CSC373H1 Summer 2014 Assignment 4

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Q1. The Mute Prison

Claim: The mute prison problem is NP-complete.

Proof:

- 1. Show the mute prison problem is NP.
- 2. Show the mute prison problem is NP-hard.
- 1. Suppose we are given a certificate S and have access to value k and matrix T. We can verify that the certificate is satisfiable in the following way. Suppose each element in S represents an inmate. Verification would involve iterating on each inmate in the following way:

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for inmate\ in\ S do |\ j=1; while j\leqslant m do |\ if\ T[inmate,\ j] then |\ for\ (other inmate,\ j] then |\ S is not a subset of inmates who do don't speak the same language; end |\ end end |\ j++; end
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Clearly, the verification that S is a subset where no two inmates speak the same language can run in polynomial time $O(mn^2)$. Once this verification if complete all that is left to do is to verify that $|S| \ge k$, which is O(1). Therefore the mute prison problem is NP.

<u>2.</u> To show that the mute prison problem is NP-hard we must perform a reduction using an NP-complete problem. We will use a reduction on NP-complete 3-SAT in CNF, in order to show 3-SAT \leq_p Mute Prison Problem.

Properties of Reduction

Suppose that ϕ is an instance of 3-SAT and C_1 , C_2 , ..., C_m are the clauses of ϕ . By construction of 3-SAT in CNF we have $C_i = (z_{i1} \lor z_{i2} \lor z_{i3})$. In the reduction each C_i 's boolean value will represent a boolean value for each language, L_i , spoken by some inmate(s), precisely, $L_i = C_i = (z_{i1} \lor z_{i2} \lor z_{i3})$. Each boolean value for L_i has a specific mean:

$$L_i = \begin{cases} 1 & \text{if } L_i \text{ is spoken by at most 1 inmate} \\ 0 & \text{if } L_i \text{ is spoken by at least 1 inmate} \end{cases}$$

Producing $L_1, L_2, ..., L_m$ will take polynomial time since we iterate through each C_i and perform a boolean or operation on each z_i in C_i which takes O(m).

Finally, the mute prison problem requires a matrix T to produce the subset of inmates S. Let T be an m x m matrix, so that no inmates are left without a language. The rows in T will represent inmates and the columns will represent languages such that column i represents L_i . The algorithm that performs the reduction will

iterate through each L_i . If $L_i = 1$ then set T[i, i] = 1, else if $L_i = 0$ then T[1, i] = T[2, i] = ... = T[m, i] = 1. Assigning all inmates to speak L_i , when $L_i = 0$, will guarantee that |S| = 0. Alternatively, \forall i, if $L_i = 1$ then |S| = m. So that if ϕ is satisfies 3-SAT, then T will satisfy the mute prison problem if we set k = m. Again this process is polynomial as it iterates through m L_i 's and assigns at most m inmates the language L_i , so it will run $O(m^2)$.

ϕ of 3-SAT is satisfiable \to L and k of mute prison problem is satisfiable

Suppose ϕ of 3-SAT is satisfiable, then each clause C_1 , C_2 , ..., C_m is satisfied. A set of L_1 , ..., L_m is produced such that $\forall L_i, L_i = 1$. Then we form matrix T of size m x m, such that T resembles the identity matrix as each T[i,i] = 1. Also, k = m, so that when S is assembled all m inmates speak a different language, then $|S| \ge k$ is satisfied.

L and k of mute prison problem is satisfiable $\rightarrow \phi$ of 3-SAT is satisfiable

Suppose that T and k of the mute prison problem are satisfiable. Also, suppose |S| is at least m=k. Choose only the first m inmates from S, and extract only their rows from T to form a new matrix T'. It will follows that in T' there will be only m columns where there is at most one entry with the value 1. We will attribute these m columns with variables L_1 , ..., L_m , such that, $1 \le i \le m$, $L_i = 1$. We then form m clauses of a 3-SAT CNF, call them C_i , ..., C_m . Each C_i relates to L_i , so that the boolean value of $C_i = (z_{i1} \lor z_{i2} \lor z_{i3}) = 1$. Thus set any one of the z_{i1} , z_{i2} , or z_{i3} to 1. It follows that all $C_i = 1$, thus $\phi = (C_1 \land C_2 \land ... \land C_m)$ Is satisfiable.

