

1. Prove that if $P = NP$, then every non-trivial problem in this class is NP-complete.

Let P be a non-trivial problem in $P = NP$. We want to show that P is NP-Complete, that is, every problem $Q \in NP$ reduces to P in polynomial time. Let I_1 be a fixed instance for P for which the answer is *Yes*, and I_2 a fixed instance for P for which the answer is *No*. Finally, let I be an input instance for Q . Since $NP = P$, there exists a polynomial-time algorithm to solve Q . Invoke this algorithm to solve Q on I . If the answer is *Yes* (respectively, *No*), the reduction algorithm generates the instance I_1 (respectively, I_2) for P . Since I_1 and I_2 are fixed instances, their sizes are constant, that is, do not depend on the size of I . Therefore the running time of the reduction algorithm is the same as that to solve Q on I . This shows that $Q \leq P$.

2. Graph isomorphism problem: Two graphs G, H (both directed or both undirected) are called isomorphic if there exists a bijection $f : V(G) \rightarrow V(H)$ such that $(u,v) \in E(G)$ if and only if $(f(u),f(v)) \in E(H)$.

(a) Prove that GI is in NP.

The map f is a certificate.

(b) Is GI NP-complete?

GI is not known to be NP-complete. Laszlo Babai proposes a pseudo-polynomial-time algorithm for this problem. See:

https://en.wikipedia.org/wiki/Graph_isomorphism_problem

3. Subgraph isomorphism problem: Given two graphs G and H , decide whether there exists an injective function $f : V(H) \rightarrow V(G)$ such that $(u,v) \in E(H)$ if and only if $(f(u),f(v)) \in E(G)$. Prove that SGI is NP-complete.

First, I show that SUBGRAPH-ISOMORPHISM is in NP. For an input (G,H) of SUBGRAPH-ISOMORPHISM, guess a subset $V' \subseteq V(G)$ together with a bijective function $f : V' \rightarrow V(H)$. Then, check whether f preserves the adjacency relations of V' in $V(G)$.

In order to show the NP-Hardness, I reduce CLIQUE to SUBGRAPH-ISOMORPHISM. Let (G,r) be an input for CLIQUE. We construct the input (G,K_r) for SUBGRAPH-ISOMORPHISM, where K_r is the complete graph on r vertices. G has an r -clique if and only if G has a subgraph isomorphic to K_r . This reduction evidently runs in polynomial time.

4. Partition problem: You are given n positive integers a_1, a_2, \dots, a_n such that

$a_1 + a_2 + \dots + a_n = A$ is even.

Decide whether the given integers can be partitioned into two subcollections such that the sum of the integers in each subcollection is $A / 2$. Prove that PARTITION is NP-complete.

Use reduction from SUBSET-SUM. Let S, t be an instance of SUBSET-SUM with $A = \sum_{a \in S} a$. Create the instance $S \cup \{2A - t, A + t\}$ for PARTITION.

5. Bin-packing problem: You are given n objects of weights w_1, w_2, \dots, w_n . You are given an infinite supply of bins each with weight capacity C . You want to pack all the objects in m bins such that m is as small as possible. Assume that each $w_i \leq C$.
- (a) Frame an equivalent decision problem, and prove that the decision problem can be solved in poly-time if and only if the optimization problem can be solved in poly-time.

Equivalent decision problem: Decide whether the objects can be packed in m bins for any given $m \geq 1$. A solution of the optimization problem clearly indicates whether m bins suffice. Conversely, if we have an oracle solving the decision problem, we can invoke the oracle with $m = 1, 2, \dots, n$ until the decision is *yes*. The running time increases by a factor of n only. We can do binary search to increase the running time by a factor of $\log n$ only.

(b) Prove that the decision version is NP-Complete.

Use reduction from PARTITION (Exercise 6.17). Let $S = (a_1, a_2, \dots, a_n)$ be an input instance for PARTITION with $\sum_{i=1}^n a_i = 2t$. Take n objects of weights $w_i = a_i$, bins each of capacity $C = t$, and $m = 2$. The partition problem has a solution if and only if two bins suffice.

6. Knapsack problem: A thief finds n objects of weights w_1, w_2, \dots, w_n and integer-valued profits p_1, p_2, \dots, p_n . The thief has a knapsack of capacity C . The goal of the thief is to pack objects in the knapsack without exceeding the capacity so as to maximize the profit of packed objects.

0,1 variant: Each object can be packed or discarded.

Fractional variant: Any fraction of any object can be packed.

(a) Prove that the fractional knapsack problem can be solved in polynomial time.

Pack the objects in the decreasing order of p_i / w_i values.

This is a greedy algorithm. Prove its correctness.

