq \mathbf{NP}$ . There are many similar unanswered questions in this area.

The class **NP** was originally defined in terms of nondeterministic Turing machine. The name **NP** comes from *nondeterministic polynomial time bounded Turing machine*.

## 1.1 NP-hard, and NP-Complete

It was observed that there is a large collection of decision problems (membership in a languages) such as the *satisfiability* of Boolean formula, *independent set* of a certain size in an undirected graph, *3-colouring* of graph etc. are in the class *NP*. All these problems are difficult to solve in the sense that there is no known polynomial time algorithm. But they have a connection, one of them can be translated to another in polynomial time. The translation or *reduction* is defined in the following way.

**Definition 2.** A language  $L \subseteq \{0, 1\}^*$  is *polynomial time mapping reducible*<sup>5</sup> to  $L' \subseteq \{0, 1\}^*$ , if there is a polynomial-time bounded computable function  $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ , such that

$$x \in L \text{ if and only if } f(x) \in L', \text{ for all } x \in \{0, 1\}^*.$$

This is denoted as  $L \leq_P L'$ .

**Definition 3.** A language  $L'$  is **NP-hard** if for every  $L \in \mathbf{NP}$ ,  $L \leq_P L'$ . A language  $L'$  is called **NP-complete** if it is **NP-hard** and also belongs to the class **NP**.

Following are a few properties of ‘ $\leq_P$ ’.

- (a) If  $L \leq_P L'$  and  $L' \in \mathbf{P}$ , then  $L \in \mathbf{P}$ .
- (b)  $L \leq_P L$ , for all  $L$  - the binary relation is reflexive.
- (c) If  $L \leq_P L'$  and  $L' \leq_P L''$ , then  $L \leq_P L''$  - the binary relation is transitive.
- (d) What can you conclude about  $L'$ , if  $L \leq_P L'$  and  $L$  is **NP-hard**?

**Proof:** Proof of these properties are simple.

- (a) Let the polynomial time bounded ( $p$ ) Turing computable function  $f$  reduces  $L$  to  $L'$  and let  $L'$  be decided by a polynomial time bounded ( $q$ ) Turing machine  $M$ . Following is the decider for  $L$ .

$N$ : input  $x$

- (i) Compute  $f(x)$ .
- (ii) Simulate  $M$  on  $f(x)$ .
- (iii) If  $M$  accepts  $f(x)$ , accept  $x$ ;

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<sup>5</sup>It is also called *polynomial time many-one reducible*. It was Richard Karp who demonstrated it for the first time [RMK].

- (iv) otherwise *reject*  $x$ .
- (b) Ex.
- (c) Ex.
- (d) If  $L'' \in \mathbf{NP}$ , then  $L'' \leq_P L \Rightarrow L'' \leq_P L'$  (transitivity of ' $\leq_P$ '). So all problems of of  $\mathbf{NP}$  are mapping reducible to  $L'$ , that makes it  $\mathbf{NP-hard}$ .

QED.

Consider the following synthetic language.

$$L_{\mathbf{NP}} = \{ \langle V, x, 1^n, 1^t \rangle : \exists u \in \{0, 1\}^n \text{ s.t. } V \text{ accepts } \langle x, u \rangle \text{ within } t \text{ steps} \},$$

where  $V$  is an encoding of a deterministic Turing machine.

**Proposition 2.**  $L_{\mathbf{NP}}$  is  $\mathbf{NP-complete}$ .

**Proof:**

$L_{\mathbf{NP}}$  is in  $\mathbf{NP}$ :

We design a verifier  $V'$  for  $L_{\mathbf{NP}}$ . Consider  $\langle V, x, 1^n, 1^t \rangle$ , if there is an  $u \in \{0, 1\}^n$  such that  $V$  accepts  $\langle x, u \rangle$  in time  $t$ , then  $\langle V, x, 1^n, 1^t \rangle \in L_{\mathbf{NP}}$ . This  $u$  can be used as a certificate of  $\langle V, x, 1^n, 1^t \rangle$ . Its length is linear with respect to the length of input, due to  $1^n$ .  $V'$  simulates  $V$  on  $\langle x, u \rangle$  for at most  $t$  steps. This can be done in polynomial time. If  $V$  accepts, then  $V'$  returns 'Y' else it returns 'N'.

Any language  $L \in \mathbf{NP}$  is polynomial time reducible to  $L_{\mathbf{NP}}$ :

If  $L$  is in  $\mathbf{NP}$ , then by definition there is a polynomial  $p : \mathbb{N}_0 \rightarrow \mathbb{N}_0$  and a polynomial time bounded Turing machine  $V'$  so that for all  $x \in \{0, 1\}^*$ ,  $x \in L$  if and only if there is a  $u \in \{0, 1\}^{p(|x|)}$  such that  $V'$  accepts  $\langle x, u \rangle$  in polynomial time. Let the running time of  $V'$  be bounded by the polynomial  $q : \mathbb{N}_0 \rightarrow \mathbb{N}_0$ . The reduction is

$$x \mapsto \langle V', x, 1^{p(|x|)}, 1^{q(|x|+p(|x|))} \rangle .$$

This mapping can be done in polynomial time as  $\langle V' \rangle$  is of fixed length, and lengths of both  $1^{p(|x|)}$  and  $1^{q(|x|+p(|x|))}$  are polynomial bounded. QED.

But this language is not very interesting or useful for reducing problems from different practical domains. S A Cook in 1971 [SAC] and L A Levin in 1973 (independently from USSR) [LAL] presented the notion of  $\mathbf{NP-completeness}$  and gave examples of  $\mathbf{NP-complete}$  problems from domains like logic etc.. Subsequently R M Karp in 1972 [RMK] showed a large collection of practical problems to be  $\mathbf{NP-complete}$ .

## 1.2 Boolean Formula

**Definition 4.** A *boolean formula* is defined as follows.

1. Boolean constants *true* and *false* are boolean formulas.
2. Boolean variables  $x_1, x_2, \dots$  (that takes values *true* or *false*) are boolean formulas.
3. If  $f_1$  and  $f_2$  are boolean formula, then so are  $(f_1 \vee f_2)$ ,  $(f_1 \wedge f_2)$  and  $\neg f_1$ .
4. Nothing else is a boolean formula.

A variable or a negation of a variable is called a *literal*. We shall use  $\overline{f}$  for  $\neg f$  for negation of a formula.

We encode *true* and *false* as 1 and 0 respectively. If  $\phi$  is a boolean formula of  $n$  variables,  $x_1, \dots, x_n$ , we can assign truth values to the variables (an element  $v \in \{0, 1\}^n$ ) and get a truth value  $\phi(v)$  for the formula. All possible assignments to the variables form the truth table. A boolean formula  $\phi$  is *satisfiable* if there is a truth assignment that make  $\phi$  *true*. It is *unsatisfiable* if for no truth assignment the formula is *true*.

**Example 2.** The formula

$$(x_1 \vee x_2 \vee \overline{x_3}) \wedge (\overline{x_1} \vee \overline{x_2} \vee x_3) \wedge (x_1 \vee \overline{x_2} \vee \overline{x_3})$$

is satisfiable with assignment  $x_1 = 1, x_2 = 0, x_3 = 1$ . But following formulas are unsatisfiable

1.  $x \wedge \overline{x}$ .
2.  $(x_1 \vee x_2) \wedge \overline{x_1} \wedge \overline{x_2}$ .
3.  $(\overline{x_1} \vee \overline{x_2} \vee \overline{x_3}) \wedge (x_1 \vee x_2 \vee x_3) \wedge (x_1 \vee \overline{x_3}) \wedge (x_3 \vee \overline{x_2}) \wedge (x_2 \vee \overline{x_1})$ .

