

## Multitrack Turing Machine

a	a	b	B	...
b	c	b	B	...
c	b	c	B	...

Finite control

$$\Sigma = \{a, b, c\} \quad \Rightarrow \quad \Sigma = \{b, a, c\}$$

a	a	b	B	...
b	c	b	B	...
c	b	c	B	...

Finite control

single Track TM

## Multitrack Turing Machine

$$\delta(q_0, [a, b, c]) = (q_1, [b, b, b], L/R)$$

\* Can be convertible to single track TM.

a. Design TM to multiply 2 integers

Sol:

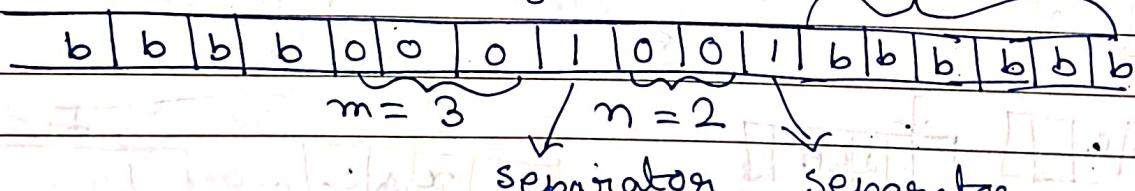
$$3 \times 2 = 2 + 2 + 2$$

$$3 \times 2 = 0 + 2 + 2 + 2$$

$$3 \times 2 = 0 + 2 + 2 + 2$$

$$= 6$$

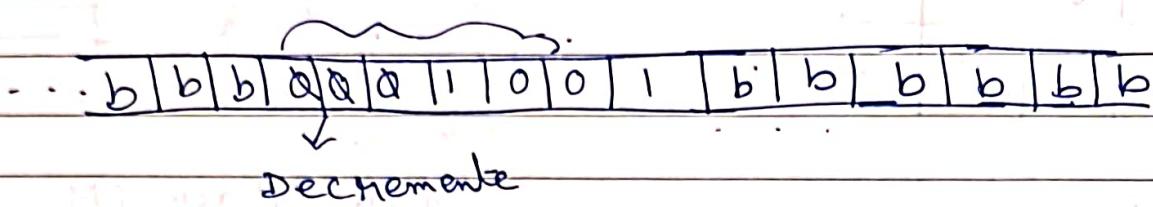
Answer



The steps

1.  $0^m | 0^n |$  is placed on the tape  
(the off will be written after the rightmost 1)
2. The leftmost '0' is erased
3. A block of  $n$  '0's is copied onto the right.  
(concept in  $3 \times 2 = 2 + 2 + 2$ )

4. Step 2 & 3 are repeated n times and 10<sup>m</sup>10<sup>n</sup> is obtained on the tape.



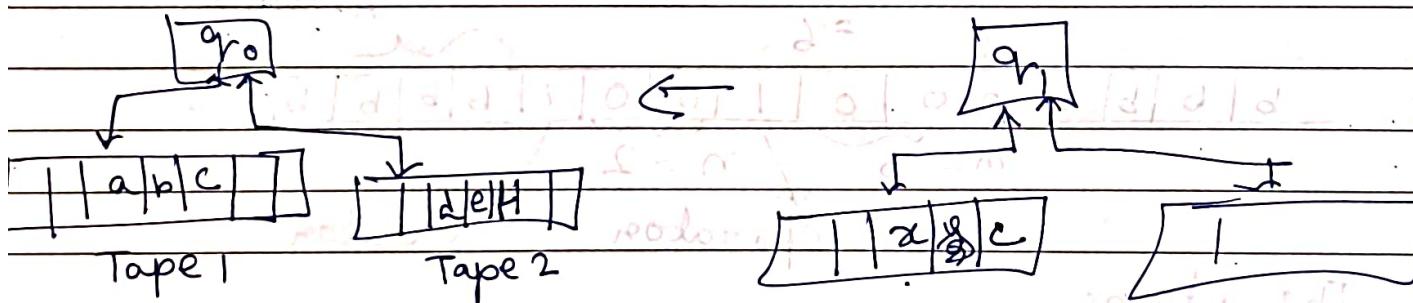
## Subroutine

This will have a initial state and a return state. After reaching the return state, there is a ~~temporary~~ temporary halt.