So, ϕ of 3-SAT is satisfiable \Leftrightarrow L and k of mute prison problem is satisfiable . Also, because the reduction was shown to be polynomial it is proven that the mute prison problem is NP-hard.

By the proofs $\underline{1}$ and $\underline{2}$ it follows that the mute prison problem is NP-complete.

Q2. The Nonsense Prerequisites

Claim: The nonsense prerequisites problem is NP-complete.

Proof:

- 1. Show the nonsense prerequisites problem is NP.
- 2. Show the nonsense prerequisites problem is NP-hard.
- $\underline{1}$. Suppose we know G(V, E) and k and we are given E' as a certificate. We verify the certificate with the following algorithm:

If there is a cycle in G'(V, E'') then setting each edge in G' to a weight -1 will produce a negative edge cycle which, after relaxations, we can identify easily. Given that G(V, E'') may or may not be connected, to locate a cycle in the graph we must perform the relaxation with Bellman-Ford |V| times. Bellman-Ford runs at O(VE), it is executed |V| times in the verifier, thus we have $O(V^2E)$ for our algorithm. Since |V| = n, and $|E| = O(n^2)$, the verifier runs $O(n^4)$. So the verifier is polynomial and then the nonsense prerequisites problem is NP.

 $\underline{2}$. To show the nonsense prerequisites problem is NP-hard, as directed by the problem set, we will perform a reduction using NP-complete problem VECTOR COVER. So, we will show VECTOR COVER \leqslant_p The Nonsense Prerequisites Problem.

Properties of Reduction

Take the G(V, E) and k given to the VECTOR COVER problem. k represents $|S| \le k$, such that $S \subseteq V$ such that if $(u, v) \in E$, then $u \in S$ or $v \in S$. However, in the nonsense prerequisites, the k corresponds to edges that when removed from the graph will make it acyclic. It follows that the reduction must somehow convert the vertices in G to represent edges. This is done by splitting each vertex in two, so given $V = \{v_1, v_2, ..., v_n\}$, produce $V' = \{v_{pre-1}, v_{post-1}, v_{pre-2}, v_{post-2}, ..., v_{pre-n}, v_{post-n}\}$, and (v_{pre-i}, v_{post-i}) is a directed edge such that $(v_{pre-i}, v_{post-i}) \in E'$. Also, we must create a circumstance in the new graph where each undirected edge $(v_i, v_j) \in E$, becomes directed edges $(v_{post-i}, v_{pre-j}) \in E'$ and $(v_{post-j}, v_{pre-i}) \in E'$. This construction guarantees in G'(V', E') that when we enter any v_{pre-j} we can walk a path $v_{pre-i} \to v_{post-i} \to v_{pre-j} \to v_{post-j} \to v_{pre-j}$, and indeed this is a cycle. Thus, we have a cycles, such that if $(v_i, v_j) \in E$, then the cycle is limited to the new vertices $\{v_{pre-i}, v_{post-i}, v_{pre-j}, v_{post-j}\}$. So the reduction is complete and can easily be performed in polynomial time. $O(n\alpha(m+n))$ to produce new directed edges from m existing edges, splitting vertices in V and creating new edges, and adding them to the new graph G' using make-set, union, and link.

