

# The P versus NP Problem

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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? 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. To attack the P versus NP problem, the NP-completeness is a useful tool. We prove the known NP-complete problem MONOTONE 1-IN-3 3SAT can be polynomially reduced to the polynomial language 2SET PACKING. In this way, MONOTONE 1-IN-3 3SAT must be in P. If any NP-complete problem can be solved in polynomial time, then every language in NP has a polynomial time algorithm. Hence, we demonstrate the complexity class P is equal to NP.

**Keywords:** Complexity Classes · Completeness · Polynomial Time · MONOTONE 1-IN-3 3SAT · 2SET PACKING.

## 1 Introduction

The  $P$  versus  $NP$  problem is a major unsolved problem in computer science [3]. This is considered by many to be the most important open problem in the field [3]. It is one of the seven Millennium Prize Problems selected by the Clay Mathematics Institute [3]. It was essentially mentioned in 1955 from a letter written by John Nash to the United States National Security Agency [1]. However, the precise statement of the  $P = NP$  problem was introduced in 1971 by Stephen Cook in a seminal paper [3].

In 1936, Turing developed his theoretical computational model [9]. The deterministic and nondeterministic Turing machines have become in two of the most important definitions related to this theoretical model for computation [9]. A deterministic Turing machine has only one next action for each step defined in its program or transition function [9]. 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 [9].

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 [4]. A complexity class is a set of problems, which

are represented as a language, grouped by measures such as the running time, memory, etc [4].

In the computational complexity theory, the class  $P$  contains those languages that can be decided in polynomial time by a deterministic Turing machine [7]. The class  $NP$  consists in those languages that can be decided in polynomial time by a nondeterministic Turing machine [7]. The biggest open question in theoretical computer science concerns the relationship between these classes: Is  $P$  equal to  $NP$ ? 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 [6].

It is fully expected that  $P \neq NP$  [8]. For that reason,  $P = NP$  is considered as a very unlikely event [8]. 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 for computer science, but for many other fields as well [1]. Whether  $P = NP$  is still a controversial possible solution to this problem [1]. However, we prove the complexity class  $P$  is equal to  $NP$ . Hence, we solve one of the most important open problems in computer science with a solution which was certainly unexpected and with stunning practical consequences [1].

## 2 Definitions

Let  $\Sigma$  be a finite alphabet with at least two elements, and let  $\Sigma^*$  be the set of finite strings over  $\Sigma$  [2]. A Turing machine  $M$  has an associated input alphabet  $\Sigma$  [2]. For each string  $w$  in  $\Sigma^*$  there is a computation associated with  $M$  on input  $w$  [2]. We say that  $M$  accepts  $w$  if this computation terminates in the accepting state, that is  $M(w) = \text{"yes"}$  [2]. 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 [2].

The language accepted by a Turing machine  $M$ , denoted  $L(M)$ , has an associated alphabet  $\Sigma$  and is defined by

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

We denote by  $t_M(w)$  the number of steps in the computation of  $M$  on input  $w$  [2]. 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  $\Sigma^n$  is the set of all strings over  $\Sigma$  of length  $n$  [2]. We say that  $M$  runs in polynomial time if there is a constant  $k$  such that for all  $n$ ,  $T_M(n) \leq n^k + k$  [2]. In other words, this means the language  $L(M)$  can be accepted by the Turing machine  $M$  in polynomial time. Therefore,  $P$  is the complexity class of languages that can be accepted in polynomial time by deterministic Turing machines [4]. A verifier for a language  $L$  is a deterministic Turing machine  $M$ , where

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

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  $w$  [2]. A verifier uses additional information, represented by the symbol  $c$ , to verify that a string  $w$  is a member of  $L$ . This information is called certificate.  $NP$  is also the complexity class of languages defined by polynomial time verifiers [8].

