

TIME
COMPLEXITY
AND BIG-O
NOTATION

"TIME COMPLEXITY"

= RUNNING TIME FOR PROGRAMS

- CONSIDER ONLY COMPUTABLE FUNCTIONS.
→ DECIDABLE (Always Halt)

- CONSIDER ONLY DETERMINISTIC MACHINES

That "guess the right thing" or "try all possibilities" is a questionable operation on a real machine.

- CONSIDER SOME INPUT, w .

FOR TMs: JUST COUNT THE TRANSITIONS.

- CONSIDER ALL INPUTS OF SIZE N .

WHAT IS THE MAXIMUM TIME THE TURING MACHINE MIGHT TAKE?

OUR GOAL: FIND A FUNCTION OF N
TO DESCRIBE THE RUNNING TIME.

$$f(N) = \dots$$

Often, the function can be ugly!

$$f(N) = 17N^3 + 5N^2 + 3\text{Log}N_3 + 29$$

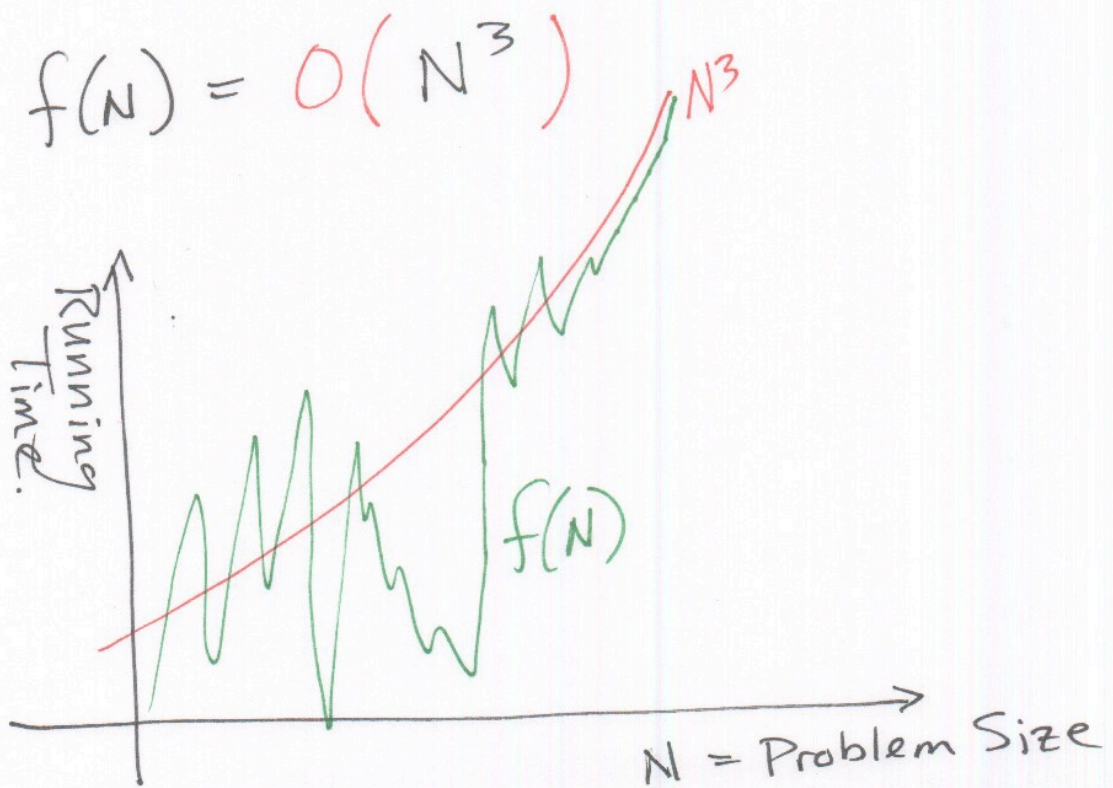
For large values of N , we only care about N^3

We want the

ASYMPTOTIC UPPER BOUND.

The "ORDER" (or "BIG-O") Notation:

$$f(N) = O(N^3)$$



Also: ignore constant factors.

\Rightarrow ignore $17N^3$

LET M BE A DETERMINISTIC TURING MACHINE THAT ALWAYS HALTS.

LET n BE THE SIZE OF AN INPUT.

DEFN

THE "TIME COMPLEXITY" (i.e., the RUNNING TIME) OF M IS A FUNCTION f .

$f(n)$ = THE MAXIMUM NUMBER OF STEPS THAT M TAKES ON ANY INPUT OF SIZE n .

NOTE

"SIZE OF INPUT" usually means the LENGTH OF THE INPUT.

... But may sometimes mean something else, such as

- Number of nodes in a graph.
- Number of rules in a CFG.
- etc.

BIG - O NOTATION

$$17N^3 + 5N^2 + 3N + 29$$

$$O(N^3)$$

FOR POLYNOMIAL FUNCTIONS.

* TAKE THE HIGHEST ORDER TERM

* IGNORE THE COEFFICIENT.

$$f(n) = 17n^3 + 5n^2 + 3n + 29$$

We say: $f(n) = O(n^3)$

Also:

$$\begin{aligned} f(n) &= O(n^4) \\ &= O(n^5) \\ &= O(2^n) \end{aligned}$$

DEFN

Let $f(n)$ be some running time function of interest.

We say $f(n) = O(n^3)$

if, for all $n \geq$ some value (n_0)
(i.e., for all n large enough)
 f (^{looks} behaves) like n^3 , ignoring
constant factors.

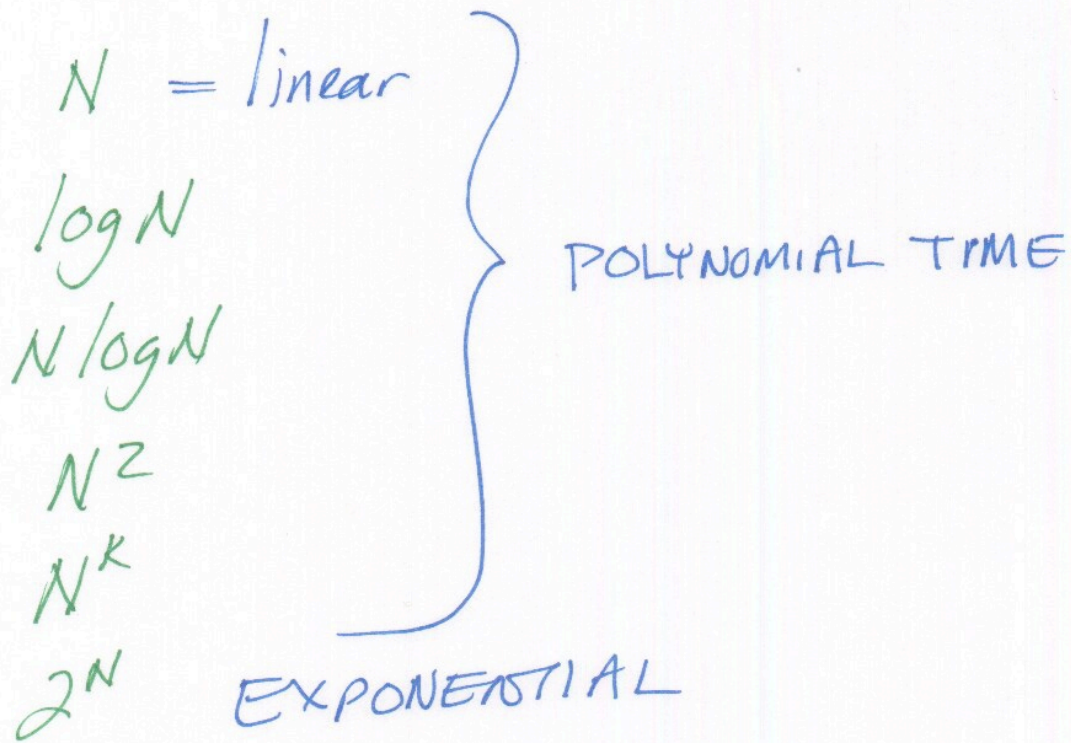
MORE PRECISELY:

