P versus NP

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- Abstract

P versus NP is considered as one of the most important open problems in computer science. This consists in knowing the answer of the following question: Is P equal to NP? It was essentially mentioned in 1955 from a letter written by John Nash to the United States National Security Agency. However, a precise statement of the P versus NP problem was introduced independently by Stephen Cook and Leonid Levin. Since that date, all efforts to find a proof for this problem have failed. Another major complexity classes are L and NL. Whether L = NL is another fundamental question that it is as important as it is unresolved. We demonstrate if L is not equal to NL, then P = NP. In addition, we show if L is equal to NL, then P = NP. In this way, we prove the complexity class P is equal to NP. Furthermore, we demonstrate the complexity class L is equal to NP.

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1 Introduction

In previous years there has been great interest in the verification or checking of computations [18]. Interactive proofs introduced by Goldwasser, Micali and Rackoff and Babi can be viewed as a model of the verification process [18]. Dwork and Stockmeyer and Condon have studied interactive proofs where the verifier is a space bounded computation instead of the original model where the verifier is a time bounded computation [18]. In addition, Blum and Kannan has studied another model where the goal is to check a computation based solely on the final answer [18]. More about probabilistic logarithmic space verifiers and the complexity class NP has been investigated on a technique of Lipton [18]. In this work, we show some results about the logarithmic space verifiers applied to the class NP and logarithmic space disqualifiers applied to the class NP which solve one of the most important open problems in computer science, that is P versus NP.

The P versus NP problem is a major unsolved problem in computer science [6]. This is considered by many to be the most important open problem in the field [6]. It is one of the seven Millennium Prize Problems selected by the Clay Mathematics Institute to carry a US\$1,000,000 prize for the first correct solution [6]. The precise statement of the P = NP problem was introduced in 1971 by Stephen Cook in a seminal paper [6]. In 2012, a poll of 151 researchers showed that 126 (83%) believed the answer to be no, 12 (9%) believed the answer is yes, 5 (3%) believed the question may be independent of the currently accepted axioms and therefore impossible to prove or disprove, 8 (5%) said either do not know or do not care or don't want the answer to be yes nor the problem to be resolved [12].

The P=NP question is also singular in the number of approaches that researchers have brought to bear upon it over the years [9]. From the initial question in logic, the focus moved to complexity theory where early work used diagonalization and relativization techniques [9]. It was showed that these methods were perhaps inadequate to resolve P versus NP by demonstrating relativized worlds in which P=NP and others in which $P\neq NP$ [4]. This shifted the focus to methods using circuit complexity and for a while this approach was deemed the one most likely to resolve the question [9]. Once again, a negative result

showed that a class of techniques known as "Natural Proofs" that subsumed the above could not separate the classes NP and P, provided one-way functions exist [23]. There has been speculation that resolving the P=NP question might be outside the domain of mathematical techniques [9]. More precisely, the question might be independent of standard axioms of set theory [9]. Some results have showed that some relativized versions of the P=NP question are independent of reasonable formalizations of set theory [13].

It is fully expected that $P \neq NP$ [22]. Indeed, if P = NP then there are stunning practical consequences [22]. For that reason, P = NP is considered as a very unlikely event [22]. Certainly, P versus NP is one of the greatest open problems in science and a correct solution for this incognita will have a great impact not only in computer science, but for many other fields as well [1]. Whether P = NP or not is still a controversial and unsolved problem [1]. We show some results that prove this outstanding problem with the unexpected solution of P = NP.

2 Theory and Methods

2.1 Preliminaries

In 1936, Turing developed his theoretical computational model [25]. The deterministic and nondeterministic Turing machines have become in two of the most important definitions related to this theoretical model for computation [25]. A deterministic Turing machine has only one next action for each step defined in its program or transition function [25]. A nondeterministic Turing machine could contain more than one action defined for each step of its program, where this one is no longer a function, but a relation [25].

Let Σ be a finite alphabet with at least two elements, and let Σ^* be the set of finite strings over Σ [3]. A Turing machine M has an associated input alphabet Σ [3]. For each string w in Σ^* there is a computation associated with M on input w [3]. We say that M accepts w if this computation terminates in the accepting state, that is M(w) = "yes" [3]. Note that M fails to accept w either if this computation ends in the rejecting state, that is M(w) = "no", or if the computation fails to terminate, or the computation ends in the halting state with some output, that is M(w) = y (when M outputs the string y on the input w) [3].