A boolean formula is in *conjunctive normal form (CNF)* or *clausal normal form* if it is conjunction (and) of clauses. A clause is a disjunction (or) of literals. It is in *disjunctive normal form (DNF)* or *sum of product* form if it is disjunction of the conjunctions of literals. It is a  $k$ -CNF if it is in CNF and every clause has at most  $k$  literals.

We define the following languages:

$$SAT^6 = \left\{ \phi = \bigwedge_{i=1}^m \left( \bigvee_{j=1}^{n_i} u_{ij} \right) : m, n_i > 0 \text{ and } \phi \text{ is satisfiable} \right\},$$

where  $u_{ij}$  is a *literal* and

$$3SAT = \left\{ \phi = \bigwedge_{i=1}^m \left( \bigvee_{j=1}^3 u_{ij} \right) : m > 0 \text{ and } \phi \text{ is satisfiable} \right\},$$

$$2SAT = \left\{ \phi = \bigwedge_{i=1}^m (l_{i1} \vee l_{i2}) : \phi \text{ is satisfiable} \right\},$$

**Proposition 3.** Any Boolean function  $f : \{0, 1\}^n \rightarrow \{0, 1\}$  can be expressed as a *disjunctive normal form (DNF)* or a *conjunctive normal form (CNF)* (functional completeness of  $\vee, \wedge, \neg$ ).

**Proof:** It is known that for a  $n$  variable formula there are  $2^n$  rows in the truth table. We take the standard convention that the truth assignment corresponding of the variables in the  $j^{th}$  row is the  $n$ -bit binary representation of  $j$ ,  $0 \leq j \leq 2^n - 1$ .

Consider the truth-table corresponding to an  $n$ -variable Boolean function  $f(x_1, \dots, x_n)$ . For the equivalent DNF formula  $\psi$ , we only consider those rows of the table where the truth values of the function is 1. Each row corresponds to

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<sup>6</sup>One may define  $SAT = \{\phi : \phi \text{ is satisfiable}\}$ .

a *conjunction* ( $\wedge$ ) of literals, and all of them are connected by disjunction ( $\vee$ ) to form the final formula  $\psi$ . Let  $j$  be one such row and the values of the variables be  $v_1, \dots, v_n$  (an  $n$ -tuple of 1's and 0's). Corresponding to this row, the conjunct of literals is  $D_j = l_1 \wedge \dots \wedge l_n$ , where  $l_i = x_i$  if  $v_i = 1$ , otherwise it is  $\overline{x_i}$ . It is clear that no other assignment of variables can make  $D_j$  true as at least one of the literals will be false. Finally the equivalent DNF formula is

$$\psi(x_1, \dots, x_n) = D_{i_1} \vee \dots \vee D_{i_k},$$

where there are  $k$  rows with truth values 1.

We observe that  $\psi(v_1, \dots, v_n) = 1$ , if one of disjuncts is 1 i.e. one of the rows of the truth table of  $\psi$  has a 1.

On the other hand if  $\psi(v_1, \dots, v_n) = 0$ , then all  $D_{i_j}$ s are false or 0. So  $f(v_1, \dots, v_n) = 1$  if and only if  $\psi(x_1, \dots, x_n) = 1$ .

As an example consider a 5-variable formulas and the  $j^{\text{th}}$  row of the truth table where the truth value 1. Let the values of the Boolean variables in the  $j^{\text{th}}$  row be (01101), then the corresponding conjunctive formula  $D_j = \overline{x_1} \wedge x_2 \wedge x_3 \wedge \overline{x_4} \wedge x_5$ . It is clear that  $D_j(01101) = 1$ , but  $D_j(k) = 0$  for any other truth assignment.

Similarly to get the equivalent CNF formula of  $f$ , we consider only those rows where the truth values are 0. If the values of the variables  $(x_1, \dots, x_n)$  are  $(v_1, \dots, v_n)$  in one such row  $j$ , we take  $C_j = l_1 \vee \dots \vee l_n$ , where  $l_i = x_i$  if  $v_i = 0$ , otherwise it is  $\overline{x_i}$ . The truth value of  $C_j$  is 1 or true if the value of one variable say,  $x_i$ , is changed to  $1 - v_i$ .

The equivalent CNF formula of  $f$  is

$$\psi(x_1, \dots, x_n) = C_{i_1} \wedge \dots \wedge C_{i_k},$$

where there are  $k$  rows of the truth table with the truth values 0.

As an example we consider a 5-variable Boolean formula. Let the variables in the  $j^{\text{th}}$  row of the truth table, where  $\phi(j) = 0$ , takes the values (10011), then  $C_j = \overline{x_1} \vee x_2 \vee x_3 \vee \overline{x_4} \vee \overline{x_5}$ . It is clear that  $C_j(10011) = 0$ , but  $C_j(k) = 1$ , if  $k \neq 10011$ . QED.

The length of such a formula may be  $O(n2^n)$ , where the length of a formula is the count of the number of  $\vee$  and  $\wedge$ . The size of a truth table is exponential in the number of variables.

Before we prove that 3SAT is NP-complete, we shall prove an interesting result that is  $2\text{SAT} \in \mathbf{P}$ .

Let  $\phi$  be a 2SAT formula. We construct a graph  $G_\phi = (V_\phi, E_\phi)$ , where  $V_\phi = \{x, \overline{x} : x \text{ is a variable in } \phi\}$ , and  $E_\phi = \{(l_1, l_2) : \text{if } (l_2 \vee \overline{l_1}) \text{ (or } \overline{l_1} \vee l_2)\} \text{ is a clause in } \phi\}$ .

Each edge in  $G_\phi$  captures a clause in  $\phi$  as a logical implication. Note that  $(v \vee \overline{u})$ ,  $(\overline{u} \vee v)$  and  $(u \Rightarrow v)$  are logically equivalent.

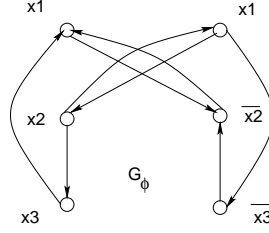
There is a symmetry in the graph. A clause  $(l_1 \vee l_2) = (\overline{\overline{l_1}}, \overline{\overline{l_2}})$  gives rise to two edges:  $(\overline{l_1} \vee l_2)$  and  $(\overline{l_2}, l_1)$ . If there is a path from some literal  $l_1 \rightarrow \dots \rightarrow l_k$ ,  $k \geq 1$  in the graph, then by the transitivity of implication we have  $(l_1 \Rightarrow l_k)$ . If there is a path from  $l_1$  to  $l_k$ , then there is a path from  $\overline{l_k}$  to  $\overline{l_1}$ .

Following the semantics of implication, if  $l_1$  is assigned the value true, then every literal reachable from  $l_1$  in  $G_\phi$  should also be true. Symmetrically, if  $l_1$  is assigned false, then all its predecessor literals will also be false.

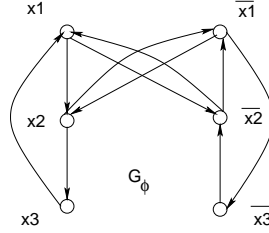
A variable  $x$  cannot be assigned any truth value in a formula  $\phi$ , if there is a path from  $x$  to  $\bar{x}$  (equivalently a path from  $\bar{x}$  to  $x$ ) in  $G_\phi$ , as it is same as  $x \Leftrightarrow \bar{x}$  - a contradiction.