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 [9]. 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* [7]. A language  $L \subseteq \{0, 1\}^*$  is *NP-complete* if

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

If  $L$  is a language such that  $L' \leq_p L$  for some  $L' \in NP\text{-complete}$ , then  $L$  is *NP-hard* [7]. Moreover, if  $L \in NP$ , then  $L \in NP\text{-complete}$  [7]. A Boolean formula  $\phi$  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  $\phi$  is a set of values for the variables in  $\phi$ . A satisfying truth assignment is a truth assignment that causes  $\phi$  to be evaluated as true. A formula with a satisfying truth assignment is a satisfiable formula. 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 [4]. 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 [4]. A Boolean formula is in 3-conjunctive normal form or *3CNF*, if each clause has exactly three distinct literals [4]. 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  $\neg x_2$ .

### 3 Summary

I have a polynomial algorithm for *MONOTONE 1-IN-3 3SAT* that no one has found before: The trick is that I noticed given an instance  $\phi$  of the language *MONOTONE 1-IN-3 3SAT*, then we can construct a Boolean formula  $\phi'$  in *CNF* such that

- If  $\phi'$  has a truth assignment with exactly one true literal for each clause, then  $\phi$  complies with the same property.
- $\phi'$  contains the clauses of  $\phi$  but modified and there are at most two clauses of two literals for each occurrence of a literal from every clause in  $\phi$ . In this way, there are at most polynomially number of clauses in  $\phi'$  in relation to  $\phi$ .
- We replace each occurrence of any variable into a single clause in  $\phi$  by some unique variable in  $\phi'$  which appears in  $\phi'$  as follows: Once negated in a clause of two literals and twice as a positive literal in a clause of two and three literals when the variable occurrences appear more than once in  $\phi$ .

We create sets of fewer than three elements assigned to each literal from every clause into  $\phi'$ , such that the set assigned to the negated literal of some variable is not mutually disjoint with the sets assigned to the two positive literals from this unique variable in  $\phi'$ . Moreover, we modify those sets with exactly two elements just adding a common element to the sets of the literals of every clause in  $\phi'$ , because this guarantee that the sets assigned to the literals of each clause are not mutually disjoint. The amount of sets in this procedure is small since there are exactly the same amount of sets which is equal to the total number of literals for each clause in  $\phi'$ . In this way, a truth assignment  $T$  for the Boolean formula  $\phi'$  of  $m'$  clauses with exactly one true literal for each clause corresponds to  $m'$  mutually disjoint sets assigned for the literals that were true in this assignment, that is the negated or positive literal when the variable in  $T$  is false or true respectively. Certainly, the possibility between  $m'$  mutually disjoint sets when every clause of two literals has exactly one set chosen from its literals makes possible the appropriated assignment of all the variables in  $\phi$  where they were replaced by several unique variables in  $\phi'$  when we evaluate as true each literal which has been assigned from every selected set. At the same time, we obtain the property between  $m'$  mutually disjoint sets when every clause of three literals has exactly one set chosen from its literals converts this into a certificate for the clauses of three literals in  $\phi$  which are equivalent to the modified clauses of three literals that contain  $\phi'$  when we evaluate as true each literal which has been assigned from every selected set. In this way, the collection of these sets is an instance of the polynomial language *2SET PACKING* assuming that we require  $m'$  mutually disjoint sets. If we cannot pick  $m'$  mutually disjoint sets from this collection, then this will imply  $\phi'$  has not a truth assignment with exactly one true literal for each clause. The reason is because if we cannot choose a single set for every clause, then there is not a truth assignment where we may select as true exactly one literal for each clause in  $\phi'$ . In addition, we cannot obtain  $m'$  mutually disjoint sets where there is a literal and its negation within the selected sets since as we explained above those assigned sets are not mutually disjoint.

## 4 Results

### Definition 1. *MONOTONE 1-IN-3 3SAT*

*INSTANCE:* A Boolean formula  $\phi$  in 3CNF such that there is no clause which contains a negated literal.

*QUESTION:* Is there a truth assignment for  $\phi$  such that each clause in  $\phi$  has exactly one true literal?

*REMARKS:* *MONOTONE 1-IN-3 3SAT* is in NP-complete [5].

### Definition 2. *2SET PACKING*

*INSTANCE:* A collection  $C$  of finite sets and a positive integer  $K \leq |C|$  such that for all  $c \in C$  we have  $|c| \leq 2$  where  $|\dots|$  is the cardinality function.