Another relevant advance in the last century has been the definition of a complexity class. A language over an alphabet is any set of strings made up of symbols from that alphabet [7]. A complexity class is a set of problems, which are represented as a language, grouped by measures such as the running time, memory, etc [7]. The language accepted by a Turing machine M, denoted L(M), has an associated alphabet Σ and is defined by:

$$L(M) = \{w \in \Sigma^* : M(w) = "yes"\}.$$

Moreover, L(M) is decided by M, when $w \notin L(M)$ if and only if M(w) = "no" [7]. We denote by $t_M(w)$ the number of steps in the computation of M on input w [3]. For $n \in \mathbb{N}$ we denote by $T_M(n)$ the worst case run time of M; that is:

$$T_M(n) = max\{t_M(w) : w \in \Sigma^n\}$$

where Σ^n is the set of all strings over Σ of length n [3]. We say that M runs in polynomial time if there is a constant k such that for all n, $T_M(n) \leq n^k + k$ [3]. In other words, this means the language L(M) can be decided by the Turing machine M in polynomial time. Therefore, P is the complexity class of languages that can be decided by deterministic Turing

machines in polynomial time [7]. A verifier for a language L_1 is a deterministic Turing machine M, where:

```
L_1 = \{w : M(w, c) = "yes" \text{ for some string } c\}.
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We measure the time of a verifier only in terms of the length of w, so a polynomial time verifier runs in polynomial time in the length of w [3]. A verifier uses additional information, represented by the symbol c, to verify that a string w is a member of L_1 . This information is called certificate. NP is the complexity class of languages defined by polynomial time verifiers [22]. If NP is the class of problems that have succinct certificates, then the complexity class coNP must contain those problems that have succinct disqualifications [22]. That is, a "no" instance of a problem in coNP possesses a short proof of its being a "no" instance [22].

▶ **Definition 1.** We will extend the definition of succinct disqualification for an element $w \in L_2$ when $L_2 \in coNP$ as the polynomially bounded string c by w such that M(w,c) = "no" and M is the polynomial time verifier of the complement of L_2 in NP.

2.2 First Hypothesis

A function $f: \Sigma^* \to \Sigma^*$ is a polynomial time computable function if some deterministic Turing machine M, on every input w, halts in polynomial time with just f(w) on its tape [25]. Let $\{0,1\}^*$ be the infinite set of binary strings, we say that a language $L_1 \subseteq \{0,1\}^*$ is polynomial time reducible to a language $L_2 \subseteq \{0,1\}^*$, written $L_1 \leq_p L_2$, if there is a polynomial time computable function $f: \{0,1\}^* \to \{0,1\}^*$ such that for all $x \in \{0,1\}^*$:

```
x \in L_1 if and only if f(x) \in L_2.
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An important complexity class is NP-complete [11]. A language $L_1 \subseteq \{0,1\}^*$ is NP-complete if:

```
■ L_1 \in NP, and
■ L' \leq_p L_1 for every L' \in NP.
```

If L_1 is a language such that $L' \leq_p L_1$ for some $L' \in NP$ -complete, then L_1 is NP-hard [7]. Moreover, if $L_1 \in NP$, then $L_1 \in NP$ -complete [7]. A principal NP-complete problem is SAT [11]. An instance of SAT is a Boolean formula ϕ which is composed of:

- 1. Boolean variables: x_1, x_2, \ldots, x_n ;
- 2. Boolean connectives: Any Boolean function with one or two inputs and one output, such as \land (AND), \lor (OR), \rightarrow (NOT), \Rightarrow (implication), \Leftrightarrow (if and only if);
- 3. and parentheses.

A truth assignment for a Boolean formula ϕ is a set of values for the variables in ϕ . A satisfying truth assignment is a truth assignment that causes ϕ to be evaluated as true. A Boolean formula with a satisfying truth assignment is satisfiable. The problem SAT asks whether a given Boolean formula is satisfiable [11]. We define a CNF Boolean formula using the following terms:

A literal in a Boolean formula is an occurrence of a variable or its negation [7]. A Boolean formula is in conjunctive normal form, or CNF, if it is expressed as an AND of clauses, each of which is the OR of one or more literals [7]. A Boolean formula is in 3-conjunctive normal form or 3CNF, if each clause has exactly three distinct literals [7].

For example, the Boolean formula:

$$(x_1 \lor \neg x_1 \lor \neg x_2) \land (x_3 \lor x_2 \lor x_4) \land (\neg x_1 \lor \neg x_3 \lor \neg x_4)$$

is in 3CNF. The first of its three clauses is $(x_1 \lor \neg x_1 \lor \neg x_2)$, which contains the three literals $x_1, \neg x_1$, and $\neg x_2$. Another relevant NP-complete language is 3CNF satisfiability, or 3SAT [7]. In 3SAT, it is asked whether a given Boolean formula ϕ in 3CNF is satisfiable.