G(V, E), k of VECTOR COVER is satisfiable \rightarrow

G*(V*, E*), k of the nonsense prerequisite is satisfiable

Suppose using undirected G(V, E) and k, VECTOR COVER is satisfied. Suppose also that we have access to $S = \{s_1, ..., s_q\}$, which is the vertex cover of G and $|S| \le k$. We perform the reduction and have $G^*(V^*, E^*)$. It follows in G^* any cycles is limited to $\{v_{pre-i}, v_{post-i}, v_{pre-j}, v_{post-j}\}$. To break a cycle in G^* we could remove any edge from the cycle, but to do this efficiently we need to remove an edges that break many cycles at once. This is precisely $E' = \{(s_{pre-1}, s_{post-1}), ..., (s_{pre-q}, s_{post-q})\}$, because in G^* the edges in E' that correspond to vertices in S, are precisely the set of edges that appear in all cycles. Thus $|E'| = |S| \le k$, and so $G^*(V^*, E^*)$ and k of the nonsense prerequisite is satisfiable.

G(V, E), k of the nonsense prerequisite is satisfiable \rightarrow

$G^*(V^*, E^*)$, k of VECTOR COVER is satisfiable

Suppose G(V, E), k when used in the nonsense prerequisite problem is satisfiable. Now, conversively suppose the original graph, $G^*(V^*, E^*)$, before reduction, and k in VERTEX COVER were not satisfiable. This would mean that the set of vertex cover $S \subseteq V$, |S| textgreater k. But since G(V, E) and k were satisfiable then $|E'| \le k$. But by the construction of the reduction this in impossible. Since every $(v_{pre-i}, v_{post-i}) \in E'$ corresponds to a vertex $v_i \in S$, this will mean that there is some $v_i \in S$ that is not represented in E', since |E'| < |S|. This means that there is some cycle left over in G when E - E' is performed. So then a contradiction is reached, and so $G^*(V^*, E^*)$, k of VECTOR COVER must be satisfiable.

By proving both directions, it follows that G(V, E), k of VECTOR COVER is satisfiable \leftrightarrow $G^*(V^*, E^*)$, k of the nonsense prerequisite is satisfiable.

Additionally, since the reduction can be performed in polynomial time, then the nonsense prerequisite problem is NP-hard. \blacksquare

Q3. T-rex Christmas

 $\underline{1}$. $\{0, 1\}$ -integer programming formulation:

Minimize:

Ρ

Subject To:

a)
$$x_q^1 \in \{0, 1\} \qquad \forall \ q \ s.t. \ q \ represents \ a \ path \ i \to j, \ such \ that \ L[i, j] = 1$$

$$x_q^0 \in \{0, 1\}$$
 b)
$$x_q^1 + x_q^0 \geqslant 1$$
 c)

 $\sum_{\forall q} \, \mathbf{x}_q^{f(q,\ m)} \leqslant \mathbf{P} \qquad \forall \ \mathbf{m} \in \{0,\,1,\,...,\,\mathbf{n}\text{-}1\}$

Explanation of Constraints:

- a) x_q^1 and x_q^0 represent the direction (x_q^1 clockwise and x_q^0 counter-clockwise) a gift q will be passed from i to j, such that i, $j \in \{0, 1, ..., n-1\}$. If $x_q^1 = 1$ then the gift will be sent clockwise, if $x_q^1 = 0$ it will not be sent clockwise (same for x_q^0 but counter-clockwise).
- b) This constraint limits \mathbf{x}_q^1 and \mathbf{x}_q^0 so both cannot be 0, and it tightly bound when only one of the variables equals 1.
- c) Assume some linear function f, and also assume m represents a pass (m, m+1[n]). f(q, m) = 1 if on the clockwise direction of passing a gift from i to j a pass in (m, m+1[n]) occurs. Conversely, f(q, m) = 0 if on the counter-clockwise direction of passing a gift from i to j a pass in (m, m+1[n]) occurs. We can assume linearity of f as the computation of the intersection of gift direction and passes will have been computed previously and values will be contained in some 2-D matrix which f access at will. Finally, each summation for each m will represent, when the IP completes, each P_m . Clearly if the pass (m, m+1[n]) is used then then $x_q^{f(q, m)} = 1$, which will be added to P_m . Additionally, the constraint is limited to P, as a tight bound, P is some value that will equal the maximum of all P_m 's.