**Example 3.** Consider the following example,

$$\phi = (x_1 \vee x_2) \wedge (x_1 \vee \bar{x}_3) \wedge (\bar{x}_1 \vee \bar{x}_2) \wedge (\bar{x}_2 \vee x_3).$$



There is an satisfying assignment,  $x_1 = 1, x_2 = 0, x_3 = 1$ . But if we include another clause,  $(\bar{x}_1 \vee x_2)$ , then there is no satisfying assignment any more, as there will be a path from  $x_1$  to  $\bar{x}_1$  and also a path from  $\bar{x}_1$  to  $x_1$ .



**Lemma 1.** A 2SAT formula  $\phi$  is *unsatisfiable* if and only if there is a variable  $x$  such that there is a path from  $x$  to  $\bar{x}$  (also a path from  $\bar{x}$  to  $x$ ).

**Proof:** Let for some variable  $x$  there are two such paths and at the same time  $\phi$  is satisfiable. So there is a truth value  $v(x)$  for  $x$ . Let  $v(x)$  is *true* and  $v(\bar{x})$  is *false*. As there is a path from  $x$  to  $\bar{x}$ , there must be an edge  $(l_1, l_2)$  such that  $v(l_1) = \text{true}$  but  $v(l_2) = \text{false}$ . The corresponding clause is  $(\bar{l}_1 \vee l_2)$  and is not satisfiable - a contradiction. Similar argument works for  $v(\bar{x}) = \text{false}$ .

In the other direction, we assume that there is no variable  $x$  with such pair of paths. The satisfying truth assignment of  $\phi$  is as follows:  
The following procedure will be repeated until all nodes are assigned truth values.

Take a literal  $l$ , a node in  $G_\phi$ , that has not been assigned any truth value and there is no path from  $l$  to  $\bar{l}$ . Assign *true* to  $l$  and every literal reachable from the node of  $l$ . Assign *false* to the negation of these literals. In other words, if a node is assigned *false* then its predecessor is also assigned *false*. If  $l'$  is reachable from  $l$ , then  $v(l') = \text{true}$ . If the node  $\bar{l}$  is reachable from  $\bar{l}$ , then both have value *false*.

We claim that the process cannot assign same truth value to  $l'$  and  $\bar{l}'$  i.e. nodes of both  $l'$  and  $\bar{l}'$  cannot be reachable from  $l$ . If that was possible then  $\bar{l}$  would have been reachable from both of them resulting a path from  $l$  to  $\bar{l}$ . QED.

**Proposition 4.** 2SAT  $\in$  P

**Proof:** The steps of the algorithm are as follows:

$M$ : input  $\phi$



1. Build the graph  $G_\phi$ .
2. For each variable  $x$ , test whether  $\bar{x}$  is reachable and vice versa.
3. Accept if no such path exist; otherwise reject.

It is an  $O(n^2)$  algorithm. QED.

It is interesting that 2SAT is of so low complexity, but there is no known polynomial time algorithm for 3SAT. In fact there is a very strong belief that it is impossible to have one.

### 1.3 Cook-Levin Theorem

Cook [SAC] and Levin [LAL] demonstrated the first **NP**-complete problem.

**Theorem 2.** Both SAT and 3SAT are **NP**-complete (Cook and Levin).

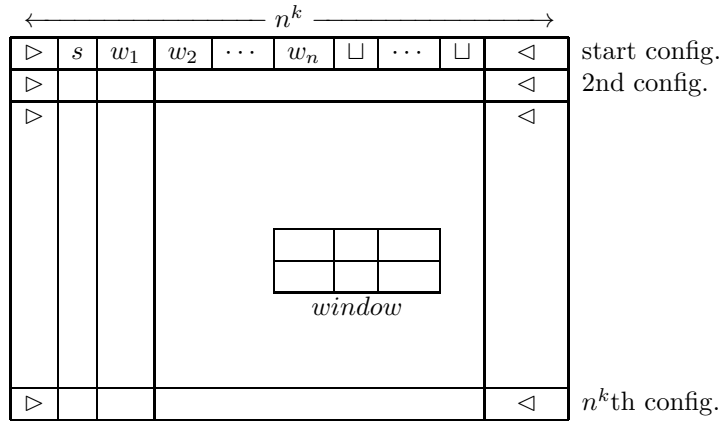
**Lemma 3.** Both SAT and 3SAT are in NP.

**Proof:** The certificate is the truth value assignment of the variables in the SAT (3SAT) formula  $\phi$ . Given an assignment it is possible to evaluate the truth value of  $\phi$  in polynomial time. QED.

**Theorem 4.** SAT is **NP**-complete.

We need to reduce any language  $L \in \mathbf{NP}$  to SAT in polynomial time. Let  $L$  be decided by an NTM  $N$  in polynomial time. The reduction of  $L$  to SAT takes a  $x \in \Sigma^*$  as input and produces a boolean formula  $\phi$  that in a sense simulates the computation of  $N$  on the input  $x$ . If  $N$  accepts  $x$  i.e.  $x \in L$ , then there is a *satisfying* truth assignment for  $\phi$ . Otherwise  $\phi$  is *unsatisfiable*.

**Proof:** Let  $N$  decide  $L$  in  $n^k$  time. The total computation of  $N$  on the input  $x = w_1w_2 \cdots w_n$  can be captured by a table of size  $n^k \times n^k$ .



QED.

We have used two end markers  $\{\triangleright, \triangleleft\}$  for every configuration. The first row is the *start configuration* of the computation on input  $x = w_1w_2 \cdots w_n$  at the start state  $s$ . The table corresponding to an input  $x \in L$  must have a row of *accepting* configuration. The problem is to determine whether there is a table with an *accepting* configuration corresponding to the nondeterministic computation of  $N$  on  $x$ .

The reduction maps  $x \mapsto \phi$ . The variables of the boolean formula  $\phi$  are defined as follows:

Let  $N = (Q, \Sigma, \delta, s)$  and  $\Sigma = \{\triangleright, \sqcup, s, \dots\}$ . For each  $p \in C = Q \cup \Sigma \cup \{\triangleleft\}$  and  $1 \leq i, j \leq n^k$  (row and column), we have a variable  $v_{i,j,p}$ .

A cell at the  $i^{\text{th}}$  row and  $j^{\text{th}}$  column is  $cell[i, j]$ . A variable  $v_{i,j,p}$  for the  $cell[i, j]$  is 1 (*true*) if its content is  $p$ , otherwise it is 0 (*false*).

The formula  $\phi$  is a conjunction of four formulas:

$$\phi = \phi_{cell} \wedge \phi_{start} \wedge \phi_{move} \wedge \phi_{accept}.$$

Each  $cell[i, j]$  contains exactly one  $p \in C$ .

$$\phi_{cell} = \bigwedge_{1 \leq i, j \leq n^k} \left( \left( \bigvee_{p \in C} v_{i,j,p} \right) \wedge \left( \bigwedge_{\substack{p, q \in C, \\ p \neq q}} (\overline{v_{i,j,p}} \vee \overline{v_{i,j,q}}) \right) \right).$$

**Example 4.** Let  $v_1, v_2, v_3$  be boolean variables. The following formula is true if and only if exactly one variable is true.

$$f(v_1, v_2, v_3) = (v_1 \vee v_2 \vee v_3) \wedge (\overline{v_1} \vee \overline{v_2}) \wedge (\overline{v_2} \vee \overline{v_3}) \wedge (\overline{v_3} \vee \overline{v_1}).$$

The *start configuration* of  $N$  on input  $x = w_1 w_2 \dots w_n$  is proper if the following formula is *satisfiable*.

$$\phi_{start} = v_{1,1,\triangleright} \wedge v_{1,2,s} \wedge v_{1,3,w_1} \wedge \dots \wedge v_{1,n+2,w_n} \wedge v_{1,n+3,\sqcup} \wedge \dots \wedge v_{1,n^k-1,\sqcup} \wedge v_{1,n^k,\triangleleft}.$$

There is an *accepting configuration* in the table of computation of  $N$  if the following formula is satisfied.

$$\phi_{accept} = \bigvee_{1 \leq i, j \leq n^k} v_{i,j,Y}.$$

The transition from configuration  $C_i$  to  $C_{i+1}$  must be compatible with the state transition relation of  $N$ , for all  $i$ ,  $1 \leq i < n^k$ . This is ensured by the *satisfiability* of  $\phi_{move}$ .

