CSOR 4231 - Introduction to Algorithms HW #6

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Problem 1 Statement. A large store has m customers and n products and maintains an $m \times n$ matrix A such that $A_{ij} = 1$ if customeri has purchased product j; otherwise, $A_{ij} = 0$. Two customers are called orthogonal if they did not purchase any products in common. Your task is to help the store determine a maximum subset of orthogonal customers. Give an efficient algorithm for this problem or state its decision version and prove it is NP-complete.

Problem 1 Solution. We posit that this problem is NP-complete. Its decision version can be stated as follows:

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CUST-ORTH(D) : find a set of orthogonal customers of size \geq k.
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Our proof requires two steps: (1) show CUST-ORTH(D) \in NP; (2) show that IS(D) \leq_p CUST-ORTH(D).

Step 1: Let us have the certification algorithm CUST-ORTH-CERT(A, S, k), where A is the original matrix, the certificate S is a collection of customers s.t. $S_i \perp S_j, \forall i, j \in m \times n$, and k is the minimum number of such sets required.

Algorithm 1 certify that the given certificate solves CUST-ORTH(D)

```
1: procedure CUST-ORTH-CERT(A, S, k)
       if size(S) \leq k then
2:
3:
          return NO
       end if
4:
       for i = 1 to n do
5:
          for j = 1 to m do
6:
              if j \neq i and S_j \not\perp S_i then
7:
                 return NO
8:
              end if
9:
          end for
10:
       end for
11:
       return YES
12:
13: end procedure
```

It is trivial to see that this algorithm runs in $O(mn^2)$ time, which is polynomial in the size of the input. Thus, $CUST - ORTH(D) \in NP$.

Step 2: It is known that the independent-set decision problem (IS(D)) is NP-complete. Input to IS(D) is the graph G(V, E), and the value k. Broadly, we'll map edges to products and vertices to customers. We define our reduction procedure, R, from IS(D) to CUST-ORTH(D) as follows:

- Let m = |E| and n = |V|; our A will be size $n \times m$ and initially zero
- For each edge $e \in E$, give a numeric label to that edge, this will be its column-index j in A
- For each vertex $u \in V$, give a numeric label to that vertex, this will be its row-index i in A
- For each edge $e = (u, v) \in E$, $A_{ue} = 1$ and $A_{ve} = 1$, where e, u, v are indices as specified above
- \bullet the k value remains the same for both problems

Clearly this takes only polynomial time to translate the inputs. Now it remains to show that - under this translation - there is an independent-set of size $\geq k$ for IS(D) \iff there is an orthogonal set of customers of size $\geq k$ for CUST-ORTH(D).

 \Longrightarrow Suppose there is an independent-set of size k for IS(D). For all pairs of vertices selected from the generated independent set, say u and v, we have that $e=(u,v)\not\in E$. Thus in A, the entries $A_{ue}=0$ and $A_{ve}=0$ will be created. Since the column's of A represent edges, and one edge connects two vertices, either both $A_{ae}=1$ and $A_{be}=1$ or both are 0. There cannot be some $A_{ij}=1$ and $A_{kj}=1$ if there is no edge (i,j); And there are no edges (i,j) between any vertices i and j of the independent-set. So, under R, there will be at least k row vectors which are all mutually orthogonal.

 \Leftarrow Suppose there is a set of k orthogonal customers for CUST-ORTH(D). As described above, there can only be two 1's in the same column of A if there was originally an edge in G connecting the two vertices corresponding to these customers. For every pair a and b of customers in the orthogonal set, $A_{aj} = 1 \implies A_{bj} = 0$ (and vice. versa) for j = 1, 2, ..., n, . This implies that there could not have been any edges connecting the corresponding vertices of a and b in G. Thus, the set of k orthogonal customers correspond exactly to k vertices in G for which no two are connected by an edge - an independent set.

Now we may conclude that CUST-ORTH(D) $\in NPC$.

Algorithms – HW6 2 Alexander Stein

Probelm 2 Statement. Suppose you had a polynomial-time algorithm that, on an input graph, answers **yes** if and only if the graph has a Hamiltonian cycle. Show how, on an input graph G = (V, E), you can return in polynomial time

- a Hamiltonian cycle in G, if one exists,
- **no**, if G does not have a Hamiltonian cycle.

