

11 Intractable Problems

Theory of computability is the study of what can or cannot be computed by a TM/computer, among all problems. Theory of complexity is the study of what can or cannot be computed efficiently by a TM/computer, among all decidable problems.

11.1 Defining Complexity Classes

- A complexity class is specified by several parameters.
 - Model of computation. (e.g., k -string TM.)
 - Mode of computation: How a machine accepts its input. (e.g., deterministic/nondeterministic modes.)
 - Resource to measure. (e.g., time and space.)
 - Bound on resource. (e.g., polynomial time and exponential space.)
- Some complexity class forms: $\mathbf{TIME}(f(n))$, $\mathbf{SPACE}(f(n))$, $\mathbf{NTIME}(f(n))$, $\mathbf{NSPACE}(f(n))$.
($\mathbf{TIME}(f(n)) \subseteq \mathbf{NTIME}(f(n))$, $\mathbf{SPACE}(f(n)) \subseteq \mathbf{NSPACE}(f(n))$)
($\mathbf{NTIME}(f(n)) \subseteq \mathbf{SPACE}(f(n))$, $\mathbf{NSPACE}(f(n)) \subseteq \mathbf{TIME}(k^{\log n + f(n)})$)

- Some complexity classes:
 - $\mathbf{P} = \cup_{k \geq 0} \mathbf{TIME}(n^k)$.
 - $\mathbf{NP} = \cup_{k \geq 0} \mathbf{NTIME}(n^k)$. ($\mathbf{P} \subseteq \mathbf{NP}$)
 - $\mathbf{PSPACE} = \cup_{k \geq 0} \mathbf{SPACE}(n^k)$. ($\mathbf{NP} \subseteq \mathbf{PSPACE}$)
 - $\mathbf{NPSPACE} = \cup_{k \geq 0} \mathbf{NSPACE}(n^k)$. ($\mathbf{PSPACE} = \mathbf{NPSPACE}$)
 - $\mathbf{EXP} = \cup_{k \geq 0} \mathbf{TIME}(2^{n^k})$. ($\mathbf{P} \subseteq \mathbf{EXP}$, $\mathbf{NP} \subseteq \mathbf{EXP}$, $\mathbf{NPSPACE} \subseteq \mathbf{EXP}$)
 - $\mathbf{L} = \mathbf{SPACE}(\log n)$.
 - $\mathbf{NL} = \mathbf{NSPACE}(\log n)$. ($\mathbf{L} \subseteq \mathbf{NL} \subseteq \mathbf{P}$)
 - $\mathbf{coP} = \{L : \bar{L} \in \mathbf{P}\}$. ($\mathbf{P} = \mathbf{coP}$)
 - $\mathbf{coNP} = \{L : \bar{L} \in \mathbf{NP}\}$. ($\mathbf{NP} = \mathbf{coNP}???$)
 - $\mathbf{coPSPACE} = \{L : \bar{L} \in \mathbf{PSPACE}\}$. ($\mathbf{PSPACE} = \mathbf{coPSPACE}$)
 - $\mathbf{coNPSPACE} = \{L : \bar{L} \in \mathbf{NPSPACE}\}$. ($\mathbf{NPSPACE} = \mathbf{coNPSPACE}$)

Remark: For any complexity class C , its complement $\mathbf{co}C = \{L : \bar{L} \in C\}$. If C is deterministic, then $C = \mathbf{co}C$. However, if C is nondeterministic, it is sometimes not known whether $C = \mathbf{co}C$.

11.2 The class P

Reading: Sipser 7.2

- Definition: \mathbf{P} is the class of problems solvable in polynomial time (number of steps) by deterministic TMs. $O(n^c)$, where c is a constant.
Tractable (not so hard).
- Why use polynomial as the criterion?
If a problem is not in \mathbf{P} , it will be extremely expensive and probably impossible to solve in practice for large sizes.
Polynomials have nice closure properties: $+$, $-$, $*$, and composition.
 \mathbf{P} is independent of models of computation: Basic TM, TMs with multi-tracks, multi-tapes, multi-cursors, multi-strings, multi-dimensional tape. (All variations of TM except nondeterministic TM.)
- Examples of problems in \mathbf{P} : Sorting, Searching, Selecting, Minimum Spanning Tree, Shortest Path, etc..