At every point in time the computation of a TM is *local*. The head can move one place to left or to right, or it may remain stationary after changing the content of the current cell. The validity of  $C_i \rightarrow_N C_{i+1}$  is checked by looking at every window of size  $2 \times 3$  on these pair of configurations.

Given an NTM  $N$ , there is a finite set of *valid windows* that are *compatible* to  $Q, \Sigma$  and  $\Delta$ .

**Example 5.** Let  $((p, a), \{(p, b, \rightarrow)\}), ((p, b), \{(q, c, \leftarrow), (q, a, -)\}) \in \Delta$ .

The state in the following positions can affect the *window*.

		Window			
1	$p$				
2		$p$			
3			$p$		
4				$p$	
5					$p$

Following are a few possible valid windows.  $\alpha, \beta$  are any tape symbols.

$$1(a) \begin{array}{|c|c|c|} \hline a & \alpha & \beta \\ \hline p & \alpha & \beta \\ \hline \end{array}, \quad 1(b) \begin{array}{|c|c|c|} \hline b & \alpha & \beta \\ \hline c & \alpha & \beta \\ \hline \end{array}, \quad 1(c) \begin{array}{|c|c|c|} \hline b & \alpha & \beta \\ \hline a & \alpha & \beta \\ \hline \end{array}, \quad \text{central cells are unchanged.}$$

$$2(a) \begin{array}{|c|c|c|} \hline p & a & \alpha \\ \hline b & p & \alpha \\ \hline \end{array}, \quad 2(b) \begin{array}{|c|c|c|} \hline p & b & \alpha \\ \hline \beta & c & \alpha \\ \hline \end{array}, \quad 2(c) \begin{array}{|c|c|c|} \hline p & b & * \\ \hline q & a & * \\ \hline \end{array},$$

And there are many more but finite and depends on  $N$  but not on input  $x$ .

Following are a few invalid windows.

$$3(a) \begin{array}{|c|c|c|} \hline \alpha & a & \beta \\ \hline \alpha & b & \beta \\ \hline \end{array}, \quad 3(b) \begin{array}{|c|c|c|} \hline p & a & \alpha \\ \hline p & b & \alpha \\ \hline \end{array}, \quad 3(c) \begin{array}{|c|c|c|} \hline \alpha & p & b \\ \hline q & \alpha & a \\ \hline \end{array}.$$

*Basis:* The formula  $\phi_{start}$  is satisfiable if and only if the first row of the table is a start configuration.

*Hypothesis:*  $C_i$  is a reachable configuration.

*Induction:* If all windows of  $(C_i, C_{i+1})$  are valid, then  $C_{i+1}$  is also a reachable configuration i.e.  $C_i \rightarrow_N C_{i+1}$ .

We call an window  $W_{ij}$  if the  $cell[i, j]$  is in its upper central position. In  $W_{ij}$ , if upper three symbols are tape symbols, then the content of  $cell[i, j]$  (upper-central) is same as the content of  $cell[i + 1, j]$  (central-lower). The central cell does not change if there is no adjacent state symbol.

If a  $W_{ij}$  contains a state symbol in  $cell[i, j]$  (top-center), it is guaranteed that the lower three cells are updated properly following the transition relation of  $N$ .

$$\phi_{move} = \bigwedge_{\substack{1 \leq i < n^k \\ 1 < j \leq n^k}} \text{valid } W_{ij}.$$

Each valid  $W_{ij}$  can be replaced by the content of its cells. Let the possible contents of 6-cells be  $a_1, \dots, a_6$ . The “valid  $W_{ij}$ ” can be replaced by

$$\bigvee_{\text{valid } a_1, \dots, a_6} (v_{i,j-1,a_1} \wedge v_{i,j,a_2} \wedge v_{i,j+1,a_3} \wedge v_{i+1,j-1,a_4} \wedge v_{i+1,j-,a_5} \wedge v_{i+1,j+1,a_6})$$

The time complexity of the reduction is as follows:

- The variables are of the form  $v_{i,j,p}$ , where  $1 \leq i, j \leq n^k$ ,  $p \in C = Q \cup \Sigma \cup \{\triangleleft\}$ . The number of variables are  $|C| \times n^k \times n^k = O(n^{2k})$  as  $|C|$  does not depend on the length of the input. Lengths of  $i, j$  and  $p$  takes  $2k \log n + \log p = O(\log n)$  bits. There is a length of  $O(\log n)$  bits for each variable.
- The formula for the validity of cells,  $\phi_{cell}$  is a conjunction over  $n^k \times n^k$  cells. The length of each conjuncts is independent of the length of input  $x$ . So the the length of  $\phi_{cell}$  is  $O(n^{2k})$ .
- The formula  $\phi_{start}$  encodes the first row with  $n^k$  variables and  $n^k - 1$  ‘ $\wedge$ ’ operators. Its length is  $O(n^k \log n)$ . The contribution
- The formula  $\phi_{accept}$  is a disjunction over all cells. Its length is  $O(n^{2k} \log n)$ .
- Similarly the formula for moves,  $\phi_{move}$  is over all windows, over all (almost) cells is also  $O(n^{2k} \log n)$ . The number of valid windows is independent of the length input  $x$ .

The total length of the formula is  $O(n^{2k} \log n)$ . The claim that it can be generated in polynomial time due to its *repetitive nature!*

## 1.4 Reduction

We have already proved that  $3SAT \in NP$ . Now we reduce **SAT** to **3SAT** in polynomial time to show the following.

**Proposition 5.** 3SAT is NP-hard

**Proof:** Following is a reduction of SAT to 3SAT. Consider a CNF formula  $\phi = C_1 \wedge \dots \wedge C_k$ . We wish to transform it in equivalent 3CNF formula  $\psi$ . Let the clause  $C_i$  has  $m > 3$  literals i.e.  $C_i = l_{i1} \vee \dots \vee l_{im}$ . We introduce a new variable  $z_{i1}$  and write  $f_1 = (l_{i1} \vee \dots \vee l_{i(m-2)} \vee z_{i1}) \wedge (\overline{z_{i1}} \vee l_{i(m-1)} \vee l_{im})$ . If there is an assignment that makes  $C_i$  false (all its literals are false), then no assignment of  $z_{i1}$  can make  $f_1$  true. On the other hand, if there is an assignment that makes  $C_i$  true, then there is an assignment of  $z_{i1}$  that makes  $f_1$  true. If in the given satisfying assignment both  $l_{i(m-1)}$  and  $l_{im}$  are false then  $z_{i1} \leftarrow 0$ , else  $z_{i1} \leftarrow 1$ .

