

Algorithms for Computational Logic

Introduction

Emmanuel Hebrard (adapted from **João Marques Silva**, Inês Lynce and Vasco Manquinho)



LAAS-CNRS
/ Laboratoire d'analyse et d'architecture des systèmes du CNRS

Laboratoire conventionné
avec l'Université Fédérale
de Toulouse Midi-Pyrénées



Outline

- 1 The Complexity of SAT
- 2 The Tractability of SAT Fragments

1 The Complexity of SAT

- P and NP
- Cook-Levin Theorem

2 The Tractability of SAT Fragments

- Tractable Fragments

Cook-Levin Theorem

SAT is NP-complete

- SAT is "*at least as hard*" as **any** problem in NP
 - ▶ If there exists a polynomial algorithm for SAT then there exists one for every problem in NP
 - ▶ If $SAT \in P$ then $NP = P$

- Recall:

P

Set of problems that are solved by a *polynomial Turing Machine* (running in $\mathcal{O}(n^c)$ time for a constant c)

NP

Set of problems that are solved by a polynomial *Non-determinist* Turing Machine (running in $\mathcal{O}(n^c)$ time for a constant c)

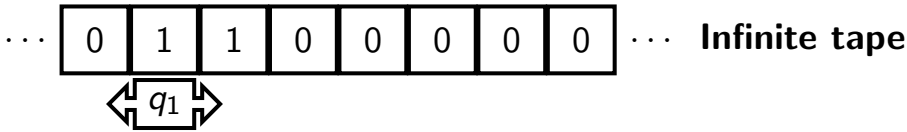
NP-hard problem

A problem Q is **NP-hard** if it is “at least as hard as the hardest problem in **NP**”: if Q can be solved in $\mathcal{O}(T)$ time then any problem in **NP** can be solved in $\mathcal{O}(Tn^c)$ time for some constant c .

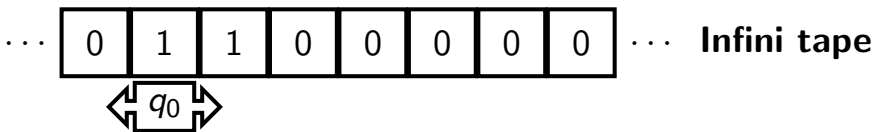
- If an **NP-hard** problem can be solved in polynomial time, then $\mathbf{P} = \mathbf{NP}$

NP-complete problem

A problem Q is **NP-complete** if it is **NP-hard** and is in **NP**



- An infinite **tape**, where we can read/write the symbols **0** and **1** and a **head**
- A “program”
 - ▶ A finite set of **states** with an initial state q_0 and a final state q_f .
 - ▶ A **transition table** associating a triplet $\langle \text{state, symbol, } \{\leftarrow, \rightarrow\} \rangle$ to every pair $\langle \text{state, symbol} \rangle$
- Meaning: “if reading symbol x in state q then write x' , change to state q' and move right/left”



| état | symbol | |
|-------|-----------------------|-----------------------|
| | 0 | 1 |
| q_0 | $q_f, 0, *$ | $q_1, 0, \rightarrow$ |
| q_1 | $q_2, 0, \rightarrow$ | $q_1, 1, \rightarrow$ |
| q_2 | $q_3, 1, \leftarrow$ | $q_2, 1, \rightarrow$ |
| q_3 | $q_4, 0, \leftarrow$ | $q_3, 1, \leftarrow$ |
| q_4 | $q_0, 1, \rightarrow$ | $q_4, 1, \leftarrow$ |

- A non-determinist Turing Machine can have several transitions in the same configuration
- We assume that it makes the right choice (or explore all possible choices in parallel)
- It is sufficient to have up two transitions for any one configuration

| état | symbol | |
|-------|--|-----------------------|
| | 0 | 1 |
| q_0 | $q_f, 0, *$ | $q_1, 0, \rightarrow$ |
| q_1 | $q_2, 0, \rightarrow$ ou $q_4, 1, \leftarrow$ | $q_1, 1, \rightarrow$ |
| q_2 | $q_3, 1, \leftarrow$ | $q_2, 1, \rightarrow$ |
| q_3 | $q_4, 0, \leftarrow$ | $q_3, 1, \leftarrow$ |
| q_4 | $q_0, 1, \rightarrow$ | $q_4, 1, \leftarrow$ |