11.3 The class NP

Reading: Sipser 7.3

- An NTM is an unrealistic (unreasonable) model of computation which can be simulated by other models with an exponential loss of efficiency. It is a useful concept that has had great impact on the theory of computation.
- NTM $N = (Q, \Sigma, \Gamma, \delta, q_0, q_{accept}, q_{reject})$, where $\delta : Q \times \Gamma \rightarrow 2^P$ for $P = Q \times \Gamma \times \{L, R\}$.
- $\delta(q, X)$ is a set a moves. Which one to choose? This is nondeterminism. The computation can be illustrated by a tree, with each node representing a configuration.
- Nondeterminism can be viewed as a kind of parallel computation wherein multiple independent processes or threads can be running concurrently. When a nondeterministic machine splits into several choices, that corresponds to a process forking into several children, each then proceeding separately. If at least one process accepts, then the entire computation accepts.
- Time complexity of nondeterministic TMs (NTMs): Let N be an NTM that is a decider (where all computation paths halt in the tree for any input). The time complexity of N , $f(n)$, is the maximum number of steps that N uses on any computation path for any input of length n . In other words, $f(n)$ is the maximum height of all computation trees for all input of length n .

- An unreasonable model of computation:

Theorem; Every $T(n)$ -time multi-tape TM has an equivalent $O(T^2(n))$ -time single-tape TM.

Theorem: Every $T(n)$ -time single-tape NTM has an equivalent $O(2^{O(T(n))})$ -time single-tape DTM.

- Definition: **NP** is the class of problems solvable in polynomial time by nondeterministic TMs.
- Another definition of nondeterministic TMs (algorithms):
 - Guessing phase: Guess a solution (always on target).
 - Verifying phase: Verify the solution.
- Example: TSP (DEC) is in **NP**.

INSTANCE: An edge-weighted, undirected, and complete graph $G(V, E, w)$ and a bound $B \geq 0$. (The weight on an edge is allowed to be $+\infty$.)

QUESTION: Is there a tour (a cycle that passes through each node exactly once) in G with total weight no more than B ?

A NTM N that decides TSP in polynomial time can be defined as follows: Assume that a coded instance of TSP is shown on the tape as input. N first generates an arbitrary permutation of the n nodes and then checks whether the permutation represents a tour with total weight no larger than the bound B . If so, N accepts the input. If not, N rejects the input, perhaps hoping that another choice of permutations may end up accepting. Note that the permutation generated can be any of the $n!$ possible permutations. The nondeterminism of the machine guarantees that if the instance is a yes-instance, the right permutation will always be selected.

Note: The acceptance of an input by a nondeterministic machine is determined by whether there is an accepting computation among all, possibly exponentially many, computations. In the above proof, if there is a solution, it will always be generated by the Turing machine. This is like that a nondeterministic machine has a guessing power. A tour can only be found by a deterministic machine in exponential time, however, it can be found by a nondeterministic machine in just one step. So any nondeterministic proof should always contain two stages: Guessing and verifying.

- Example: Graph Coloring is in **NP**. Consider the decision problem: For a given graph, is there a coloring scheme of the nodes that uses no more than B colors such that no two nodes connected by an edge are given the same color? A nondeterministic TM first guesses a coloring scheme (in polynomial time) and then verifies if the coloring uses no more than B colors and if any two adjacent nodes have different colors (in polynomial time).

- Theorem: $\mathbf{P} \subseteq \mathbf{NP}$.
Any deterministic TM is a special case of a nondeterministic TM.
- Theorem: Any $\Pi \in \mathbf{NP}$ can be solved by a deterministic TM in time $O(c^{p(n)})$ for some $c > 0$ and polynomial $p(n)$.
- Open problem: $\mathbf{NP} \subseteq \mathbf{P}$? or $\mathbf{P} = \mathbf{NP}$?

11.4 Polynomial reduction

Reading: Sipser 7.4

- Definition: Let Π_1 and Π_2 be two decision problems, and $\{I_1\}$ and $\{I_2\}$ be sets of instances for Π_1 and Π_2 , respectively. We say there is a polynomial reduction from Π_1 to Π_2 , or $\Pi_1 \leq_p \Pi_2$ if there is $f: \{I_1\} \rightarrow$ subset of $\{I_2\}$ such that (1) f can be computed in polynomial time and (2) I_1 has a “yes” solution if and only if $f(I_1)$ has a “yes” solution.
- Theorem: If $\Pi_1 \leq_p \Pi_2$, then $\Pi_2 \in \mathbf{P}$ implies $\Pi_1 \in \mathbf{P}$.
- Theorem: If $\Pi_1 \leq_p \Pi_2$ and $\Pi_2 \leq_p \Pi_3$, then $\Pi_1 \leq_p \Pi_3$.
- Remark: \leq_p means “no harder than”.

