1) RBT-Sorting.

The input is a sequence of $n$ integers with many duplications, such that the number of distinct integers in the sequence is $O(\log n)$. Design a sorting algorithm (based on comparisons only) to sort such sequences using at most $O(n \log \log n)$ comparisons in the worst case (justify the running time).

2) RBT+Means. (Duke final spring 2002)

The mean $M$ of a set of $k$ integers $\{x_1, x_2, \ldots, x_k\}$ is defined as

$$M = \frac{\sum_{i=1}^{k} x_i}{k}.$$

We want to maintain a data structure $\mathcal{D}$ on a set of integers under the normal $\text{Init}$, $\text{Insert}$, $\text{Delete}$, $\text{Find}$ operations, as well as a $\text{Mean}$ operation, defined as follows:

- $\text{Init}(\mathcal{D})$: Create an empty structure $\mathcal{D}$.
- $\text{Insert}(\mathcal{D}, x)$: Insert $x$ in $\mathcal{D}$.
- $\text{Delete}(\mathcal{D}, x)$: Delete $x$ from $\mathcal{D}$.
- $\text{Find}(\mathcal{D}, x)$: Return pointer to $x$ in $\mathcal{D}$.
- $\text{Mean}(\mathcal{D}, a, b)$: Return the mean of the set $x \in \mathcal{D}$ with $a \leq x \leq b$.

(a) What does $\text{Mean}(\mathcal{D}, 7, 17)$ return if $\mathcal{D}$ contains integers

$$(2, 3, 5, 7, 11, 13, 17, 19, 23, 29, 31, 27)?$$

(b) Describe how to modify a standard red-black tree in order to implement $\mathcal{D}$ such that $\text{Init}$ is supported in $O(1)$ time and $\text{Insert}$, $\text{Delete}$, $\text{Find}$, and $\text{Mean}$ are supported in $O(\log n)$ time.

3) Interval Partitioning.

The input is a sequence of $n$ pairs of start and finish times $(s_j, f_j)$ already sorted in increasing order of $s_j$. Design an algorithm to schedule the corresponding intervals using a minimum number $k$ of lecture-rooms in at most $O(n \log k)$ time in the worst case (justify the running time).
4) Completion times. (4.13)

13. A small business—say, a photocopying service with a single large machine—faces the following scheduling problem. Each morning they get a set of jobs from customers. They want to do the jobs on their single machine in an order that keeps their customers happiest. Customer $i$'s job will take $t_i$ time to complete. Given a schedule (i.e., an ordering of the jobs), let $C_i$ denote the finishing time of job $i$. For example, if job $j$ is the first to be done, we would have $C_j = t_j$; and if job $j$ is done right after job $i$, we would have $C_j = C_i + t_i$. Each customer $i$ also has a given weight $w_i$ that represents his or her importance to the business. The happiness of customer $i$ is expected to be dependent on the finishing time of $i$'s job. So the company decides that they want to order the jobs to minimize the weighted sum of the completion times, $\sum_{i=1}^{n} w_i C_i$.

Design an efficient algorithm to solve this problem. That is, you are given a set of $n$ jobs with a processing time $t_i$ and a weight $w_i$ for each job. You want to order the jobs so as to minimize the weighted sum of the completion times, $\sum_{i=1}^{n} w_i C_i$.

**Example.** Suppose there are two jobs: the first takes time $t_1 = 1$ and has weight $w_1 = 10$, while the second job takes time $t_2 = 3$ and has weight $w_2 = 2$. Then doing job 1 first would yield a weighted completion time of $10 \cdot 1 + 2 \cdot 4 = 18$, while doing the second job first would yield the larger weighted completion time of $10 \cdot 4 + 2 \cdot 3 = 46$.

5) ClubNet. (4.28)

28. Suppose you're a consultant for the networking company CluNet, and they have the following problem. The network that they're currently working on is modeled by a connected graph $G = (V, E)$ with $n$ nodes. Each edge $e$ is a fiber-optic cable that is owned by one of two companies—creatively named $X$ and $Y$—and leased to CluNet.

Their plan is to choose a spanning tree $T$ of $G$ and upgrade the links corresponding to the edges of $T$. Their business relations people have already concluded an agreement with companies $X$ and $Y$ stipulating a number $k$ so that in the tree $T$ that is chosen, $k$ of the edges will be owned by $X$ and $n - k - 1$ of the edges will be owned by $Y$.

CluNet management now faces the following problem. It is not at all clear to them whether there even exists a spanning tree $T$ meeting these conditions, or how to find one if it exists. So this is the problem they put to you: Give a polynomial-time algorithm that takes $G$, with each edge labeled $X$ or $Y$, and either (i) returns a spanning tree with exactly $k$ edges labeled $X$, or (ii) reports correctly that no such tree exists.
6) Kruskal’s variant. (4.31)

31. Let’s go back to the original motivation for the Minimum Spanning Tree Problem. We are given a connected, undirected graph \( G = (V, E) \) with positive edge lengths \( \{\ell_e\} \), and we want to find a spanning subgraph of it. Now suppose we are willing to settle for a subgraph \( H = (V, F) \) that is “denser” than a tree, and we are interested in guaranteeing that, for each pair of vertices \( u, v \in V \), the length of the shortest \( u-v \) path in \( H \) is not much longer than the length of the shortest \( u-v \) path in \( G \). By the length of a path \( P \) here, we mean the sum of \( \ell_e \) over all edges \( e \) in \( P \).