$$f(n) = O(g(n))$$

IF $\exists c$ and $\exists n_0$ such that

$$f(n) \leq c \cdot g(n) \text{ for all } n \geq n_0.$$

TYPICAL COMPLEXITY CLASSES



$O(N)$ = linear time algorithms

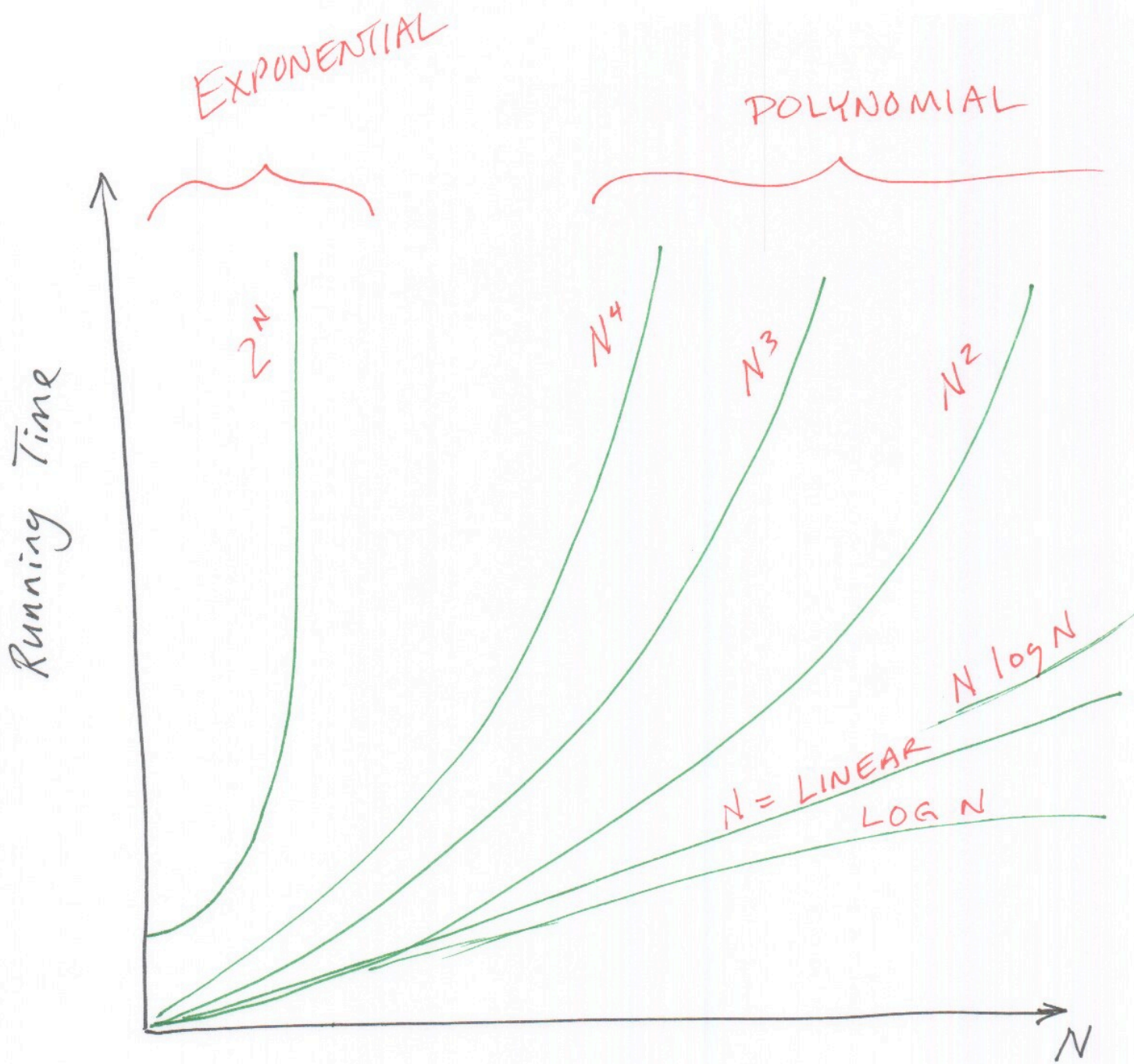
$O(N \log N)$

$O(N^2)$

$O(N^3)$

$O(N^4)$

$O(2^N)$



Q: Why aren't there many
 $O(\log N)$
algorithms?

A: The input has size n .
Just to read all the input
requires $O(n)$

TIME COMPLEXITY CLASSES

TIME(n)

The set of all languages/problems that can be DECIDED in $O(n)$ time.

TIME(n^2)

... that can be DECIDED in $O(n^2)$ time.

TIME($n \log n$) ... in $O(n \log n)$

TIME(n^3) ... in $O(n^3)$

TIME(2^N) ... in exponential time.

etc.

NOTE:

TIME(n) \subset TIME($n \log n$) \subset TIME(n^2) \subset

TIME(n^3) \subset TIME(n^k) \subset TIME(2^N)

THE TIME COMPLEXITY
OF AN EXAMPLE
ALGORITHM

ALGORITHM TO DECIDE $\{0^k 1^k \mid k \geq 0\}$

- SCAN INPUT TO MAKE SURE IT IS IN THE FORM $0^* 1^*$.

n STEPS TO SCAN
 n STEPS TO REPOSITION TO LEFT END
 $2n$ STEPS $O(n)$

- REPEAT WHILE TAPE CONTAINS AT LEAST ONE 0 AND AT LEAST ONE 1...

- SCAN ACROSS TAPE AND CHANGE A 0 TO X AND A 1 TO X.

$2n$ STEPS $O(n)$

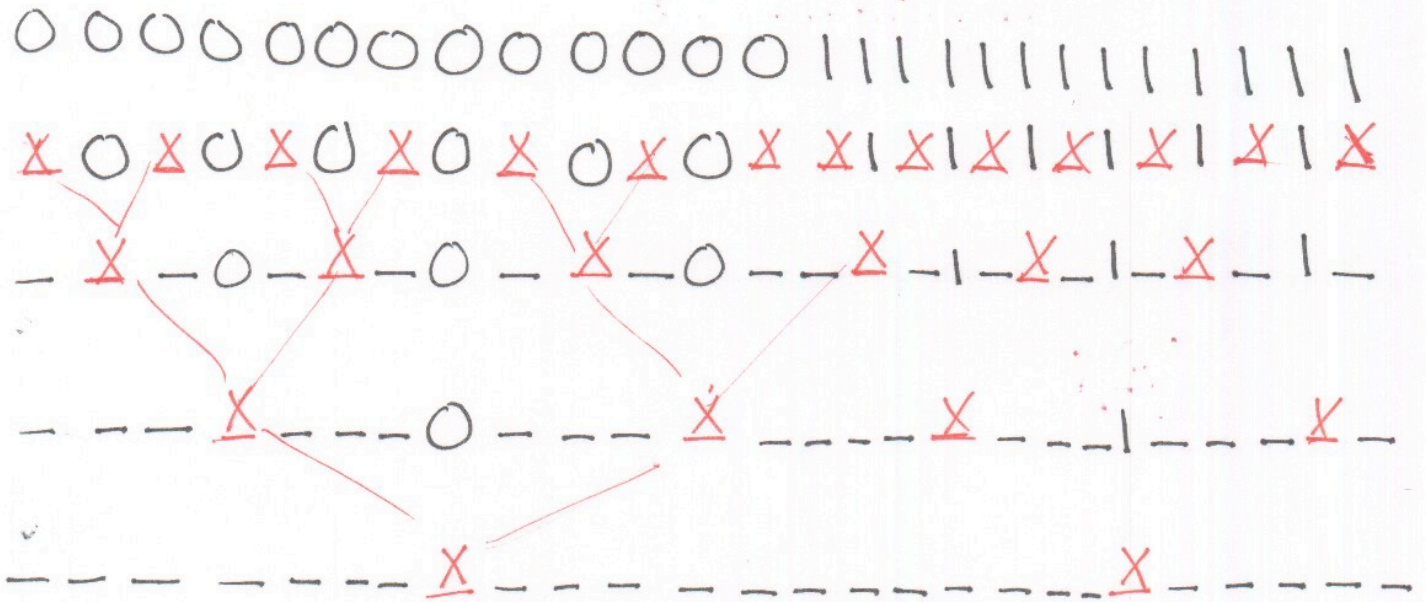
END LOOP

$n/2$ REPETITIONS
WHOLE LOOP TAKE $\frac{n}{2} \cdot O(n)$ $O(n^2)$

- IF TAPE CONTAINS ALL Xs, THEN ACCEPT ELSE REJECT.

n STEPS $O(n)$

$$O(n) + O(n^2) + O(n) \implies O(n^2)$$



Cross off every other 0

Cross off every other 1

Repeat until nothing remains.