11.5 The class NPC

Reading: Sipser 7.4

- Definition: **NPC** (**NP**-complete) is the class of the hardest problems in **NP**, or equivalently, $\Pi \in \mathbf{NPC}$ if $\Pi \in \mathbf{NP}$ and $\forall \Pi' \in \mathbf{NP}, \Pi' \leq_p \Pi$.
- Theorem: If $\Pi_1 \in \mathbf{NPC}, \Pi_2 \in \mathbf{NP}$, and $\Pi_1 \leq_p \Pi_2$, then $\Pi_2 \in \mathbf{NPC}$.
- Theorem: If $\exists \Pi \in \mathbf{NPC}$ such that $\Pi \in \mathbf{P}$, then $\mathbf{P} = \mathbf{NP}$.
- Theorem: If $\exists \Pi \in \mathbf{NPC}$ such that $\Pi \notin \mathbf{P}$, then $\mathbf{P} \neq \mathbf{NP}$.
- Satisfiability (SAT):
 INSTANCE: A boolean formula α in *CNF* with variables x_1, \dots, x_n (e.g., $\alpha = (x_1 \vee x_2) \wedge (\neg x_2 \vee x_3)$).
 QUESTION: Is α satisfiable? (Is there a truth assignment to x_1, \dots, x_n such that α is true?)
 Cook’s Theorem: $\text{SAT} \in \mathbf{NPC}$. (Need to prove (1) $\text{SAT} \in \mathbf{NP}$ and (2) $\forall \Pi \in \mathbf{NP}, \Pi \leq_p \text{SAT}$.)
- Proof of Cook’s Theorem:
 SAT is clearly in **NP** since a NTM exists that guesses a truth assignment and verifies its correctness in polynomial time. Now we wish to prove $\forall \Pi \in \mathbf{NP}, \Pi \leq_p \text{SAT}$, or equivalently, for any polynomial-time NTM $M, L(M) \leq_p L_{\text{SAT}}$.
 For any NTM M , assume $Q = \{q_0, q_1(\text{accept}), q_2(\text{reject}), \dots, q_r\}$ and $\Gamma = \{s_0, s_1, s_2, \dots, s_v\}$. Also assume that the time is bounded by $p(n)$, where n is the length of the input. We wish to prove that there is a function $f_M: \Sigma^* \rightarrow \{\text{instances of SAT}\}$ such that $\forall x \in \Sigma^*, x \in L(M)$ iff $f_M(x)$ is satisfiable. In other words, we wish to use a Boolean expression $f_M(x)$ to describe the computation of M on x .
 Variables in $f_M(x)$:
 — State: $Q[i, k]$. M is in q_k after the i th step of computation (at time i).
 — Head: $H[i, j]$. Head points to tape square j at time i .
 — Symbol: $S[i, j, l]$. Tape square j contains s_l at time i . (Assume the tape is one-way infinite and the leftmost square is labeled with 0.)

For example, initially $i = 0$. Assume the configuration is q_0abba . Let $s_0 = B$, $s_1 = a$, and $s_2 = b$. Therefore, we set the following Boolean variables to be true: $Q[0,0]$, $H[0,0]$, $S[0,0,1]$, $S[0,1,2]$, $S[0,2,2]$, $S[0,3,1]$ and $S[0,j,0]$ for $j = 4,5,\dots$. A configuration defines a truth assignment, but not vice versa.

Clauses in $f_M(x)$:

— At any time i , M is in exactly one state.

$$Q[i,0] \vee \dots \vee Q[i,r] \text{ for } 0 \leq i \leq p(n).$$

$$\neg Q[i,k] \vee \neg Q[i,k'] \text{ for } 0 \leq i \leq p(n) \text{ and } 0 \leq k < k' \leq r.$$

— At any time i , head is scanning exactly one square.

$$H[i,0] \vee \dots \vee H[i,p(n)] \text{ for } 0 \leq i \leq p(n).$$

$$\neg H[i,j] \vee \neg H[i,j'] \text{ for } 0 \leq i \leq p(n) \text{ and } 0 \leq j < j' \leq p(n).$$

— At any time i , each square contains exactly one symbol.