2. LP relaxation:

Minimize:

Ρ

Subject To:

a)
$$0\leqslant x_q^1\leqslant 1\qquad\forall\ q\ s.t.\ q\ represents\ a\ path\ i\to j,\ such\ that\ L[i,\,j]=1$$

$$0\leqslant x_q^0\leqslant 1$$
 b)
$$x_q^1+x_q^0\geqslant 1$$
 c)
$$\sum_{\forall q}x_q^{f(q,\ m)}\leqslant P\qquad\forall\ m\in\{0,\,1,\,...,\,n\text{-}1\}$$

Rounding Scheme:

Let all x_0^{1*} , x_0^{0*} , x_1^{1*} , x_1^{0*} , ..., x_s^{1*} , x_s^{0*} be the solution returned by the relaxed LP. Allow us to round all x_q^{1*} 's and x_q^{0*} 's to produce y_0^1 , y_0^0 , y_1^1 , y_1^0 , ..., y_s^1 , y_s^0 , and let this represent the solution to the integer programming such that:

$$y_q^1$$
 or $y_q^0 = \begin{cases} 1 & \text{if corresponding } \mathbf{x}_q^{1*} \text{ or } \mathbf{x}_q^{0*} \geqslant \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$

Proof that Rounding Scheme works:

All constraints in the interger program will be satisfied. For example a) is satisfied because y_q^1 or $y_q^0 \in \{0, 1\}$. Constraint b) is satisfied even in the unfortunate case where x_q^{1*} and x_q^{0*} equals 1/2 or greater, then y_q^1 and y_q^0 will equal 1. Additionally both y_q^1 and y_q^0 cannot both equal 0 by the constraint b). Finally, in constraint c), P is a bound specified by the algorithm, and indeed with the specified values for y_q^1 and y_q^0 the constraint will be satisfied. Therefore we conclude that the result of our rounding scheme will be feasible in the integer program. Proof of 2-approximation:

In both the relaxed linear program and rounding scheme function f will return the same results, such that the result of the relaxed linear program, since P is minimized, P will equal some P_m . Indeed suppose $P_m = \sum_{\forall q} x_q^{f(q, m)} *$. It follows, from our rounding scheme, that the rounding solution would be $P'_m = \sum_{\forall q} y_q^{f(q, m)}$. Then the proof is such that:

$$\begin{array}{l} \mathbf{x}_q^{1*}\geqslant\frac{1}{2}\rightarrow\mathbf{y}_q^1=1\\ \mathbf{x}_q^{0*}\geqslant\frac{1}{2}\rightarrow\mathbf{y}_q^0=1\\ \text{then,}\\ 2\mathbf{x}_q^{1*}\geqslant1\rightarrow2\mathbf{x}_q^{1*}\geqslant\mathbf{y}_q^1\\ 2\mathbf{x}_q^{0*}\geqslant1\rightarrow2\mathbf{x}_q^{0*}\geqslant\mathbf{y}_q^0\\ \text{finally,}\\ \mathbf{P'}_m=\sum_{\forall q}\mathbf{y}_q^{f(q,\ m)}\leqslant2^*\mathbf{P}_m=2^*\sum_{\forall q}\mathbf{x}_q^{f(q,\ m)}\\ \mathbf{P'}_m=\sum_{\forall q}\mathbf{y}_q^{f(q,\ m)}\leqslant2^*\mathbf{P}_m=2^*\mathbf{x}_q^{f(q,\ m)}\\ \mathbf{P'}_m=\sum_{\forall q}\mathbf{y}_q^{f(q,\ m)}\leqslant2^*\mathbf{P}\leqslant2^*\mathbf{LPOPT}\leqslant2^*\mathbf{IPOPT}\\ \text{Therefore we have that the rounding scheme attains a 2-approximation.} \end{array}$$

Q4. Vertex Cover