## Non-Deterministic Turing Machine

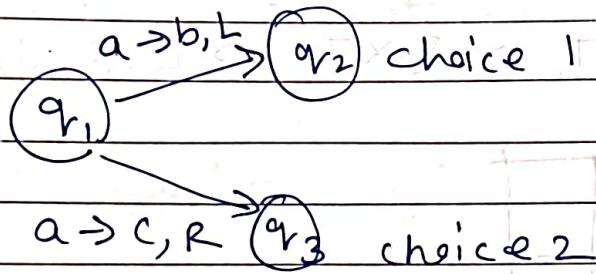
e.g. if  $n = 2$  with current configuration

$$\delta(q_0, a, e) = (q_1, x, y, L, R) \quad N = 2$$



## Non-Deterministic Turing Machine

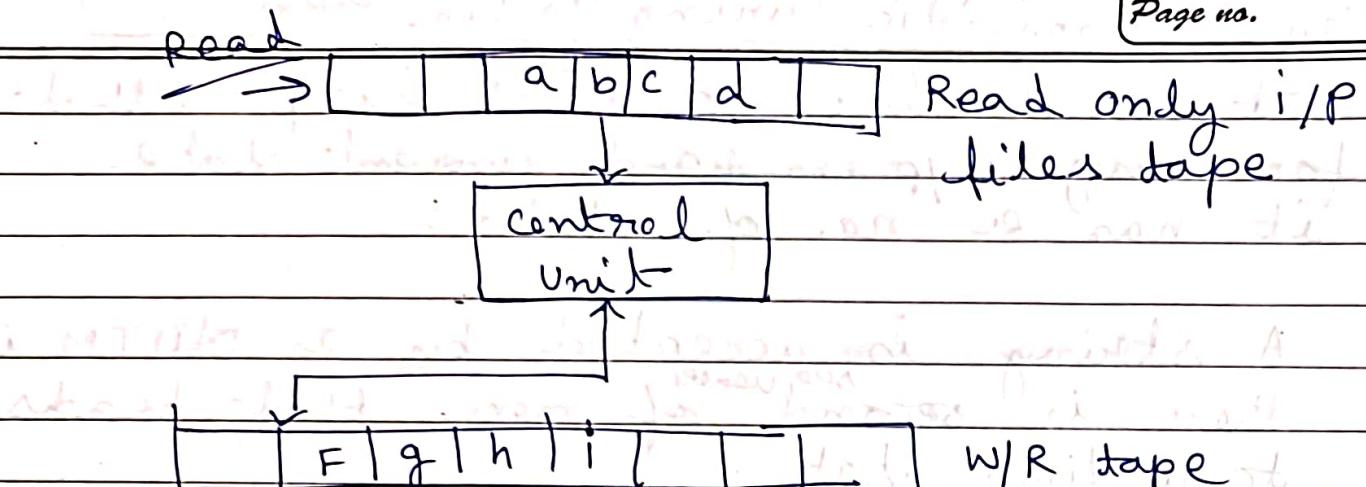
- It is similar to DTM except that for any i/p example and current state it has a no. of choices.
- A string is accepted by a NDTM if there is <sup>sequence</sup> of moves that leads to a final state.
- The transition ~~fn.~~ <sup>multiple choices</sup> fn.
- A NDTM is allowed to have more than one transition fn. a given tape symbol.



## OFFline turing machine

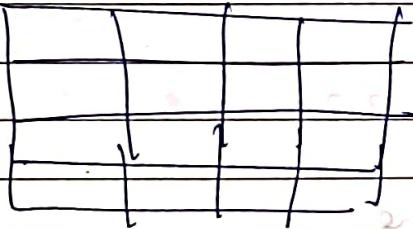
Has 2 tapes.

- ① One tape is read only and contains the i/p.
- ② The other is read-write and is initially blank.



