



Aalto University  
School of Science

# CS-E4530 Computational Complexity Theory

## Lecture 6: The Cook–Levin Theorem

Aalto University  
School of Science  
Department of Computer Science

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# Agenda

- Boolean satisfiability
- CNF formulas and Boolean functions
- The Cook–Levin theorem

# NP-complete Problems

- *Last lecture:*

- ▶ We saw that TMSAT is NP-complete
- ▶ Definition tied directly to the definition of NP
- ▶ Does not really tell us anything new about NP

- *This lecture:*

- ▶ Prove that a problem called *CNF-SAT* is NP-complete
- ▶ First example of a *natural* NP-complete problem
- ▶ Starting point for further NP-completeness proofs

# Cook–Levin Theorem

- One of the founding results of computational complexity
  - ▶ CNF-SAT is NP-complete
  - ▶ Named after Stephen Cook and Leonid Levin
  - ▶ Both independently proved the theorem around 1971



Stephen Cook



Leonid Levin

# Boolean Formulas

- **Boolean formula is built from the following primitives:**
  - ▶ *Variables*  $x_1, x_2, \dots, x_n$
  - ▶ *Operators* AND ( $\wedge$ ), OR ( $\vee$ ), and NOT ( $\neg$ )
  - ▶ **Example:**  $\varphi = (x_1 \wedge x_2) \vee (x_2 \wedge x_3) \vee (x_3 \wedge x_1)$

## Definition (Boolean formulas, recursive definition)

The set of *Boolean formulas* over variables  $x_1, x_2, \dots, x_k$  is defined as follows:

- $x_i$  is a Boolean formula for any  $i = 1, 2, \dots, n$ .
- If  $\varphi$  is Boolean formula, then  $\neg\varphi$  is Boolean formula.
- If  $\varphi$  and  $\psi$  are Boolean formulas, then  $\varphi \wedge \psi$  and  $\varphi \vee \psi$  are Boolean formulas.

# Value of a Boolean Formula

- An assignment gives value 1 (true) or 0 (false) to each variable
  - ▶ Semantics of NOT, AND and OR are defined in the obvious way

## Definition (Value of a Boolean formula)

Let  $z = (z_1, z_2, \dots, z_n) \in \{0, 1\}^n$  be an assignment. The *value  $\varphi(z)$  of formula  $\varphi$  under assignment  $z$*  is defined as follows:

- If  $\varphi = x_i$ , when  $\varphi(z) = z_i$ .
- If  $\varphi = \neg\psi$ , then  $\varphi(z) = 1 - \psi(z)$ .
- If  $\varphi = \psi_1 \wedge \psi_2$ , then  $\varphi(z) = 1$  if  $\psi_1(z) = \psi_2(z) = 1$ , and  $\varphi(z) = 0$  otherwise.
- If  $\varphi = \psi_1 \vee \psi_2$ , then  $\varphi(z) = 1$  if  $\psi_1(z) = 1$  or  $\psi_2(z) = 1$ , and  $\varphi(z) = 0$  otherwise.

# Value of a Boolean Formula

- **Assignment  $z$  *satisfies* formula  $\varphi$  if  $\varphi(z) = 1$** 
  - ▶ A formula is *satisfiable* if there is a satisfying assignment
  - ▶ A formula is *unsatisfiable* otherwise
  
- **Examples:**
  - ▶  $\varphi = (x_1 \wedge x_2) \vee (x_2 \wedge x_3) \vee (x_3 \wedge x_1)$
  - ▶  $\varphi$  is satisfiable
  
  - ▶  $\psi = (x_1 \vee \neg x_2) \wedge \neg x_1 \wedge x_2$
  - ▶  $\psi$  is unsatisfiable

# Conjunctive Normal Form

- A formula in *conjunctive normal form* is a formula that is an AND of ORs:
  - ▶ **Example:**  $(x_1 \vee x_2) \wedge (x_2 \vee \neg x_3) \wedge (x_3 \vee \neg x_1)$
- **Formally:**
  - ▶ A CNF formula is a formula of form

$$\bigwedge_{i=1}^m \left( \bigvee_{j=1}^k \ell_{i,j} \right),$$

where each  $\ell_{i,j}$  is either an  $x$  or  $\neg x$  for some variable  $x$

- ▶ Terms  $\ell_{i,j}$  are called *literals*
- ▶ Terms  $\bigvee_{j=1}^k \ell_{i,j}$  are called *clauses*



# CNF-SAT and k-SAT

## Definition (CNF-SAT)

- **Instance:** A CNF formula  $\varphi$ .
- **Question:** Is  $\varphi$  satisfiable?