$$S[i,j,0] \vee \dots \vee S[i,j,v] \text{ for } 0 \leq i \leq p(n) \text{ and } 0 \leq j \leq p(n).$$

$$\neg S[i,j,l] \vee \neg S[i,j,l'] \text{ for } 0 \leq i \leq p(n), 0 \leq j \leq p(n) \text{ and } 0 \leq l < l' \leq v.$$

— At time 0, M is in its initial configuration. Assume $x = s_{l_1} \dots s_{l_n}$.

$$Q[0,0].$$

$$H[0,0].$$

$$S[0,0,l_1], \dots, S[0,n-1,l_n].$$

$$S[0,j,0] \text{ for } n \leq j \leq p(n).$$

— By time $p(n)$, M has entered q_1 (accept). (If M halts in less than $p(n)$ steps, additional moves can be included in the transition function.)

$$Q[p(n),1].$$

— Configuration at time $i \rightarrow$ configuration at time $i+1$. Assume $\delta(q_k, s_l) = (q_{k'}, s_{l'}, D)$, where $D = -1, 1$.

If the head does not point to square j , symbol on j is not changed from time i to time $i+1$.

$$H[i,j] \vee \neg S[i,j,l] \vee S[i+1,j,l] \text{ for } 0 \leq i \leq p(n), 0 \leq j \leq p(n), \text{ and } 0 \leq l \leq v.$$

If the current state is q_k , the head points to square j which contains symbol s_l , then changes are made accordingly.

$$\neg H[i,j] \vee \neg Q[i,k] \vee \neg S[i,j,l] \vee H[i+1,j+D],$$

$$\neg H[i,j] \vee \neg Q[i,k] \vee \neg S[i,j,l] \vee Q[i+1,k'], \text{ and}$$

$$\neg H[i,j] \vee \neg Q[i,k] \vee \neg S[i,j,l] \vee S[i+1,j,l'], \text{ for } 0 \leq i \leq p(n), 0 \leq j \leq p(n), 0 \leq k \leq r, \text{ and } 0 \leq l \leq v.$$

Let $f_M(x)$ be the conjunction of all the clauses defined above. Then $x \in L(M)$ iff there is an accepting computation of M on x iff $f_M(x)$ is satisfiable. f_M can be computed in polynomial time since $|f_M(x)| \leq (\text{number of clauses}) * (\text{number of variables}) = O(p(n)^2) * O(p(n)^2) = O(p(n)^4)$.

So there is a polynomial reduction from any language in **NP** to SAT. So SAT is **NP**-complete.

11.6 NP-complete problems

Reading: Sipser 7.5

- How to prove Π is **NP**-complete:
 - Show that $\Pi \in \mathbf{NP}$.
 - Choose a known **NP**-complete Π' .
 - Construct a reduction f from Π' to Π .
 - Prove that f is a polynomial reduction by showing that (1) f can be computed in polynomial time and (2) $\forall I'$ for Π' , I' is a yes-instance for Π' if and only if $f(I')$ is a yes-instance for Π .

- Seven basic **NP**-complete problems.

- 3-Satisfiability (3SAT):

INSTANCE: A formula α in CNF with each clause having three literals.

QUESTION: Is α satisfiable?

- 3-Dimensional Matching (3DM):

INSTANCE: $M \subseteq X \times Y \times Z$, where X, Y, Z are disjoint and of the same size.

QUESTION: Does M contain a matching, which is $M' \subseteq M$ with $|M'| = |X|$ such that no two triples in M' agree in any coordinate?

- PARTITION:

INSTANCE: A finite set A of numbers.

QUESTION: Is there $A' \subseteq A$ such that $\sum_{a \in A'} a = \sum_{a \in A - A'} a$?

- Vertex Cover (VC):

INSTANCE: A graph $G = (V, E)$ and $0 \leq k \leq |V|$.

QUESTION: Is there a vertex cover of size $\leq k$, where a vertex cover is $V' \subseteq V$ such that $\forall (u, v) \in E$, either $u \in V'$ or $v \in V'$?

- Hamiltonian Circuit (HC):

INSTANCE: A graph $G = (V, E)$.

QUESTION: Does G have a Hamiltonian circuit, i.e., a tour that passes through each vertex exactly once?

- CLIQUE:

INSTANCE: A graph $G = (V, E)$ and $0 \leq k \leq |V|$.

QUESTION: Does G contain a clique (complete subgraph) of size $\geq k$?

- A proof that 3SAT is NP-complete:

First, 3SAT is obvious in **NP**.

Next, we show that $\text{SAT} \leq_p 3\text{SAT}$.