This process increases the length of the formula by 4 (increase in the number of  $\vee$  and  $\wedge$ ) and reduce the clause size to  $m-1$ . If the transformation is repeated for  $m-3$  times, the increase in length is by  $4(m-3)$ . QED.

We reduce 3SAT to the following set to prove that it is NP-hard.

**Proposition 6.**

$$INDSET = \{ \langle G, k \rangle : \exists S \subseteq V(G) \text{ s.t. } |S| \geq k \text{ and } \forall u, v \in S, \{u, v\} \notin E(G) \}$$

is NP-complete.

**Proof:** We show two different reductions.

**First reduction:** Let there be  $m$  number of 3-literal clauses in the 3CNF Boolean formula. Each clause  $C$  gives a triangle  $T$  with the vertices labelled by the literals. If two clauses  $C_i$  and  $C_j$  has a variable  $x_k$  and its negation  $\overline{x_k}$ , we join the corresponding vertices by an edge (edge for inconsistency).

If there is a satisfying assignment  $v : Var \rightarrow \{0, 1\}$ , then each clause is also satisfied, so there is a vertex in each triangle whose literal value is 1. These  $m$  vertices will form an independent set. There cannot be any edge between a pair of such vertices. An edge between two triangles is between a variable and its negation.

We cannot form an independent set by taking two vertices from a triangle. Also we cannot take two vertices of two triangles that are connected by an edge (inconsistent). So, if there is an independent set of size  $m$ , assigning 1 to corresponding literals gives a satisfying assignment. There may be some extra variables, that can be assigned any value.

**Second Reduction:** Associate a complete graph of 7 vertices to every clause. So there are  $7m$  vertices. Among the eight possible assignments,  $\{000, \dots, 111\}$ , one will make a clause false e.g. if the clause is  $x_1 \vee \overline{x_4} \vee \overline{x_{11}}$ , then the assignment  $x_1 \leftarrow 0, x_4 \leftarrow 1, x_{11} \leftarrow 1$  will make it false. Associate remaining seven satisfying assignments to seven nodes of the clause. If two nodes in two different clauses have a common variable assigned to different values, 0 and 1, join them by an edge (inconsistency).

If there is a satisfying assignment  $v : Var \rightarrow \{0, 1\}$  of  $\phi$ , then pick-up a vertex from the seven nodes of a clause  $C$  which has the restriction of  $v$  to the variables of the clause. This selected vertex cannot have any edge going out of the clause (7-node complete subgraph) to another selected vertex of a different clause, as they are selected using a satisfying assignment. So there is an independent set of size  $m$ .

An independent set cannot take more than one vertex from the 7-vertices of any clause. If there is an independent set of size  $m$ , their vertices are coming from  $m$  different clauses. There cannot be any edge between these vertices as they form an independent set. The assignment given to the variables *locally* to every clause gives a consistent *global* assignment. As an example corresponding to the clause  $x_1 \vee \overline{x_4} \vee \overline{x_{11}}$ , if the vertex with the assignment  $x_1 \leftarrow 1$ ,  $x_4 \leftarrow 0$ ,  $x_{11} \leftarrow 1$ , is an element of the independent set, then there is a satisfying assignment that is an extension of this local assignment. QED.

**Example 6.** Consider the following 3SAT formula and show both the reductions.

$$(x_1 \vee x_2 \vee x_3) \wedge (\overline{x_1} \vee \overline{x_2} \vee \overline{x_3}) \wedge (\overline{x_1} \vee x_2 \vee x_3)$$

**Proposition 7.** SUBSET-SUM =  $\{ \langle S, t \rangle : S = \{x_1, \dots, x_k : x_i \in \mathbb{N}\}$  is a multiset and for some  $\{y_1, \dots, y_l\} \subseteq S$ ,  $\sum_{i=1}^l y_i = t \}$ . is **NP**-complete.

**Proof:** The certificate for SUBSET-SUM is  $C$ , a collection of elements of  $S$ . Following is a verifier.

$V =$  "On input  $\langle \langle S, t \rangle, C \rangle$

1. Test whether  $C \subseteq S$ .
2. Test whether  $\sum C = t$ .
3. *Accept* if both are *true*, else *reject*."

So SUBSET-SUM  $\in$  **NP**.

We reduce a 3SAT formula  $\phi$  to an instance of a SUBSET-SUM problem in polynomial time to show that it is **NP**-complete.

Let the variables of  $\phi$  be  $x_1, \dots, x_l$  and the clauses be  $C_1, \dots, C_k$ . Following table shows the elements of  $S$  and the value  $t$  constructed from the formula  $\phi$  such that  $\langle S, t \rangle \in$  SUBSET-SUM if and only if  $\phi$  is satisfiable.

Table ( $T$ )

	Variables						Clauses				
	$x_1$	$x_2$	$x_3$	$x_4$	$\dots$	$x_l$	$c_1$	$c_2$	$c_3$	$\dots$	$c_k$
$y_1$	1	0	0	0	$\dots$	0	1	0	0	$\dots$	0
$z_1$	1	0	0	0	$\dots$	0	0	1	1	$\dots$	0
$y_2$		1	0	0	$\dots$	0	0	1	1	$\dots$	0
$z_2$		1	0	0	$\dots$	0	1	0	0	$\dots$	0
$y_3$			1	0	$\dots$	0	0	0	0	$\dots$	0
$z_3$			1	0	$\dots$	0	1	1	1	$\dots$	0
$\vdots$	$\vdots$		$\vdots$		$\vdots$		$\vdots$		$\vdots$	$\vdots$	
$y_l$					$\dots$	1	0	0	0	$\dots$	0
$z_l$					$\dots$	1	0	0	0	$\dots$	0
$g_1$					$\dots$	0	1	0	0	$\dots$	0
$h_1$					$\dots$	0	1	0	0	$\dots$	0
$g_2$					$\dots$	0	0	1	0	$\dots$	0
$h_2$					$\dots$	0	0	1	0	$\dots$	0
$g_3$					$\dots$	0	0	0	1	$\dots$	0
$h_3$					$\dots$	0	0	0	1	$\dots$	0
$\vdots$	$\vdots$		$\vdots$		$\vdots$		$\vdots$		$\vdots$	$\vdots$	
$g_k$					$\dots$	0	0	0	0	$\dots$	0
$h_k$					$\dots$	0	0	0	0	$\dots$	0
$t$	1	1	1	1	$\dots$	1	3	3	3	$\dots$	3

Each row of the table (other than  $t$ ) corresponds to a decimal number, member of  $S$ . These decimal numbers use digits 0 and 1. The decimal number  $t$  uses digits 1 and 3. Blanks correspond to zeros.

- (a) For each variable  $x_i$  there are a pairs of numbers  $y_i$  and  $z_i$ . The digits of each of them is partitioned in to two parts, the *variable* part (left side) and the *clause* part.
- (b) The digit in  $T[y_i, x_i] = T[z_i, x_i] = 1$ . All other digits in the *variable* part are 0's. We select  $y_i$  from  $S$  if the truth value of  $x_i \leftarrow 1$ . Otherwise select  $z_i$ .
- (c) The digits in  $T[y_i, c_j] = 1$  if the clause  $c_j$  has the literal  $x_i$ . The digit in  $T[z_i, c_j] = 1$  if the clause  $c_j$  has the literal  $\overline{x_i}$ . Other digits are 0's.
- (d)  $S$  also contains a pair of numbers  $g_j$  and  $h_j$  for each clause  $c_j$ . The digit in  $T[g_j, c_j] = T[h_j, c_j] = 1$ . All other digits of these numbers are 0's.
- (e) The digits in the *variable* part of  $t$  are all 1's and the digits in the *clause* part of  $t$  are all 3's.
- (f) The target is to get the value of  $t$  after adding the selected numbers  $y_i$  or  $z_i$  for  $i = 1, 2, \dots, l$  (each variable) and zero, one or both of  $g_j, h_j$  for  $j = 1, 2, \dots, k$  (each clause).