## Multidimensional turing machine

- It has a multidimensional tape  
eg. A two dimensional T. M. would read and write on an infinite tape divided into ~~sqrs.~~ squares, like a checkboard.
- $\delta: Q \times T \rightarrow Q \times T \times \{L, R, U, D\}$



## Recursive & Recursively Enumerable

Language :- ~~one of sixes~~ M.T. accepts A ~~in front~~

⇒ The T.M. may do any of the following

- Halt & accept the i/p

- Halt & Reject the i/p, or

- Never halt / loop

### Recursively Enumerable lang :-

There is a T.M. for a language which accept every string otherwise

not. ~~and other incomprehensible word~~

### Recursive lang :-

There is a T.M. for a language which halt on every string.

### Halting Problem :-

Halting problem is undecidability, it is not a problem, it just asks questions :- "is it possible to tell whether a given m/c will halt for same given i/p".

e.g. Input :- A T.M. & i/p string w

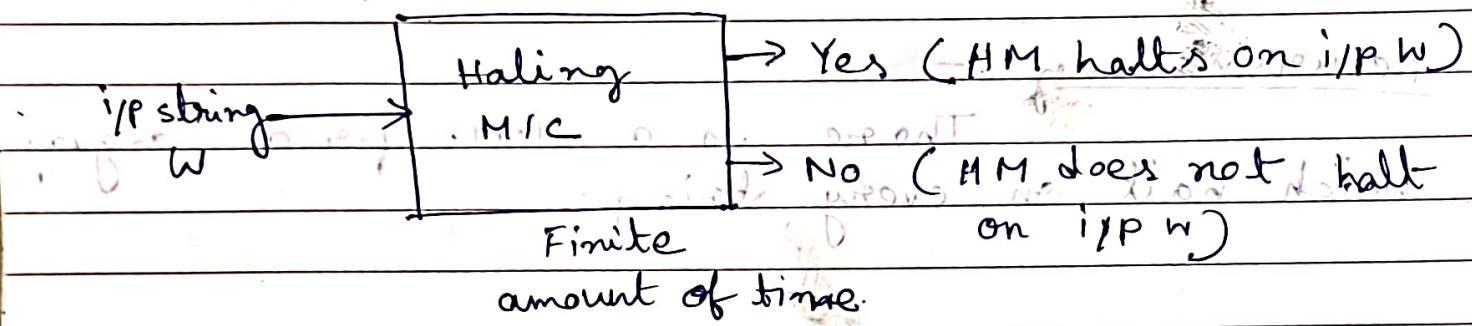
problem :- Does the TM finish computing of the ~~string~~ string w in a finite no. of steps? The answer must be yes or no.

Proof: Assume T.M. exists to solve this problem & then we will show it is contradicting itself.

We will call this TM as a Halting machine that produces a 'yes' or 'no' in a finite amount of time.

→ If the halting m/c finishes in a finite amount of time, and o/p comes as 'yes' otherwise 'no'.

The block diagram of Halting m/c -

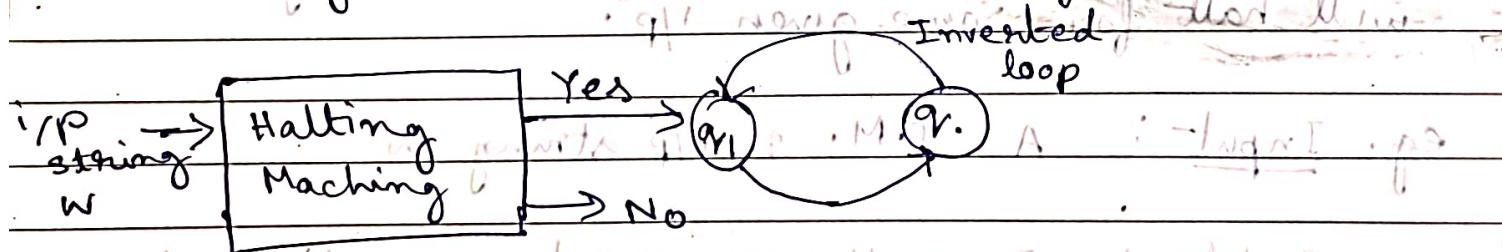


Now, we will design an inverted halting M/C as

If it returns 'yes', then loop forever

If it returns 'No', then halt

The diagram for inverted halting m/c -



all for contradiction, since if NT exist some

for contradiction, so in w must exist

so there is contradiction, so S is not

# Complexity Classes

Bittoo  
TRADE MARK RECO.

Date :

Page no.