Given any instance of SAT, $f(x_1, \dots, x_n) = c_1 \wedge \dots \wedge c_m$, where c_i is a disjunction of literals. To construct an instance for 3SAT, we need to convert any c_i to an equivalent c'_i , a conjunction of clauses with exactly 3 literals.

Case 1. If $c_i = z_1$ (one literal), define y_i^1 and y_i^2 . Let $c'_i = (z_1 \vee y_i^1 \vee y_i^2) \wedge (z_1 \vee y_i^1 \vee \neg y_i^2) \wedge (z_1 \vee \neg y_i^1 \vee y_i^2) \wedge (z_1 \vee \neg y_i^1 \vee \neg y_i^2)$.

Case 2. If $c_i = z_1 \vee z_2$ (two literals), define y_i^1 . Let $c'_i = (z_1 \vee z_2 \vee y_i^1) \wedge (z_1 \vee z_2 \vee \neg y_i^1)$.

Case 3. If $c_i = z_1 \vee z_2 \vee z_3$ (three literals), let $c'_i = c_i$.

Case 4. If $c_i = z_1 \vee z_2 \vee \dots \vee z_k$ ($k > 3$), define $y_i^1, y_i^2, \dots, y_i^{k-3}$. Let $c'_i = (z_1 \vee z_2 \vee y_i^1) \wedge (\neg y_i^1 \vee z_3 \vee y_i^2) \wedge (\neg y_i^2 \vee z_4 \vee y_i^3) \wedge \dots \wedge (\neg y_i^{k-3} \vee z_{k-1} \vee z_k)$.

If c_i is satisfiable, then there is a literal $z_l = T$ in c_i . If $l = 1, 2$, let $y_i^1, \dots, y_i^{k-3} = F$. If $l = k - 1, k$, let $y_i^1, \dots, y_i^{k-3} = T$. If $3 \leq l \leq k - 2$, let $y_i^1, \dots, y_i^{l-2} = T$ and $y_i^{l-1}, \dots, y_i^{k-3} = F$. So c'_i is satisfiable.

If c'_i is satisfiable, assume $z_l = F$ for all $l = 1, \dots, k$. Then $y_i^1, \dots, y_i^{k-3} = T$. So the last clause $(\neg y_i^{k-3} \vee z_{k-1} \vee z_k) = F$. Therefore, c'_i is not satisfiable. Contradiction.

The instance of 3SAT is therefore $f'(x_1, \dots, x_n, \dots) = c'_1 \wedge \dots \wedge c'_m$, and f is satisfiable if and only if f' is satisfiable.

- A proof that CLIQUE is **NP**-complete.

INSTANCE: $G = (V, E)$ and $k \leq |V|$.

QUESTION: Does G contain a clique (a complete subgraph) of size k or more?

Proof. CLIQUE is obviously in **NP**. Now we show that $\text{SAT} \leq_p \text{CLIQUE}$. For any Boolean expression $f(x_1, \dots, x_n) = c_1 \wedge \dots \wedge c_m$, we define a graph G , where $V = \{ \langle y, i \rangle \mid y \text{ is a literal in } c_i \}$ and $E = \{ \langle y, i \rangle \langle z, j \rangle \mid i \neq j$

j and $y \neq \neg z$. Finally, let $k = m$. Since $|V| = O(mn)$ and $|E| = O(m^2n^2)$, the graph can be constructed in polynomial time. Next we show that f is satisfiable iff there is a clique in G with size m .

Assume there is a truth assignment for f such that at least one literal, say l_i , in c_i ($i = 1, \dots, m$) is true. Consider a subgraph of G consisting of m nodes $\langle l_i, i \rangle$. For any $i \neq j$, we have $l_i \neq \neg l_j$. So edge $\langle l_i, i \rangle \langle l_j, j \rangle$ is in G . Therefore the subgraph is a clique of size m .

Assume G contains a clique of size m or more. Let V' be the set of any m nodes in the clique. Since no edge in G connects $\langle y, i \rangle$ and $\langle z, j \rangle$ when $i = j$, and nodes in V' are fully connected, then V' contains $\langle l_1, 1 \rangle, \dots, \langle l_m, m \rangle$, where l_i is a literal in clause c_i . Further $l_i \neq \neg l_j$ for any $i \neq j$. Then there is a truth assignment such that $l_i = \text{true}$ for $i = 1, \dots, m$. Obviously, $f(x_1, \dots, x_n) = c_1 \wedge \dots \wedge c_m$ is satisfiable under the truth assignment.