- Consider a problem Q and a Turing machine that solves it in polynomial time: $\mathcal{O}(n^c)$ pour une donnée de taille n
- This machine executes $\mathcal{O}(n^c)$ instructions and therefore requires a tape of length $\mathcal{O}(n^c)$
- We build the propositional logic formula with the following variables:
 - ▶ A variable $R_{i,t}$ for every cell i of the tape, every symbol k and every time step t : **true iff the symbol v written on cell i at time t is k** ($\mathcal{O}(1)$ symbols, hence $\mathcal{O}(n^{2c})$ variables)
 - ▶ A variable $L_{i,t}$ for every cell i of the tape and every time step t : **true iff the head is at position i at time t** ($\mathcal{O}(n^{2c})$ variables)
 - ▶ A variable $Q_{j,t}$ for every state q_j of the program and every time step t : **true iff the machine is in state q_j at time t** ($\mathcal{O}(1)$ states, hence $\mathcal{O}(n^c)$ variables)

| état | symbol | |
|-------|-----------------------|-----------------------|
| | 00 | 1 |
| q_0 | $q_f, 0, *$ | $q_1, 0, \rightarrow$ |
| q_1 | $q_2, 0, \rightarrow$ | $q_1, 1, \rightarrow$ |
| q_2 | $q_3, 1, \leftarrow$ | $q_2, 1, \rightarrow$ |
| q_3 | $q_4, 0, \leftarrow$ | $q_3, 1, \leftarrow$ |
| q_4 | $q_0, 1, \rightarrow$ | $q_4, 1, \leftarrow$ |

- For a transition $(q_2, 0) \Rightarrow (q_3, 1, \leftarrow)$, we add the following clauses, for all i and all t :
 - ▶ $Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow Q_{3,t+1}$
 - ▶ $Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow L_{i-1,t+1}$
 - ▶ $Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow R_{i,1,t+1}$
- $\Theta(n^c)$ other clauses

- Consider a problem $Q \in \mathbf{P}$
- Q admits a Turing machine that runs in $\mathcal{O}(|x|^{c_1})$ time
- For any input x , there exists a Horn Formula $\phi(Q, x)$ such that:
 - ▶ $\phi(Q, x)$ is satisfiable if and only if $Q(x) = \text{true}$
 - ▶ $|\phi(Q, x)| \in \mathcal{O}(|x|^{c_2})$
- An algorithm for *Horn-SAT* can solve any problem in \mathbf{P} in polynomial time
 - ▶ Not so useful in itself (though *Horn-SAT* is \mathbf{P} -complete for log space reductions)

- Can we come up with a similar encoding for *non-deterministic* machines ?

| état | symbol | |
|-------|-----------------------|-----------------------|
| | 0 | 1 |
| q_0 | $q_f, 0, *$ | $q_1, 0, \rightarrow$ |
| q_1 | $q_2, 0, \rightarrow$ | $q_1, 1, \rightarrow$ |
| q_2 | $q_3, 1, \leftarrow$ | $q_2, 1, \rightarrow$ |
| | $q_4, 0, \rightarrow$ | |
| q_3 | $q_4, 0, \leftarrow$ | $q_3, 1, \leftarrow$ |
| q_4 | $q_0, 1, \rightarrow$ | $q_4, 1, \leftarrow$ |

- There are $\mathcal{O}(1)$ non-deterministic transitions (in the program)
- We add a variable $X_{l,t}$ for every non-deterministic transition l and for every time t
- The transition clauses become:
 - ▶ $X_{l,t} \wedge Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow Q_{3,t+1}$
 - ▶ $X_{l,t} \wedge Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow L_{i-1,t+1}$
 - ▶ $X_{l,t} \wedge Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow R_{i,1,t+1}$
 - ▶ $\neg X_{l,t} \wedge Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow Q_{4,t+1}$
 - ▶ $\neg X_{l,t} \wedge Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow L_{i+1,t+1}$
 - ▶ $\neg X_{l,t} \wedge Q_{2,t} \wedge L_{i,t} \wedge R_{i,0,t} \Rightarrow R_{i,0,t+1}$
- They are not Horn anymore
Otherwise we would have shown $\mathbf{P} = \mathbf{NP}$!