→ Time complexity

- How long computation takes to execute
- In TM, this could be measured, as no. of moves which are required to perform computation

— no. of m/c cycles

→ Space complexity

- How much storage is required for computation
- In TM no. of cells are used
- No. of bytes used.

— Types of complexity classes :

- ① P-class : set of decision problems which are solvable in polynomial time or in the class P. at most n<sup>log n</sup> time.
- If there exists an algorithm A such that A is an algo.

→ A takes instance of problem if P

→ A always op's the correct answer

"yes" or "no" answer in now and - no wrong answer most fast algorithm

→ There exists a polynomial P such that the execution of A on the i/p of size n always terminates in p(n) or fewer steps.

eg. The minimum spanning tree problem is in class P.

The class P is often considered as synonymous with the class of computationally feasible problems, although in practice this is somewhat unrealistic.

### NP class:

A decision problem is non-deterministically polynomial time solvable or in the class NP if there exists an Algorithm A such that,

→ There exists a polynomial p such that for each potential witnesses of each instance of size of  $D$ , the execution of the algorithm A takes at most  $p(n)$  steps.

→ Think of a non-deterministic computer as a computer that magically 'guesses' a solution, then has to verify that it is correct.

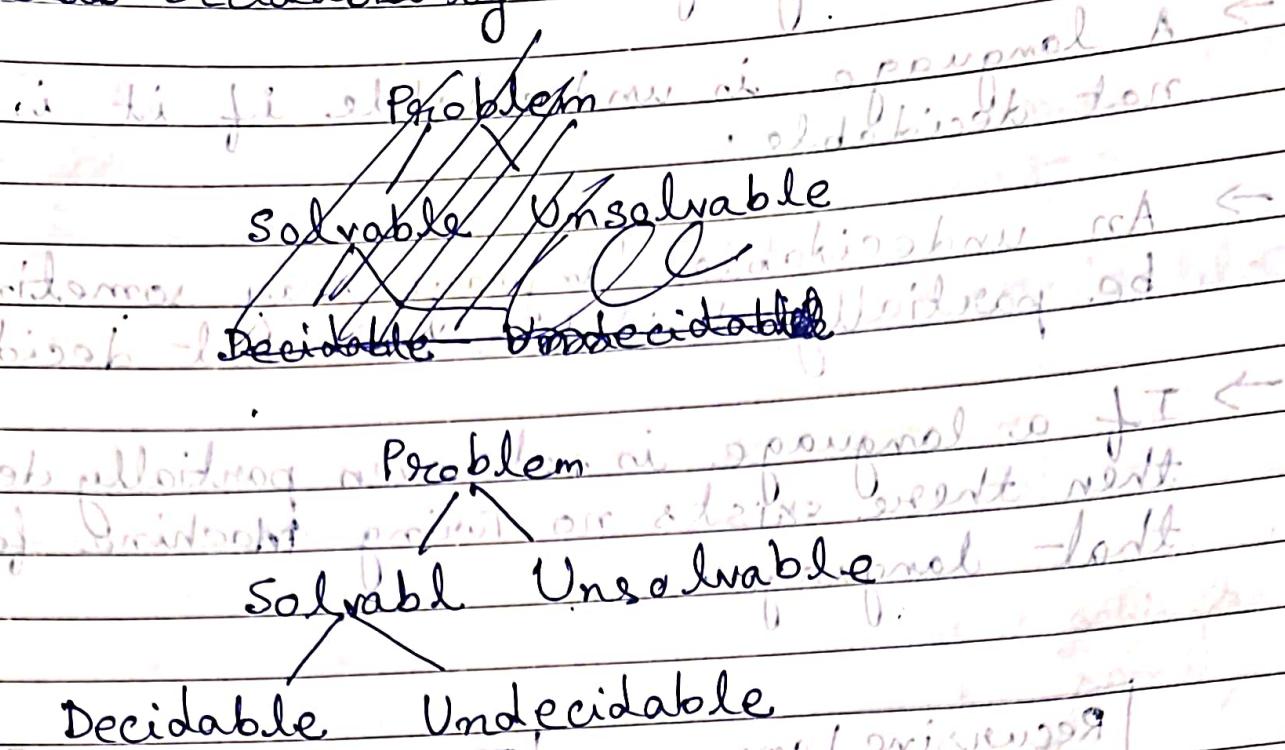
- If solution exists, computer always guess it.