A logarithmic space Turing machine has a read-only input tape, a write-only output tape, and read/write work tapes [25]. The work tapes may contain at most $O(\log n)$ symbols [25]. In computational complexity theory, L is the complexity class containing those decision problems that can be decided by a deterministic logarithmic space Turing machine [22]. NL is the complexity class containing the decision problems that can be decided by a nondeterministic logarithmic space Turing machine [22].

A logarithmic space transducer is a Turing machine with a read-only input tape, a write-only output tape, and read/write work tapes [25]. The work tapes must contain at most $O(\log n)$ symbols [25]. A logarithmic space transducer M computes a function $f: \Sigma^* \to \Sigma^*$, where f(w) is the string remaining on the output tape after M halts when it is started with w on its input tape [25]. We call f a logarithmic space computable function [25]. We say that a language $L_1 \subseteq \{0,1\}^*$ is logarithmic space reducible to a language $L_2 \subseteq \{0,1\}^*$, written $L_1 \leq_l L_2$, if there exists a logarithmic space computable function $f: \{0,1\}^* \to \{0,1\}^*$ such that for all $x \in \{0,1\}^*$:

$$x \in L_1$$
 if and only if $f(x) \in L_2$.

The logarithmic space reduction is used in the definition of the complete languages for the classes L and NL [22]. A Boolean formula is in 2-conjunctive normal form, or 2CNF, if it is in CNF and each clause has exactly two distinct literals. There is a problem called 2SAT, where we asked whether a given Boolean formula ϕ in 2CNF is satisfiable. 2SAT is complete for NL [22]. Another special case is the class of problems where each clause contains XOR (i.e. exclusive or) rather than (plain) OR operators. This is in P, since an XOR SAT formula can also be viewed as a system of linear equations mod 2, and can be solved in cubic time by Gaussian elimination [21]. We denote the XOR function as \oplus . The XOR 2SAT problem will be equivalent to XOR SAT, but the clauses in the formula have exactly two distinct literals. XOR 2SAT is in L [2], [24].

We can give a certificate-based definition for NL [3]. The certificate-based definition of NL assumes that a logarithmic space Turing machine has another separated read-only tape [3]. On each step of the machine, the machine's head on that tape can either stay in place or move to the right [3]. In particular, it cannot reread any bit to the left of where the head currently is [3]. For that reason this kind of special tape is called "read-once" [3]. Besides, in the certificate-based definition of NL, we assume the certificate string is appropriated for the instance [22]. For example, a truth assignment for a Boolean formula ϕ is appropriated for the instance when every possible variable in ϕ could be evaluated in that truth assignment string, but we cannot affirm the same for every possible binary string.

▶ **Definition 2.** A language L_1 is in NL if there exists a deterministic logarithmic space Turing machine M with an additional special read-once input tape polynomial $p : \mathbb{N} \to \mathbb{N}$ such that for every $x \in \{0,1\}^*$:

$$x \in L_1 \Leftrightarrow \exists \ appropriated \ u \in \{0,1\}^{p(|x|)} \ such \ that \ M(x,u) = "yes"$$

where by M(x,u) we denote the computation of M where x is placed on its input tape and the certificate u is placed on its special read-once tape, and M uses at most $O(\log |x|)$ space on

its read/write tapes for every input x where $|\ldots|$ is the bit-length function [3]. M is called a logarithmic space verifier [3].

We state the following Hypothesis:

 \triangleright Hypothesis 3. Given a nonempty language $L_1 \in L$, there is a language L_2 in NP-complete with a deterministic Turing machine M, where:

```
L_2 = \{w : M(w, u) = y, \exists \text{ appropriated } u \text{ such that } y \in L_1\}
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when M runs in logarithmic space in the length of w, u is placed on the special read-once tape of M, and u is polynomially bounded by w. In this way, there is an NP-complete language defined by a logarithmic space verifier M such that when the input is an element of the language with its certificate, then M outputs a string which belongs to a single language in L.

From the early days of automata and complexity theory, two different models of Turing machines are considered, the offline and online machines [17]. Each model has a read-only input tape and some work tapes [17]. The offline machines may read their input two-way while the online machines are not allowed to move the input head to the left [17]. In the terminology of the (generalized) Turing machine models are called two-way and one-way Turing machines, respectively [17].

Hartmanis and Mahaney have investigated the classes 1L and 1NL of languages recognizable by deterministic one-way logarithmic space Turing machine and nondeterministic one-way logarithmic space Turing machine, respectively [14]. They have shown that $1L \neq 1NL$ (by looking at a uniform variant of the string non-equality problem from communication complexity theory) and have defined a natural complete problem for 1NL under deterministic one-way logarithmic space reductions [14]. Furthermore, they have proven that $1NL \subseteq L$ if and only if L = NL [14].

▶ **Theorem 4.** If the Hypothesis 3 is true, therefore when $L \neq NL$, then P = NP.

Proof. We can simulate the computation M(w, u) = y in the Hypothesis 3 by a nondeterministic logarithmic space Turing machine N such that N(w) = y, since we can read the certificate string u within the read-once tape by a work tape in a nondeterministic logarithmic space generation of symbols contained in u [22]. Certainly, we can simulate the reading of one symbol from the string u into the read-once tape just nondeterministically generating the same symbol in the work tapes using a logarithmic space [22].

If we suppose that $L \subset 1NL$, then we can accept the elements of the language $L_1 \in L$ by a nondeterministic one-way logarithmic space Turing machine M'. In this way, there is a nondeterministic logarithmic space Turing machine M''(w) = M'(N(w)) which will accept when $w \in L_2$. Consequently, M'' is a nondeterministic logarithmic space Turing machine which decides the language L_2 . The reason is because we can simulate the output string of N(w) within a read-once tape and thus, we can compute in a nondeterministic logarithmic space the logarithmic space composition using the same techniques of the logarithmic space composition reduction, but without any reset of the computation [22]. Certainly, we do not need to reset the computation of N(w) for the reading at once of a symbol in the output string of N(w) by the nondeterministic one-way logarithmic space Turing machine M'. Therefore, L_2 is in NL and thus, $L_2 \in P$ due to $NL \subseteq P$ [22]. If any single NP-complete problem can be solved in polynomial time, then P = NP [7]. Since $L_2 \in P$ and $L_2 \in NP$ -complete, then we obtain the complexity class P is equal to NP under the assumption that $L \subset 1NL$.

Hartmanis and Mahaney have also shown with their result that if $1NL \subseteq L$ or even $1NL \subset L$, then L = NL, because they proved there is a complete problem for both 1NL and NL at the same time [14]. In this way, if $L \neq NL$, then $L \subset 1NL$ by contraposition [22]. Since we already obtained that P = NP under the assumption that $L \subset 1NL$, therefore if $L \neq NL$, then P = NP.

2.3 Second Hypothesis

An important complexity class is coNP-complete [11]. A language $L_1 \subseteq \{0,1\}^*$ is coNP-complete if:

- $L_1 \in coNP$, and
- $L' \leq_p L_1$ for every $L' \in coNP$.

If L_1 is a language such that $L' \leq_p L_1$ for some $L' \in coNP$ -complete, then L_1 is coNP-hard [7]. Moreover, if $L_1 \in coNP$, then $L_1 \in coNP$ -complete [7]. A principal coNP-complete problem is UNSAT [11]. A Boolean formula without any satisfying truth assignment is unsatisfiable. The problem UNSAT asks whether a given Boolean formula is unsatisfiable [11].

coNL is the complexity class containing the languages such that their complements belong to NL [22]. We can give a disqualification-based definition for coNL [3]. The disqualification-based definition of coNL assumes that a logarithmic space Turing machine has another separated read-only tape, that is the same kind of special tape called "read-once" that we use in the certificate-based definition for NL [3]. Besides, in the disqualification-based definition of coNL, we assume the disqualification string is appropriated for the instance [22].

▶ **Definition 5.** A language L_1 is in coNL if there exists a deterministic logarithmic space Turing machine M with an additional special read-once input tape polynomial $p : \mathbb{N} \to \mathbb{N}$ such that for every $x \in \{0,1\}^*$:

```
x \in L_1 \Leftrightarrow \forall \ appropriated \ u \in \{0,1\}^{p(|x|)} \ then \ M(x,u) = "yes"
```

where by M(x, u) we denote the computation of M where x is placed on its input tape and the disqualification u is placed on its special read-once tape, and M uses at most $O(\log |x|)$ space on its read/write tapes for every input x where $|\ldots|$ is the bit-length function. M is called a logarithmic space disqualifier.

For example, there is a well-known coNL problem that states: Given a directed graph G = (V, E) and two nodes $s, t \in V$, is there no possible path from s to t? In that problem, an appropriated disqualification u is a sequence of nodes contained in V when s is the first node and t is the last one such that this sequence of nodes is not a path: There is at least a consecutive pair of nodes in the sequence where they are not connected by an edge.

We state the following Hypothesis:

 \triangleright Hypothesis 6. Given a nonempty language $L_1 \in 1NL$, there is a language L_2 in coNP-complete with a deterministic Turing machine M, where:

```
L_2 = \{w : M(w, u) = y, \forall \text{ appropriated } u \text{ such that } y \in L_1\}
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when M runs in logarithmic space in the length of w, u is placed on the special read-once tape of M, and u is polynomially bounded by w. In this way, there is a coNP-complete language defined by a logarithmic space disqualifier M such that when the input is an element of the language with any of its appropriated disqualification, then M always outputs a string which belongs to a single language in 1NL.