Consider  $\phi = C_1 \wedge C_2 \wedge C_3$  where  $C_1 = x_1 \vee \overline{x_2} \vee \overline{x_3}$ ,  $C_2 = \overline{x_1} \vee x_2 \vee x_3$  and  $C_3 = \overline{x_1} \vee \overline{x_2} \vee \overline{x_3}$ . A satisfying assignment is  $x_1 \leftarrow 1$ ,  $x_2 \leftarrow 1$ , and  $x_3 \leftarrow 0$ . We choose  $y_1, y_2, z_3$  (ignore other rows and columns of the table). So far the sum is  $100100 + 010011 + 001111 = 111222$ . We also choose  $g_1, g_2, g_3$  to make the final sum equal to 111333 ( $t$ ).

If  $\phi$  is satisfiable: there is a truth assignment for each variable. If  $x_i \leftarrow 1$ , we choose the number  $y_i$ . If  $x_i \leftarrow 0$ , we choose the number  $z_i$ . Whatever be the case, when added we get 1 in first  $l$  digits of  $t$ .

At least one of the three literals of a clause  $C_j$  must be true. It may be due to  $l_i$ . If  $l_i = x_i$  i.e.  $x_i \leftarrow 1$ , we have already chosen  $y_i$  which has 1 in its  $c_j$  column. If  $l_i = \overline{x_i}$ , we have chosen  $z_i$  and it has 1 in its  $c_j$  column. The sum of the digits of the column  $c_j$  for a satisfying assignment can be 1, 2, or 3. They can all be brought to 3 by adding  $g_j, h_j$ . But that is not possible if a clause is unsatisfiable.

If subset of  $S$  gives the sum  $t$ : for every  $i$  either  $y_i$  or  $z_i$  is chosen, but not both, as first  $l$  digits of  $t$  are all 1's. In column  $c_j$  at most 2 can be supplied from  $g_j$  and  $h_j$ . So 1 must come from the literal of a clause. So the clause is satisfied. QED.

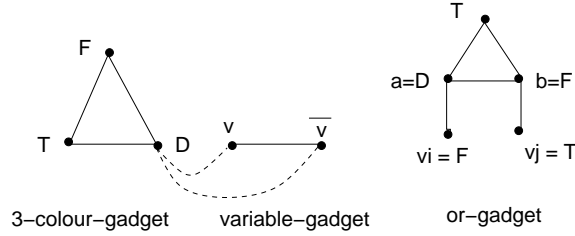
**Proposition 8.** 3COL = { $\langle G \rangle$ : graph  $G$  has a vertex colouring with at most three colours} is NP-complete.

**Proof:** The certificate of 3COL is colouring of different vertices. A polynomial time verifier can check validity of colouring in polynomial time. So 3COL is in NP.

We reduce 3SAT to 3COL. Let  $\phi$  be a 3CNF formula with  $m$  clauses  $c_1, \dots, c_m$  and  $n$  variables  $x_1, \dots, x_n$ . The construction is as follows:

1. There is a pair of vertices  $v_i, \overline{v_i}$  for every variable  $x_i$  and its negation  $\overline{x_i}$ .

2. Five vertices  $u_{i1}, \dots, u_{i5}$  for each clause  $c_i$ .
3. Three special vertices  $T, F, D$  for three colours *true*, *false* and  $D$ .

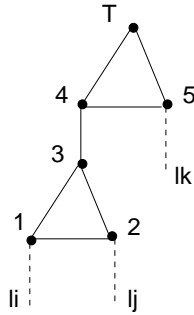


Form a triangle with  $T, F, D$  to force three colours to colour them.  
 Form a triangle with  $v_i, \bar{v}_i$  and  $D$  so that a variable can take either colour  $T$  (*true*) or  $F$  (*false*) and not  $D$ .

The difficult part is to ensure that at least one literal in every clause is *true* if and only if the graph is 3-colourable.

We start with a graph of 3-vertices,  $a, b,$  and  $T$  forming a triangle. The vertex  $a$  is connected to a literal-vertex  $v_i$  and  $b$  to a literal-vertex  $v_j$ . In the triangle of  $a, b, T$ ,  $a$  and  $b$  can be coloured with  $F$  and  $D$  only. Literal vertices can be coloured only with  $T$  and  $F$ . So one literal must be coloured with  $T$ . This is called an “or gadget”.

Now we look into the five vertices  $u_{i1}, \dots, u_{i5}$  corresponding to the clause  $c_i$ . The corresponding graph has following edges:  $\{\{u_{i1}, u_{i2}\}, \{u_{i1}, u_{i3}\}, \{u_{i2}, u_{i3}\}\}, \{\{u_{i3}, u_{i4}\}, \{u_{i4}, u_{i5}\}, \{u_{i5}, T\}, \{u_{i4}, T\}\}$  and  $\{\{u_{i1}, l_i\}, \{u_{i2}, l_j\}, \{u_{i5}, l_k\}\}$ , where  $v_i, v_j, v_k$  are vertices corresponding to literals  $l_i, l_j, l_k$  respectively.

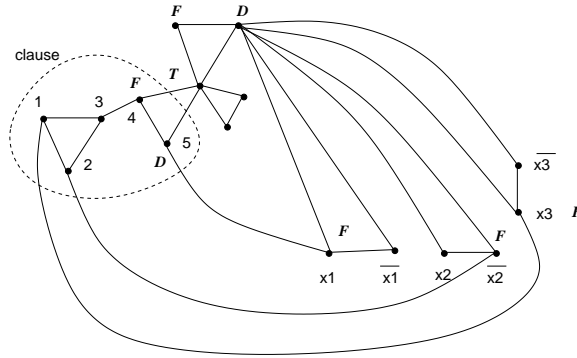


Following are the possible colour assignments:

$u_{i5}$	$u_{i4}$	$u_{i3}$	$u_{i1}$	$u_{i2}$	Literal coloured $T$
$F$	$D$				$l_k$
$D$	$F$	$T$	$F$	$D$	$l_i$
$D$	$F$	$T$	$D$	$F$	$l_j$
$D$	$F$	$D$	$T$	$F$	$l_j$
$D$	$F$	$D$	$F$	$T$	$l_i$

So one literal must be coloured  $T$ . The claim is that 3CNF formula  $\phi$  is satisfiable if and only if the graph is 3-colourable.

Following figure shows an example with a clause  $C = x_1 \vee \bar{x}_2 \vee x_3$ .



If all three literals are *false*, then node 1 and 2 are coloured with *T* and *D*. But that needs a 4th colour for node 3. But the table shows that if one of the literal is *true* i.e. coloured with *T*. then the graph is 3-colourable. QED.

**Proposition 9.**  $\text{dHAMPATH} = \{ \langle G, s, d \rangle : G \text{ is a directed graph with a Hamiltonian path from } s \text{ to } d \}$ .

**Proof:** It is clear that  $\text{dHAMPATH}$  is in **NP**. A sequence of vertices on the path is a certificate. This can be verified in polynomial time.