- One way to imagine it: "a parallel computer that can freely spawn an infinite no. of processes."

It will do giant in a no. of ways, but each one of them is deterministic.

- All processors attempts to verify that their sol. works.
  - If a processor finds it has a working sol.  
so.  $NP =$  Problem verifiable in polynomial time
- Every problem in this class can be solvable in exponential time using exhaustive search.

## ~~Decidability~~ Decidability



## Decidability & Undecidability

### Decidable Languages :-

A language  $L$  is decidable if it is a recursive language. All decidable languages are recursive languages and vice-versa.

### Partially Decidable language :-

A language ' $L$ ' is partially decidable if ' $L$ ' is a recursively enumerable language.

## Undecidable language:-

- A language is undecidable if it is not decidable.
- An undecidable language may sometime be partially decidable but not decidable.
- If a language is not even partially decidable then there exists no Turing Machine for that language.

Recursive Language	TM will always <u>Halt</u>
Recursively Enumerable Language	TM will halt sometimes but may not halt sometimes.
Decidable Language	Recursive Language
Partially Decidable Language	Recursively Enumerable Language
UNDECIDABLE	No. TM for that lang.

## DECIDABILITY Overview

- Every question about regular languages is decidable.
- Some questions about CFLs are decidable, but some are not.
- The "Halting Problem" is undecidable.
- Some languages are not turing recognizable.
- Many questions about turing machines are not even decidable, and some are not even turing recognizable.

Q: Given a DFA and a string, will the DFA accept?

Is this problem decidable?

Sol: Languages are decidable if we must express the problem in terms of languages. i.e. A  $\in$  A  $\in$  L

Given a string at any position MT can go to state  $q_1$  or  $q_2$ .

Ignore this MT only depends on what happens after this MT.

Theorem: The language:  $\{B, w \mid B \text{ is a DFA that accepts string } w\}$

$A_{DFA} = \{B, w \mid B \text{ is a DFA that accepts string } w\}$   
is decidable.

$\langle \rangle \rightarrow$  denotes something abstract / value / coding

Possible Confusion:

1. String  $w \rightarrow$  Regular Lang.

2. String  $\langle B, w \rangle$

Language  $A_{DFA} \leftarrow$  Not Regular, Not CFG,

But it is decidable.

Proof:

→ Provide a TM that decides it.

→ The TM is given as  $i/p: \langle B, w \rangle$

→ A DFA and a string  $w$ .

→ The TM checks  $B$  to make sure that  $B$  is a valid representation.

→ The TM then simulates  $B$  on  $w$ .

→ If  $B$  reaches a final state at the end of  $w$ .

→ Otherwise the TM will reject.

→ This TM will always halt.

# Cook's Theorem:

~~Cook's~~

The satisfiability problem (SAT) is NP complete.

or

L

What SAT?

A propositional logic formula  $\phi$  is called satisfiable if there is some assignment to its variables that makes it evaluate to true.

→ ~~Proposition~~  $p \wedge q$  is satisfiable  $p=1, q=1$

→  $p \wedge \neg p$

operator: NOT  $\neg$  (high)  
AND  
OR +

3SAT

A language  $3SAT = \{ \phi_1, \phi_2, \dots \}$

## Undecidability of the Halting Problem

Given a program, will it 'halt'?

Can we design a machine which if given a program can find out or decide if that program will always halt or not, halt on a particular i/p?

Let us assume that we can:

$H(P, I) \leftarrow H$  can take 2 arguments.  
 If  $H(P, I)$   $\rightarrow$  Halt, No to this.  
 $\rightarrow$  Not Halt  $\rightarrow$  Halt, yes to this.

This allows us to write another program:

$C(x)$   
 On  $\leq m$  No  $\{$  if  $\{ H(x, x) = \text{Halt} \}$   
 Loop forever;

Also close in 'No pid' in L. Create

Return;

If we run 'C' on itself

$C(C)$

$H(C, C) = \text{Halt}$   $H(C, C) = \text{Not Halt}$

This means we cannot actually design a machine which can say that a machine can tell us a program can halt or not.

## Complexity:

When a problem / language is decidable, it simply means that the problem is computationally solvable in principle.