At each stage we should have
the same number of 0's ~~as~~
~~as~~ as 1's.

So: $\{0^k 1^k \mid k \geq 0\} \in \text{TIME}(n^2)$

But there is a better algorithm!

- Scan input to make sure it is in the form $0^* 1^*$ ← $O(n)$

- Repeat ~~while~~ while the tape contains at least one 0 and at least one 1...

- Scan tape to see if number of 0's plus number of 1's is ODD or EVEN

- If ODD then REJECT. ← $O(1)$

- Scan across the entire tape.

Cross off every other 0, starting with the first 0.

Cross off every other 1, starting with the first 1.

END ←

Number of reps is $1 + \log_2 n$
 $(1 + \log_2 n) \cdot O(n) = O(n \log n)$

- If ~~no~~ no 0's and no 1's remain then ACCEPT, else REJECT ← $O(n)$

So: $\{0^k 1^k \mid k \geq 0\} \in \text{TIME}(n \log n)$

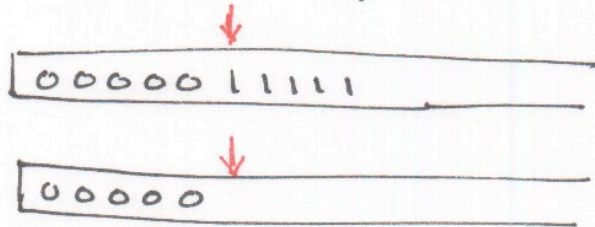
DIFFERENT MODELS OF COMPUTATION

What about a different model of computation?

Assume we have multiple tapes.

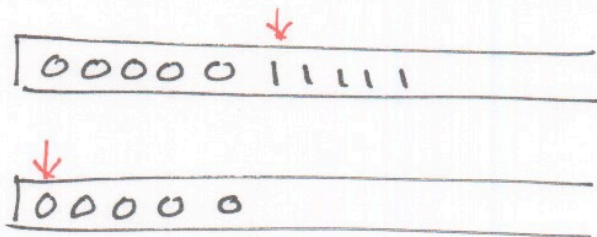
ALGORITHM USING 2 TAPES. $\{0^k 1^k \mid k \geq 0\}$

- Copy all 0's to tape 2.



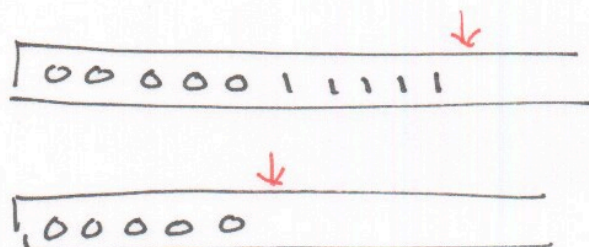
$O(n)$

- Reposition tape 2 to beginning.



$O(n)$

- Scan both tapes simultaneously.
- Make sure both heads hit \sqcup at the same time.



$O(n)$

THEOREM

For every multitape Turing machine algorithm that takes time $t(n)$,
There is an equivalent single tape Turing machine that takes time $O(t^2(n))$.

PROOF

In time $t(n)$, the longest the tapes can be is $t(n)$.

You can simulate the multitape algorithm on a machine with one tape.

Each step of the simulation can be done in $O(t(n))$ time.

To simulate the entire algorithm:

$$t(n) \cdot O(t(n)) = O(t^2(n))$$

Bottom Line

The model of computation matters!

However, the differences are "relatively small".

A polynomial-time algorithm will remain polynomial-time, regardless of the details of the model of computation!

AS LONG AS THE MACHINES ARE DETERMINISTIC!

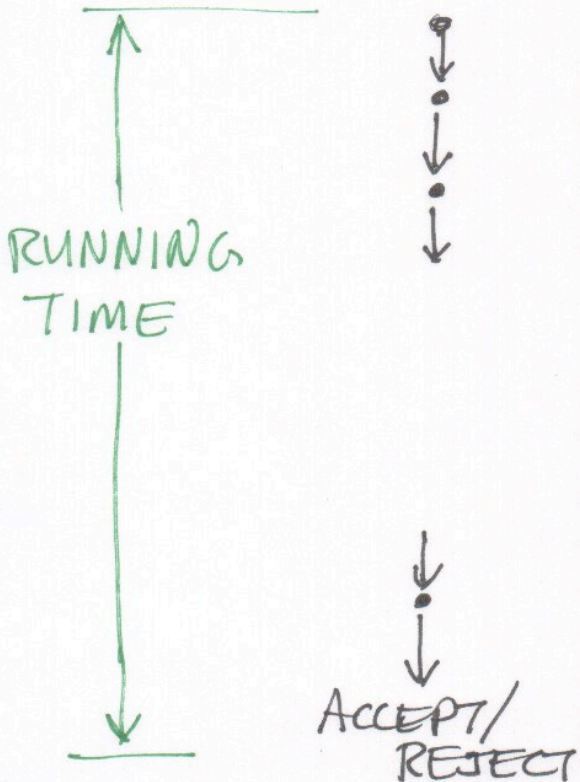
The class of Polynomial-time Problems seems quite ROBUST.
(Details of the computer don't matter.)

NON-DETERMINISTIC T.M.S.

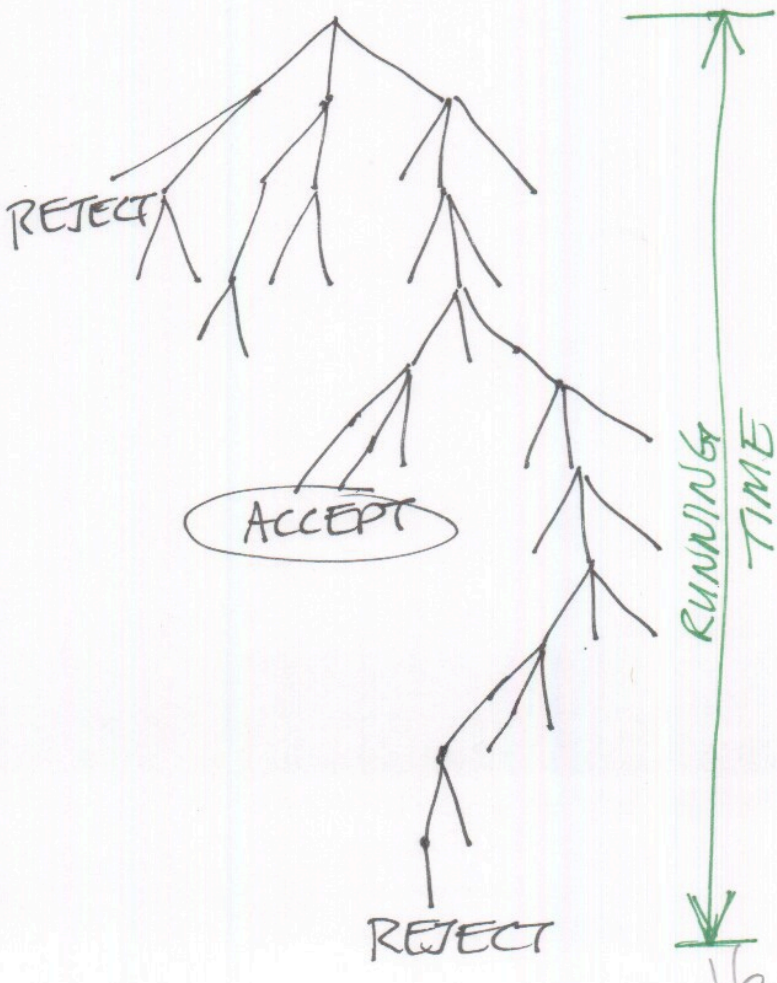
RUNNING TIME:

The number of steps the TM uses on the longest branch of computation.

DETERMINISTIC COMPUTATION HISTORY



NONDETERMINISTIC COMPUTATION HISTORY



EVERY NONDETERMINISTIC TM CAN
BE SIMULATED ON A DETERMINISTIC
TM, USING EXPONENTIALLY
MANY MORE STEPS.

NONDET TM

TAKES 419 STEPS ON INPUT w

DET SIMULATION

CAN BE DONE IN 2^{419} STEPS

NONDET TM

TAKES $O(N^2)$ TIME

DET SIMULATION

CAN BE DONE IN 2^{N^2} TIME

THE COMPLEXITY
CLASSES

P AND NP

THE CLASS P

All reasonable deterministic models
of computation are
POLYNOMIALLY EQUIVALENT.

