

CSCE 629-601 Analysis of Algorithms

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Solutions to Assignment # 6

1. A *vertex cover* in an undirected graph G is a set C of vertices in G such that every edge in G has at least one end in C . Consider the following two versions of the Vertex-Cover problem:

VC-D: Given a graph G and an integer k , decide whether G contains a vertex cover of at most k vertices.

VC-O: Given a graph G , construct a minimum vertex cover for G

Prove: VC-D is in \mathcal{P} if and only if VC-O is in \mathcal{P} .

Solutions. Suppose that the VC-O problem is in \mathcal{P} , i.e., is solvable in polynomial time. Thus, there is a polynomial time algorithm \mathcal{A}_O that solves the VC-O problem. We construct a polynomial time algorithm \mathcal{A}_D for the VC-D problem as follows: given an instance (G, k) of the VC-D problem, run \mathcal{A}_O on G to obtain an minimum vertex cover C in G , then return “yes” if and only if $|C| \leq k$. The algorithm \mathcal{A}_D is correct because the graph G contains a vertex cover of at most k vertices if and only if the minimum vertex cover of G contains no more than k vertices. Since the algorithm \mathcal{A}_O runs in polynomial time, the algorithm \mathcal{A}_D runs in polynomial time. Thus, the problem VC-D is solvable in polynomial time, i.e., it is in \mathcal{P} .

Conversely, suppose that the VC-D problem is in \mathcal{P} so there is a polynomial time algorithm \mathcal{A}'_D for the VC-D problem. We can construct a polynomial time algorithm \mathcal{A}'_O for the VC-O problem, which is given in Figure 1

By our assumption, the algorithm \mathcal{A}'_D runs in polynomial time. Because the algorithm \mathcal{A}'_O calls the algorithm \mathcal{A}'_D at most $n + n^2$ times (at most n times in step 1 and at most n times in step 3.1.1 for each i), the algorithm \mathcal{A}'_O also runs in polynomial time.

To see the correctness of the algorithm \mathcal{A}'_O , first note that step 1 find the smallest integer i_0 such that $\mathcal{A}'_D(G, i_0) = \text{yes}$. Thus, after step 1, $i = i_0$ is size of a minimum vertex cover of the graph G . In general, for each execution of step 3.1, we know that the minimum vertex cover is of size i , and we are looking for a vertex cover of i vertices in the graph G . In particular, if $\mathcal{A}'_D(G - v, i - 1) = \text{yes}$, then the graph $G - v$ has a vertex cover of size $i - 1$, which, plus the vertex v , gives a minimum vertex cover of i vertices in the graph G . This means that the vertex v is in a minimum vertex cover of the graph G . As a result, if this is true at step 3.1.1, then we can include v in the minimum vertex cover C (at step 3.1.2), then recursively look for a (minimum) vertex cover of $i - 1$ vertices in the graph $G - v$. Step 3.1.2 of the algorithm thus correctly updates the graph G and the value i . In particular, at end of step 3, the set C contains

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Algorithm  $\mathcal{A}'_O$ 
Input: a graph  $G$ 
Output: a minimum vertex cover  $C$  in  $G$ 
1. for ( $i = 0$ ;  $i \leq n$ ;  $i++$ ) do if ( $\mathcal{A}'_D(G, i) = \text{yes}$ ) break;
2.  $C = \emptyset$ ;
3. while ( $i > 0$ )
3.1 for (each vertex  $v$  in  $G$ ) do
3.1.1 if ( $\mathcal{A}'_D(G - v, i - 1) = \text{yes}$ )
3.1.2  $C = C \cup \{v\}$ ;  $G = G - v$ ;  $i = i - 1$ ;
3.1.3 break;  $\backslash\backslash$  break the for-loop
4. return( $C$ ).

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Figure 1: The algorithm \mathcal{A}'_O for VC-O

a vertex cover of i_0 vertices for the graph G , where i_0 is the value i after step 1, i.e., i_0 is the size of a minimum vertex cover of the input graph G . In conclusion, the algorithm \mathcal{A}'_O runs in polynomial time and returns a minimum vertex cover C of the input graph G (at step 4). In other words, the VC-O problem is solvable in polynomial time, i.e., it is in \mathcal{P} .

This completes the proof of the question. \square

2. Using the fact that the INDEPENDENT SET problem is \mathcal{NP} -complete, prove that the following problem is \mathcal{NP} -complete:

CLIQUE: Given a graph G and an integer k , is there a set C of k vertices in G such that for every pair v and w of vertices in C , v and w are adjacent in G ?

Solutions. We first show that CLIQUE is in \mathcal{NP} . An instance of the CLIQUE problem takes the form (G, k) , where G is a graph and k is an integer. Consider the algorithm given in Figure 2.