## Growth Rate of Functions:

When we have 2 algorithms for the same problem, we may require a comparison between the running time of these 2 algorithms.

Definition: Let  $f(n), g: N \rightarrow R^+ (R^+ \text{ being the set of all possible +ve real nos.})$  we say that  $f(n) = O(g(n))$  if there exist +ve integers c. and  $N_0$  such that

$$f(n) \leq c g(n) \text{ for all } n \geq N_0$$

Here, f is 'big Oh' of g.

\*\*  $f(n) = O(g(n))$  is not an eq<sup>n</sup>, it only expresses a relation btwn. 2 fn's. f and g.

Q. Let  $f(n) = 4n^3 + 5n^2 + 7n + 3$ .

P.T.  $f(n) = O(n^3)$

Sol: Let us take  $C = 5$  and  $N_0 = 10$

$$\therefore f(n) = 4n^3 + 5n^2 + 7n + 3 \text{ for } n \geq 10$$

take  $n = 10$ ,  $f(n) = 573 \leq 10^3 \therefore f(n) = O(n^3)$

e.g.  $n = 11$   $\therefore f(n) = 4(11)^3 + 5(11)^2 + 7(11) + 3 \leq 4(11)^3 = 5324$

Theorem:

9th June 9 AM Analysis IIT

If  $p(n) = a_k n^k + a_{k-1} n^{k-1} + \dots + a_1 n + a_0$  is a polynomial degree like over  $n^k$  and  $a_k > 0$ , then  $p(n) = O(n^k)$

Proof: If  $p(n) = a_k n^k + a_{k-1} n^{k-1} + \dots + a_1 n + a_0$

Let us assume

$\because a_k$  is +ve int.

No such that  $n \geq N$  s.t.  $a_k n^k + \dots + a_1 n + a_0 \leq 1$

$\therefore a_k \geq 1$  (C)

Now  $a_k \geq 1$  implies  $a_k n^k \geq n^k$  for all  $n \geq N$

$\therefore \frac{a_{k-1}}{n} + \frac{a_{k-2}}{n^2} + \dots + \frac{a_1}{n^{k-1}} + \frac{a_0}{n^k} \leq \frac{1}{n^k}$

$$\therefore \left| \frac{a_{k-1}}{n} + \frac{a_{k-2}}{n^2} + \dots + \frac{a_1}{n^{k-1}} + \frac{a_0}{n^k} \right| \leq \frac{1}{n^k}$$

As  $a_k \geq 1$ ,  $\frac{p(n)}{n^k} = \frac{a_k + a_{k-1} n^{-1} + \dots + a_1 n^{k-1} + a_0 n^{-k}}{n^k} \geq 1$

Also

$$\frac{p(n)}{n^k} = a_k + \left( \frac{a_{k-1}}{n} + \dots + \frac{a_1}{n^{k-1}} + \frac{a_0}{n^k} \right)$$

But we have  $a_k + 1$  also  $n \geq N$

So

## The classes P and NP

### P-class

→ A TM is said to be of time complexity  $T(n)$  if the following holds: Given any input  $w$  of length  $n$ , M halts after making at most  $T(n)$  moves.

→ A language  $L$  is in class P if there exists some polynomial  $T(n)$  such that  $L = T(M)$  for some deterministic TM  $M$  of time complexity  $\leq T(n)$ .

~~construct the time complexity for the TMs, given~~

Q: Find the running time for the Euclidean Algo. for evaluating  $\text{gcd}(a, b)$  where  $a \geq b$  are +ve integers expressed in binary representation.

~~soln~~ ← Polynomial algorithm

The Euclidean algo. is based on the below facts

1. If we subtract smaller number from larger no., GCD doesn't change. So, if we keep subtracting repeatedly the larger of 2, we end up with GCD.

2. Instead of subtraction, if we divide smaller no., the algorithm stops when we find remainder '0'.

## The classes P and NP

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~~sod~~  $\xrightarrow{\text{Polynomial algorithm}}$

The Euclidean algo. is based on the below facts

1. If we subtract smaller number from larger no., GCD doesn't change. So, if we keep subtracting repeatedly the larger of 2, we end up with GCD.