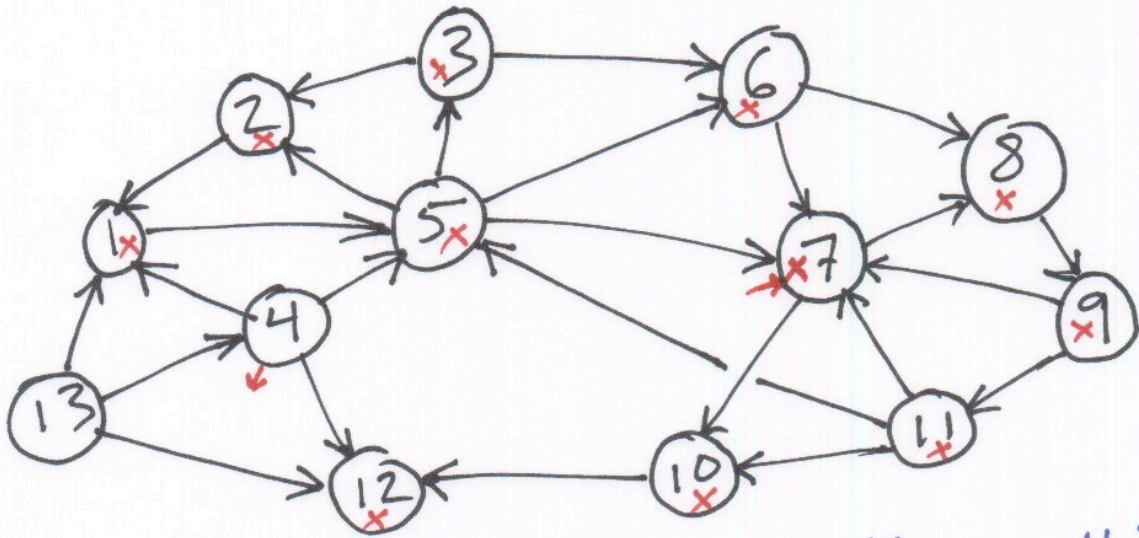
The class of languages that
can be decided...
[i.e., the set of problems that can
be solved...]
in POLYNOMIAL TIME on a
DETERMINISTIC TURING MACHINE.

$$P = \bigcup_k \text{TIME}(n^k)$$

THE "PATH" PROBLEM

INPUT: A DIRECTED GRAPH G
AND TWO NODES, s and t ...

IS THERE A PATH FROM s TO t ?



IS THERE A PATH FROM 7 TO 4?

PATH $\in P$

PROOF

- PROVIDE AN ALGORITHM
- SHOW ITS RUNNING TIME.

USE A MARKING ALGORITHM

$O(m^2)$ where m = number of nodes

THEOREM

EVERY CONTEXT-FREE LANGUAGE IS IN P.

PROOF

PROVIDE AN $O(n^3)$ ALGORITHM.

A "DYNAMIC PROGRAMMING" ALGORITHM.

- USE A TABLE TO STORE PARTIAL RESULTS.
- AVOID HAVING TO RECOMPUTE THINGS OVER AND OVER.
- BUILD BIGGER RESULTS OUT OF SMALLER RESULTS.

```
FOR i = 1 TO N.  
    FOR COMPUTE ALL RESULTS  
        OF SIZE i  
    STORE EACH RESULT.  
    MAKE USE OF RESULTS  
        OF SIZE < i.  
END
```

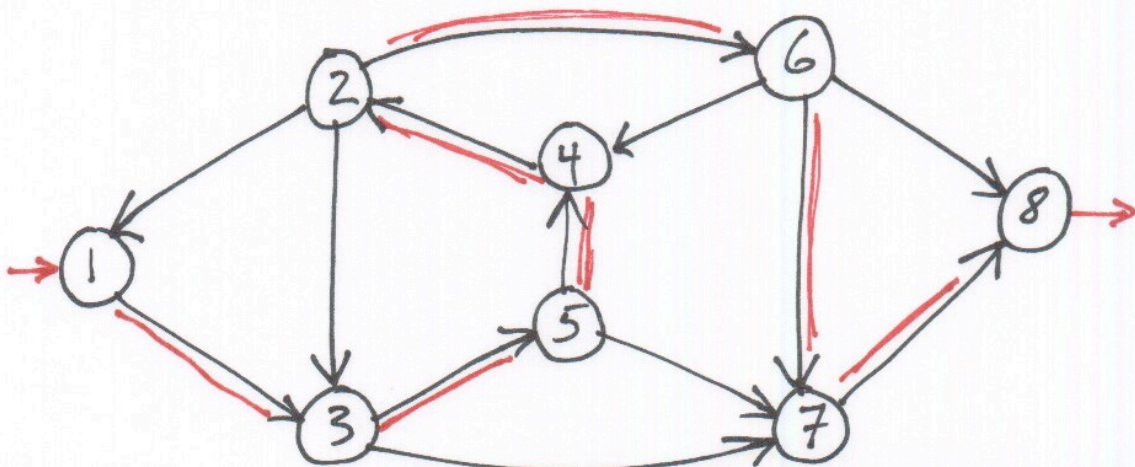

HAMPATH: THE "HAMILTONIAN" PATH PROBLEM

GIVEN A DIRECTED GRAPH, IS THERE A PATH THAT GOES THROUGH EVERY NODE EXACTLY ONCE?

HAMPATH = $\left\{ \langle G, s, t \rangle \mid \left. \begin{array}{l} G \text{ is a directed} \\ \text{graph and there is} \\ \text{a "HAMILTONIAN"} \\ \text{Path from } s \text{ to } t \end{array} \right\}$

WE ARE GIVEN THE STARTING AND ENDING NODES.

EXAMPLE: $G, 1, 8$



13542678

EXPONENTIAL ALGORITHM

GENERATE ALL POSSIBLE PATHS.

1 2 3 4 5 6 7 8
1 4 3 2 8 7 5 6
⋮

TEST EACH PATH. TO SEE IF IT
IS LEGAL.

NOTE: This "test" can be done
quickly! ← IN POLYNOMIAL TIME

This problem is in class NP.

It ~~seems~~ to require exponential
time.

But given the answer, we can
VERIFY it in polynomial
time.

DEFINITION
OF NP



POLYNOMIAL
VERIFIABILITY

POLYNOMIAL VERIFIABILITY

Given a language A ,

A "VERIFIER" is an algorithm that is given some extra information, " c ", which it can use to check (in polynomial time) to verify that w is in A .

EXAMPLE: HAMPATH

Given a problem, such as

$$w = \langle G, s, t \rangle$$

is there a Hamiltonian Path?

EXPONENTIALLY HARD [Probably]

But the verifier algorithm is passed

some info: $c = "13542678"$

and can then CONFIRM that

$w \in \text{HAMPATH}$

IN POLYNOMIAL TIME.

DEFINITION

A "VERIFIER" for a language A is an algorithm V where

$$A = \left\{ w \mid V \text{ accepts } \langle w, c \rangle \text{ for some string } c \right\}$$

A "POLYNOMIAL-TIME VERIFIER" runs in polynomial time in the length of w .

A language is "POLYNOMIALLY VERIFIABLE" if it has a polynomial-time verifier.

The string c is called the "CERTIFICATE" (or "PROOF").

We don't care about the length of c ; but note that a polynomial-time verifier ~~can~~ does not have time to read a certificate that is longer than polynomial in the length of w .

DEFINITION

"NP" is the class of languages that have polynomial-time verifiers.

THEOREM

A language is in NP iff it is decided by some NONDETERMINISTIC POLYNOMIAL-TIME Turing Machine

Sometimes this is given as the definition of "NP".

PROOF

- Convert a Polynomial-time Verifier into an equivalent polynomial-time nondeterministic Turing Machine.

The TM:

INPUT: w (of length n .)

ALGORITHM:

- Nondeterministically guess string c (length at most n^k)
- Run V on $\langle w, c \rangle$
- If V accepts, accept. Else reject.

- Assume ~~that~~ you have a polynomial-time non-deterministic TM. Construct a ~~polynomial-time~~ polynomial-time verifier.

The Verifier:

INPUT: $\langle w, c \rangle$

ALGORITHM:

Simulate the Non-deterministic TM.
Use c as a guide about which choice to make. at each step.
If this branch accepts, then ACCEPT
else REJECT.