*QUESTION:* Does  $C$  contain at least  $K$  mutually disjoint sets?

*REMARKS:* *2SET PACKING* is solvable in polynomial time by matching techniques [5].

**Theorem 1.** *MONOTONE 1-IN-3 3SAT*  $\leq_p$  *2SET PACKING*.

*Proof.* Consider a Boolean formula  $\phi$  in 3CNF with  $m$  clauses such that there is no clause which contains a negated literal in which the positive literal  $x$  appears  $k$  times. We replace the first occurrence of  $x$  by  $x_1$ , the second by  $x_2$ , and so on, where  $x_1, x_2, \dots, x_k$  are  $k$  new variables. Later, we add

$$(\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_3) \wedge \dots \wedge (\neg x_k \vee x_1)$$

to a new Boolean formula  $\phi'$  in CNF which contains the modified clauses of  $\phi$  plus these above clauses of fewer than 3 literals where we do this procedure for each positive literal  $x$  in  $\phi$ . Note, this is logically equivalent to

$$x_1 \Rightarrow x_2 \Rightarrow \dots \Rightarrow x_k \Rightarrow x_1$$

such that the resulting expression in  $\phi'$  satisfies the condition for a truth assignment on  $x$ . In this new formula  $\phi'$  every variable appears at most 3 times. If  $k = 1$ , then the literal  $x_1$  appears once. If  $k > 1$ , for every integer  $i$  between 1 and  $k$ , we have that the literal  $x_i$  appears twice and  $\neg x_i$  appears once. Suppose we have the following instance  $\phi$  of *MONOTONE 1-IN-3 3SAT*

$$\dots (x \vee w \vee g) \wedge \dots \wedge (x \vee y \vee z) \dots$$

then the transformed expression  $\phi'$  in CNF over the variable  $x$  is

$$\dots (x_1 \vee w \vee g) \wedge \dots \wedge (x_2 \vee y \vee z) \dots (\neg x_1 \vee x_2) \wedge (\neg x_2 \vee x_1)$$

where

- the variable  $x_1$  appears thrice and,
- the literal  $x_1$  appears twice and,
- the literal  $\neg x_1$  appears once.