For a Yes-instance $x = (G, k)$ of CLIQUE, i.e., if there is a set C of k vertices in G in which every pair of vertices are adjacent, then when the string y is this set C , the algorithm **VerifyClique**(x, y) will pass the tests in steps 1-2, and return in step 3 with an answer "Yes." On the other hand, if $x = (G, k)$ is a No-instance of CLIQUE, i.e., if there is no set of k vertices in which all vertices are pairwise adjacent, then the algorithm **VerifyClique**(x, y) for any y will either return in step 1 with a "No" answer (if y does not encode a set of k vertices in the graph G) or return in step 2 with a "No" answer (since the graph G does not have k pairwise adjacent vertices). That is, on a No-instance x of CLIQUE, the algorithm **VerifyClique**(x, y) returns "No" for all y . Finally, the algorithm **VerifyClique**(x, y) obviously runs in time $O(|x|^2)$ (assuming the graph G is given in an adjacency matrix). By the definition of problems in \mathcal{NP} , the algorithm **VerifyClique** shows that the problem CLIQUE is in \mathcal{NP} .

Now, we show that the INDEPENDENT SET problem is polynomial-time reducible to the CLIQUE problem, which will give the \mathcal{NP} -hardness of the CLIQUE problem. This reduction uses the notion of "complement graphs". Given an undirected graph $G = (V, E)$, the *complement graph* \overline{G} of G is defined as $\overline{G} = (V, \overline{E})$, which has same set V of vertices, and the edge set \overline{E} is defined as $\overline{E} = \{[u, v] \mid u, v \in V, u \neq v, \text{ and } [u, v] \notin E\}$.

The reduction algorithm takes as input an instance (G, k) of INDEPENDENT SET, constructs the complement graph \overline{G} of G , then outputs (\overline{G}, k) as an instance of CLIQUE. The algorithm

Algorithm. VerifyClique($x = (G, k), y$)
Input: $x = (G, k)$, where G is a graph and k is an integer k , and a string y
Output: verify if y is a solution to the CLIQUE instance $x = (G, k)$
1. **if** (y is not a set of k vertices in G) return("No");
2. **if** (any two vertices in y are not adjacent) return("No");
3. return("Yes").

Figure 2: CLIQUE is in \mathcal{NP}

obviously runs in polynomial-time (more precisely, in time $O(n^2)$). Moreover, it is easy to see that the graph G has a set C of k vertices in which no two are adjacent if and only if the same set C of k vertices in the complement graph \overline{G} has all the k vertices pairwise adjacent. That is, (G, k) is a Yes-instance of INDEPENDENT SET if and only if (\overline{G}, k) is a Yes-instance of CLIQUE. Thus, INDEPENDENT SET \leq_m^P CLIQUE. Since the INDEPENDENT SET problem is \mathcal{NP} -complete (thus, is \mathcal{NP} -hard), this reduction shows that the CLIQUE problem is \mathcal{NP} -hard.

Summarizing the discussions, we conclude that the CLIQUE problem is \mathcal{NP} -complete. \square

3. Using the fact that the PARTITION problem is \mathcal{NP} -complete, prove that the following problem is \mathcal{NP} -complete:

KNAPSACK: given n items of sizes s_1, s_2, \dots, s_n and values v_1, v_2, \dots, v_n , respectively, a knapsack of size S , and a value objective V , can we select some of these items to fit into the knapsack so that the total value of the selected items is at least V ?

Solutions. We first show that KNAPSACK is in \mathcal{NP} . An instance of the KNAPSACK problem is a tuple of $2n + 2$ integers $x = (s_1, v_1, s_2, v_2, \dots, s_n, v_n; S; V)$. Consider the algorithm in Figure 3.

For a Yes-instance $x = (s_1, v_1, \dots, s_n, v_n; S; V)$ of KNAPSACK, i.e., if there is a set A of items (note that the items are given by the integers $\{1, 2, \dots, n\}$). Thus, a set A of items is a subset of $\{1, 2, \dots, n\}$, which can fit into the knapsack (i.e., $\sum_{i \in A} s_i \leq S$) with the value at least V (i.e., $\sum_{i \in A} v_i \geq V$), then when the string y is this item set A , the algorithm **VerifyKnapsack**(x, y) will pass the tests in steps 1-2, and return in step 3 with an answer "Yes." On the other hand, if $x = (s_1, v_1, \dots, s_n, v_n; S; V)$ is a No-instance of KNAPSACK, i.e., if there is no item set A that satisfies both $\sum_{i \in A} s_i \leq S$ and $\sum_{i \in A} v_i \geq V$, then the algorithm **VerifyKnapsack**(x, y) for any y will either return in step 1 with a "No" answer (if y does not encode a set of items) or return in step 2 with a "No" answer (since there is no item set A that satisfies both $\sum_{i \in A} s_i \leq S$ and $\sum_{i \in A} v_i \geq V$). That is, on a No-instance x of KNAPSACK, the algorithm **VerifyKnapsack**(x, y) returns "No" for all y . Finally, the algorithm **VerifyKnapsack**(x, y) obviously runs in time $O(|x|)$ (for first checking if y is a subset of $\{1, 2, \dots, n\}$, then verifying the conditions for the size sum and value sum). By the definition of problems in \mathcal{NP} , the algorithm **VerifyKnapsack** shows that the problem KNAPSACK is in \mathcal{NP} .

