

# CoCoNuT - Complexity - Lecture 2

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# Outline

Non Deterministic Turing Machines

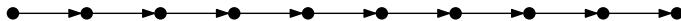
Cook-Levin Proof

More NP-complete Languages

# Non-deterministic Turing machines

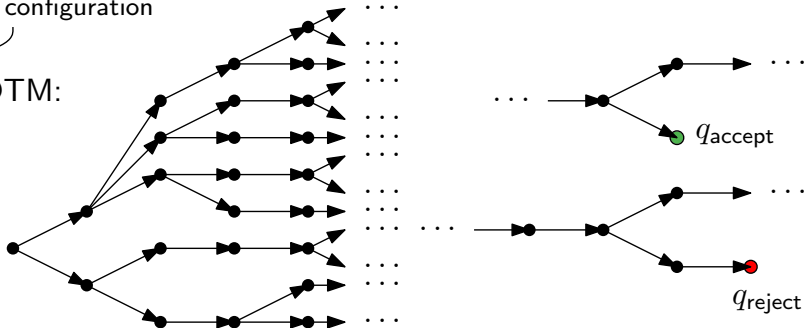
A node  $\bullet$  is a configuration

DTM:



Start configuration

NDTM:



# Time complexity for NTMs

**Definition.** Let  $N$  be a NTM which is a decider. The **running time** of  $N$  is  $f : \mathbb{N} \rightarrow \mathbb{N}$ , where  $f(n)$  is the maximum number of steps that  $N$  uses *on any branch of its computation on any input of length  $n$ .*

**Theorem.** (*Time-complexity of NTM simulation.*) Every  $t(n)$ -time NTM  $N$  has an **equivalent**  $2^{O(t(n))}$ -time TM  $M$ .

(equivalent:  $N$  and  $M$  decide the same language.)

# Nondeterministic polynomial-time

## Definitions

Let  $t : \mathbb{N} \rightarrow \mathbb{R}^+$  be a function. Then  $\text{NTIME}(t(n))$  is the set of all languages which are decidable by a **nondeterministic** Turing machine running in time  $O(t(n))$ .

- ▶ We say that an NDTM runs in time  $O(t(n))$  if **all of its computational paths** halt in time  $O(t(n))$ .

We will now see that there is a close connection between nondeterministic Turing machines and verification of proofs.

This will explain the name **NP**, which stands for “Nondeterministic Polynomial-time” (and **not** “non-polynomial time”).

# Nondeterministic polynomial-time

## Theorem

$$\text{NP} = \bigcup_{k \geq 0} \text{NTIME}(n^k).$$

**Proof (sketch)** We first show that  $\mathcal{L} \in \text{NP}$  implies that  $\mathcal{L}$  is decided by an NDTM running in polynomial time (i.e.  $\text{NP} \subseteq \bigcup_{k \geq 0} \text{NTIME}(n^k)$ ).

- ▶ From the definition of NP, there exists a verifier  $V$  for  $\mathcal{L}$  running in time  $O(n^k)$  for some fixed  $k$ .
- ▶ We define an NDTM  $N$  which behaves as follows:
  1. Nondeterministically guess a string  $c$  of length at most  $O(n^k)$ .
  2. Run  $V$  on input  $x, c$ . If  $V$  accepts, accept; otherwise, reject.

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- ▶ From the definition of NTIME, there exists an NDTM  $N$  which decides  $\mathcal{L}$  and runs in time  $O(n^k)$  for some fixed  $k$ .
- ▶ We define a verifier  $V$  which takes as input a string  $x$  and a witness  $c$ .  $c$  is the description of a sequence of computational path choices made by  $N$ .
- ▶  $V$  simulates the computation of  $N$  on input  $x$  according to  $c$ , checking whether each transition made is valid.
- ▶ If  $N$  accepts in the end,  $V$  accepts; otherwise,  $V$  rejects.

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# Cook-Levin Proof

We are now going to prove **the** major theorem in Complexity Theory; that *SAT* is NP-complete.

Let  $L$  be an language in NP.

We need to show this can be encoded as a SAT problem.

Then if we have an algorithm which solves SAT we can use it to decide  $L$ .

Hence SAT is NP-complete

# Cook-Levin

Think of NP as NDTM's

There is an execution path, i.e. a DTM execution, which contains an ACCEPT state.

