Computational complexity – Homework

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1 NP-Hardness

1.1 Halting problem

Let be φ an instance of SAT problem. We denote by n the number of variables.

Let be M a TURING machine which tests in a cycle all the 2^n possible assignations of the previous formula : when M has tested all assignations, it starts again. This machine halts if and only if φ is satisfiable. This reduction is polynomial, therefore $SAT \leqslant_p HALT$, ie. HALT is NP-hard since SAT is NP-hard.

HALT is not *NP*-complete otherwise it was decidable by a TURING-machine, but HALT is undecidable.

1.2 *TQBF*

All instance of SAT problem is an instance of TQBF. Without transformation, we have a polynomial reduction, ie. $SAT \leq_p TQBF$ so TQBF is NP-hard.

This problem is known for being PSPACE-complete. It's not NP-complete.

1.3 NAE - 3 - SAT

Let be φ an instance of NAE - 3 - SAT.

$$\varphi = \bigwedge_{i=1}^{n} (x_{i,1} \vee x_{i,2} \vee x_{i,3})$$

We will describe the "not all equal" condition is term of formula.

$$\bigwedge_{i=1}^{n} \neg (x_{i,1} \wedge x_{i,2} \wedge x_{i,3}) \wedge \neg (\neg x_{i,1} \wedge \neg x_{i,2} \wedge \neg x_{i,3})$$

using DE MORGAN's law:

$$\psi := \bigwedge_{i=1}^{n} (\neg x_{i,1} \vee \neg x_{i,2} \vee \neg x_{i,3}) \wedge (x_{i,1} \vee x_{i,2} \vee x_{i,3})$$

$$|\psi| \sim 2|\varphi| \Rightarrow |\psi| = \mathcal{O}(|\varphi|).$$

Let $\omega := \varphi \wedge \psi$. ω is an instance of *SAT* and the reduction is polynomial. If there is an solution to the *SAT* problem ξ , then φ is satisfied and, thanks to ψ , the "not all equal" condition is true.

Reciprocally, if there is a solution of the NAE - 3 - SAT problem φ , then this assignation makes ξ true.

Consequently, $SAT \leq_p NAE - 3 - SAT$ and NAE - 3 - SAT is NP-hard.

Moreover, NAE - 3 - SAT is clearly NP: a valid assignation is a sufficient witness. We can check in polynomial time if this assignation makes the formula true and if the "not all equal" condition is satisfied. Then NAE - 3 - SAT is NP-complete.

1.4 MAXCUT

Let *F* be an instance of NAE - 3 - SAT

$$F = \bigwedge_{i=1}^{m} C_i$$

We produce a graph G = (V, E) which has a vertex for each literal of F. There is a edge between two vertices if there is a clause which contains this two literals. So each clause is described by a triangle. Moreover, we add $|F|_{x_i}$ (the number of occurrences of x_i in F) edges between x_i and $\neg x_i$. The size of the cut we search is at least 5m.

If we have an assignment of the NAE - 3 - SAT, we take the vertices which are true in S and the other in \overline{S} . So, we have 2m from the triangles due to the clauses and 3m from the edges between all pair $(x_i, \neg x_i)$.

Reciprocally, if we have a cut of size $\geq 5m$.

If we have no pair $(x_i, \neg x_i)$ on the same size, we have a valid assignment.

If there is a such pair, we can move one of them on the opposite side without decreasing the size of the cut. Let n_i the number of edges between x_i and $\neg x_i$. We note a the number of edges which x_i is an extremity and which the other is in the opposite side. We note b the number of edges between b and a vertex of the opposite size. We know that $a + b \le 2n_i$. If we move x_i in the opposite size, the cut gains $n_i - a$ edges. If $\neg x_i$ go to the opposite side, it gains $n_i - b$. $\max(n_i - a, n_i - b) \ge 0$, so we can move one of these vertices to the opposite side without decreasing the size of the cut. We redo this transformation until we reach a cut of the first case (at most m times).

We proved that $NAE - 3 - SAT \leq_p MAXCUT$.