P = The class of languages for which membership can be **DECIDED** quickly.*

NP = The class of languages for which membership can be **VERIFIED** quickly.

↪ That is, given some information [the "certificate/proof"], you can quickly confirm that w is in the language.

* "quickly" means "in Polynomial time"

DEFINITION

$$\text{NTIME}(t(n)) = \left\{ L \mid \begin{array}{l} L \text{ is a language} \\ \text{decided by an} \\ O(t(n)) \text{ time} \\ \text{nondeterministic T.M.} \end{array} \right\}$$

$\text{TIME}(n^2)$ = The set of languages that can be decided by a DETERMINISTIC T.M. in $O(n^2)$ time.

$\text{NTIME}(n^2)$ = The set of languages that can be decided by a NONDETERMINISTIC T.M. in $O(n^2)$ time.

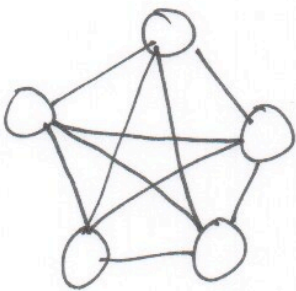
$$\text{NP} = \bigcup_k \text{NTIME}(n^k)$$

The "CLIQUE" Problem

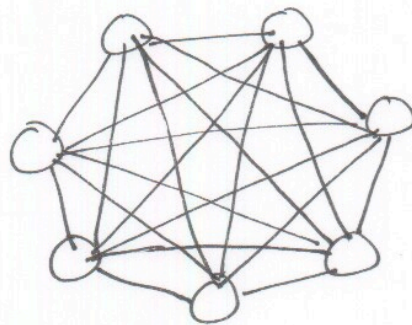
Given an undirected graph...

A "clique" is a set of nodes such that every node in the clique is connected to every other node in the clique.

A K -clique is a clique with K members.

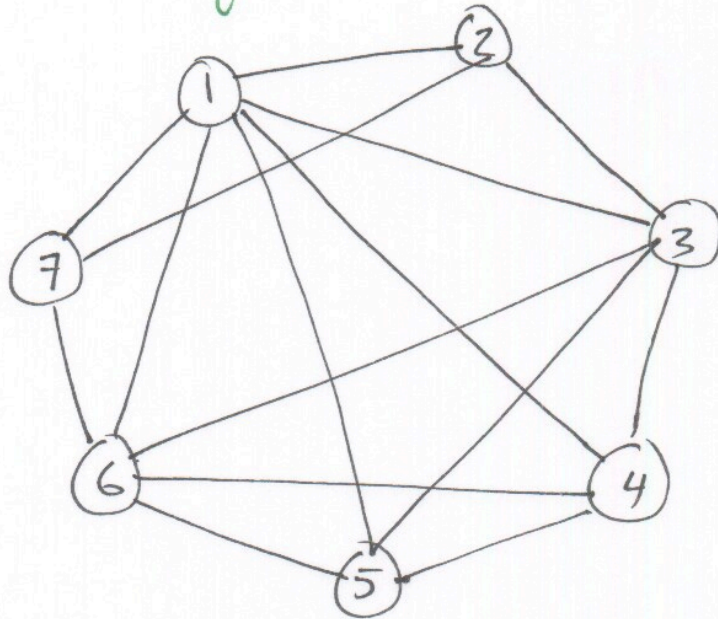


A 5-CLIQUE



A 7-CLIQUE.

Does this graph contain a 5-clique?



CLIQUE = $\{ \langle G, k \rangle \mid G \text{ is an undirected graph with a } k\text{-clique} \}$

THEOREM

CLIQUE \in NP

PROOF

- PROVIDE A POLYNOMIAL-TIME VERIFIER
- OR —
- PROVIDE A POLYNOMIAL-TIME NONDETERMINISTIC TURING MACHINE.

THE CLASS "P"

THE CLASS OF LANGUAGES THAT
CAN BE DECIDED...

[THE SET OF PROBLEMS THAT
CAN BE SOLVED...]

... IN POLYNOMIAL TIME ON
A DETERMINISTIC TURING
MACHINE

THE CLASS "NP"

THE CLASS OF LANGUAGES THAT
CAN BE DECIDED...

[THE SET OF PROBLEMS THAT
CAN BE SOLVED...]

... IN POLYNOMIAL TIME ON
A NONDETERMINISTIC TURING MACHINE.

UNSOLVED QUESTION:

$P = NP$ } Which is it?
 $P \subset NP$ }

There are lots of problems known to be in NP.

- NONE of these problems can be solved in poly. time ~~on~~ on a deterministic T.M.

These problems seem to ~~require~~ require exponential time to solve.

EXPONENTIAL-TIME PROBLEMS

$$\text{EXPTIME} = \bigcup_k \text{TIME}(2^{n^k})$$

RESULTS

$$P \subseteq NP \subseteq \text{EXPTIME}$$

APPARENTLY:

$$P \subset NP = \text{EXPTIME}$$

BUT THIS IS ALSO POSSIBLE:

$$P = NP \subset \text{EXPTIME}$$

ALL PROBLEMS/LANGUAGES

TURING RECOGNIZABLE

DECIDABLE

EXPTIME (=NP?)

P

$O(n^3)$

CFGs

REGULAR

NP-COMPLETE PROBLEMS

NP-COMPLETENESS

- An interesting subset of NP problems. **THE "NP-COMplete PROBLEMS."**
- If a polynomial time algorithm is ever found (on a deterministic machine) for any "NP-Complete" problem, then $P=NP$ follows!
- ... And polynomial time algorithms exist for all problems in NP!

Many interesting problems are NP-complete.

They seem to require exponential time.

THE SATISFIABILITY Problem "SAT"

Boolean variables; x_1, x_2, x_3, \dots

TRUE, FALSE

Boolean operations: $\wedge \vee \neg$

$$\neg \neg x = x$$

Boolean formulas, e.g.,

$$\phi = (\bar{x} \wedge y) \vee (\bar{y} \wedge z)$$

"Satisfiable" = If there is an assignment to the variables to make the formula true.

$x = \text{FALSE}, y = \text{TRUE}, z = \text{FALSE}.$

$$\text{SAT} = \{ \langle \phi \rangle \mid \phi \text{ is a satisfiable Boolean formula} \}$$

SAT \in NP

Nondeterministically guess the solution (eg. $X=FALSE, Y=TRUE, \dots$)
CHECK that it satisfies the Boolean formula in Polynomial time.

THEOREM

$SAT \in P$ iff $P=NP$.

OR EQUIVALENTLY...

SAT is "NP-COMPLETE".

Finding a polynomial time algorithm to solve a Boolean formula on a deterministic machine, would:

- Prove that ALL problems in NP have polynomial time algorithms.
- Rock the world.

PROOF THAT

SAT IS

NP-COMPLETE

RECALL ...

- The Turing Machine Acceptance Problem, $A_{TM} = \{ \langle M, w \rangle \mid M \text{ accepts } w \}$
- A_{TM} is undecidable.
- We "REDUCED" A_{TM} to an instance of the POST CORRESPONDENCE PROBLEM.
- This proved that the PCP was undecidable.
- We showed how to simulate the execution of a TM with the tiles of a PCP instance.
- The "computation history" was a sequence of "configurations."
- Finding a solution to the PCP was equivalent to finding an accepting computation history.

PROOF That SAT \in P iff P=NP

- A problem is in NP if there is a NONDETERMINISTIC TURING MACHINE ~~that~~ that will solve it in Polynomial time.
- Got a Problem? $\langle N, w \rangle$
A non-det T.M. \swarrow
An input. \searrow
- Convert it into an ~~instance~~ instance of the SAT problem.*
(A huge Boolean formula)
- Do this conversion in Polynomial time.
- If you can solve this SAT problem in Polynomial time, i.e. if SAT \in P, Then, you can solve any problem in NP ~~in~~ in Polynomial time.

* Such that, ~~there~~ there is a branch in the Nondeterministic computation that ACCEPTS IFF the Boolean Formula is SATISFIABLE.

THEOREM ("COOK-LEVIN")

SAT IS NP-COMPLETE

- ANY NP problem can be reduced into a SAT problem.
- This reduction ~~is~~ can be done in poly-time.
- So if you can solve ~~a~~ SAT in poly-time on a det. Machine, you can solve any problem in NP on a det. Machine in poly-time; i.e., $P=NP$.

GIVEN A PROBLEM IN NP...

