1. Let *G* be a directed acyclic graph where each node has a label from some finite alphabet; different nodes may have the same label. Any directed path in *G* has a *signature*, which is the string defined by concatenating the labels of its vertices. A *subsequence* of *G* is a subsequence of the signature of some directed path in *G*. For example, the strings DYNAMIC, PROGRAM, and DETPHFIRST are all subsequences of the dag shown below; in fact, PROGRAM is the signature of a path.



Describe and analyze an algorithm to compute the length of the longest common subsequence of two given directed acyclic graphs, that is, the longest string that is a subsequence of both dags.

- 2. Suppose you are given a graph G with weighted edges and a minimum spanning tree T of G.
  - (a) Describe an algorithm to update the minimum spanning tree when the weight of a single edge e is decreased.
  - (b) Describe an algorithm to update the minimum spanning tree when the weight of a single edge *e* is increased.

In both cases, the input to your algorithm is the edge *e* and its new weight; your algorithms should modify *T* so that it is still a minimum spanning tree. [Hint: Consider the cases  $e \in T$  and  $e \notin T$  separately.]

3. When there is more than one shortest paths from one node *s* to another node *t*, it is often most convenient to choose the shortest path with the fewest edges; call this the **best path** from *s* to *t*. For instance, if nodes represent cities and edge lengths represent costs of flying between cities, there could be many ways to fly from city *s* to city *t* for the same cost; the most desirable these schedules is the one with the fewest flights.

Suppose we are given a directed graph *G* with positive edge weights and a source vertex *s* in *G*. Describe and analyze an algorithm to compute *best* paths in *G* from *s* to every other vertex. [*Hint:* What is the actual output of your algorithm? If possible, use one of the standard shortest-path algorithms as a black box.]