## Definition (k-SAT)

- **Instance:** A CNF formula  $\varphi$  such that each clause in  $\varphi$  has at most  $k$  literals.
- **Question:** Is  $\varphi$  satisfiable?

- **2-SAT instance:**  $(x_1 \vee x_2) \wedge (x_2 \vee \neg x_3) \wedge (x_3 \vee \neg x_1)$

- **3-SAT instance:**  $(x_1 \vee x_2 \vee x_3) \wedge (x_2 \vee \neg x_3 \vee x_4)$

- **4-SAT instance:**  $(x_1 \vee x_2 \vee x_3 \vee \neg x_4) \wedge (x_2 \vee \neg x_3 \vee x_4)$

# CNF-SAT and NP

## Theorem

*CNF-SAT is in NP.*

- **Proof:** CNF-SAT has a polynomial-time verifier
  - ▶ *Input:* a formula  $\varphi$  over variables  $x_1, x_2, \dots, x_n$
  - ▶ *Certificate:* an assignment  $z \in \{0, 1\}^n$
  - ▶ *Verification algorithm:* compute the value  $\varphi(z)$ , accept if  $\varphi(z) = 1$

## Corollary

*For any fixed  $k \geq 1$ ,  $k$ -SAT is in NP.*

# Universality of CNF Formulas

- **CNF formulas can express all Boolean functions**
  - ▶ May require exponential number of clauses
  - ▶ This does not matter: we want to use this construction for *constant* number of variables

## Lemma

*Let  $f: \{0, 1\}^n \rightarrow \{0, 1\}$  be a Boolean function. Then there is a CNF formula  $\varphi$  over  $n$  variables with at most  $2^n$  clauses such that  $\varphi(z) = f(z)$  for all  $z \in \{0, 1\}^n$ .*

# Universality of CNF Formulas: Proof

- **For each  $z \in \{0, 1\}^n$ , we construct clause  $C_z$ :**

- ▶ Let  $\ell_i = x_i$  if  $z_i = 0$ , and  $\ell_i = \neg x_i$  if  $z_i = 1$
- ▶ Let  $C_z = \bigvee_{i=1}^n \ell_i$
- ▶ We now have  $C_z(y) = 0$  if  $z = y$ , and  $C_z(y) = 1$  if  $z \neq y$

- **For any  $f: \{0, 1\}^n \rightarrow \{0, 1\}$ , we construct formula  $\phi$ :**

- ▶ Let  $\phi_f = \bigwedge_{z: f(z)=0} C_z$
- ▶ If  $f(y) = 0$ , then  $y$  does not satisfy the clause  $C_y$  in  $\phi_f$
- ▶ If  $f(y) = 1$ , then  $y$  satisfies all clauses  $C_y$  in  $\phi_f$
- ▶ Thus, we have  $\phi_f(y) = f(y)$  for all  $y \in \{0, 1\}^n$

# Cook–Levin Theorem

## Theorem

*CNF-SAT is NP-complete.*

- **We have:** CNF-SAT is in NP
- **Next:** CNF-SAT is NP-hard

# Cook–Levin Theorem: Proof

- **General template for the proof:**
  - ▶ Let  $L \in \text{NP}$  be a language
  - ▶ We prove that there is a polynomial-time reduction from  $L$  to CNF-SAT
- **The only thing we know about  $L$  is that it is in NP**
  - ▶ There exists a verifier  $M$  for  $L$
  - ▶ For any  $x \in L$ , there is a certificate for  $x$  of length at most  $q(|x|)$ , for some polynomial  $q$
  - ▶  $M$  runs on input  $(x, u)$  in time  $p(|x|)$  for some polynomial  $p$  with  $q(n) \leq p(n)$
  - ▶  $M$  uses at most  $p(|x|)$  positions on each tape
  - ▶ We may assume  $M$  has one working tape, uses alphabet  $\{\triangleright, \square, 0, 1\}$