GIVEN: N = A nondeterministic TM.

w = An input to that TM.

CONVERT IT INTO A BOOLEAN FORMULA, ϕ .

Such that ~~the~~ ϕ is satisfiable

iff N accepts w .

(iff N has an accepting
computation history for w).

The accepting computation history on N can take at most n^k steps, for some k .

Therefore, it can use at most n^k tape cells.

Step 1 in the reduction:

CREATE AN $n^k \times n^k$ "TABLEAU"

TO MODEL THE ACCEPTING COMPUTATION HISTORY.

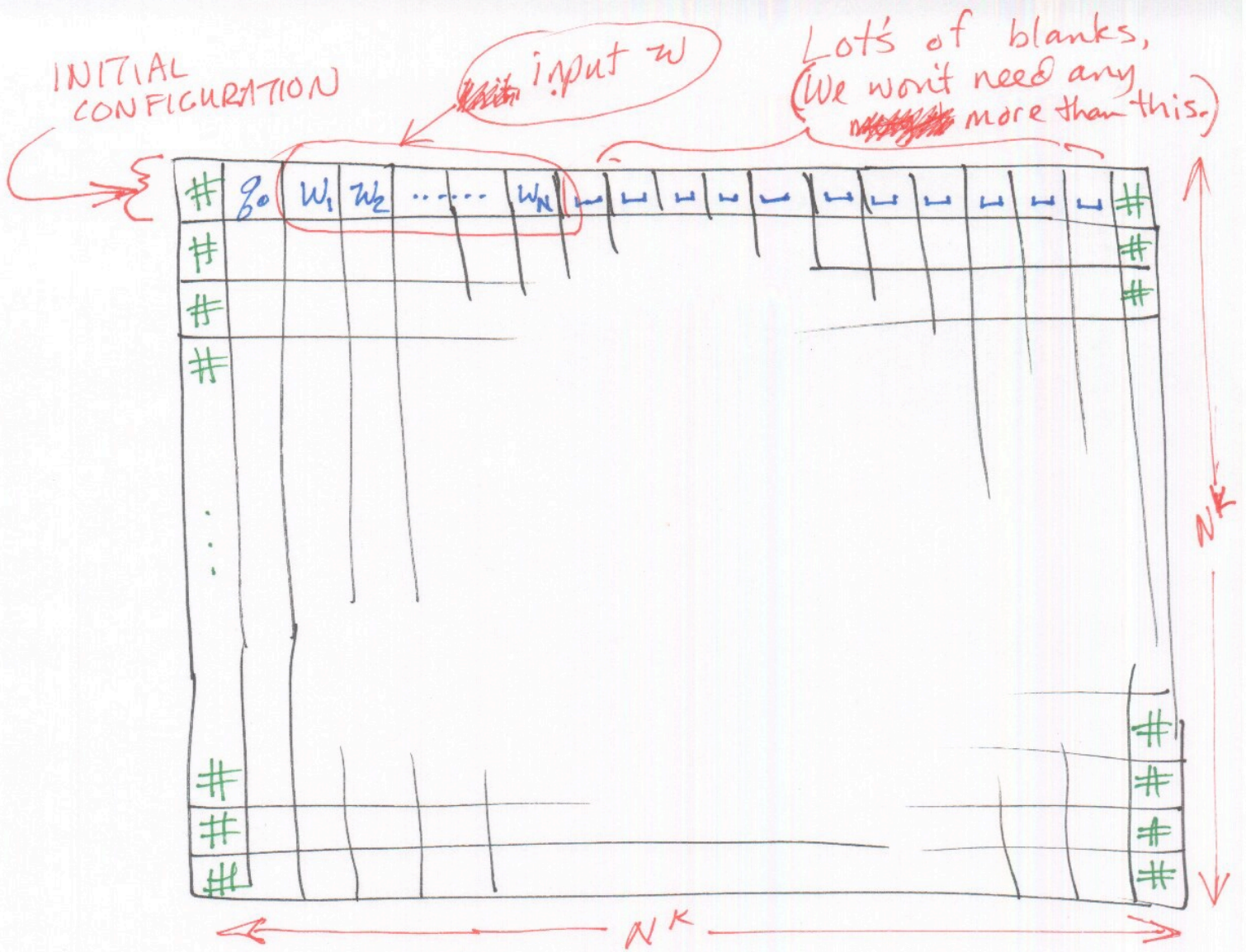
(It's big, but polynomial in size)

Step 2:

Create lots of boolean variables to model what could be in each cell of the TABLEAU.

Step 3:

Create a formula to express all the constraints on the TABLEAU to guarantee it models a legal, accepting computation history.



Each cell contains a single symbol.

← to mark end-of-tape

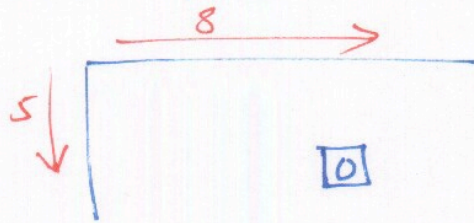
State ← Each row should have a state, q_i

Tape Symbol ~~...~~

$$Q \times \Gamma \times \{\#\}$$

What is in each cell?

What is in cell 5,8?



It could be 0, or 1 or \sqcup or # or

Create a ~~#~~ boolean variable for each possibility.

94...

$X_{5,8,0} = \text{TRUE}$ iff the cell contains "0"

$X_{5,8,1} = \text{TRUE}$ iff the cell contains "1"

$X_{5,8,\sqcup} = \text{TRUE}$ iff the cell contains " \sqcup "

$X_{5,8,\#} = \text{TRUE}$ iff the cell contains "#"

$X_{5,8,94} = \text{TRUE}$ iff the cell contains "94"

$X_{i,j,s}$ for all ~~###~~ $1 \leq i, j \leq N^k$
and $s \in Q \cup \Gamma \cup \{\#\}$

Now build the formula.

Goal: Add all constraints to assure

' the TABLEAU is a legal computational history that accepts.

CONSTRAINT #1 ← Φ_{CELL}

Every cell contains exactly one symbol.

CONSTRAINT #2 ← Φ_{START}

~~Every~~ The First Row is the starting configuration.

CONSTRAINT #3 ← Φ_{ACCEPT}

Some cell contains the symbol Φ_{ACCEPT}

CONSTRAINT #4 ← Φ_{MOVE}

Each Row (i.e. each "configuration") can legally follow the previous configuration, according to the transitions in the Nondeterministic TM we are modelling.

Construct the entire Boolean formula:

$$\phi = \phi_{\text{CELL}} \wedge \phi_{\text{START}} \wedge \phi_{\text{ACCEPT}} \wedge \phi_{\text{MOVE}}$$

This formula, while really huge,
is polynomial in size (of ~~n~~ w)

If ϕ has a solution, then
there is an accepting
computation history.

If there is an accepting
computation history, then this
formula has a solution.

If you can determine whether this
formula ϕ has a solution in poly-time,
then you can ~~be~~ determine in poly-time
whether a Non-deterministic TM \mathbb{N}
will accept w .

$$\text{SAT} \in \text{P} \Rightarrow \text{P} = \text{NP}$$

$$\Phi = \Phi_{\text{CELL}} \wedge \Phi_{\text{START}} \wedge \Phi_{\text{ACCEPT}} \wedge \Phi_{\text{MOVE}}$$

CONSTRAINT #1

Every cell contains exactly one symbol.