Moreover, MAXCUT is in NP. Indeed, a witness is the list of the vertices of S (or \overline{S}). The size is actually polynomial with respect of the size of G and we can check the solution in a polynomial time : we check easily that the cut has a size $\geqslant k$ in a quadratic time.

So, $MAXCUT \in NP$.

2 Reductions

Proposition 1. The set of the recursive languages is closed under polynomial-time KARP reduction.

Proof. That is obvious, isn't it?

Proposition 2. DLOGTIME is not closed under polynomial-time KARP reduction.

Proof. Let $\mathcal{L} \in \mathsf{DLOGTIME}$ such that $|\mathcal{L}| \geqslant 2$. For all $\mathcal{L}' \in \mathsf{PTIME} \setminus \mathsf{DLOGTIME}$ (there is at least one), there is a polynomial time function $g : \mathcal{L}' \to \{0,1\}$, $x \in \mathcal{L}' \Leftrightarrow g(x) = 1$. Let φ a bijection between $\{0,1\}$ and $\{a,b\}$ where $(a,b) \in \mathcal{L} \times \overline{\mathcal{L}}$. We have $x \in \mathcal{L}' \Leftrightarrow \varphi \circ g(x) = \varphi(a)$ ie. $x \in \mathcal{L}' \Leftrightarrow \varphi \circ g(x) \in \mathcal{L}$.

So, DLOGTIME is not close under polynomial-time KARP reduction.

Proposition 3. TAUTOLOGY is NP-hard.

Proof. Let *M* a TURING machine with oracle TAUTOLOGY. We will make that machine decide *SAT* in polynomial time.

This machine take a formula φ compute the negation $\neg \varphi$ and test with its oracle if $\neg \varphi$ is a tautology. If it is not, there is a assignment which make $\neg \varphi$ false, so this assignment make φ true and φ is satisfiable. If $\neg \varphi$ is a tautology, φ is not satisfiable.

So *M* decide *SAT* and TAUTOLOGY is NP-hard.

3 Difference of NP problems

Proposition 4. EXACTINDSET is in DP.

Proof. Let *A* be the set of all pairs (G, k) such that *G* has an independent set of size at least k, and let *B* be the set of all pairs (G, k) such *G* has a independent set of size at least k + 1. Then EXACTINDSET $= A \setminus B$ and *A* is in NP and *B* is in NP. Hence by definition of *DP*, EXACTINDSET is in *DP*.

Proposition 5. $\forall L \in DP$, L is polynomial-time reducible to EXACTINDSET.

Proof.

Lemma 6. INDSET $\geqslant_p 3 - SAT$

Proof. Suppose we have an instance F of 3 - SAT problem where $F = \bigwedge_{i=1}^{m} C_i$ where C_i is the disjunction of 3 variables. We note x_1, \ldots, x_n the variables. We create the graph G as follows:

- For each variable in each clause, create a vertex, which we will label with the name of the variable. Therefore there may be multiple vertexs with the label x_i or $\neg x_i$, if these variables appear in multiple clauses.
- For each clause, add an edge between the three vertices corresponding the variables from that clause.
- For all i, add an edge between every pair of vertexs with one is labelled with x_i and the other labelled with x_i .

There is a independent set of size *m* in *G* if and only if *F* is satisfiable.

We note that this reduction from 3-SAT to INDSET took an instance φ of 3-SAT consisting of m clauses each of three literals and produced a graph G_{φ} with 3m vertexs such that if φ is satisfiable then the largest independent set in G has m vertexs, and if G is unsatisfiable then the largest independent set of G has at most m-1 vertices.

Now suppose that A is in DP. We want to show that $A \leq_p EXACTINDSET$. By definition of DP, $A = L_1 \setminus L_2$ for $(L_1, L_2) \in NP^2$. Since 3 - SAT is NP-complete, there are polytime functions f_1 , f_2 such that for i = 1, 2 and for all $x \in \{0,1\}^*$ we have $x \in L_i \Leftrightarrow f_i(x) \in 3SAT$. Hence for each fixed $x \in \{0,1\}^*$, setting $\varphi_i = f_i(x)$, we have $x \in L_i \Leftrightarrow \varphi_i$ is satisfiable Thus from the above reduction to INDSET, there is a polytime function which takes x to a pair of graphs G_1 , G_2 such that if m_i is the number of clauses in φ_i , then for i = 1, 2, no independent set in G_i has more than m_i vertices, and $x \in L_i \Leftrightarrow$ the largest independent set in G_i has size M_i .