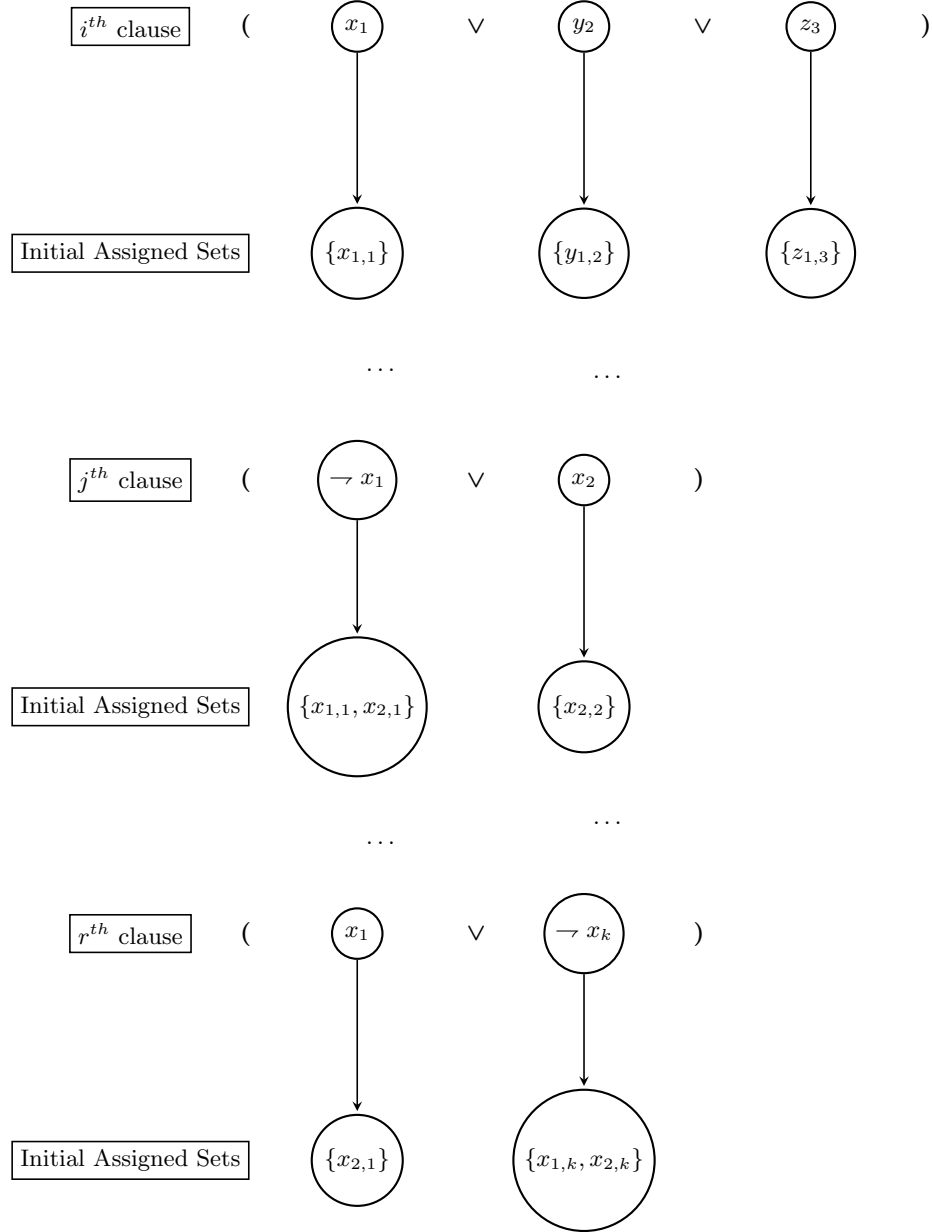
Now from the expression  $\phi'$  in *CNF*, we create an instance of *2SET PACKING* in the following two steps:

1. In the first step, we create a set for each literal that appears in every clause. For every variable  $x_i$ , we assign the set  $\{x_{1,i}, x_{2,i}\}$  to the negated literal  $\neg x_i$ , for the positive literal  $x_i$  in the clause of three literals the set  $\{x_{1,i}\}$  and for the positive literal  $x_i$  in the clause of two literals the set  $\{x_{2,i}\}$ .
2. In the second step, for each clause  $c_i$  of three literals we add to the set assigned to its literals a single element  $d_i$  which is unique for every clause  $c_i$  of three literals obtaining three sets of two elements. Moreover, for each clause  $c_j$  of two literals, there is a negated literal  $\neg x_i$  with a set  $\{x_{1,i}, x_{2,i}\}$  and another positive literal  $x_j$  with a set of a single element  $\{x_{2,j}\}$ , therefore we add to the set of cardinality equal to one the another element  $x_{1,i}$  obtaining the set  $\{x_{2,j}, x_{1,i}\}$  of two elements.