Problem 2 Solution. The general idea here is to examine one edge of G at a time. We look at the graph $G' = (V, E - \{e\})$ and if HAS-HAM(G') = yes, then we can throw that edge away because G' still has some Hamiltonian cycle. If it returns no, then we conclude that the edge we just removed must have been in every remaining Hamiltonian cycle - put it back in the graph, but mark it as visited so we do not repeat this check.

Algorithm 2 use the HAS-HAM(G) "magic" algorithm to return a Hamiltonian cycle efficiently

```
1: procedure GET-HAM(G = (V, E))
      if HAS-HAM(G) = no then
3:
          return no
       end if
 4:
       Let crit be a bit-vector of length |E| initialized to zeros
 5:
       while |E| > |V| do
 6:
          Let e be some edge s.t. e \in E and crit[e] = 0
 7:
          if HAS-HAM(G' = (V, E - \{e\})) = yes then
 8:
             remove e from E
9:
          else
10:
             crit[e] = 1
11:
          end if
12:
       end while
13:
14:
       return E
15: end procedure
```

First notice that we may terminate when only |V| edges remain, because a simple cycle has exactly |V| edges. Next, note that the crit indicator for an edge is only set to 1 if that edge must be in a Hamiltonian cycle. Upon termination, the E set returned will contain only those edges that are in a Hamiltonian cycle, if one exists.

If we let the running of time for HAS-HAM(G) be indicated by H(t), which we know to be polynomial, then the running time of GET-HAM(G) is O(mH(t)), which is also polynomial by composition.

Problem 3 Statement. There is a set of ground elements $E = e_1, e_2, ..., e_n$ and a collection of m subsets $S_1, S_2, ..., S_m$ of the ground elements (that is, $S_i \subseteq E$ for $1 \le i \le m$). The goal is to select a minimum cardinality set A of ground elements such that A contains at least one element from each subset S_i . State the decision version of this problem and prove that it is NP-complete.

Problem 3 Solution. Its decision version can be stated as follows:

```
SUBSET-COVER(D): find a subset A as stated above of size \leq k
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Our proof requires two steps: (1) show SUBSET-COVER(D) \in NP; (2) show that VC(D) \leq_p CUST-ORTH(D).

Step 1: Let us have the certification algorithm SUBSET-COVER-CERT(E, S, A, k), where E is the original ground-element-set, S the original collection of subsets, the certificate is $A \subseteq E$, and k is the maximum size of A allowed.

Algorithm 3 certify that the given certificate solves SUBSET-COVER(D)