$X_{5,8,\#} = \text{TRUE}$ iff cell contains "#"
 $X_{5,8,q_7} = \text{TRUE}$ iff cell contains "q₇"

$$\bigvee X_{i,j,s}$$

At least one of the variables for this cell is TRUE

$$s \in (Q \cup \{\#\})$$

$$\bigwedge_{s \neq t} (\overline{X_{i,j,s}} \vee \overline{X_{i,j,t}})$$

For every pair of variables for this cell, at least one is FALSE

Combining (\wedge): Exactly one ~~variable~~ variable is true.
 And this is true of all cells, (i,j) :

$$\Phi_{\text{CELL}} = \bigwedge_{1 \leq j, k \leq N^k} \left[\left(\bigvee_{s \in \dots} X_{i,j,s} \right) \wedge \left(\bigwedge_{\substack{s \neq t \\ s \in \dots \\ t \in \dots}} (\overline{X_{i,j,s}} \vee \overline{X_{i,j,t}}) \right) \right]$$

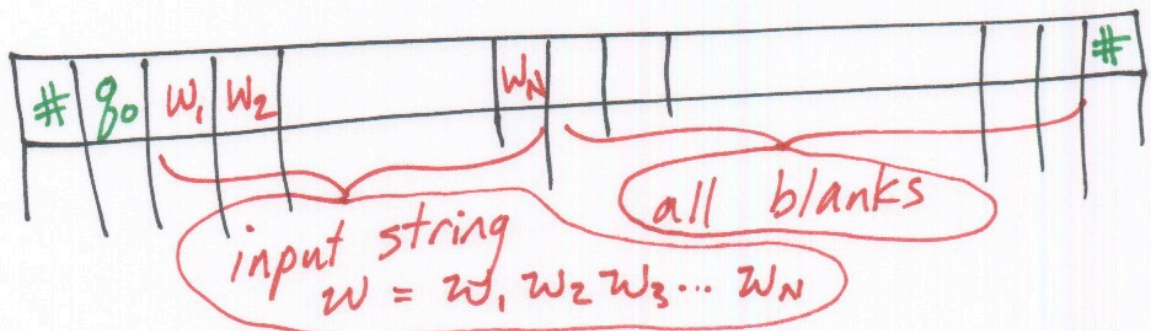
$$\phi = \phi_{\text{CELL}} \wedge \phi_{\text{START}} \wedge \phi_{\text{ACCEPT}} \wedge \phi_{\text{MOVE}}$$

CONSTRAINT #2:

The first row describes the initial configuration.

$$\phi_{\text{START}} = \dots \wedge \dots \wedge \dots \wedge \dots \wedge \dots$$

$$= x_{1,1,\#} \wedge x_{1,2,q_0} \wedge x_{2^N,2^N,\#} \wedge \dots$$



$$\dots \wedge x_{1,3,w_1} \wedge x_{1,4,w_2} \wedge x_{1,5,w_3} \dots \wedge x_{1,N+2,w_N}$$

w is in the first N cells of the tape.

$$\dots \wedge x_{1,N+3,_} \wedge x_{1,N+4,_} \wedge \dots \wedge x_{1,N^k+2,_}$$

The remaining cells of the tape contain the blank symbol, $_$.

$$\phi = \phi_{\text{CELL}} \wedge \phi_{\text{START}} \wedge \phi_{\text{ACCEPT}} \wedge \phi_{\text{MOVE}}$$

CONSTRAINT #3:

The ACCEPT state is reached.
in the computation history.

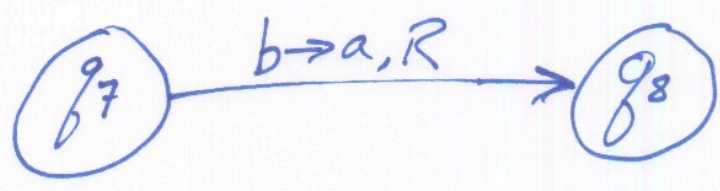
Some cell somewhere ~~represents~~
contains q_{ACCEPT} .

$$\phi_{\text{ACCEPT}} = \bigvee_{1 \leq i, j \leq 2^N} \chi_{i, j, q_{\text{ACCEPT}}}$$

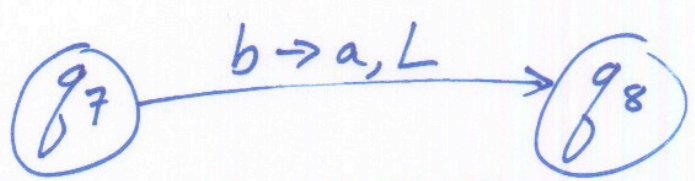
$$\Phi = \Phi_{\text{CELL}} \wedge \Phi_{\text{START}} \wedge \Phi_{\text{ACCEPT}} \wedge \Phi_{\text{MOVE}}$$

CONSTRAINT #4:

EVERY CONFIGURATION CAN LEGALLY FOLLOW THE PREVIOUS CONFIGURATION, ACCORDING TO THE DETAILS OF THE NONDET. TM'S TRANSITION FUNCTION.



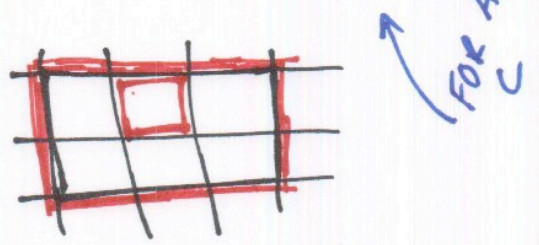
	c	q_7	b
	c	a	q_8



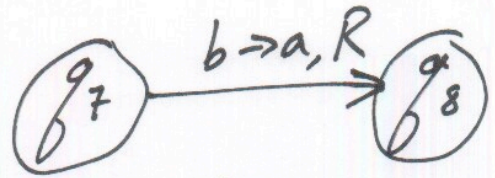
	c	q_7	b
	q_8	c	a

THE "WINDOW" CENTERED ON CELL i, j .

THE TRANSITION FUNCTION TELLS US WHAT THE LEGAL WINDOWS ARE.

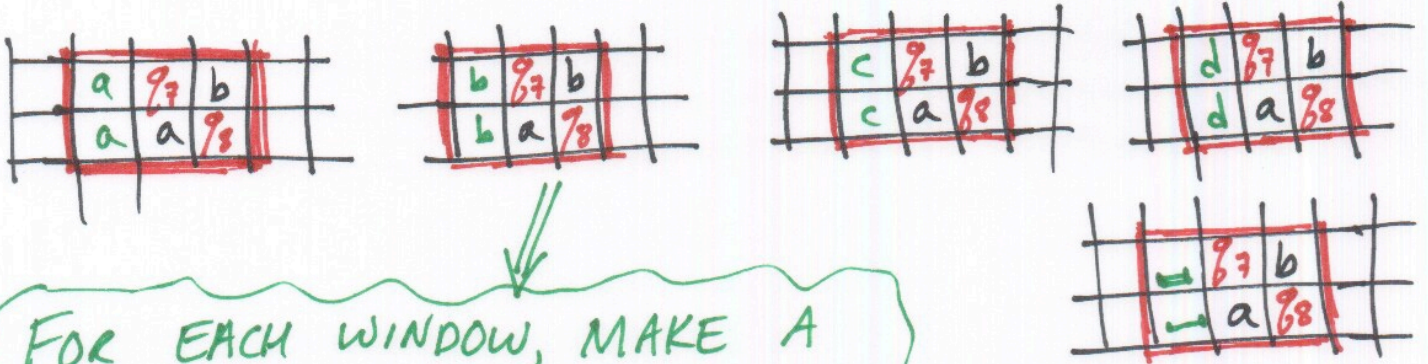


THIS IS A TRANSITION:

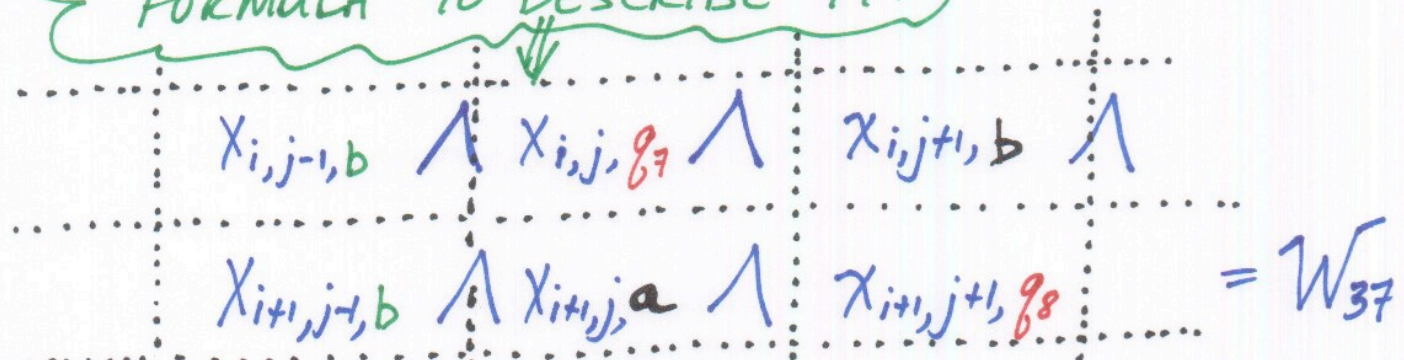


SO THESE ARE LEGAL "WINDOWS":

$$\Gamma = \{a, b, c, d, _ \}$$



FOR EACH WINDOW, MAKE A FORMULA TO DESCRIBE IT.



