

1 May 7th, 2019

1.1 Well-Characterized Problems

Definition 1.1. NP: Class of decision problems for which there is a “yes-certificate” (i.e. there for yes-instances, there is a ”short” proof that the answer is yes)

Note that this is not symmetric, as such, we have the CO-NP class

Definition 1.2. CO-NP: Class of decision problems for which there is a “no-certificate”

We know that:

Hamiltonian Cycle Problem \in NP.

However, we still don’t know if the Hamiltonian Cycle Problem is in CO-NP. The belief among experts is that:

Hamiltonian Cycle Problem \in CO-NP.

Note that $P \subset$ CO-NP, since we can just solve the problem for a no-certificate.

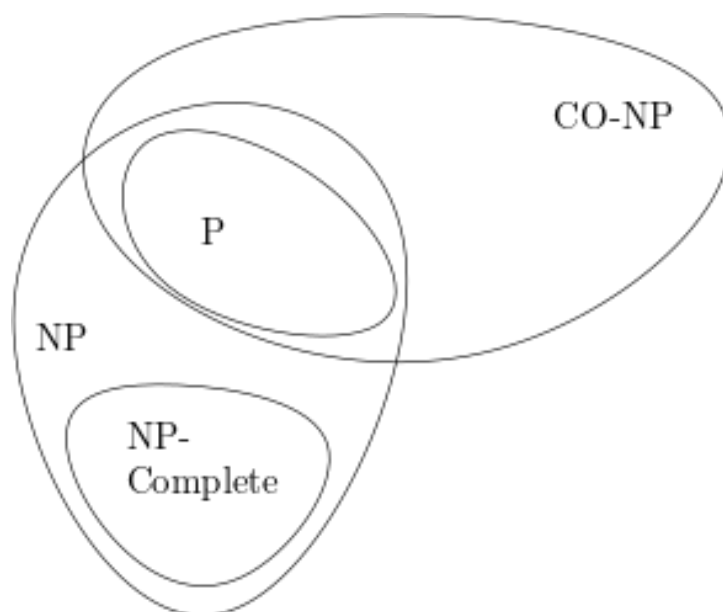


Figure 1: $NP \cap CO-NP = P$?

Definition 1.3. Problem that have both YES and NO certificates, i.e. problem in $NP \cap CO-NP$ are said to be **well characterized** or said to have a **good characterization**

Let us consider the decision version of LP: Given a feasible and bounded maximization LP, is there a feasible solution whose value is at least k ?

We already know that there is a YES-certificate, as we can just check the feasible solution whose value is at least k , i.e. $LP \in NP$.

Remark 1.4 — Note that we did not have to know that there is a poly-time solution for LP, as we are only looking at verifying the YES-certificates

To see whether the LP problem lies in CO-NP, we can consider the dual of the LP. A NO-certificate would be any feasible solution of the dual whose value is less than k , meaning that $LP \in CO-NP$.

This means that LP is a well-characterized problem. Once again, note that we did not have to use the fact that there is a poly-time algorithm for LP. Historically, when a problem is found to be well-characterized, it is likely that it belongs to P.

Decision Problem for Max Flow: Given a flow network, is there a feasible flow of value greater than or equal to k ?

- Feasible flow is a yes-certificate, so this problem belongs to NP.
- The no-certificate is cut, as if there does not exist a flow greater than or equal to k , then there is a cut of value less than k .
- Note that once again we have shown that this problem is well-characterized, without using the knowledge of having a poly-time solution.

Decision Problem for Bipartite Perfect Matching: Given a bipartite graph G with bipartition (A, B) such that $|A| = |B|$, does G have a perfect matching?

- Yes-certificate is just the perfect matching which can be easily checked
- No-certificate: we can use Hall's Theorem, i.e. show that:

$$\exists X \subseteq A, \text{ s.t. } |N(X)| < |X|.$$

We could also use Konig's Theorem:

$$\exists \text{ vertex cover of size } < n.$$

- As such this problem is well-characterized.

Problems that are well-characterized typically have an associated min-max relationship. These relationships are:

- beautiful and powerful combinatorial results
- lead to poly-time algorithms that are designed around it
- special cases of LP-duality.

We will now look at a well-characterized problem for which we do not know if there is a poly-time algorithm yet - factorization.

1.2 Well-characterization of Factorization

The decision problem for factorization is:

Given two integers x and y , does x have a factor less than y and greater than 1.

To find its first factor, find less than x and greater than 1. If no, then it is prime, if yes, then we can do binary search, but with $y = \frac{x}{2}$. After we find a factor, we can divide it and find the next factor.

Note that it takes $\log x$ calls to find a factor and since 2 is the smallest factor, there can be at most $\log x$ factors ($2 \times 2 \times \dots \times 2$). As such this will take $\log^2 x$ calls. Since the input is of the order $\log x$, we would be able to check the yes-certificate.

The no-certificate is the prime factorization of x .

1. Check each prime factor and see if it is $\geq y$.
2. We must check if each factor is indeed prime. We can do this with a black box since there is a poly-time algorithm for checking whether a number is prime (Manindra Agarwal).
3. Multiply them together to see if their product is x .
4. After doing this, you should be convinced that the answer is no.

1.3 MAX-SAT Problem

SAT (Satisfiability) was the first problem to be shown was NP-Complete (Proved by Stephen Cook in the early 1970's)

Input:

- n boolean variables (each can be set True or False)
- m clauses: C_1, C_2, \dots, C_m (each is a disjunction of literals)

Goal: Find an assignment of true/false to the variables that maximizes the number of satisfied clauses.

Remark 1.5 — The original satisfiability problem: Given any boolean formula include “and”, “or”, “not”, can the formula be “satisfied”?

Example 1.6

Let the boolean formula be:

$$(x_1 \vee x_2 \vee \overline{x_3}) \wedge (\overline{x_2} \vee x_4)$$

Can we assign x_i true or false to have the formula be true?

It turns out there this is equivalent (poly-time transformation) from this problem to MAX-SAT.

Definition 1.7. x_1 is a **variable**, and $\overline{x_1}$ is the **negation of a variable**. Collectively, they are **literals**

Definition 1.8. \vee : or, **disjunction**

\wedge : and, **conjunction**

Definition 1.9. A **clause** is one or more literals combined with disjunctions.

Definition 1.10. A **formula** is one or more clauses combined with conjunctions

We can rewrite SAT as: Given a set of clauses, is there an assignment of true/false to variable, such that all the clauses are satisfied.

Remark 1.11 — This is clearly in the class NP, as if the answer is yes, we can just give the solution.

MAX-SAT is the maximization version of this problem (note that this is also NP-Hard, as we can easily use this to solve SAT).

1.4 Randomized Algorithm 1

Algorithm

Set each variable x_i to true independently with probability $\frac{1}{2}$.

Analysis

- Let $C_j = x_1 \vee x_2 \vee \dots \vee x_k$.
- The probability that the clause is satisfied is $1 - \frac{1}{2^k} \geq \frac{1}{2}$.
- The expected number of clauses satisfied is:

$$\geq \frac{1}{2}m \geq \frac{1}{2}OPT.$$

Remark 1.12 — Consider the variable of the problem in which each clause has exactly 3 literals (MAX-3SAT). For this, we have:

$$\text{EXP \# of clauses satisfied} \geq \frac{7}{8}m,$$

meaning that the approximation factor would be $\frac{7}{8}$.

Theorem 1.13

If there is a $(\frac{7}{8} + \epsilon)$ - approximation algorithm for MAX-3SAT for any constant $\epsilon > 0$, then P=NP.

This means that our simple algorithm is the best possible for MAX-3SAT.