(b) Formulate an equivalent decision version of the 0,1 knapsack problem.

Decision problem: Given $w_1, w_2, \dots, w_n, p_1, p_2, \dots, p_n, C$, and a (positive) integer P , decide whether a profit of $\geq P$ is achievable without exceeding the knapsack capacity C .

(a) [If] Let M be a polynomial-time algorithm for solving the maximization problem. Using M , we determine the maximum profit P^* , and return *true* if and only if $P^* \geq P$.

[Only if] Let D be a polynomial-time algorithm for solving the decision problem. We invoke D multiple times with separate profit bounds P in order to determine the maximum profit P^* . Initially, we start with $L = 0$ and $R = \sum_{i=1}^n p_i$, since we definitely know that P^* must lie between these two values. We compute $P = \lfloor (L + R) / 2 \rfloor$, and call D with this profit bound P . If D returns *true*, we conclude that P^* is between P and R , so we set $L = P$. On the other hand, if D returns *false*, we set $R = P - 1$, since P^* must be smaller than P . This binary search procedure is repeated until we have $L = R$. We output this value ($L = R$) as P^* .

The total number of invocations of D is $O(\log \sum_{i=1}^n p_i)$ which is $O(\log(np_{\max}))$. Since each invocation runs in polynomial time, the total running time is polynomial in n and $\log p_{\max}$.

(c) Prove that the decision version of the 0,1 knapsack problem is NP-complete.

(b) Clearly, the decision version of the knapsack problem is in NP.

In order to prove its NP-hardness, we reduce PARTITION to it. Let a_1, a_2, \dots, a_n be an input instance for PARTITION with $A = \sum_{i=1}^n a_i$.

We consider n objects O_1, O_2, \dots, O_n such that the weight of O_i is $w_i = 2a_i$ and the profit of O_i is $p_i = 2a_i$. Finally, we take the knapsack capacity $C = A$ and the profit bound $P = A$. Clearly, this reduction can be done in polynomial time.

Suppose that $\sum_{j=1}^k a_{i_j} = A/2$ for some subcollection $a_{i_1}, a_{i_2}, \dots, a_{i_k}$ of a_1, a_2, \dots, a_n . This implies that $\sum_{j=1}^k w_{i_j} = 2 \times (A/2) \leq C$ and $\sum_{j=1}^k p_{i_j} = 2 \times (A/2) \geq P$, that is, the objects $O_{i_1}, O_{i_2}, \dots, O_{i_k}$ satisfy the capacity constraint and the profit bound.

Conversely, suppose that the objects $O_{i_1}, O_{i_2}, \dots, O_{i_k}$ satisfy $\sum_{j=1}^k w_{i_j} \leq C$ and $\sum_{j=1}^k p_{i_j} \geq P$. These, in turn, imply that $\sum_{j=1}^k 2a_{i_j} \leq A$ and $\sum_{j=1}^k 2a_{i_j} \geq A$, that is, $\sum_{j=1}^k a_{i_j} = A/2$. Therefore, the integers $a_{i_1}, a_{i_2}, \dots, a_{i_k}$ satisfy the requirement of the PARTITION problem.

A is assumed to be even. If so, you can take $w_i = p_i = a_i$, and $C = P = A/2$.

7. Let $G = (V, E)$ be an undirected graph, and k a positive integer. You want to determine whether the removal of some k or fewer edges from E makes G bipartite. Prove that this problem is NP-complete.

The problem is clearly in NP, because the list of edges to remove from E in order to make it bipartite is a succinct certificate for the problem. It is easy to verify whether the list contains $\leq k$ edges, and their removal from G leaves a bipartite graph.

For NP-Hardness, we make a reduction from MAX-CUT. Let (G, r) be an instance for MAX-CUT. If m is the number of edges in G , generate the instance (G, k) of the given problem, where $k = m - r$. G has a cut of size $s \geq r$ if and only if the removal of the remaining $m - s$ edges makes G bipartite. The number of edges removed is $m - s \leq m - r = k$.

8. **Weighted MAX-CUT problem:** Let $G = (V, E)$ be an undirected graph with each edge e carrying a positive weight w_e (assume to be a positive integer). The weight of a cut X, Y of V is the sum of the weights of the edges connecting X and Y . Let k be a positive integer. Decide whether G has a cut of weight $\geq k$. Prove that the weighted MAX-CUT problem is NP-complete.

Weighted MAX-CUT is a generalized version of the (unweighted) MAX-CUT problem. More specifically, reduce MAX-CUT to wt-MAX-CUT as follows. Let (G, k) be an instance of MAX-CUT. Take the weight of each edge of G as 1, and pass the same G and k along with these weights to the output.