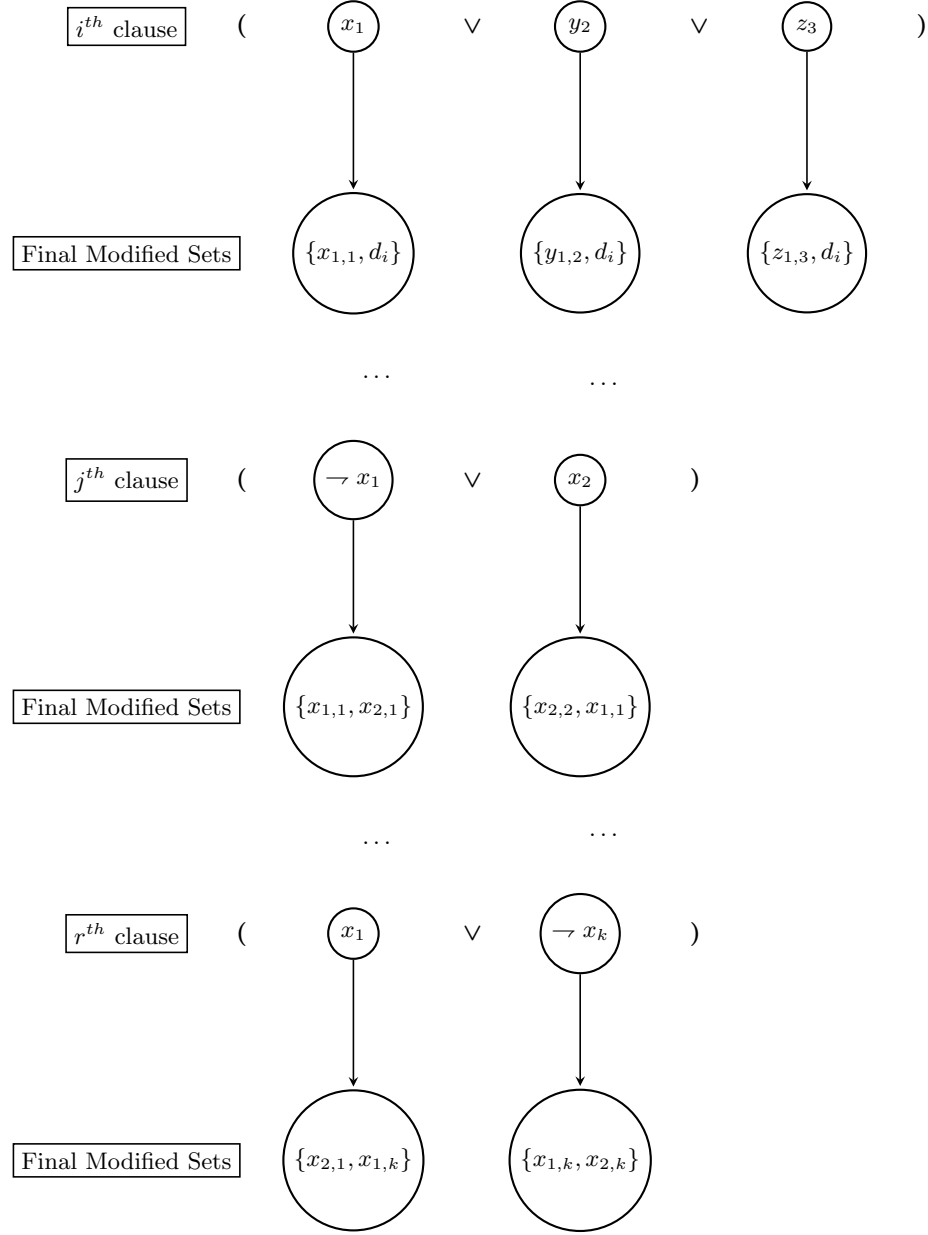
If we consider an instance of *2SET PACKING* created from the sets of each literal from every clause of  $\phi'$ , then the first step guarantee that we should analyze separately the sets from the positive literals in relation to the set of the single negated literal. In this way, in the first step there is no a possibility where we could choose the three sets for a single variable in  $\phi'$  at the same time because they are not mutually disjoint sets. For that reason, we can assure if we choose some  $K$  mutually disjoint sets from this instance, then we can assign as true the literals which are represented by these sets, since there is no a violation when for some variable  $x_i$  the literals  $x_i$  and  $\neg x_i$  could be both true or false at the same time.

In the second step, we guarantee this truth assignment created from a  $K$  mutually disjoint sets of this instance complies that every clause of three literals in  $\phi'$  has exactly one true literal since the sets from the literals of a clause of three literals are not mutually disjoint since they contain the single element  $d_i$  which is unique for every clause  $c_i$  of three literals. Moreover, this second step makes possible that exactly one literal for a clause of two literals in  $\phi'$  can be true just making possible the condition of assignment for every variable  $x$  in the original  $\phi$ . Note, that we add to the set of the positive literal of clauses of two literals in  $\phi'$ , the element  $x_{1,i}$  and not  $x_{2,i}$  from the negated literal  $\neg x_i$ , because that guarantee there should not be a violation against the same truth assignment between the variables  $x_j$  and  $x_i$  that represents the variable  $x$  over two distinct clauses of three literals that contain  $x_j$  and  $x_i$  in  $\phi'$  respectively.