▶ **Theorem 7.** If the Hypothesis 6 is true, therefore when L = NL, then P = NP.

Proof. We can accept the elements of the language $L_1 \in 1NL$ by a nondeterministic one-way logarithmic space Turing machine M'. In this way, there is a nondeterministic logarithmic space Turing machine M''(w,u) = M'(M(w,u)) which will accept when $w \in L_2$ for all the appropriated disqualification u, where u is placed on the special read-once tape of M''. The reason is because we can simulate the output string of M(w, u) within a read-once tape and thus, we can compute in a nondeterministic logarithmic space the logarithmic space composition using the same techniques of the logarithmic space composition reduction, but without any reset of the computation [22]. Certainly, we do not need to reset the computation of M(w,u) for the reading at once of a symbol in the output string of M(w,u)by the nondeterministic one-way logarithmic space Turing machine M'. Consequently, M''can be converted into a logarithmic space disqualifier for the language L_2 just assuming that L = NL, because of the nondeterministic logarithmic space Turing machine M'' could be simulated by a deterministic logarithmic space Turing machine. Therefore, L_2 is in coNLand thus, $L_2 \in P$ due to $coNL \subseteq P$ [22]. If any single coNP-complete problem can be solved in polynomial time, then P = NP [22]. Since $L_2 \in P$ and $L_2 \in coNP$ -complete, then we obtain the complexity class P is equal to NP under the assumption that L = NL.

3 Results

3.1 First Hypothesis

We show a previous known NP-complete problem:

▶ Definition 8. *NAE 3SAT*

INSTANCE: A Boolean formula ϕ in 3CNF.

QUESTION: Is there a truth assignment for ϕ such that each clause has at least one true literal and at least one false literal?

REMARKS: NAE $3SAT \in NP$ -complete [11].

We define a new problem:

▶ Definition 9. MAXIMUM EXCLUSIVE-OR 2SAT

INSTANCE: A positive integer K and a Boolean formula ϕ that is an instance of $XOR\ 2SAT$.

QUESTION: Is there a truth assignment in ϕ such that at most K clauses are unsatisfied? REMARKS: We denote this problem as $MAX \oplus 2SAT$.

▶ Theorem 10. $MAX \oplus 2SAT \in NP$ -complete.

Proof. It is trivial to see $MAX \oplus 2SAT \in NP$ [22]. Given a Boolean formula ϕ in 3CNF with n variables and m clauses, we create three new variables a_{c_i} , b_{c_i} and d_{c_i} for each clause $c_i = (x \lor y \lor z)$ in ϕ , where x, y and z are literals, in the following formula:

$$P_i = (a_{c_i} \oplus b_{c_i}) \wedge (b_{c_i} \oplus d_{c_i}) \wedge (a_{c_i} \oplus d_{c_i}) \wedge (x \oplus a_{c_i}) \wedge (y \oplus b_{c_i}) \wedge (z \oplus d_{c_i}).$$

We can see P_i has at most one unsatisfied clause for some truth assignment if and only if at least one member of $\{x,y,z\}$ is true and at least one member of $\{x,y,z\}$ is false for the same truth assignment. Hence, we can create the Boolean formula ψ as the conjunction of the P_i formulas for every clause c_i in ϕ , such that $\psi = P_1 \wedge \ldots \wedge P_m$. Finally, we obtain that:

 $\phi \in NAE \ 3SAT \ if \ and \ only \ if \ (\psi, m) \in MAX \oplus 2SAT.$

Consequently, we prove NAE $3SAT \leq_p MAX \oplus 2SAT$ where we already know the language NAE $3SAT \in NP$ -complete [11]. To sum up, we show $MAX \oplus 2SAT \in NP$ -hard and $MAX \oplus 2SAT \in NP$ and thus, $MAX \oplus 2SAT \in NP$ -complete.

▶ **Theorem 11.** There is a deterministic Turing machine M, where:

```
MAX \oplus 2SAT = \{w : M(w, u) = y, \exists \text{ appropriated } u \text{ such that } y \in XOR \text{ } 2SAT\}
```

when M runs in logarithmic space in the length of w, u is placed on the special read-once tape of M, and u is polynomially bounded by w.