Let  $\mathbb{T}$  be the DTM corresponding to this path.

- ▶ The input  $x_1, \dots, x_n$  is the problem instance
- ▶ Assume to be bits.
- ▶ The alphabet is  $\{\triangleright, \square, 0, 1\}$
- ▶ The state space is  $\Sigma$ , which is of constant size  $S$ .
- ▶ We assume that the states ACCEPT, START  $\in \Sigma$ .
- ▶ Let  $\bar{\Sigma} = \Sigma \cup \{\triangleright, \square, 0, 1\}$ .



# Cook-Levin

Since  $L \in \text{NP}$  there is a path of the NDTM:

- ▶ Whose run time is bounded by some constant  $T = n^c$
- ▶ Which accepts the input

We can describe the execution of the NDTM via a tableau  
Which for a *specific execution path* could look like:

▷	START	$x_1$	$x_2$	...	$x_n$	<input type="checkbox"/>	...	<input type="checkbox"/>
▷	0	$s$	$x_2$	...	$x_n$	<input type="checkbox"/>	...	<input type="checkbox"/>
⋮								
▷	0	1	ACCEPT	...	1	0	...	<input type="checkbox"/>

where  $s \in \Sigma$ .

# Tableau

▷	START	$x_1$	$x_2$	...	$x_n$	□	...	□
▷	0	$s$	$x_2$	...	$x_n$	□	...	□
⋮								
▷	0	1	ACCEPT	...	1	0	...	□

The tableau will have at most  $T$  rows, and  $T + 2$  columns.

The total number of possible tableau is given by

$$T \times (T + 2) \times S \times 3^T$$

since each row can have at most one state symbol which comes from  $\Sigma$  and all the other symbols are on the tapes and so (bar the first column) are from  $\{0, 1, \square\}$ .

The  $3^T$  term looks like a problem....

- ▶ But we know at least one such tableau accepts.

# Obtaining a CNF

Let the variables for the CNF be  $x_{i,j,s}$ , where  $s \in \overline{\Sigma}$  variable for each cell.

- ▶ The variable  $x_{i,j,s}$  is true if cell  $(i,j)$  contains  $s$ .
- ▶ Total number of variables is  $T \cdot (T + 1) \cdot (S + 4)$ , i.e. polynomial in  $n$ .

We want a (poly-size) CNF which is satisfiable iff *there is a tableau* which is valid.

# Obtaining a CNF

Define four sub-formulae:

- ▶  $\phi_{\text{accept}}$  evaluates to true if the tableau contains an accepting state.
- ▶  $\phi_{\text{start}}$  evaluates to true if the first row is the correct starting configuration;
- ▶  $\phi_{\text{cell}}$  evaluates to true if all squares in the tableau are uniquely filled;
- ▶  $\phi_{\text{move}}$  evaluates to true if the tableau is a valid computational path of  $M$  (according to its transition rules);

Then our desired CNF is  $\phi$ ,

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

This is the easiest one we just require at least one cell to contain an ACCEPT symbol.

So

$$\phi_{\text{accept}} = \bigvee_{i,j} c_{i,j,\text{ACCEPT}}.$$

Note this is a poly-size CNF.

Need the first row to contain:

- ▶  $\triangleright$  in the first column.
- ▶ START in the second column.
- ▶ Then the  $n$  input values.
- ▶ Then blank values.

$$\begin{aligned}\phi_{\text{start}} = & c_{1,1,\triangleright} \wedge c_{1,2,\text{START}} \wedge c_{1,3,x_1} \wedge c_{1,4,x_2} \wedge \cdots \wedge c_{1,n+2,x_n} \\ & \wedge c_{1,n+3,\square} \wedge \cdots \wedge c_{1,T+2,\square}.\end{aligned}$$

Note this is a poly-size CNF.

- ▶ So the CNF depends on the problem instance

We need to encode

- ▶ Every cell has a value
- ▶ No cell has two values

So we have

$$\phi_{\text{cell}} = \bigwedge_{i,j} \left( \left( \bigvee_{s \in \bar{\Sigma}} c_{i,j,s} \right) \bigwedge_{t \neq u, t, u \in \bar{\Sigma}} (\neg c_{i,j,t} \vee \neg c_{i,j,u}) \right).$$

Note this is a poly-size CNF.

