

# P $\neq$ NP proof

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We prove  $P \neq NP$  by contradiction by showing that there is no polynomial-time algorithm to solve the set partition problem.

## 1 Proof

### 1.1 The set partition problem

Given integers  $a_1, \dots, a_n$  and target  $c$ , determine whether there exist  $s_i \in \{-1, 1\}$  such that

$$\sum_{i=1}^n a_i s_i = c.$$

This problem is in NP since, given witness  $s$ , the verifier  $\sum_{i=1}^n a_i s_i - c$  is computable in  $O(n)$ .

### 1.2 Indistinguishable instance generation

Define generators  $G_1, G_2$ :

1.  $G_1$ : sample  $a'_i \sim U(-2^n, 2^n]$ , set  $a_i = |a'_i|$ ,  $c = \sum_i a'_i$ .
2.  $G_2$ : sample  $a'_i \sim U(-2^n, 2^n]$ , set  $a_i = |a'_i|$ ,  $c = \sum_i a'_i + 2$ .

Then every  $G_1$  instance is satisfiable with exactly two solutions (w.h.p.) and  $G_2$  instances are unsatisfiable (w.h.p). Without solving the NP problem, no polynomial-time algorithm can distinguish  $G_1$  from  $G_2$  with more than negligible advantage, since the two distributions are statistically indistinguishable.

Suppose  $P = NP$ . Then a polynomial-time solver  $S$  exists. Feeding instances from  $G_1, G_2$  to  $S$  yields a distinguisher:

$$S(I) = \begin{cases} G_1, & \text{if a solution found,} \\ G_2, & \text{otherwise.} \end{cases}$$

This distinguisher succeeds, contradicting indistinguishability.

Thus,

$$P \neq NP. \quad \square$$

## A Support and Total Variation Analysis

### A.1 Distributions of $X$ and $Y$ (sum-only version)

For the purpose of analyzing the generators  $G_1$  and  $G_2$ , define

$$X = \sum_{i=1}^n a_i \quad \text{under } G_1, \quad Y = \sum_{i=1}^n a_i + 2 \quad \text{under } G_2,$$

where  $a_i$  are sampled independently from  $U[0, 2^n]$ .

The distributions are then

$$\mathcal{L}(X) : \Pr[X = t] = \Pr\left[\sum_{i=1}^n a_i = t\right], \quad \mathcal{L}(Y) : \Pr[Y = t] = \Pr[X = t-2].$$

### A.2 Support Preservation of $G_2$

For fixed  $a_1, \dots, a_n$ , consider the lattice span

$$h(a) = \gcd(2a_1, 2a_2, \dots, 2a_n).$$

The support of  $X$  lies in a single residue class modulo  $h(a)$ . Since  $h(a) = 2 \cdot \gcd(a_1, \dots, a_n)$ , a shift of  $+2$  preserves the support iff  $\gcd(a_1, \dots, a_n) = 1$ .

For  $a_i$  sampled uniformly at random from  $\{0, \dots, 2^n\}$ ,

$$\Pr[\gcd(a_1, \dots, a_n) = 1] = \frac{1}{\zeta(n)} \rightarrow 1 \quad \text{as } n \rightarrow \infty.$$

Hence with overwhelming probability,  $h(a) = 2$ , and the shift  $+2$  remains within the same support.

### A.3 Total Variation Distance

By the central local limit theorem,

$$\text{TV}(\mathcal{L}(X), \mathcal{L}(Y)) = \frac{1}{2} \sum_{k \in \mathbb{Z}} |\Pr[X = k] - \Pr[X = k - 2]| = O\left(\frac{1}{\sigma\sqrt{n}}\right).$$

As  $\sigma = \Theta(2^n \sqrt{n})$ , the total variation distance between the sums of generator  $G_1$  and  $G_2$  is negligible in  $n$ .

### A.4 Conclusion: Statistical Indistinguishability

The distributions  $\mathcal{L}(X)$  and  $\mathcal{L}(Y)$  satisfy:

1. They are supported on the same lattice (span 2).

2. Their total variation distance obeys

$$\text{TV}(\mathcal{L}(X), \mathcal{L}(Y)) \leq \text{negl}(n).$$

Hence  $\mathcal{L}(X)$  and  $\mathcal{L}(Y)$  are *statistically indistinguishable*. That is, for any distinguisher  $D$  (even unbounded),

$$|\Pr[D(X) = 1] - \Pr[D(Y) = 1]| \leq \text{negl}(n).$$

Therefore, no algorithm (efficient or not) can distinguish  $G_1$  from  $G_2$  with more than negligible advantage.

## B Solutions of the Generators

In the previous sections we have already sufficiently analyzed the magnitude of  $\Pr[X = k]$  and similar statements.