Here’s a variant of Kruskal’s Algorithm designed to produce such a subgraph.

- First we sort all the edges in order of increasing length. (You may assume all edge lengths are distinct.)
- We then construct a subgraph \( H = (V, F) \) by considering each edge in order.
- When we come to edge \( e = (u, v) \), we add \( e \) to the subgraph \( H \) if there is currently no \( u-v \) path in \( H \). (This is what Kruskal’s Algorithm would do as well.) On the other hand, if there is a \( u-v \) path in \( H \), we let \( d_{uv} \) denote the length of the shortest such path; again, length is with respect to the values \( \{\ell_e\} \). We add \( e \) to \( H \) if \( 3\ell_e < d_{uv} \).

In other words, we add an edge even when \( u \) and \( v \) are already in the same connected component, provided that the addition of the edge reduces their shortest-path distance by a sufficient amount.

Let \( H = (V, F) \) be the subgraph of \( G \) returned by the algorithm.

(a) Prove that for every pair of nodes \( u, v \in V \), the length of the shortest \( u-v \) path in \( H \) is at most three times the length of the shortest \( u-v \) path in \( G \).

(b) Despite its ability to approximately preserve shortest-path distances, the subgraph \( H \) produced by the algorithm cannot be too dense. Let \( f(n) \) denote the maximum number of edges that can possibly be produced as the output of this algorithm, over all \( n \)-node input graphs with edge lengths. Prove that

\[
\lim_{{n \to \infty}} \frac{f(n)}{n^2} = 0.
\]
7) Mobile wireless devices. (6.14)

14. A large collection of mobile wireless devices can naturally form a network in which the devices are the nodes, and two devices \( x \) and \( y \) are connected by an edge if they are able to directly communicate with each other (e.g., by a short-range radio link). Such a network of wireless devices is a highly dynamic object, in which edges can appear and disappear over time as the devices move around. For instance, an edge \((x, y)\) might disappear as \( x \) and \( y \) move far apart from each other and lose the ability to communicate directly.

In a network that changes over time, it is natural to look for efficient ways of maintaining a path between certain designated nodes. There are two opposing concerns in maintaining such a path: we want paths that are short, but we also do not want to have to change the path frequently as the network structure changes. (That is, we'd like a single path to continue working, if possible, even as the network gains and loses edges.) Here is a way we might model this problem.

Suppose we have a set of mobile nodes \( V \), and at a particular point in time there is a set \( E_0 \) of edges among these nodes. As the nodes move, the set of edges changes from \( E_0 \) to \( E_1 \), then to \( E_2 \), then to \( E_3 \), and so on, to an edge set \( E_b \). For \( i = 0, 1, 2, \ldots, b \), let \( G_i \) denote the graph \((V, E_i)\). So if we were to watch the structure of the network on the nodes \( V \) as a "time lapse," it would look precisely like the sequence of graphs \( G_0, G_1, G_2, \ldots, G_{b-1}, G_b \). We will assume that each of these graphs \( G_i \) is connected.

Now consider two particular nodes \( s, t \in V \). For an \( s-t \) path \( P \) in one of the graphs \( G_i \), we define the length of \( P \) to be simply the number of edges in \( P \), and we denote this \( \ell(P) \). Our goal is to produce a sequence of paths \( P_0, P_1, \ldots, P_b \) so that for each \( i \), \( P_i \) is an \( s-t \) path in \( G_i \). We want the paths to be relatively short. We also do not want there to be too many changes—points at which the identity of the path switches. Formally, we define \( \text{changes}(P_0, P_1, \ldots, P_b) \) to be the number of indices \( i \) (\( 0 \leq i \leq b - 1 \)) for which \( P_i \neq P_{i+1} \).

Fix a constant \( K > 0 \). We define the cost of the sequence of paths \( P_0, P_1, \ldots, P_b \) to be

\[
\text{cost}(P_0, P_1, \ldots, P_b) = \sum_{i=0}^{b} \ell(P_i) + K \cdot \text{changes}(P_0, P_1, \ldots, P_b).
\]

(a) Suppose it is possible to choose a single path \( P \) that is an \( s-t \) path in each of the graphs \( G_0, G_1, \ldots, G_b \). Give a polynomial-time algorithm to find the shortest such path.

(b) Give a polynomial-time algorithm to find a sequence of paths \( P_0, P_1, \ldots, P_b \) of minimum cost, where \( P_i \) is an \( s-t \) path in \( G_i \) for \( i = 0, 1, \ldots, b \).