

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. The answer $P = NP$ is considered as a very unlikely event. However, we prove the complexity class P 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 Babai 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 have 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 $coNP$ 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 [22]. 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 $P \neq NP$ [21]. Indeed, if $P = NP$ then there are stunning practical consequences [21]. 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

In 1936, Turing developed his theoretical computational model [24]. The deterministic and nondeterministic Turing machines have become in two of the most important definitions related to this theoretical model for computation [24]. A deterministic Turing machine has only one next action for each step defined in its program or transition function [24]. 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 [24].

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 $M(w) = \text{"yes"}$ [3]. Note that M fails to accept w either if this computation ends in the rejecting state, that is $M(w) = \text{"no"}$, 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:

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

Moreover, $L(M)$ is decided by M , when $w \notin L(M)$ if and only if $M(w) = \text{"no"}$ [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:

$$L_1 = \{w : M(w, c) = \text{"yes"} \text{ for some string } c\}.$$

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 [21]. If NP is the class of problems that have succinct certificates, then the complexity class $coNP$ must contain those problems that have succinct disqualifications [21]. That is, a “no” instance of a problem in $coNP$ possesses a short proof of its being a “no” instance [21].

► **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 $M(w, c) = \text{“no”}$ and M is the polynomial time verifier of the complement of L_2 in NP .*

A function $f : \Sigma^* \rightarrow \Sigma^*$ 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 [24]. 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 $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ such that for all $x \in \{0, 1\}^*$:

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

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, \dots, x_n ;
2. Boolean connectives: Any Boolean function with one or two inputs and one output, such as \wedge (AND), \vee (OR), \neg (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:

$$(x_1 \vee \neg x_1 \vee \neg x_2) \wedge (x_3 \vee x_2 \vee x_4) \wedge (\neg x_1 \vee \neg x_3 \vee \neg x_4)$$

is in $3CNF$. The first of its three clauses is $(x_1 \vee \neg x_1 \vee \neg x_2)$, 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 [24]. The work tapes may contain at most $O(\log n)$ symbols [24]. In computational complexity theory, L is the complexity class containing those decision problems that can be decided by a deterministic logarithmic space Turing machine [21]. NL is the complexity class containing the decision problems that can be decided by a nondeterministic logarithmic space Turing machine [21].

A logarithmic space transducer is a Turing machine with a read-only input tape, a write-only output tape, and read/write work tapes [24]. The work tapes must contain at most $O(\log n)$ symbols [24]. A logarithmic space transducer M computes a function $f : \Sigma^* \rightarrow \Sigma^*$, where $f(w)$ is the string remaining on the output tape after M halts when it is started with w on its input tape [24]. We call f a logarithmic space computable function [24]. 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 $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ such that for all $x \in \{0, 1\}^*$:

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

The logarithmic space reduction is used in the definition of the complete languages for the classes L and NL [21]. 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 [21]. 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 [20]. 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], [23].

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 [21]. 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 $p : \mathbb{N} \rightarrow \mathbb{N}$ such that for every $x \in \{0, 1\}^*$:

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

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 $|\dots|$ is the bit-length function [3]. M is called a logarithmic space verifier [3].

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

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

If L_1 is a language such that $L' \leq_p L_1$ for some $L' \in \text{coNP-complete}$, then L_1 is *coNP-hard* [7]. Moreover, if $L_1 \in \text{coNP}$, then $L_1 \in \text{coNP-complete}$ [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* [21]. 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 [21].

► **Definition 3.** *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 $p : \mathbb{N} \rightarrow \mathbb{N}$ such that for every $x \in \{0, 1\}^*$:*

$$x \in L_1 \Leftrightarrow \forall \text{ appropriated } u \in \{0, 1\}^{p(|x|)} \text{ then } M(x, u) = \text{“yes”}$$

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 $|\dots|$ 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.

3 Hypothesis

The two-way Turing machines may move their head on the input tape into two-way (left and right directions) while the one-way Turing machines are not allowed to move the input head on the input tape to the left [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].

We state the following Hypothesis:

▷ **Hypothesis 4.** Given a nonempty language $L_1 \in 1NL$, there is a language L_2 in *coNP-complete* with a deterministic Turing machine M , where:

$$L_2 = \{w : M(w, u) = y, \forall \text{ appropriated } u \text{ such that } y \in L_1\}$$

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 5.** *A language L_1 is in $coNL$ if and only if L_1 complies with there is a non-deterministic logarithmic space Turing machine M_1 , such that for every appropriate instance x , then all the paths on $M_1(x)$ accept when $x \in L_1$ or at least one of the paths on $M_1(x)$ rejects when $x \notin L_1$.*

Proof. The proof is simple, since we can interchange the acceptance and rejection states of M_1 into a new Turing machine M'_1 , such that $M'_1(x) = \text{"yes"}$ when $M_1(x) = \text{"no"}$ and $M'_1(x) = \text{"no"}$ when $M_1(x) = \text{"yes"}$. Certainly, the language $L(M'_1)$ is the complement of L_1 and it is in NL according to the definition of NL and because of the string x is an appropriate instance [21]. ◀

► **Theorem 6.** *When the Hypothesis 4 is true, then $P = NP$.*

Proof. We can simulate the computation $M(w, u) = y$ in the Hypothesis 4 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 [21]. 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 [21]. We remove each symbol generated in the work tapes, when we try to generate the next symbol contiguous to the right on the string u . In this way, the generation will always be in logarithmic space. In addition, we can guarantee that the generation of symbols in the work tapes always produces an appropriated string u , since we can check this by symbol per symbol from the generated string u .

Since we have that $L_1 \in 1NL$, then we can accept the elements of the language L_1 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 $w \in L_2$ for all the possible paths and will reject when $w \notin L_2$ for at least one of the paths.

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 [21]. 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' . Actually, the logarithmic space reduction is possible, because of M' is in one way. Indeed, it is not necessary to reset the computation of N in the composition $M'(N(w))$ on the input w , because M' never moves to the left the head on the input tape (that would be the output tape of N).

As a consequence of Theorem 5, then L_2 would be in $coNL$ and thus, $L_2 \in P$ due to $coNL \subseteq P$ [21]. If any single $coNP$ -complete problem can be solved in polynomial time, then $P = NP$ [21]. 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 the Hypothesis 4 is true. ◀

4 Results

We show a previous known $coNP$ -complete problem:

► **Definition 7. 3UNSAT**

INSTANCE: A Boolean formula ϕ in 3CNF.

QUESTION: Is ϕ unsatisfiable?

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

We define a new problem:

► **Definition 8. 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 9.** 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 10.** There is a deterministic Turing machine M , where:

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

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 10 when:

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

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 1. We assume whether a value does not exist in the array A into the cell of some position i when $A[i] = \text{undefined}$. In addition, we reject immediately when the following comparisons:

$$A[i] < 1 \vee A[i] > 3$$

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. ◀

► **Theorem 11.** *The Hypothesis 4 is true.*

Proof. This is a direct consequence of Theorems 9 and 10. ◀

► **Theorem 12.** $P = NP$.

Proof. This is a direct consequence of Theorems 6 and 11. ◀

5 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 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

Algorithm 1 Logarithmic space disqualifier

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

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[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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