

CME305 Sample Midterm II

1. Matchings and Vertex Covers

- (a) Define what a matching in G is.
- (b) Define what a vertex cover of G is.
- (c) Let M be a maximum matching and C a minimum vertex cover.
Show that $|M| \leq |C| \leq 2|M|$.

Solution:

1. A matching M is a subset of E such that no two edges share an endpoint.
2. A vertex cover C is a subset of V such that all edges are incident to at least one element of C .
3. Let M be a maximum matching and C a minimum vertex cover. It is easy to see that we need at least $|M|$ nodes to cover all the edges of M . Hence $|M| \leq |C|$. Now let us prove that $V(M)$, the vertices in the matching M , is a vertex cover. Assume it isn't, then there is at least one edge e that is not covered by $V(M)$. It is easy to see that this implies that $M \cup \{e\}$ is a matching so M is not maximum. Contradiction. Furthermore, we know that $|V(M)| = 2|M|$, hence $|C| \leq 2|M|$.

2. Traveling Salesman Problem

Assume that deciding whether a graph has a Hamiltonian cycle is NP-Complete. Prove that the Traveling Salesman Problem is NP-Hard.

Solution: As defined in class the TSP problem defines a complete graph K_n with a cost function $c : E \rightarrow \mathfrak{R}^+$ and asks to find a cycle that visits all vertices exactly once and such that the cost of the cycle is minimized.

We are now going to show how to solve Hamiltonian Cycle on G by using TSP and a polynomial amount of work. For this, we will use the decision version of TSP, which is “is there a TSP of cost at most k ?”.

We reduce Hamiltonian cycle HC to TSP i.e. show $HC \leq_P TSP$. Given $G(V, E)$ with $|V| = n$ we define $c(e) = 1$ for all $e \in E$. Then we add edges E' to G to make G a complete graph and assign $c(e) = 2$ for all $e \in E'$. We can do this in polynomial time. Now given this cost, if the answer to is there a TSP cycle of cost at most n is “yes”, then we know there is a cycle that visits all nodes exactly once. The edges it uses are from E (given the cost function we created), hence we can say for sure that there is a Hamiltonian Cycle in G . Given that Hamiltonian Cycle is NP-Complete, the decision version of TSP is NP-Complete, hence TSP is NP-Hard.

3. Lecture Attendance Planning

A group of students want to minimize their lecture attendance by sending only one of the group to each of the n lectures. We have the following constraints:

- Each of the n lectures should be covered.
- Lecture i starts at time a_i and ends at time b_i .
- It takes r_{ij} time to commute from lecture i to lecture j .
- Assume all times r_{ij} as well as the duration of the lectures are in minutes and integers.

Minimize the number of students that will attend lectures i.e. develop a flow based algorithm to identify the minimum number of students needed to cover all n lectures.

Solution: We are going to solve this problem using a maximum matching on a bipartite graph. First observe that minimizing the number of students attending courses is equivalent to maximizing the number of classes that a given student can attend. If classes i and j are such that one can go to class i , commute to class j and still be on-time, then we only need one student to go to both classes. Let us now build the graph:

For each lecture i , set two nodes x_i and y_i . $X = \{x_1, \dots, x_n\}$ and $Y = \{y_1, \dots, y_n\}$ are our partitions. The edge (x_i, y_j) exists if $i \neq j$ and one can go to class i and then to class j . More formally $i \neq j$ and $a_j \geq b_i + r_{ij}$.

Building such graph takes at most $2n \times 3n^2$ steps, polynomial in n .

Let M be a maximum matching in $G(X, Y, E)$. We claim that the minimum number of students needed is $n - |M|$. We can prove this by contradiction.

Assume you can go to all the lectures with $n - p < n - |M|$ students. Then that means that we can “reuse” p students. Let $I = \{i_1, \dots, i_p\}$ be the set of lectures where we are reusing a student (i.e. the set of lectures that at the end the student will go to another lecture). Let $J = \{j_1, \dots, j_p\}$ be the set of lectures they are attending afterwards. It is easy to see that $M' = \{(x_{i_1}, y_{j_1}), \dots, (x_{i_p}, y_{j_p})\}$ is a matching in G . But $p > |M|$, which is a contradiction.

We can find such matching using a flow algorithm, just like we saw in class.