We reduce  $3\text{SAT}^7$  to  $\text{dHAMPATH}$  in polynomial time. Consider a 3CNF formula with  $m$  clauses and  $n$  variables,  $x_1, \dots, x_n$ .

$$\phi = (l_{11} \vee l_{12} \vee l_{13}) \wedge \dots \wedge (l_{m1} \vee l_{m2} \vee l_{m3}).$$

where  $l_{ij}$ ,  $1 \leq i \leq m$  and  $1 \leq j \leq 3$ , is  $x_k$  or  $\bar{x}_k$ , for some  $k$ ,  $1 \leq k \leq n$ .

There is a starting vertex labelled with  $s$  and an end vertex labelled with  $d$ .

For every variable  $x_k$  there is a *doubly linked chain-graph* of  $3m + 1$  vertices.

There is a vertex  $s_{i(i+1)}$  between every pair of doubly linked chain-graphs corresponding to variables  $x_i$  and  $x_{i+1}$ ,  $1 \leq i < n$ . There are directed edges, from  $s$  to the two ends of the *doubly linked graph* of  $x_1$ , from  $s_{i(i+1)}$  to the two ends of the *doubly linked graph* of  $x_{i+1}$ , from the two ends of the *doubly linked graph* of  $x_i$  to  $s_{i(i+1)}$ ,  $1 \leq i < n$ , and from two ends of the *doubly linked graph* of  $x_n$  to  $d$ .

For every clause there is a vertex. Call them  $c_1, \dots, c_m$ . Each doubly linked graph corresponding to a variable has a pair of nodes corresponding to a clause. Every such pair is separated by a node, and there are two terminal nodes. This accounts for the number  $3m + 1$ ,

$$\bigcirc \bigcirc_1 \bigcirc_1 \bigcirc \bigcirc_2 \bigcirc_2 \bigcirc \dots \bigcirc \bigcirc_m \bigcirc_m \bigcirc.$$

If a clause  $c_j$  has  $x_i$ , then there is a directed edge from the left node of the pair  $\bigcirc_i \bigcirc_i$  of the variable graph of  $x_i$  to  $c_j$  and a directed edge from  $c_j$  to the right node of the pair. If it is  $\bar{x}_i$  then these two directed edges are reversed.

If there is a satisfying assignment of a 3CNF formula, then every variable  $x_i$  is either 1 or 0. If  $x_i \leftarrow 1$ , then the path starts from the left end of the *doubly linked graph* of  $x_i$ . If  $x_i \leftarrow 0$ , then the path starts from the right end of the *doubly linked graph* of  $x_i$ .

So there is a path from  $s$  through different variable nodes to  $d$ . To cover the nodes corresponding to the clauses, take one literal per clause that makes

<sup>7</sup>Actually we reduce SAT formula.



it true. Let the literal  $l_i$  ( $x_i$  or  $\bar{x}_i$ ) is true for the clause  $c_j$ . Break the path of the *doubly linked graph* of  $x_i$  and include  $c_j$  in it.

In the other direction, if there is a Hamiltonian path from  $s$  to  $d$ , then there is a *truth value* assignment for the formula.

The number of vertices of the formula is  $2 + m + (3m + 1) + (n - 1)$ . So the encoding of the graph is a polynomial over the encoding of the formula.

QED.

## 1.5 Search Problem

We have asked membership question about the languages in **NP** e.g. whether the formula is satisfiable, whether the graph has an independent set of size  $k$ , whether the directed graph has a Hamiltonian path etc. These are decision problems.

We may search for solution, if there is one, for each such problems e.g. give a satisfying assignment of the formula, give an independent set of size  $k$ , give a Hamiltonian path etc.

Search problems are in general more difficult than the corresponding decision problem. It is easier to answer whether a positive integer ( $> 1$ ) is composite, but more difficult to get its factorization. But if an **NP-complete** problems can be solved in polynomial time i.e.  $\mathbf{P} = \mathbf{NP}$ , then the certificate of any **NP** language can be generated in polynomial time.

**Proposition 10.** If  $\mathbf{P} = \mathbf{NP}$ , then for each  $L \in \mathbf{NP}$  and its verifier  $V$ , there is a polynomial time Turing machine  $M$  that can generate a certificate  $w$  with respect to  $V$ , when run on  $x \in L$ .

**Proof:** We need to show that, if  $\mathbf{P} = \mathbf{NP}$ , then for each polynomial time bounded Turing machine  $M$  and for each polynomial  $p(n)$ , there is a polynomial time bounded Turing machine  $M'$  with the following property.

For every  $x \in \{0, 1\}^n$ , if there is a  $w \in \{0, 1\}^{p(n)}$  such that  $M$  accepts  $\langle x, w \rangle$  i.e.  $M(x, w) = 1$ , then  $M'$  on input  $x$  produces  $w$  as the output i.e.  $M'(x) = w$ .

We consider the case of SAT. We assume that a Turing machine  $A$  decides the membership of SAT in polynomial time (this amounts to saying  $\mathbf{P} = \mathbf{NP}$ ). We show that there is polynomial time Turing machine  $B$ , that on input of a satisfiable CNF formula  $\phi$  of  $n$  variables,  $\phi(x_1, \dots, x_n)$ , produces a satisfying assignment.

The Turing machine  $B$  works as follows:

1. Run  $A$  on  $\phi$  to check whether it is satisfiable or not.
2. If  $\phi$  is satisfiable, then for  $i \leftarrow 1, \dots, n$  do the following steps.
3. Assign  $x_i$  to 0, simplify the formula to  $n - i$  variables., and run  $A$  to check whether  $\phi(v_i, \dots, v_{i-1}, 0, x_{i+1}, \dots, x_n)$  is satisfiable, where  $v_1, \dots, v_{i-1}$  are already known assignments.
4. If it is, continue; otherwise continue with  $\phi(v_i, \dots, v_{i-1}, 1, x_{i+1}, \dots, x_n)$  (simplified).
5. At the end either it is known that  $\phi$  is unsatisfiable, or we have the satisfying assignment.

$B$  is clearly polynomial time Turing machine.

Any  $L \in \mathbf{NP}$  is *Levin reducible* to SAT, so a satisfying assignment of  $f(x) = \phi_x \in \text{SAT}$  can be mapped back to the witness of  $x \in L$ . QED.

## 1.6 Reduction to SAT

The set of  $\mathbf{NP}$ -complete problems is closed under Karp-reduction. An obvious question is how do we reduce *INDSET* to SAT. This time the input is  $\langle G, k \rangle$ , where  $G = (V, E)$  is an undirected graph and  $k$  is a positive integer. The element  $\langle G, k \rangle \in \text{INDSET}$  if  $G$  has a independent set of size  $k$ . We define a computable map  $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$  such that  $f(G, k) = \phi$  and

$\langle G, k \rangle \in \text{INDSET}$  if and only if  $\phi$  is satisfiable.

We need to choose the boolean variables and encode the independent set constraints as a boolean formula. This is in the similar line of encoding the computation of an NTM as a boolean formula.

Let  $V = \{v_1, \dots, v_n\}$  and an independent set of size  $k$  be  $I = \{u_1, \dots, u_k\}$ . We introduce variables  $x_{ij}$ , where  $1 \leq i \leq n$  and  $1 \leq j \leq k$ . The variable  $x_{ij}$  is *true* if  $v_i = u_j$ . Following is the set of constraints.