Proof. Given a valid instance (ψ, K) for $MAX \oplus 2SAT$ when ψ has m clauses, we can create a certificate array A which contains K different natural numbers in ascending order which represents the indexes of the clauses in ψ that we are going to remove from the instance. We read at once the elements of the array A and we reject whether this is not an appropriated certificate: That is when the numbers are not sorted in ascending order, or the array A does not contain exactly K elements, or the array A contains a number that is not between 1 and m. While we read the elements of the array A, we remove the clauses from the instance (ψ, K) for $MAX \oplus 2SAT$ just creating another instance ϕ for XOR 2SAT where the Boolean formula ϕ does not contain the K different indexed clauses ψ represented by the numbers in A. Therefore, we obtain the array A would be valid according to the Theorem 11 when:

```
(\psi, K) \in MAX \oplus 2SAT \Leftrightarrow (\exists appropriated array A such that \phi \in XOR 2SAT).
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Furthermore, we can make this verification in logarithmic space such that the array A is placed on the special read-once tape, because we read at once the elements in the array A and we assume the clauses in the input ψ are indexed from left to right. Hence, we only need to iterate from the elements of the array A to verify whether the array is an appropriated certificate and also remove the K different clauses from the Boolean formula ψ when we write the final clauses to the output. This logarithmic space verification will be the Algorithm 1. We assume whether a value does not exist in the array A into the cell of some position i when A[i] = undefined. In addition, we reject immediately when the following comparisons:

$$A[i] \le max \lor A[i] < 1 \lor A[i] > m$$

hold at least into one single binary digit. Note, in the loop j from min to max - 1, we do not output any clause when max - 1 < min.

▶ Theorem 12. The Hypothesis 3 is true.

Proof. This is a consequence of Theorems 10 and 11.

3.2 Second Hypothesis

We show a previous known coNP-complete problem:

▶ Definition 13. 3UNSAT

INSTANCE: A Boolean formula ϕ in 3CNF.

QUESTION: Is ϕ unsatisfiable?

REMARKS: $3UNSAT \in coNP$ -complete [11].

We define a new problem:

◀

Algorithm 1 Logarithmic space verifier

```
1: /*A valid instance for MAX \oplus 2SAT with its certificate*/
 2: procedure VERIFIER((\psi, K), A)
        /*Initialize minimum and maximum values*/
3:
4:
        min \leftarrow 1
        max \leftarrow 0
5:
        /*Iterate for the elements of the certificate array A^*/
6:
        for i \leftarrow 1 to K+1 do
7:
           if i = K + 1 then
8:
                /*There exists a K+1 element in the array*/
9:
               if A[i] \neq undefined then
10:
                   /*Reject the certificate*/
11:
                   return "no"
12:
               end if
13:
               /*m is the number of clauses in \psi^*/
14:
               max \leftarrow m+1
15:
           else if A[i] = undefined \lor A[i] \le max \lor A[i] < 1 \lor A[i] > m then
16:
                /*Reject the certificate*/
17:
               \mathbf{return} \ "no"
18:
19:
           else
               max \leftarrow A[i]
20:
21:
           end if
           /*Iterate for the clauses of the Boolean formula \psi^*/
22:
           for j \leftarrow min \text{ to } max - 1 \text{ do}
23:
                /*Output the indexed j^{th} clause in \psi^*/
24:
               output "\wedge c_i"
25:
26:
           end for
27:
           min \leftarrow max + 1
28:
        end for
29: end procedure
```

▶ Definition 14. SUM ZERO

INSTANCE: A collection of integers C such that $0 \notin C$ and every integer in C has the same bit-length of the number that represents the cardinality of C multiplied by 3 (we do not take into account the symbol minus in counting the bit-length of the negative integers).

QUESTION: Are there two elements $a, b \in C$, such that a + b = 0? REMARKS: We denote this problem as SUM-ZERO.

▶ Theorem 15. SUM– $ZERO \in 1NL$.

Proof. Given a collection of integers C, we can read its elements from left to right, verify that every element is not equal to 0, check that every element in C has the same bit-length and count the amount of elements in C to finally multiply it by 3 and compare its bit-length with the single bit-length from the elements in C. In addition, we can nondeterministically pick two elements a and b from C and accept in case of a+b=0 otherwise we reject. We can make all this computation in a nondeterministic one-way using logarithmic space. Certainly, the calculation and store of the bit-length of the elements in C could be done in logarithmic space since this is a unique value. On the one hand, we can count and store the number of elements that we read from the input and multiply it by 3 to finally compare its bit-length with the stored unique bit-length from the elements of the collection, since the cardinality of C multiplied by 3 could be stored in a binary number of bit-length that is logarithmic in relation to the encoded length of C. On the other hand, the two elements a and b that we pick from C have a logarithmic space in relation to the encoded length of C, because of every integer in C has the same bit-length of the number that represents the cardinality of C multiplied by 3. Indeed, we never need to read to the left on the input for the acceptance of the elements in SUM-ZERO in a nondeterministic logarithmic space.

▶ **Theorem 16.** There is a deterministic Turing machine M, where:

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3UNSAT = \{w : M(w, u) = y, \forall \text{ appropriated } u \text{ such that } y \in SUM\text{-}ZERO\}
```

