Theory Problems

Course: Coursera Algorithms Specialization

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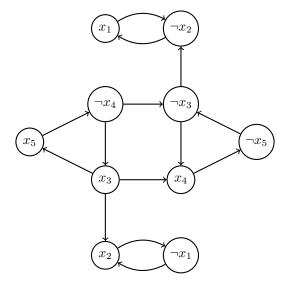
Problem 1

In the 2SAT problem, you are given a set of clauses, where each clause is the disjunction of two literals (a literal is a Boolean variable or the negation of a Boolean variable). You are looking for a way to assign a value "true" or "false" to each of the variables so that all clauses are satisfied — that is, there is at least one true literal in each clause. For this problem, design an algorithm that determines whether or not a given 2SAT instance has a satisfying assignment. (Your algorithm does not need to exhibit a satisfying assignment, just decide whether or not one exists.) Your algorithm should run in O(m+n) time, where m and n are the number of clauses and variables, respectively. [Hint: strongly connected components.]

Each clause, a disjunction of 2 literals, can be represented as 2 implications. These implications can be used to create an implication graph. If the implication graph has any cycles containing a variable and its negation, a contradiction exists and therefore no satisfying assignment exists. Kosaraju's algorithm can be used to find the strongly connected components of a graph. It uses two depth-first searches, so it has O(m+n) complexity. If a variable and its negation are present in the same strongly connected component, no satisfying assignment exists.

Example: Given $(x_1 \lor x_2) \land (\neg x_2 \lor x_3) \land (x_3 \lor x_4) \land (\neg x_3 \lor x_5) \land (\neg x_4 \lor \neg x_5) \land (\neg x_3 \lor x_4) \land (\neg x_1 \lor \neg x_2)$

$$\begin{aligned} &(x_1 \vee x_2) \leftrightarrow (\neg x_1 \to x_2) \wedge (\neg x_2 \to x_1) \\ &(\neg x_2 \vee x_3) \leftrightarrow (x_2 \to x_3) \wedge (\neg x_3 \to \neg x_2) \\ &(x_3 \vee x_4) \leftrightarrow (\neg x_3 \to x_4) \wedge (\neg x_4 \to x_3) \\ &(\neg x_3 \vee x_5) \leftrightarrow (x_3 \to x_5) \wedge (\neg x_5 \to \neg x_3) \\ &(\neg x_4 \vee \neg x_5) \leftrightarrow (x_4 \to \neg x_5) \wedge (x_5 \to \neg x_4) \\ &(\neg x_3 \vee x_4) \leftrightarrow (x_3 \to x_4) \wedge (\neg x_4 \to \neg x_3) \\ &(\neg x_1 \vee \neg x_2) \leftrightarrow (x_1 \to \neg x_2) \wedge (x_2 \to \neg x_1) \end{aligned}$$



A satisfying assignment exists.

Problem 2

In lecture we define the length of a path to be the sum of the lengths of its edges. Define the bottleneck of a path to be the maximum length of one of its edges. A mininum-bottleneck path between two vertices s and t is a path with bottleneck no larger than that of any other s-t path. Show how to modify Dijkstra's algorithm to compute a minimum-bottleneck path between two given vertices. The running time should be $O(m \log n)$, as in lecture.

Dijkstra's algorithm can be modified so that the value stored in the priority queue (minimum heap) is the bottleneck instead of the total path length. Dijkstra's running time is $O(m \log n)$.

Problem 3

We can do better. Suppose now that the graph is undirected. Give a linear-time O(m) algorithm to compute a minimum-bottleneck path between two given vertices.

Camerini's algorithm finds the minimum bottleneck spanning tree in O(m). The minimum bottleneck path between any two vertices will follow the minimum bottleneck spanning tree between those two vertices. See Wikipedia article on Minimum Bottleneck Spanning Trees.

Problem 4

What if the graph is directed? Can you compute a minimum-bottleneck path between two given vertices faster than $O(m \log n)$?

Camerini's algorithm for directed graphs finds the minimum bottleneck spanning arborescence, a directed graph from source vertex to each of the other connected vertices, in $O(m \log m)$. Gabow and Tarjan's algorithm for minimum bottleneck spanning arborescence uses the modified version of Dijkstra's algorithm mentioned in Problem 2 and implements the priority queue using a Fibonacci heap. It runs in $O(m+n \log n)$. See Wikipedia article on Minimum Bottleneck Spanning Trees.

Problem 5

Recall that a set H of hash functions (mapping the elements of a universe U to the buckets $0,1,2,\ldots,n-1$) is universal if for every distinct $x,y\in U$, the probability Prob[h(x)=h(y)] that x and y collide, assuming that the hash function h is chosen uniformly at random from H, is at most $\frac{1}{n}$. In this problem you will prove that a collision probability of $\frac{1}{n}$ is essentially the best possible. Precisely, suppose that H is a family of hash functions mapping U to $0,1,2,\ldots,n-1$, as above. Show that there must be a pair $x,y\in U$ of distinct elements such that, if h is chosen uniformly at random from H, then $Prob[h(x)=h(y)]\geq \frac{1}{n}-\frac{1}{|U|}$

Answer...