This is the big problem and where we need to cope with the  $3^T$  term above.

We use the *locality* of Turing machines.

In particular that a TM can only affect cells around the head tape.

Consider a  $2 \times 3$  “window” (submatrix) in a tableau.

- ▶ There are  $(T - 1) \times (T - 1)$  possible window locations,
- ▶ i.e. a poly number.

Each window can contain  $(S + 4)^6$  possible values

- ▶ i.e. a poly number.



However, some of the windows are “illegal”

The illegal ones correspond to transitions which cannot take place.

### Example

The head cannot move two positions in one step,

So if  $s$  is a head state the following window is illegal

$s$	0	0
0	0	$s$

The constraint that a window  $w$  is legal can be written as

$$\phi_w = \bigvee_{\text{legal windows } v} [w_{ij} = v_{ij} \text{ for } i \in \{1, 2\}, j \in \{1, 2, 3\}].$$

Note that this constraint is a boolean formula with a fixed, finite number of terms, which can be written in CNF.

Note as we are talking about a NDTM there are multiple paths, and hence multiple legal tableau for any specific point we have reached in the computation path.

- ▶ We only need one path to exist
- ▶ i.e. one assignment to the variables in our SAT problem
- ▶ But we encode *all* paths in the legal moves.

We set

$$\phi_{\text{move}} = \bigwedge_{\text{windows } w} \phi_w.$$

Need to show that if all windows in a tableau are legal, then each row is a valid configuration which follows from the previous one.

Do this by induction:

- ▶ For the base case,  $\phi_{\text{start}}$  ensures that the first row is valid.
- ▶ If row  $r$  is a valid configuration, then row  $r + 1$  is also a valid configuration.
  - ▶ Need to show each row only contains one head configuration.
  - ▶ We don't know where it is, but we do know there is only one.
  - ▶ See notes for how this is done (it's easy).

## Obtaining a CNF

That the resulting CNF is poly-size is trivial to see as all components are poly-size

### So what have we done?

Given a problem  $L \in \text{NP}$  we have

- ▶ Used the fact *there exists* a NDTM accepting  $L$  to encode a tableau.
- ▶ The tableau defines a SAT formula. The variables being the cells in the tableau.
- ▶ If there is an accepting state then for  $L$  then there is a solution to this SAT formulae.
- ▶ Calling a sub-procedure to solve SAT will decide if there is an accepting assignment.
- ▶ Thus if we can solve SAT then we can decide any language in NP.
- ▶ Thus SAT is NP-complete.

# All is not as it seems

There is a rather cool logical issue with the previous slide

We started with a problem which we knew was in NP

Then we showed that we could decide whether it was in NP.

The issue is that NP is defined by the fact that **there exists** a NDTM (or verifier)

It does not say we can find it!

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# Constructive

The point is the proof is non-constructive.

- ▶ Given an NP language you cannot write down the SAT formula via the proof.
- ▶ Many parts of math are non-constructive, e.g. mean value theorem.

However, **in practice** for **most** problems in NP we know an **explicit** verifier.

Given this **explicit** verifier writing down the SAT formulae is relatively easy.

So in practice if we have an efficient procedure to solve SAT, then we can solve **any** problem in NP efficiently.

# More NP-complete Languages

## 3-SAT

- ▶ **Literal:**  $x$  or  $\bar{x}$
- ▶ **Clause:** Disjunction of literals, e.g.  $(x_1 \vee \bar{x}_2 \vee x_3)$
- ▶  $\phi$  is in **conjunctive normal form** if  $\phi$  is a conjunction of clauses
- ▶ **3-CNF formula:** A CNF formula with all clauses having 3 literals, e.g.  $(x_1 \vee \bar{x}_2 \vee \bar{x}_3) \wedge (x_2 \vee \bar{x}_5 \vee x_6) \wedge (x_3 \vee \bar{x}_6 \vee x_4)$ .

$$3\text{-SAT} := \{ \langle \phi \rangle \mid \phi \text{ is a satisfiable 3-CNF formula} \}$$

**Theorem.** 3-SAT is NP-complete.

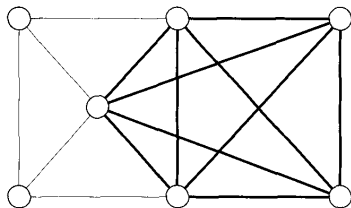
**Theorem.** 2-SAT  $\in$  P.