```
1: procedure SUBSET-COVER-CERT(E, S, A, k)
       if size(A) \ge k then
2:
          return NO
 3:
       end if
 4:
       Let X be an array of |S| booleans, false-initialized
 5:
       for i = 1 to |S| do
 6:
          for j = 1 to |A| do
 7:
              if A[j] \in S_i then
 8:
                 X[i] = true
9:
              end if
10:
          end for
11:
       end for
12:
       if any element of X = false then
13:
          return NO
14:
       end if
15:
       return YES
16:
17: end procedure
```

Note that m is the number of sets in S, and there are at most n of them. The size of each set is obviously at most n. So the for-loop on line 4 executes n times. The inner for-loop on line 5 executes at most k times (which is at most n). The check for membership on line 6 is also at most n. All other work in this algorithm is dominated by lines 4-7, so the overall running time is $O(n^3)$, which is polynomial. Thus: SUBSET-COVER(D) \in NP.

Step 2: It is known that the vertex-cover decision problem (VC(D)) is NP-complete. Input to VC(D) is the graph G(V, E), and the value k. We define our simple reduction procedure, R, from VC(D) to SUBSET-COVER(D) as follows:

- the vertices of G will be the elements of E
- For each edge $e = (u, v) \in G.E$, create a new subset $S_i = \{u, v\}$ and add it to S. (The value of the index doesn't matter)
- \bullet the k value remains the same for both problems

Clearly this takes only polynomial time to translate the inputs. Now it remains to show that - under this translation - there is a vertex-cover of size $\leq k$ for VC(D) \iff there is a subset-cover of size $\leq k$ for

SUBSET-COVER(D).

 \Longrightarrow Suppose there is a vertex-cover of size k for VC(D). Under R, every set $S_i \in S$ we create will have at least one element e_v corresponding to v from that vertex-cover, because there is one set in S for every edge in G.E, and the vertex-cover will "touch" every edge at least once (by definition). Thus, the set A of only those elements corresponding to vertices of the vertex-cover constitutes a subset-cover of size k for SUBSET-COVER(D).

 \Leftarrow Suppose there is a subset-cover A of size k for SUBSET-COVER(D). Every set $S_i \in S$ must correspond to an edge in G.E. Every element $e \in A$ must be correspond to a vertex in G.V. So, if A contains at least one element from every set, then the vertices corresponding to those elements must "touch" every edge in G.E at least once. Thus, G must have had a vertex-cover of size k.

Now we may conclude that SUBSET-COVER(D) $\in NPC$.

Algorithms – HW6 5 Alexander Stein

Problem 4 Statement. A paper mill manufactures rolls of paper of standard width 3 meters. Customers watn to buy paper rolls of shorter width, and the mill has to cut such rolls from the 3m rolls. For example, one 3m roll can be cut into 2 rolls of width 93cm and one roll of width 108 cm; the remaining 6cm goes to waste. The mill receives an order of

- 97 rolls of width 135 cm
- 610 rolls of width 108 cm
- 395 rolls of width 93 cm
- 211 rolls of width 42 cm

Form a linear program to compute the smallest number of 3m rolls that have to be cut to satisfy this order, and explain how they should be cut.

Problem 4 Solution. Let us define some terms for clarity:

- Let $c_0 = 135$, $c_1 = 108$, $c_2 = 93$, and $c_3 = 42$. These are our *cut-types*.
- Let x_i represent the total number of 3m rolls which we will cut by cut-method i (defined later).
- Let A be our coefficient matrix, where $a_{ij} = \#$ of cuts of type j for cut-method i (for a single roll)
- Let $k_0 = 97$, $k_1 = 610$, $k_2 = 395$, and $k_3 = 211$. These are the number of resized rolls ordered by each of the 4 customers.

Calculating the coefficients of A is a straightforward (albeit tedious) process. What we did to achieve this was as follows:

- List all the possible combinations of each of the cut-types. Since there are 4 types, there are $2^4 1 = 15$ situations to examine.
- Each such combination will be of the bounded form $a_0c_0 + a_1c_1 + a_2c_2 + a_3c_3 \leq 300$.
- We are considering unique types of combinations or cut-methods, i.e. one such method says "we are cutting this 300cm roll into a_0 135-cm pieces, and a_3 42-cm pieces. As they are unique, writing out a binary-counting-decimal matrix was helpful in keeping track of what combinations were left to be calculated.
- We solved for the a_i of each cut-method by minimizing the wasted material (300 sum), and enforcing that each coefficient take on an integer value > 0. It had to be greater than zero because otherwise it devolved into a different cut-method.
- In certain cases, there were more than one set of non-zero integer solutions. In the case that only one a-value changed from set to set, we just took the set with the largest varying value. However in the case where multiple a-values change from set to set, we added a new row to A with those constants, because such a set could potentially be used for a more optimal solution during SIMPLEX.