# Execution Tables

- **Execution of  $M$  on input  $(x, u)$  can be viewed as a table:**

- ▶ Row  $i$  describes the state of  $M$ , the positions of heads and the contents of the tapes after step  $i$
- ▶ Since  $M$  runs in time  $p(|x|)$ , each row needs to store at most  $1 + 3 \cdot 2 \cdot p(|x|)$  entries
  - three tapes, one (head,symbol)-pair per position on a tape
- ▶ The number of rows is at most  $p(|x|)$ , and wlog we may assume exactly  $p(|x|)$  (no moves after  $M$  enters halting state)

- **Table can be encoded as binary:**

- ▶  $|Q|$  bits for state
- ▶ 3 bits per each tape position on each of 3 tapes
  - 1 bit for head marker (location indicator)
  - 2 bits for current symbol encoding

# Execution Tables

- Execution table is *accepting* if:

- ▶ State entry on the last row corresponds to the halting state
- ▶ The encoding of the output tape on the last row corresponds to  
▷  $1□□\dots$

- By definition:

- ▶  $M$  has accepting execution table if and only if  $M$  accepts



# Cook–Levin Theorem: Proof

- **Proof overview:**

- ▶ Let  $x$  be an instance of  $L$
- ▶ We construct a CNF-SAT formula  $\varphi_x$  over  $S = p(|x|) \cdot (|Q| + 9p(|x|))$  variables
- ▶ Assignment  $z$  to  $\varphi_x$  encodes an execution table of  $M$
- ▶ Formula  $\varphi_x$  is construed so that a given assignment  $z$  satisfies  $\varphi_x$  if and only if:
  - (i)  $z$  encodes a valid execution table
  - (ii)  $z$  encodes an execution table on input  $(x, u)$  for some  $u \in \{0, 1\}^*$
  - (iii)  $z$  encodes an accepting execution table

# Cook–Levin Theorem: Proof

- **Clauses of  $\varphi_x$  are constructed to *locally* enforce the constraints**
  - ▶ We could use the universality lemma to directly construct a CNF formula to enforce that the variables encode an accepting execution table
  - ▶ This would give *exponential* size in terms of  $|x|$
  - ▶ Need to be more careful to get polynomial size
  
- **Basic idea: encode *local* constraints**

# Cook–Levin Theorem: Proof

- **Clauses of  $\varphi_x$  that enforce the starting and halting conditions:**
  - ▶ Contents of the input tape on the first row of the execution table is  $\triangleright x \square \square \dots$  and of the other tapes  $\triangleright \square \square \dots$
  - ▶ All heads start at position 1 and the first state is  $q_0$
  - ▶ State on the last row of the execution table is  $q_h$
  - ▶ Contents of the output tape on the last row of the execution table is  $\triangleright 1 \square \square \dots$
- **These can be encoded by a conjunction of  $O(p(|x|))$  single-literal clauses**

# Cook–Levin Theorem: Proof

- **Clauses of  $\varphi_x$  that enforce consistency of the table:**
  - ▶ Only single head position and state for each row
  - ▶ If head *is not* at position  $j$ , then the tape symbols at position  $j$  do not change between steps
  - ▶ If head *is* at position  $j$ , then the tape symbols and the head markers around position  $j$  change correctly between steps
  - ▶ The machine state changes correctly between steps
- **Each of these conditions can be viewed as a Boolean function on a *constant* number of variables**
  - ▶ At most  $c = 2|Q| + 3 \cdot 2 \cdot 6$  variables per constraint
  - ▶ Encode as a CNF with  $2^c$  clauses using universality lemma
  - ▶ About  $O(p(|x|)^2)$  constraints needed

# Cook–Levin Theorem: Proof

- **The final CNF formula  $\phi_x$  is conjunction of all constraints:**
  - ▶ A conjunction of CNF formulas is a CNF formula
- **By construction, this gives us a reduction from  $L$  to CNF-SAT**
  - ▶  $x \in L$  if and only if  $\phi_x$  is satisfiable
  - ▶  $\phi_x$  can be constructed in polynomial time

- **CNF-SAT is a relevant problem in practice**
  - ▶ *Problem-specific* reductions to CNF-SAT can be much more compact than the general reduction given by Cook–Levin theorem
  - ▶ E.g. Intel has used CNF-SAT solvers to verify and optimise processor designs
  - ▶ Highly efficient CNF-SAT solvers are available as open-source software
  - ▶ For many difficult optimisation problems, reducing to CNF-SAT and applying off-the-shelf solvers can be much faster than anything you implement yourself

# Lecture 6: Summary

- CNF-SAT and  $k$ -SAT
- CNF-SAT is NP-complete