# More NP-complete Languages

A  $k$ -**clique** in a graph is a set of  $k$  nodes in which every two nodes are connected by an edge.

$CLIQUE := \{(G, k) \mid G \text{ is an undirected graph with a } k\text{-clique}\}$

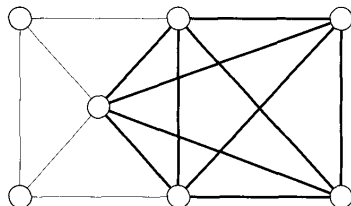


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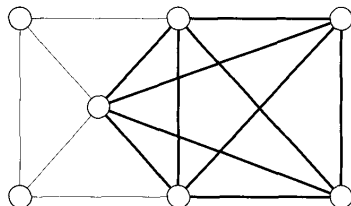
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More NP-complete languages:

- ▶  $HAMPATH$

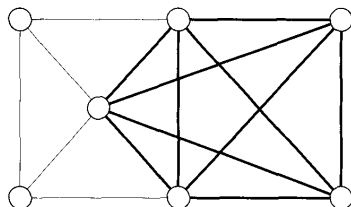


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More NP-complete languages:

- ▶  $HAMPATH$
- ▶  $SUBSET-SUM = \{\{x_1, \dots, x_k\} \mid$   
for some  $\{y_1, \dots, y_\ell\} \subseteq \{x_1, \dots, x_k\}$  we have:  $\sum y_i = 0\}$

# CLIQUE is NP-complete: Proof Example

Clearly  $CLIQUE \in NP$

To show  $CLIQUE$  is NP-complete, we reduce 3-SAT to  $CLIQUE$ .

Given a 3-SAT formulae  $\phi$  with  $m$  clauses create a graph  $G$ :

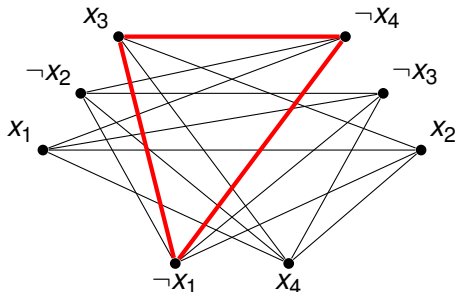
- ▶ Have at most  $3 \cdot m$  vertices, one for each term in a clause.
- ▶ Associate at most three vertices with each clause in  $\phi$ .
- ▶ Connect every pair of vertices with an edge
  - ▶ Except those in the same clause
  - ▶ Except vertices associated to a variable and its negation

# CLIQUE is NP-complete: Proof Example

The graph for the formulae

$$(x_1 \vee \neg x_2 \vee x_3) \wedge (\neg x_1 \vee x_4) \wedge (x_2 \vee \neg x_3 \vee \neg x_4)$$

is...



One 3 clique is highlighted corresponding to the witness

$$\{x_3 = 1, \neg x_1 = 1, \neg x_4 = 1\}.$$

# CLIQUE is NP-complete: Proof Example

If  $\phi$  is satisfiable then  $G$  has an  $m$ -clique:

- ▶ Let  $w$  be the witness to satisfiability
- ▶ Form set  $S$  of variables by picking  $x_i$  in each clause  $C$  such that
  - ▶  $x_i = w_i$
  - ▶  $x_i$  satisfies  $C$ .
- ▶ There is an edge from every element in  $S$  to every other element
- ▶ Thus  $S$  is a clique

# CLIQUE is NP-complete: Proof Example

If  $G$  has an  $m$ -clique then  $\phi$  is satisfiable:

- ▶ Any  $m$ -clique must contain exactly one vertex from each clause
- ▶ The vertices are consistent (as  $x$  and  $\neg x$  not connected)
- ▶ Thus taking the labels for the vertices as being true we get a satisfying assignment

**Summary:** A procedure to determine decidability of CLIQUE implies a procedure to determine decidability of 3-SAT.

As 3-SAT is NP-complete (proved in notes), so is CLIQUE.