Design and Analysis of Algorithms (CS345A)

Practice-sheet: Fobinacci Heap and NP-completeness

Fibonacci Heaps

1. Tinkering with Fibonacci heap

In the Fibonacci heap discussed in the class, as soon as a marked node v loses its second child, the subtree rooted at v is cut from its parent and added to the root list. What if the subtree rooted at a marked node v is cut from its parent and added to the root list only when it loses its 3rd child? Will all the bounds still hold on the amortized time complexity of various operations on Fibonacci heap? Give rigorous mathematical arguments to support your claim.

2. Core property of Fibonacci heap

This problem is directly from the lectures. Let v be any node in Fibonacci heap. Show that the degree of v is $O(\log(s(v))$, where s(v) is the size of the subtree rooted at v.

3. A short and clean code for Decrease-key in Fibonacci Heap

Write a neat pseudo code for the Decrease-key(H, x) in a Fibonacci Heap?

4. Delete-key in a Fibonacci heap

Design an efficient algorithm for deleting an element from a Fibonacci Heap. The amortized cost must be $O(\log n)$.

5. A surprising property for Fibonacci Heap

Let v be any node in a Fibonacci heap. We showed that if the size of the subtree rooted at v is m, then the degree of v is $O(\log m)$. Can we say the same thing about the height as well? That is, will the height of v be bounded by $O(\log m)$? Note that all operations, including merging of Fibonacci heaps is allowed.

Hint: There exists a sequence of operations that may result in a Fibonacci heap which will be a single tree that is just a vertical chain of m elements. Invent one such sequence.

1 NP-completeness

1. Polynomial reduction \leq_P

Let A and B be any two computational problems. Let χ be any algorithm for solving B. Problem A is said to be reducible to problem B in polynomial time if each instance I of A can be solved by

- A polynomial number of executions of χ on instances (of B) each of which are also polynomial of size of I,
- and, if required, basic computational steps (each taking O(1) time) which are also polynomial in the size of I.

Convince yourself that this definition of \leq_P subsumes the definition of polynomial time reducibility discussed in the class.

2. Feedback set

Given an undirected graph G = (V, E), a feedback set is a set $X \subseteq V$ with the property that G - X has no cycle. The *Undirected Feedback Set Problem* asks: Given G and k, does there exist a feedback set of size at most k? Prove that *Undirected Feedback Set Problem* is NP-complete.

Hint: Reduce vertex cover problem to Feedback set problem. The reduction will be similar or same as used in some example discussed in the class.

3. Subgraph Isomorphism

Let G = (V, E) and G' = (V', E') be two graphs. G is said to be isomorphic to G' if we can obtain G' from G by renaming its vertices suitably. In formal words, it means the following.

A 1-1 and onto function $f: V \to V'$ is said to be an isomorphism if for each pair of vertices $u, v \in V$, $(u, v) \in E$ if and only if $(f(u), f(v)) \in E'$.

Subgraph-Isomorphism Problem is defined as follows. Given any two graphs G = (V, E) and G' = (V', E'), does there exist any subgraph of G which is isomorphic to G'. Show that Subgraph-Isomorphism Problem is NP-complete.

Hint: Reduce independent set problem or Hamiltonian cycle problem to subgraph isomorphism problem.

4. Clique Problem

A clique is a complete graph (edge exists between each pair of its vertices). Consider the following problem: Given an undirected graph G = (V, E) and an integer k, does G contain a clique of size k?

Show that this problem is NP-complete.

Hint: Use the fact that *Independent Set* is NP-complete.

5. Approximation Algorithm for Vertex Cover

In the course, we discussed bipartite-matching problem. The notion of matching can be extended naturally to any arbitrary undirected graph.

Now consider the following algorithm for computing vertex cover for a given graph G = (V, E):

- (a) $S \leftarrow \emptyset$;
- (b) Compute maximum matching \mathcal{M} of G;
- (c) For each edge $(u, v) \in \mathcal{M}$ do : $S \leftarrow S \cup \{u, v\}$
- (d) return S.

Show that the above algorithm computes a vertex cover. Prove that the size of vertex cover returned by the algorithm is at most twice the size of the optimal (minimum size) vertex cover.

Important Note: We discussed the area of approximation algorithm very briefly. So only simple exercises on approximation algorithms and NP-completeness, if at all, may be expected in the exam.