Preuve

- Consider a problem $Q \in \mathbf{P}$
- Q admits a non-determinist Turing machine that runs in $\mathcal{O}(|x|^{c_1})$ time
- For any input x there exists a Boolean formula $\phi(Q, x)$ such that:
 - ▶ $\phi(Q, x)$ is satisfiable if and only if $x \in \text{true}(Q)$ et $|\phi(Q, x)| \in \mathcal{O}(|x|^{c_2})$
- All problems in \mathbf{NP} reduce to SAT
 - ▶ If SAT is in \mathbf{P} , then all problems in \mathbf{NP} can be solved in polynomial time and therefore $\mathbf{P} = \mathbf{NP}$
 - ▶ If SAT is not in \mathbf{P} , then $\mathbf{P} \neq \mathbf{NP}$
- Si $\text{SAT} \in \mathbf{P}$ alors on peut trouver une interprétation de $\phi(Q, x)$ en temps polynomial, et donc résoudre Q en temps polynomial, quel que soit $Q \in \mathbf{NP}$
- Donc $\text{SAT} \in \mathbf{P}$ implique $\mathbf{P} = \mathbf{NP}$!

- 1 The Complexity of SAT
 - \mathbf{P} and \mathbf{NP}
 - Cook-Levin Theorem
- 2 The Tractability of SAT Fragments
 - Tractable Fragments

- SAT is **NP**-complete (Cook's theorem)

- 3-SAT is hard: **Exercise**

- ▶ Encoding:

$$(p_1 \vee p_2 \vee x) \wedge (\neg x \vee p_3 \vee \dots \vee p_k) \iff (p_1 \vee p_2 \vee \dots \vee p_k)$$

- 2-SAT is easy (Resolution)

- *Horn*-SAT is easy (Unit propagation)

Ladner's Theorem

If $\mathbf{P} = \mathbf{NP}$, then there are problems in \mathbf{NP} that are neither in \mathbf{P} nor \mathbf{NP} -complete.

- For instance GRAPHISOMORPHISM may be such problem; or FACTORISATION
- What about fragments of SAT?
 - ▶ We know some are easy (2-SAT, *Horn*-SAT), are there others?
 - ▶ How do we know which ones are hard and which ones are easy?
 - ▶ Are there some in the intermediate class?

Constraint Satisfaction Problem (CSP)

Data: a triplet $(\mathcal{X}, \mathcal{D}, \mathcal{C})$ where:

- \mathcal{X} is a ordered set of *variables*
- \mathcal{D} is a *domain*
- \mathcal{C} is a set of *constraints*, where for $c \in \mathcal{C}$:
 - ▶ its *scope* $S(c)$ is a list of variables
 - ▶ its *relation* $R(c)$ is a subset of $\mathcal{D}^{|S(c)|}$

Question: does there exist a solution $\sigma \in \mathcal{D}^{|\mathcal{X}|}$ such that for every $c \in \mathcal{C}$, $\sigma(S(c)) \in R(c)$?