Look at the example of the Figure 1 and see how is applied the first step and finally the second step which is represented in the Figure 2 over this reduction. We have the clause  $c_i = (x_1 \vee y_2 \vee z_3)$  of three literals and the clauses  $c_j = (\neg x_1 \vee x_2)$  and  $c_r = (x_1 \vee \neg x_k)$  of two literals. Note, the variable  $x_1$  only appears in these three clauses. According to the first step in the Figure 1, we add the set  $\{x_{1,1}, x_{2,1}\}$  to the negated variable  $\neg x_1$  and the another sets  $\{x_{1,1}\}$  and  $\{x_{2,1}\}$  to the positive literals in the clause  $c_i$  of three literals and the clause  $c_r$  of two literals respectively. We also see how is applied the second step in the Figure 2 where it is added the element  $d_i$  to all the sets of the literals in the



**Fig. 1.** First step



**Fig. 2.** Second step



clause  $c_i$  of three literals. The transformation in the clauses of two literals in the second step is also visible in the Figure 2 where it is added the element  $x_{1,1}$  to the set  $\{x_{2,2}\}$  assigned to the positive literal  $x_2$  in the clause  $c_j$  of two literals. With these two steps, we obtain the sets of each literal from every clause have a cardinality equal to 2.

Now, if the Boolean formula  $\phi'$  has  $m'$  clauses, then the collection  $C$  of sets for each literal from every clause complies that

$$\phi \in \text{MONOTONE 1-IN-3 SAT} \text{ if and only if } (C, m') \in \text{2SET PACKING}.$$

Is it the case that  $C$  contains  $m'$  mutually disjoint sets and  $\phi$  is not a Boolean formula in *MONOTONE 1-IN-3 SAT*? The answer is no. Think for example in some  $m'$  mutually disjoint sets, then we can transform them in a truth assignment  $T$  evaluating every literal assigned in these sets as true. In this truth assignment  $T$  from the clauses of two literals exactly one literal would be true and from the clauses of three literals exactly one literal would be true in the Boolean formula  $\phi'$ . Certainly, we guarantee that since the literals of every clause have not mutually disjoint sets in  $\phi'$  after the reduction. In this way, the truth assignment  $T$  will be a satisfying assignment for the Boolean formula  $\phi'$  where every clause has exactly one true literal which also implies that the original Boolean formula  $\phi$  will be in *MONOTONE 1-IN-3 SAT*. Take into account that we assume that  $K = m'$  since at least the  $m'$  clauses which contains  $\phi'$  should be satisfiable with this truth assignment  $T$  in order to satisfy the original Boolean formula  $\phi$  as well.

This is a polynomial time reduction:

- We can create the Boolean formula  $\phi'$  in time  $O(m^2)$  just replacing the modified clauses of three literals and adding the clauses of two literals. Certainly, we can go each time through every clause  $\phi$  just adding the new clauses of two literals and modifying the other clauses of three literals which are at most  $m - 1$ .
- After that, we create the instance for *2SET PACKING* in time  $O(m)$  just creating the new sets in the first step and modifying them in the second step. Notice there at most two clauses of two literals for each variable in  $\phi'$ . Since the amount of variables in  $\phi'$  is linear in relation to the number of clauses in  $\phi$ , then we can assume the creation and the modification of these sets can be done in time  $O(m)$ .

Certainly, we can create the instance  $(C, m')$  just running the whole reduction in time  $O(m^2)$  and thus, the proof is completed.

**Theorem 2.**  $P = NP$ .

*Proof.* The known *NP-complete* problem *MONOTONE 1-IN-3 SAT* can be reduced in polynomial time to *2SET PACKING* where *2SET PACKING*  $\in P$  [5]. Consequently, *MONOTONE 1-IN-3 SAT*  $\in P$ . If any *NP-complete* problem can be solved in polynomial time, then every language in *NP* has a polynomial time algorithm [4]. In conclusion, we finally prove that  $P = NP$ .

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