when M runs in logarithmic space in the length of w, u is placed on the special read-once tape of M, and u is polynomially bounded by w.

Proof. Given a Boolean formula ϕ in 3CNF with n variables and m clauses, we can create a disqualification array A which contains m positive integers between 1 and 3 which represents the literals of the clauses in ϕ which appear from left to right. We read at once the elements of the array A and we reject whether this is not an appropriated disqualification: That is when the array A does not contain exactly m elements, or the array A contains a number that is not between 1 and 3. While we read the elements of the array A, we select from the clauses ϕ the literals such that these ones occupy the position that represents the number between 1 and 3, that is the first, second or third place within the clause from left to right. In this way, we output the selected literals that are represented by a positive or negative (in case of a negated variable) integer just creating another instance C for SUM–ZERO where the collection C contains those integers which are the selected literals for each clause in ϕ . Therefore, we obtain that all the appropriated array A would be valid according to the Theorem 16 when:

```
\phi \in 3UNSAT \Leftrightarrow (\forall \ appropriated \ array \ A \ such \ that \ C \in SUM{\rm -}ZERO)
```

since we assume the positive and negated literals of some variable in the input ϕ correspond to a positive integer and its negative value, respectively. Furthermore, we can make this

disqualification in logarithmic space such that the array A is placed on the special read-once tape, because we read at once the elements in the array A. Hence, we only need to iterate from the elements of the array A to verify whether the array is an appropriated disqualification and pick the m literals from the Boolean formula ϕ when we write the final integers that represent these literals to the output. This logarithmic space disqualification will be the Algorithm 2. We assume whether a value does not exist in the array A into the cell of some position i when A[i] = undefined. In addition, we reject immediately when the following comparisons:

```
A[i] < 1 \lor A[i] > 3
```

hold at least into one single binary digit. Note, that every possible literal in ϕ could have a representation by an integer between $-3 \times m$ and $3 \times m$ with the exception of 0, where m is the cardinality of the collection C. In this way, we guarantee the output collection C is an appropriated instance of SUM–ZERO just filling with zeroes to the left the elements with bit-length lesser than $|3 \times m|$ where $|\dots|$ is the bit-length function.

Algorithm 2 Logarithmic space disqualifier

```
1: /*A valid instance for 3UNSAT with its disqualification*/
 2: procedure DISQUALIFIER(\phi, A)
 3:
        /*Initialize an index*/
 4:
       i \leftarrow 0
        /*m is the number of clauses in \phi^*/
 5:
        /*Iterate for the elements of the disqualification array A^*/
 6:
        for i \leftarrow 1 to m+1 do
 7:
           if i = m + 1 then
 8:
               /*There exists an m+1 element in the array*/
 9:
               if A[i] \neq undefined then
10:
                   /*Reject the disqualification*/
11:
                   {f return} "no"
12:
               end if
13:
               /*Break the for loop*/
14:
               break
15:
16:
           else if A[i] = undefined \lor A[i] < 1 \lor A[i] > 3 then
               /*Reject the disqualification*/
17:
               return "no"
18:
           else
19:
               j \leftarrow A[i]
20:
           end if
21:
           /*From the indexed i^{th} clause c_i = (x_j \vee y_k \vee z_r) in \phi^*/
22:
           /*Where x, y and z are literals with local indexes \{j, k, r\} = \{1, 2, 3\} in c_i^*/
23:
           /*Output the integer representation of the j^{th} literal, that is n(x_i)^*/
24:
           /*Filled with zeroes to the left until a total of |3 \times m| bits including the literal*/
25:
           /*But, the bit-length of the symbol minus is ignored in filling the negated literals*/
26:
27:
           output ", n(x_i)"
        end for
28:
29: end procedure
```

◀

▶ **Theorem 17.** The Hypothesis 6 is true.

Proof. This is a consequence of Theorems 15 and 16.

3.3 Consequences

▶ Theorem 18. If $L \neq NL$ then P = NP.