Projection

The projection $\sigma(X)$ of a tuple σ on a set of variables $X = (x_{i_1}, \dots, x_{i_k}) \subseteq \mathcal{X}$ as the tuple $(\sigma(x_{i_1}), \dots, \sigma(x_{i_k}))$

- Example: the constraint $x + y = z$ (on the Boolean ring)

| x | y | z | $S(x + y = z)$ |
|-----|-----|-----|----------------|
| 0 | 0 | 0 | $R(x + y = z)$ |
| 0 | 1 | 1 | |
| 1 | 0 | 1 | |
| 1 | 1 | 0 | |

- A relation $R(c)$ over some variables can easily be expressed in clausal form
- Each clause excludes exactly one tuple, example: $x + y + z \neq 2$

| $x + y + z \neq 2$ | $x + y + z = 2$ | \iff | CNF |
|--------------------|-----------------|------------------------------------|--|
| x y z | x y z | | |
| 0 0 0 | 0 1 1 | $(\bar{x} \wedge y \wedge z) \vee$ | $(x \vee \bar{y} \vee \bar{z}) \wedge$ |
| 0 0 1 | 1 0 1 | $(x \wedge \bar{y} \wedge z) \vee$ | $(\bar{x} \vee y \vee \bar{z}) \wedge$ |
| 0 1 0 | 1 1 0 | $(x \wedge y \wedge \bar{z}) \vee$ | $(\bar{x} \vee \bar{y} \vee z) \wedge$ |
| 1 0 0 | | | |
| 1 1 1 | | | |

- A clause is a particular case of relation on the Boolean domain

- We can define fragments of CSP via restrictions on the *domain*, the *structure* or on the *language*
 - ▶ **Domain:** Boolean CSPs: $\mathcal{D} = \{0, 1\}$, Three-valued CSPs, CSP on \mathbb{Z} , etc.
 - ▶ **Structure:** e.g., the incidence graph (bipartite graph variables / constraints) is a *tree* or has a bounded *treewidth*
 - ▶ **Language:** the library of relations is restricted to a given set Γ

Language fragment

$\text{CSP}(\Gamma)$ is the problem of deciding the satisfiability of a CSP whose constraints all have relations in Γ .

- For instance Three-valued $\text{CSP}(\{\neq\})$ is **NP-hard** since 3-COLORATION is **NP-hard**

pp-definability

A relation R over x_1, \dots, x_k on domain \mathcal{D} is (*pp*-)definable from a set of relation Γ if and only if there exists a CSP $\mathcal{N} = (\mathcal{X}, \mathcal{D}, \mathcal{C})$ such that:

- $\{x_1, \dots, x_k\} \subseteq \mathcal{X}$
- $c \in \mathcal{C} \implies R(c) \in \Gamma \cup \{=\}$
- $R(x_1, \dots, x_k) \iff (x_1, \dots, x_k)$ can be extended to a solution of \mathcal{N}

- i.e., the relation R can be encoded using relations in Γ
 - ▶ $<$ is definable from $\{\leq, \neq\}$
 - ▶ A k -clause $(p_1 \vee \dots \vee p_k)$ is definable from 3-clauses
 - ▶ All k -ary relations are definable from k -clauses

Closure

$\ll \Gamma \gg$ is the set of relations that are definable from Γ

- $\text{CSP}(\Gamma)$ and $\text{CSP}(\ll \Gamma \gg)$ have the same complexity
- Boolean CSPs whose incidence graph is such that constraints vertices have degree 2 (constraints are on at most 2 variables) is in \mathbf{P}
 - ▶ Any binary relation is definable by binary clauses
 - ▶ If Γ is the languages composed of 2-clauses, $\{(x \vee y), (\bar{x} \vee y), (\bar{x} \vee \bar{y})\}$, then:
 - ★ $\text{CSP}(\Gamma)$ is 2-SAT
 - ★ $\text{CSP}(\ll \Gamma \gg)$ is "Boolean binary CSP"

Schaefer's Theorem

Boolean $\text{CSP}(\ll \Gamma \gg)$ is in \mathbf{P} if:

- Γ are 2-clauses
- Γ are Horn-clauses
- Γ are dual Horn-clauses
- $\Gamma = \{\oplus\}$ (i.e., XOR. Also known as "AFFINE-SAT")
- Every relation in Γ accepts the tuple with only 0
- Every relation in Γ accepts the tuple with only 1

and is \mathbf{NP} -hard otherwise

- *Dichotomy*: we know the complexity of *all* the language-based fragments of SAT, and none of them is an intermediate problem