Now we use the notation $G \sqcup H$ for the disjoint union of graphs G and H. That is, the vertices in $G \sqcup H$ are the disjoint union of those in G and H, and similarly for the edges. Now let $G_1' = G_1 \sqcup G_1$ Then a maximum independent set in G_1' is the union of maximum independent sets in the two copies of G_1 . Thus $x \in L_1 \Rightarrow$ maximum independent set of G_1' is $2m_1$ and $x \in \overline{L_1} \Rightarrow$ maximum independent set of G_1' is $2m_1 = 2m_1 =$

and the edges consist of those of G_2 together with an edge from each of the new vertices to every vertex of G_2 . Then we have designed G'2 so that no independent set can contain both vertices of G_2 and new vertices, so $x \in L_2 \Rightarrow$ maximum independent set of G'_2 is m_2 , $x \in \overline{L_2} \Rightarrow$ maximum independent set of G'_2 is m_2 , m_2 , m_2 and let m_2 and let

Lemma 7.

$$x \in A \Leftrightarrow (G_3, k) \in DP$$

(⇒): Suppose $x \in A$. Then by (2) $x \in L_1 \cap L_2$, so by (3) and (6) we conclude the maximum independent set of $G_3 = G_1' \sqcup G_2'$ is $2m_1 + m_2 - 1 = k$.

(\Leftarrow): Suppose $x \notin A$. There are three cases:

- $x \in L_1 \cap L_2 \Rightarrow maxindset(G_3) = 2m_1 + m_2 > k$.
- $x \in L_1 \cap L_2 \Rightarrow maxindset(G_3) \leq 2m_1 + m_2 2 < k$
- $x \in L_1 \cap L_2 \Rightarrow maxindset(G_2) = 2m_1 + m_2 3 < k$

Theorem 8. EXACTINDSET is DP-complete.

Proof. EXACTINDSET is in DP (proposition 4) and it is DP-hard (proposition 5).

4 Classes with exponential resources

We name BOUNDEDHALT the language of 3-tuples $\langle M, x, k \rangle$ where the machine M halts on input x in k steps.

Theorem 9. BOUNDEDHALT is EXP-complete.

Proof.

Lemma 10. BOUNDEDHALT \in EXP.

Proof. Let $\langle M, x, k \rangle$ an instance of BOUNDEDHALT.

We simulate M on x for k steps and accepts if and only if M halts and rejects otherwise. The running time is $m^{\mathcal{O}(1)}$. Let $n = |\langle \alpha, x, k \rangle| \geqslant \log k$. Therefore, the running time $\leqslant m^c = 2^{c \log m} \leqslant 2^{cn}$.

Lemma 11. BOUNDEDHALT is EXP-hard.

Proof. For each language $\mathcal{L} \in \text{EXP}$, we need to give polytime reduction from \mathcal{L} to BOUNDEDHALT. For a given language $\mathcal{L} \in \text{EXP}$, we know there is a TURING MACHINE $M_{\mathcal{L}}$ that decides A in time $g(n) \leq 2^{n^k}$ for a $c \in \mathbb{N}$.

Let f such that $f(w) = \langle M_{\mathcal{L}}, w, m \rangle$ where $m = 2^{|w|^c}$.

f is polytime computable and $w \in \mathcal{L} \Rightarrow \langle M_{\mathcal{L}}, w, m \rangle \in \text{BOUNDEDHALT}$ and $w \notin \mathcal{L} \Rightarrow \langle M_{\mathcal{L}}, w, m \rangle \notin \text{BOUNDEDHALT}$

BOUNDEDHALT is in EXP (lemma 10) and it is EXP-hard (lemma 11).

Theorem 12.

$$L = P \Rightarrow PSPACE = EXP$$

Proof. Take an arbitrary $\mathcal{L} \in \text{EXP}$, decided by a TURING machine in time 2^{n^c} .