GIVEN AN i, j POSITION, IT MUST CONTAIN (OR MATCH) ONE OF THE LEGAL WINDOWS.

$$W_1 \vee W_2 \vee W_3 \vee \dots \vee W_{37} \vee \dots \vee W_{592}$$

NOW MAKE SURE EVERY WINDOW IN THE TABLE IS LEGAL.

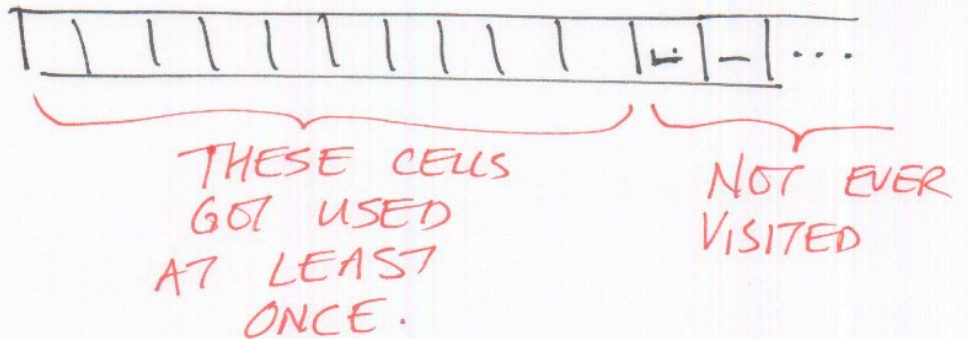
$$\emptyset_{\text{MOVE}} = \bigwedge_{1 \leq i, j \leq N^k} \left(\bigvee_{\text{ALL 592 LEGAL WINDOWS}} (X \wedge X \wedge X \wedge X \wedge X) \right)$$

SPACE
COMPLEXITY

SPACE COMPLEXITY

How to measure?

The number of cells on the tape that we visit.



THE CLASS P-SPACE

QUESTION: What is the relationship between P ~~SPACE~~ AND PSPACE?

- An algorithm that uses 30 tape cells must use at least 30 time steps.
- An algorithm that uses 30 tape cells may use many more steps.

$$P \subseteq PSPACE$$

Most problems are in NP

BUT...

There are problems in PSPACE
for which there is no known
NP algorithm!

Game

- 2 Players; they alternate.
- Each says ~~the~~ the name of a geographic place.
- Kids play this in the car.

Player 1: Portland

Player 2: Denver

Player 1: Rio

Until one
player
gets stuck.

- There is a list/dictionary of valid words. Each word can only be used once.

The Problem: Given the dictionary,
~~can~~ can the 1st player win if
he chooses carefully?

AND-OR TREE (MIN-MAX SEARCH)

$$\exists x_1 \forall x_2 \exists x_3 \forall x_4 \exists x_5 \dots$$

Non-determinism doesn't seem to help trim the search time.

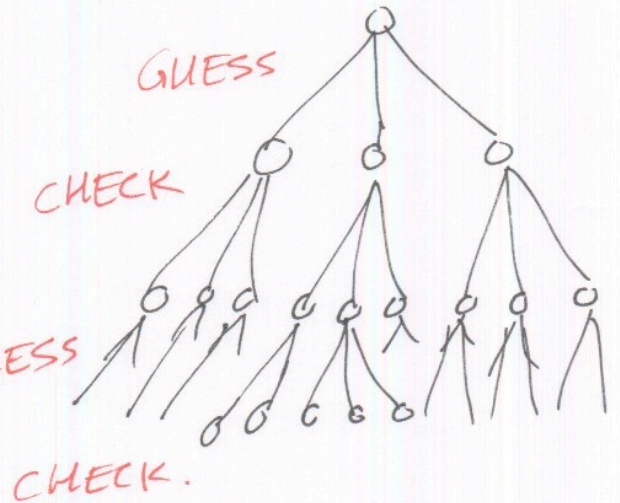
- Guess a good move for me.
- Check all his possible moves.
- Guess another good move.
- Check all his possible moves.
- \vdots

Non-determinism helps here

But does not help here.

A P-SPACE ALGORITHM

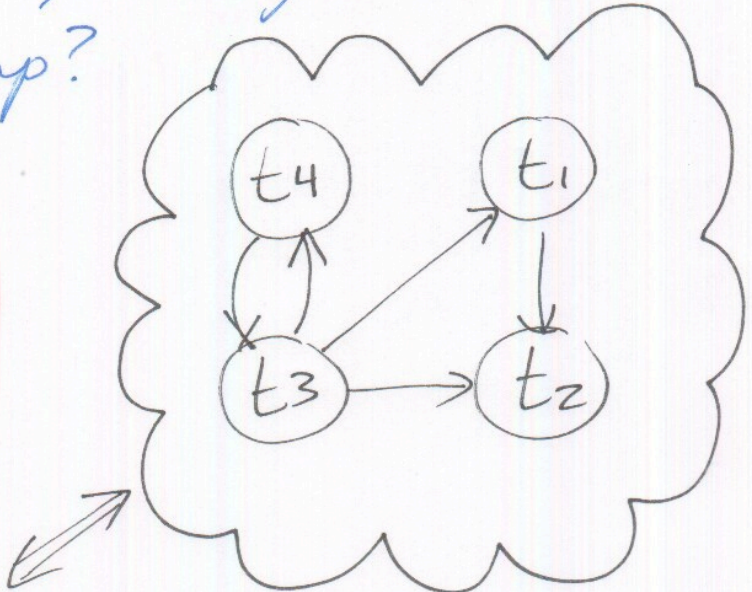
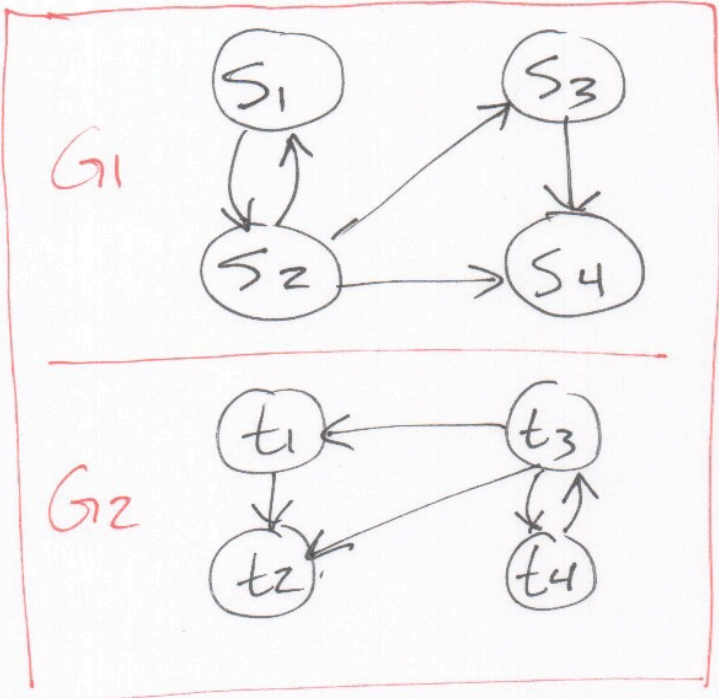
- This is a search of a tree.
- The tree is exponential in size.
- ⇒ We cannot store the tree.
- Do a depth first search of this tree.
- Time taken to search the tree: EXPONENTIAL.



GRAPH ISOMORPHISM

s_1	s_2	s_3	s_4	
t_1	t_2	t_3	t_4	← NO
4	3	1	2	← YES

Given two graphs, can you match them up?



PROBLEM:

Are 2 graphs isomorphic?

This problem is in NP.

Given an answer/correspondence, it can be checked in Polynomial time.

BTW: This problem is NOT NP-complete

Are 2 graphs NOT isomorphic?

This problem is NOT in NP.

There are $N!$ different possible correspondences.

You have to check each of them.

P	Det. TM - in Poly time
NP	NonDet. TM in Poly time
PSPACE	Det/NonDet TM - Poly space
EXPTIME	Det. TM in exponential time
EXPSPACE	Det. TM - exponential space.

WE KNOW:

$$P \subseteq NP \subseteq PSPACE \subseteq EXPTIME \subseteq EXPSPACE$$

$$P \subset EXPTIME$$

$$PSPACE \subset EXPSPACE$$