2. Instead of subtraction, if we divide smaller no., the algorithm stops when we find remainder '0'.

e.g.  $36, 6^0$

$$36 = 2 \times 2 \times 3 \times 3$$

$$60 = 2 \times 2 \times 3 \times 5$$

GCD = Multiplication of common factors

$$\text{common} = 2 \times 2 \times 3$$

$$= 12$$

In other words,

$\Rightarrow$  The algo has the following steps:

1. The ip is  $(a, b)$

2. Repeat until  $b = 0$

3. Assign  $a \leftarrow a \text{ mod } b$

4. Exchange  $a$  &  $b$  if  $a < b$ .  $a \leftarrow$

5. Output  $a$  if  $a \neq 0$ . If  $a = 0$  repeat

step 3 replaces  $a$  by  $a \text{ mod } b$

If  $a/2 \geq b$ , then  $a \text{ mod } b \leq b \leq a/2$

If  $a/2 < b$ , then  $a < 2b$

then  $a < 2b$

Write  $a = b + r$  for some  $0 \leq r < b$

Then  $a \% b = r < b \leq a/2$

$\therefore a \% b \leq a/2$

So,  $a$  is reduced by at least half

Final size on the application of step 3

$\therefore$  One iteration of step 3 reduces  $a$  &  $b$  by at least half in size.

$\therefore$  The max<sup>n</sup> no. of execution of step 3 & 4 are  $(\log a)$  and  $(\log b)$

The no. of iterations of step 3 & 4 is  $O(n)$ .

$\Rightarrow$  we have to perform step 2 at most  $\min \{[\log_2 a], [\log_2 b]\}$  times or n times.

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

$(d, n)$  is qf. out

NP Class:

$D = d$  times  $\rightarrow$   $d$  moves  $\rightarrow$   $D$  moves

$\rightarrow$  A language L is in class NP if there is a non-deterministic TM, M and a polynomial time complexity  $T(n)$  such that  $L = \{w \mid M \text{ accepts } w \text{ in } T(n) \text{ moves}\}$  for every input  $w$  of length  $n$ .

Polynomial time Reduction & NP completeness

Theorem:

If there is a polynomial time reduction from  $P_1$  to  $P_2$  and if  $P_2$  is in P, then  $P_1$  is also in P.

2. Let  $L$  be a language or problem in NP. Then  $L$  is NP-complete if

i)  $L$  is in NP

ii) For every language  $L'$  in NP there exists a polynomial-time reduction of  $L'$  to  $L$ .

iii) The class  $\Theta$  of NP complete language is a subclass of NP.

3. If  $P_1$  is NP-complete and there is a polynomial time reduction of  $P_1$  to  $P_2$ , then  $P_2$  is NP-complete.

4. If some NP-complete problem is in P, then  $P = NP$ .

Proof.  $L$  is any lang. in NP.

\*\* We will show that there is a polynomial time reduction of  $L$  to  $P_2$ .

As;  $P_1$  is NP-complete

so, time reduction of  $L$  to  $P_1$

suppose n-size input string w

& so time taken to convert this ~~is~~

~~string~~  $x \rightarrow P_1$

$y \rightarrow P_2$

i. Time taken for transforming  $w$  to  $y$  is at most  $P_1(n) + P_2(P_1(n))$

As  $P_1(n) + P_2(P_1(n))$  is a polynomial

we get a polynomial-time reduction of  $L$  to  $P_2$ . That is for which  $P_2$  is NP-complete.

Given  $L$  is a language in  $\text{NP-Complete}$ .  
 $P_2$  is a language in  $\text{NP-Complete}$ .  
 $P_2$  is a language in  $\text{NP-Complete}$ .

$L$  is a language in  $\text{NP-Complete}$ .  
 $P_2 = L \cdot P_1$  is a language in  $\text{NP-Complete}$ .

$P_1$  is a language in  $\text{NP-Complete}$ .  
 $P_2$  is a language in  $\text{NP-Complete}$ .  
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Time taken for transforming  $w$  to  $y$   
 is at most  $P_1(n) + P_2(P_1(n))$

As  $P_1(n) + P_2(P_1(n))$  is a polynomial  
 least time is polynomial time  $\leq n^k$   
 Hence get a polynomial-time reduction  
 of  $P$  to  $P_2$ .  
 $P_2$  is NP-complete now