1. Each  $u_j \in I$  must be some vertex of the graph. This is captured the following set of clauses.

$$\bigwedge_{j=1}^k \left( \bigvee_{i=1}^n x_{ij} \right).$$

2. No vertex should occur in  $I$  twice.

$$\bigwedge_{i=1}^n \bigwedge_{1 \leq j < m \leq k} (\neg x_{ij} \vee \neg x_{im}).$$

3. No element of  $I$  can be associated to two vertices of the graph.

$$\bigwedge_{j=1}^k \bigwedge_{1 \leq i < m \leq n} (\neg x_{ij} \vee \neg x_{mj}).$$

4. If two vertices are connected by an edge, then both of them cannot be in  $I$ .

$$\bigwedge_{1 \leq j < m \leq k} \bigwedge_{(v_i, v_l) \in E} (\neg x_{ij} \vee \neg x_{lm}).$$

This construction (reduction) can be done in time polynomial of the input length. The time complexity of the reduction is  $O(nk + nk^2 + kn^2 + k^2e) = O(k^2n^2)$ , where  $e = |E|$

## 1.7 coNP, EXP, and NEXP

The class  $\mathbf{coNP}$  was defined as follows:

$$\mathbf{coNP} = \{L \subseteq \{0, 1\}^* : \bar{L} \in \mathbf{NP}\}.$$

The class  $\mathbf{P}$  is closed under complementation, so  $\mathbf{P} \subseteq \mathbf{NP} \cap \mathbf{coNP}$ . We already know that the following language are in  $\mathbf{coNP}$ .

- Any language in **P** e.g. PRIME.
- $\overline{SAT} = \{\phi : \phi \text{ is unsatisfiable}\}$ .
- $\overline{INDSET}, \overline{VERTEX - COVER}, \overline{CLIQUE}$  etc.

We may define the class **coNP** using a deterministic verifier.

**Definition 5.** A language  $L \subseteq \{0, 1\}^*$  is in **coNP** if and only if there is a polynomial  $p : \mathbb{N}_0 \rightarrow \mathbb{N}_0$  and a polynomial time Turing machine so that for all  $x \in \{0, 1\}^*$ ,

$$x \in L \text{ if and only if } \forall w \in \{0, 1\}^{p(|x|)}, M \text{ accepts } \langle x, w \rangle.$$

This is actually negation of the definition of **NP**.  
Let  $L \in \mathbf{coNP}$ . So  $\overline{L} \in \mathbf{NP}$ . For all  $x \in \{0, 1\}^*$ ,

$$x \notin L \text{ if and only if } x \in \overline{L}.$$

There is a polynomial time bounded deterministic Turing machine  $V$ , a polynomial  $p(n)$ , and a witness  $w \in \{0, 1\}^{p(|x|)}$ , such that  $V$  accepts  $\langle x, w \rangle$  if and only if  $x \in \overline{L}$  i.e.

$$x \notin L \text{ if and only if } \exists w \in \{0, 1\}^{p(|x|)}, V \text{ accepts } \langle x, w \rangle.$$

Equivalently,

$$x \in L \text{ if and only if } \neg(\exists w \in \{0, 1\}^{p(|x|)}, V \text{ accepts } \langle x, w \rangle),$$

i.e.

$$x \in L \text{ if and only if } \forall w \in \{0, 1\}^{p(|x|)}, \overline{V} \text{ accepts } \langle x, w \rangle,$$

where  $\overline{V}$  is same as  $V$  in all respect, but the *accept* and *reject* states exchanged.

A language  $L$  is **coNP** complete if (i) it is in **coNP**, and (ii) every language  $L'$  in **coNP** is Karp reducible to  $L$ .

**Proposition 11.** Following language is **coNP**-complete.

$$TAUTOLOGY = \{\phi : \phi \text{ is a Boolean formula satisfiable by any assignment}\}.$$

Note that a formula  $\phi \in TAUTOLOGY$  if and only if  $\neg\phi$  is *unsatisfiable*.

**Proof:** If  $\phi$  is a Boolean formula with  $n$  variables, then it is a *tautology* if and only if it is satisfied by any assignment of  $n$  variables. So there is a polynomial time Turing machine  $V$  such that for any  $x \in \{0, 1\}^n$ ,  $V$  will evaluate  $\phi$  with  $x$  as assignment to its variables. The formula  $\phi$  is a *tautology* if it evaluates to *true* (i.e. 1) for all  $x$ . So  $TAUTOLOGY \in \mathbf{coNP}$ .

We now show that every language  $L \in \mathbf{coNP}$  is Karp reducible to TAUTOLOGY. We take  $\overline{L}$ , the complement of  $L$ . If  $L \in \mathbf{coNP}$ , then  $\overline{L} \in \mathbf{NP}$ . So by Cook-Levin reduction we get  $\phi_x$ . We know,  $x \in L$  if and only if  $x \notin \overline{L}$  if and only if  $\phi_x$  is *unsatisfiable*. So,  $x \in L$  if and only if  $\neg\phi_x$  is a *tautology*.

The reduction is, for all  $x \in \{0, 1\}^*$ , create  $\phi_x$  by Cook-Levin reduction and take the negation of the formula. QED.

It is clear that if  $\mathbf{P} = \mathbf{NP}$ , then  $\mathbf{NP} = \mathbf{coNP} = \mathbf{P}$ .

**Definition 6.** We define the class  $\mathbf{NEXP} = \bigcup_{c \geq 1} \mathbf{NTIME}(2^{n^c})$ . By definition we have  $\mathbf{P} \subseteq \mathbf{NP} \subseteq \mathbf{EXP} \subseteq \mathbf{NEXP}$ . We prove the following proposition.

**Proposition 12.** If  $\mathbf{EXP} \neq \mathbf{NEXP}$ , then  $\mathbf{P} \neq \mathbf{NP}$ .

**Proof:** We prove the contrapositive statement. We assume  $\mathbf{P} = \mathbf{NP}$  and prove that  $\mathbf{EXP} = \mathbf{NEXP}$ .

Let  $L \in \mathbf{NTIME}(2^{n^c})$ . So a non-deterministic Turing machine  $N$  decides  $L$  in time  $2^{n^c}$ . We define the language

$$L_{pad} = \{ \langle x, 1^{2^{|x|^c}} \rangle : x \in L \},$$

and claim that  $L_{pad} \in \mathbf{NP}$ . The non-deterministic Turing machine  $N_{pad}$  for  $L_{pad}$  is as follows.

$N_{pad}$ : input  $y$

1. Nondeterministically it guesses a  $z$ , and computes  $2^{|z|^c}$ , so that  $y = \langle z, 1^{2^{|z|^c}} \rangle$ . It *rejects* the input if no such  $z$  is found.
2. Otherwise, simulate  $N$  on  $z$  for  $2^{|z|^c}$  steps.
3. If  $N$  accepts  $z$ , then *accept*, else *reject*.

The running time of  $N_{pad}$  is polynomial in  $|y|$ , so  $L_{pad} \in \mathbf{NP}$ . But according to our assumption  $L_{pad} \in \mathbf{P}$ . But then  $z \in L$  if and only if  $\langle z, 1^{2^{|z|^c}} \rangle \in L_{pad}$ . The padding string can be attached to  $z$  in exponential time and membership of  $\langle z, 1^{2^{|z|^c}} \rangle \in L_{pad}$  in  $L_{pad}$  can be tested in polynomial (on the length of  $\langle x, 1^{2^{|x|^c}} \rangle$ ) time.

Therefore the membership of  $x$  in  $L$  is determined in exponential (on the length of  $x$ ). So  $L \in \mathbf{EXP}$  i.e.  $\mathbf{NEXP} \subseteq \mathbf{EXP}$ . QED.

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