Proof. This is a direct consequence of Theorems 4 and 12.

▶ Theorem 19. If L = NL then P = NP.

Proof. This is a direct consequence of Theorems 7 and 17.

▶ Theorem 20. P = NP.

Proof. Since we have either $L \neq NL$ or L = NL is true, then the complexity class P is equal to NP.

▶ Theorem 21. NL = NP.

Proof. Since the polynomial time reduction in Theorem 10 could be easily transformed in a logarithmic space reduction, then the NP-complete problem in Theorem 4, that would be $MAX \oplus 2SAT$, is necessarily in NL and thus all the problems in NP when $L \neq NL$, because of the Cook's Theorem can also be transformed in a logarithmic space reduction [11]. Certainly, every NP problem could be logarithmic space reduced to SAT by the Cook's Theorem algorithm and SAT can be indeed logarithmic space reduced to NAE 3SAT [11]. For the same reason, the coNP-complete problem in Theorem 7, that would be 3UNSAT, is necessarily in coNL and thus all the problems in coNP when L = NL, because of the problem UNSAT can be actually logarithmic space reduced to 3UNSAT [11]. Since we have either $L \neq NL$ or L = NL is true, then the complexity class NL is equal to NP.

▶ Theorem 22. L = NP.

Proof. Under the assumption of $L\subseteq 1NL$, then we would obtain that 1NL=NL. Certainly, the logarithmic space verifier of Theorem 11 works into one-way when $L\subseteq 1NL$. Indeed, the logarithmic space verifier through the certificate generates the output into one-way according to the arguments provided by the proof in Theorem 11. Moreover, we know that $XOR\ 2SAT\in 1NL$ under the assumption of $L\subseteq 1NL$, since $XOR\ 2SAT$ is in $L\ [2]$, [24]. In this way, we obtain that $MAX\oplus 2SAT\in 1NL$. Since $MAX\oplus 2SAT$ is NP-complete under logarithmic space reductions and NL=NP due to Theorem 21, then we would deduce that 1NL=NL, because of every problem in NP could be reduced to $MAX\oplus 2SAT$ under 1NL-reductions. However, it is a known result that $1NL\ne NL$ [20]. Consequently, we obtain that $1NL\subset L$ as a consequence of this contradiction. Since $1NL\subset L$ implies L=NL [14], then we have the complexity class L is equal to NP.

4 Conclusions

No one has been able to find a polynomial time algorithm for any of more than 300 important known NP-complete problems [11]. A proof of P = NP will have stunning practical consequences, because it leads to efficient methods for solving some of the important problems in NP [6]. The consequences, both positive and negative, arise since various NP-complete problems are fundamental in many fields [6]. The following consequences

are assuming that we have a practical solution for the *NP-complete* problems where such existence was proven with our nonconstructive result:

Cryptography, for example, relies on certain problems being difficult. A constructive and efficient solution to an NP-complete problem such as 3SAT will break most existing cryptosystems including: Public-key cryptography [15], symmetric ciphers [19] and one-way functions used in cryptographic hashing [8]. These would need to be modified or replaced by information-theoretically secure solutions not inherently based on P-NP equivalence.

There are enormous positive consequences that will follow from rendering tractable many currently mathematically intractable problems. For instance, many problems in operations research are *NP-complete*, such as some types of integer programming and the traveling salesman problem [11]. Efficient solutions to these problems have enormous implications for logistics [6]. Many other important problems, such as some problems in protein structure prediction, are also *NP-complete*, so this will spur considerable advances in biology [5].

Since all the *NP-complete* optimization problems become easy, everything will be much more efficient [10]. Transportation of all forms will be scheduled optimally to move people and goods around quicker and cheaper [10]. Manufacturers can improve their production to increase speed and create less waste [10]. Learning becomes easy by using the principle of Occam's razor: We simply find the smallest program consistent with the data [10]. Near perfect vision recognition, language comprehension and translation and all other learning tasks become trivial [10]. We will also have much better predictions of weather and earthquakes and other natural phenomenon [10].

There would be disruption, including maybe displacing programmers [16]. The practice of programming itself would be more about gathering training data and less about writing code [16]. Google would have the resources to excel in such a world [16].

But such changes may pale in significance compared to the revolution an efficient method for solving *NP-complete* problems will cause in mathematics itself [6]. Research mathematicians spend their careers trying to prove theorems, and some proofs have taken decades or even centuries to find after problems have been stated [1]. For instance, Fermat's Last Theorem took over three centuries to prove [1]. A method that is guaranteed to find proofs to theorems, should one exist of a "reasonable" size, would essentially end this struggle [6].

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