Algorithms – HW6 6 Alexander Stein

Here is the matrix we calculated:

Note that there are 15 rows, thus we will have 15 variables X_i . For example, X_{14} represents the number of rolls on which we will make 1×108 cm cut, 1×93 cm cut, and 2×42 cm cuts. Now we may finally formulate our Linear Program:

Objective:

$$Minimize \sum_{j=1}^{n} x_j$$

Subject to:

(i)
$$A^T \vec{x} \ge \vec{k}$$

(ii)
$$\vec{x} \geq \vec{0}$$

The first constraint just says "for each cut type c_i , enforce that we create at least k_i such cuts." Note that by the first constraint we are allowing for the possibility of creating more cuts of a particular type than required. This is necessary because requiring exact equivalence will very likely cause the LP to return "no solution."

Problem 5 Statement. Formulate linear or integer programs for the following optimization problems. (Full-credit will be given to LP solutions, when they are possible.)

- (a) Min-cost flow: Given a flow network with capacities c_e and costs a_e on every edge e, and supplies s_i on every vertex i, find a feasible flow $f: E \to R_+$ that is, a flow satisfying edge capacity constraints and node supplies that minimizes total cost of the flow.
- (b) The assignment problem: There are n persons and n objects that have to be matched on a one-to-one basis. There's a given set A of ordered pairs (i,j), where a pair (i,j) indicates that person i can be matched with object j. For every pair $(i,j) \in A$, there's a value a_{ij} for matching person i with object j. Our goal is to assign persons to objects so as to maximize the total value of the assignment.
- (c) Uncapacitated facility location: There is a sret F of m facilities and a set D of n clients. For each facility $i \in F$ and each client $j \in D$, there is a cost c_{ij} of assigning client j to facility i. Further, there is a one-time cost f_i associated with opening and operating facility i. Find a subset F' of facilities to open that minimizes the total cost of (i) operating the facilities in F' and (ii) assigning every client j to one of the facilities in F'.

Problem 5 Solution.

(a) Min-cost flow: For this problem we can formulate a linear program. Our variables are the amount of flow f_e on each edge. Let \vec{x} be an $1 \times m$, m = |E| vector representing the flows. Similarly, \vec{a} is an $1 \times m$ vector representing the edge costs, \vec{c} a $1 \times m$ vector representing the capacities, and \vec{s} an $1 \times n$, n = |V| vector representing the supplies of the vertices.

Objective:

Minimize $\vec{a}\vec{x}^T$

Subject To:

(i)
$$\sum_{j:e_i \text{ out of } i} x_j - \sum_{j:e_i \text{ in to } i} x_j = s_i \ , \ i=1,2,...,n$$

(ii)
$$\sum_{i \in V} \sum_{j:e_j \text{ out of } i} x_j - \sum_{i \in V} \sum_{j:e_j \text{ in to } i} x_j = 0 \ , \ i=1,2,...,n$$

(iii)
$$\vec{x} < \vec{c}$$

(iv)
$$\vec{x} > \vec{0}$$

(b) Assignment Problem: This problem too can be formulated as a linear program, but first we must transform the inputs a little bit - into a graph! The set A we will move directly to G.E. The persons i and objects j will all be added to G.V as individual vertices. Now it should be apparent that we have a bipartite graph, because there will be no edge between any person nodes, and no edge between any object nodes – only between person and object. The weights of each edge will equal $-a_{ij}$, because we wish to maximize "cost" instead of minimize it. Furthermore, since the problem requires an exact one-to-one correspondence, this graph problem is now a "maximum bipartite matching" problem.

Unfortunately, we're not done yet! We will translate the problem into Min-cost-flow, for which we already formulated an LP in part (a). To do so we simply (1) add the source and sink nodes s and t to G.V; (2) for every person node i add the zero-cost-edge (s,i) to G.E, and for every object node j add the zero-cost-edge (j,t) to G.E; (3) let the capacity of every edge c_e in the graph be 1; (4) let all supplies $s_i = 0$, $s_j = 0$ except $s_s = n$ and $s_t = -n$. When this new input is solved for min-cost-flow, we can take the person-object edges selected by that flow as our maximum value assignment.

(c) Uncapacitated facility location: For this one we are stuck with an (NP-complete) integer program formulation. We'll use $m \times n$ indicator variables:

$$x_{ij} = \begin{cases} 1 & \text{if client } j \text{ is assigned to facility } i \\ 0 & \text{otherwise} \end{cases}$$

If we define cost as "cost of opening facilities" + "cost of assigning customers" then we can define our integer program as follows:

Objective:

Minimize:
$$\sum_{i=1}^{m} f_i * \max_{1 \le j < n} x_{ij} + \sum_{i=1}^{m} \sum_{j=1}^{n} x_{ij} * c_{ij}$$

Subject To:

(i)
$$\sum_{i=1}^{m} x_{ij} = 1 \text{ for } j = 1, 2, ..., n$$

$$(ii) x_{ij} \in \{0, 1\}$$

Note that the quantity $\max_{1 \le j < n} x_{ij} \in \{0, 1\}$ because though a facility may have multiple customers assigned to it, it is only every opened once!