Now, we show that the PARTITION problem is polynomial-time reducible to the KNAPSACK problem, which will give the \mathcal{NP} -hardness of KNAPSACK. The reduction algorithm takes as input an instance $x = (a_1, a_2, \dots, a_n)$ of PARTITION, and outputs a tuple $y = (s_1, v_1, \dots, s_n, v_n; S; V)$ as an instance of KNAPSACK, where $s_i = v_i = a_i$ for $1 \leq i \leq n$, and $S = V = (\sum_{i=1}^n a_i)/2$. The algorithm obviously runs in polynomial-time (more precisely, in time $O(n)$).

Algorithm. VerifyKnapsack($x = (s_1, v_1, \dots, s_n, v_n; S; V), y$)
Input: a KNAPSACK instance $x = (s_1, v_1, \dots, s_n, v_n; S; V)$, and a string y
Output: verify if y is a solution to the KNAPSACK instance x

1. **if** (y is not a subset of $\{1, 2, \dots, n\}$) **return**("No");
2. **if** ($\sum_{i \in y} s_i > S$ or $\sum_{i \in y} v_i < V$) **return**("No");
3. **return**("Yes").

Figure 3: KNAPSACK is in \mathcal{NP}

We show that x is a Yes-instance of PARTITION if and only if y is a Yes-instance of KNAPSACK.

Suppose that $x = (a_1, a_2, \dots, a_n)$ is a Yes-instance of PARTITION. Then, we can divide the set $\{1, 2, \dots, n\}$ into two disjoint subsets L and R such that $\sum_{l \in L} a_l = \sum_{r \in R} a_r$. In this case, we must have $\sum_{l \in L} a_l = \sum_{r \in R} a_r = (\sum_{i=1}^n a_i)/2$. Now consider the item subset L (again a subset of items is given by a subset of $\{1, 2, \dots, n\}$). Then we have $\sum_{l \in L} s_l = \sum_{l \in L} a_l = (\sum_{i=1}^n a_i)/2 = S$ and $\sum_{l \in L} v_l = \sum_{l \in L} a_l = (\sum_{i=1}^n a_i)/2 = V$. Therefor, L is a item set that satisfies both $\sum_{l \in L} s_l \leq S$ and $\sum_{l \in L} v_l \geq V$, i.e., the instance $y = (s_1, v_1, \dots, s_n, v_n; S; V)$ is a Yes-instance of the KNAPSACK problem.

On the other hand, if $y = (s_1, v_1, \dots, s_n, v_n; S; V)$ is a Yes-instance of KNAPSACK, i.e., if there is an item set L that satisfies both $\sum_{t \in L} s_t \leq S$ and $\sum_{t \in L} v_t \geq V$, then since $s_t = v_t = a_t$ for all t and $S = V = (\sum_{i=1}^n a_i)/2$, we must have

$$(\sum_{i=1}^n a_i)/2 = V \leq \sum_{l \in L} v_l = \sum_{l \in L} a_l = \sum_{l \in L} s_l \leq S = (\sum_{i=1}^n a_i)/2.$$

Thus, the subset L of $\{1, 2, \dots, n\}$ satisfies $\sum_{l \in L} a_l = (\sum_{i=1}^n a_i)/2$. If we let $R = \{1, 2, \dots, n\} - L$, then obviously we also have $\sum_{r \in R} a_r = \sum_{i=1}^n a_i - \sum_{l \in L} a_l = (\sum_{i=1}^n a_i)/2$. That is, the set of integers $\{a_1, a_2, \dots, a_n\}$ can be divided into two subsets L and R that have the same sum. This shows that $x = (a_1, a_2, \dots, a_n)$ is a Yes-instance of PARTITION.

This completes the proof that PARTITION is polynomial-time reducible to KNAPSACK. Since PARTITION is \mathcal{NP} -complete (thus, is \mathcal{NP} -hard), the problem KNAPSACK is \mathcal{NP} -hard.

Summarizing the discussions, we conclude that the KNAPSACK problem is \mathcal{NP} -complete. \square