Let $L_{pad} := \left\{ x \diamondsuit^{2^{|x|^c}} \mid x \in \mathcal{L} \right\}$. We have $L_{pad} \in P$ and, by the assumption P = L, we found that $L_{pad} \in L$. So, there is some TURING machine M_0 deciding L_{pad} in space $\log n$. M_0 can be used to decide L: on input x, simulate M_0 on the padded input $x \diamondsuit^{2^{|x|^c}}$ accept if and only if M_0 accepts.

The space we used on input x (|x| = n) is the space used by M_0 on the padded input $x \diamondsuit^{2^{|x|^c}}$ of size $n + 2^{n^c}$, which is at most $\log \left(2^{n^c+1}\right) = n^c + 1$, a polynomial. Hence, we have that $\mathcal{L} \in \mathsf{PSPACE}$.

If we actually write the padded input $x \diamondsuit^{2^{|x|^c}}$ on the work tape, this would make the space usage exponential in n. But, we don't need to write entirely this padded input. We know what it looks like to the right of x. So we can simulate M_0 on the virtual padded input $x \diamondsuit^{2^{|x|^c}}$ by using a counter which tells the position on the tape of M_0 . If M_0 enquires about the position to the right of x, we respond with the symbol \diamondsuit (or the blank, if M_0 goes to the right of the entire padded input). Since the counter can assume the value at most 2^{n^c} , we need at most n^c bits for the counter. Thus, this new TURING machine uses space at most polynomial in n.

5 Downward self-reducibility

6 Space hierarchy theorem

Theorem 13. There is a universal TURING machine \mathcal{U} such that $\mathcal{U}(\alpha, x) = M_{\alpha}(x)$ and \mathcal{U} halts on every input (α, x) after using at most cS(|x|) space, where S is the computation space of M_{α} and c is some constant not depending on x.

Theorem 14. For every space-constructible function $f: \mathbb{N} \to \mathbb{N}$,

DSPACE
$$(o(f(n))) \subseteq DSPACE(f(n))$$

Proof. Let $f : \mathbb{N} \to \mathbb{N}$ be a space constructible function and $g : \mathbb{N} \to \mathbb{N}$ such that g = o(f) We note M a deterministic TURING machine. M's computation:

- M writes $1^{f(n)}$ symbols on a work tape, where n is the input's size.
- Simulate $\mathcal{U}(\alpha,(\alpha,x))$, and stops at most when the machine go out of the previous marks.
- If the computation of $\mathcal{U}(\alpha, (\alpha, x))$ is not terminated M rejects.
- Otherwise, outputs the opposite result of $\mathcal{U}(\alpha,(\alpha,x))$

The first step takes a space f(n). The second step takes also a space f(n) (thanks to the restriction). Finally, M runs in space $\mathcal{O}(f(n))$.

We note $\mathcal{L}(M)$ the language recognized by M ($\mathcal{L}(M) \in \mathsf{DSPACE}(f(n))$) and assume that there is a deterministic machine N such that $\mathcal{L}(M) = \mathcal{L}(N)$ and N runs in space g(n). We denote by α_0 the code of N.

According to the theorem 13, for every word x, $\mathcal{U}(\alpha_0, (\alpha_0, x))$ uses less than $cg(|(\alpha_0, x)|)$ space. And, by hypothesis, $cg(|(\alpha_0, x)|) \le f(|(\alpha_0, x)|)$ for large enough x.

Thus, for large enough x we have $M(\alpha_0, x) = \neg \mathcal{U}(\alpha_0, (\alpha_0, x)) = \neg N(\alpha_0, x)$. This is a contradiction since we have assumed $\mathcal{L}(M) = \mathcal{L}(N)$.

In conclusion, there exists no machine N that runs in space $\mathcal{O}(g(n))$ and recognizes $\mathcal{L}(M)$. Since we have proved $\mathcal{L}(M) \in \mathsf{DSPACE}(f(n))$, we obtained $\mathsf{DSPACE}(g(n)) \subsetneq \mathsf{DSPACE}(f(n))$.

7 Polynomial hierarchy