

Homework 1 (due March 3)

Teams of up to three people can turn in joint homework solutions. Complete solutions for each numbered problem (excluding starred subproblems) should take at most 3 typeset pages. You may use any resource at your disposal—printed, electronic, or human—but cite/acknowledge your sources, just as though you were writing a research paper. *Stars indicate problems that I don't know how to solve; these problems may or may not actually be open (or interesting).

1. In class we described an extension of Sleator and Tarjan's link-cut trees that support *path operations* via lazy propagation. Specifically, the data structure maintains a *value* at each node in a dynamic forest, and supports the following operations in $O(\log n)$ amortized time, in addition to the core operations LINK, CUT, and FINDROOT:
 - GETVALUE(v): Return the value associated with v .
 - SETVALUE(v, x): Change the value associated with v to x .
 - MINPATH(v): Return the *ancestor* of v with minimum value.
 - ADDPATH(v, Δ): Add Δ to the value of every *ancestor* of v .

Extend link-cut trees further, to support the following *subtree operations*:

- MINSUBTREE(v): Return the *descendant* of v with minimum value.
- ADDSUBTREE(v, Δ): Add Δ to the value of every *descendant* of v .

The algorithms for LINK, CUT, FINDROOT, GETVALUE, SETVALUE, MINPATH, and ADDPATH should require only *minor* modifications; all operations should run in $O(\log n)$ amortized time. You may assume that every node in the *represented* forest has degree at most 3. (This assumption is not actually necessary, but it dramatically simplifies the solution.)

2. In the first week of class, we saw a method to transform a data structure that supports queries and insertions into a data structure that supports queries, insertions, and deletions, when the binary operation \diamond used to compose query results has an inverse $\bar{\diamond}$. The basic idea is to maintain both a main data structure D and an auxiliary *anti-data* structure \bar{D} . To delete an item from the modified data structure, we actually insert it into \bar{D} ; then, if the size of \bar{D} exceeds half the size of D , we rebuild the entire data structure from scratch, discarding the items in \bar{D} .

However, as one student observed in class, this modified structure has issues with repeated insertions and deletions of the same item. After inserting and deleting an item, we can't insert that item again until after the next global rebuild! The standard dodge is to assume either that repeated insertions just never happen, or that the underlying data structure actually stores a *multiset* and therefore directly supports multiple insertions.

But suppose the original data structure really maintains a *set*? Specifically, let's assume the data structure maintains a set A and supports the following operations:

- **INSERT**(a) — Insert item a into the data set A . If $a \in A$ before the operation, the operation has *no effect*. This operation takes $I(n)$ time. Assume that $I(n) = \Omega(\log n)$.
- **MEMBER?**(a) — Return **TRUE** if and only if $a \in A$. This operation takes less than $I(n)$ time.
- **QUERY**(x) — Answer the decomposable query $Q(x, A)$, in $Q(n)$ time.

Let $P(n)$ denote the preprocessing time to build a new data structure containing n items.

To add support for deletions, we can maintain an exponential sequence $D_0, D_1, D_2, D_3, \dots$ of alternating data and anti-data structures, where each structure D_i stores a subset of at most half the data in the preceding structure D_{i-1} . Thus, the actual data set stored in the exponential structure is

$$A := D_0 \setminus (D_1 \setminus (D_2 \setminus (D_3 \setminus \dots))) = (D_0 \setminus D_1) \cup (D_2 \setminus D_3) \cup \dots,$$

and therefore a query can be answered using the alternating 'sum'

$$\begin{aligned} Q(x, A) &= Q(x, D_0) \bar{\diamond} (Q(x, D_1) \bar{\diamond} (Q(x, D_2) \bar{\diamond} (Q(x, D_3) \bar{\diamond} \dots))) \\ &= (Q(x, D_0) \bar{\diamond} Q(x, D_1)) \diamond (Q(x, D_2) \bar{\diamond} Q(x, D_3)) \diamond \dots \end{aligned}$$

- What is the query time for this alternating exponential structure?
- Describe **INSERT** and **DELETE** algorithms for this alternating exponential structure that run in $O(P(n)/n + I(n))$ amortized time. **INSERTING** an item that is already in A should have no effect; similarly, **DELETING** an item that is not in A should have no effect. [Hint: The **INSERT** and **DELETE** algorithms should be essentially identical. $O(P(n)/n + I(n) \log n$ amortized time is relatively straightforward, and even $O(P(n)/n + I(n) \log \log n$ is not hard, but how to get rid of that last $\log \log n$ factor?]
- Sketch** the modifications necessary to make **INSERT** and **DELETE** run in $O(P(n)/n + I(n))$ **worst-case** time.

3. After the Great Academic Meltdown of 2020, you get a job as a cook's assistant at Jumpin' Jack's Flapjack Stack Shack, which sells arbitrarily-large stacks of pancakes for just four bits (50 cents) each. Jumpin' Jack insists that any stack of pancakes given to one of his customers must be sorted, with smaller pancakes on top of larger pancakes. Also, whenever a pancake goes to a customer, its top side should not be burned.

The cook provides you with a unsorted stack of n perfectly round pancakes, of n different sizes, possibly burned on one or both sides. Your task is to throw out the pancakes that are burned on both sides (and *only* those) and sort the remaining pancakes so that their burned sides (if any) face down. Your only tool is a spatula. You can insert the spatula under any pancake and then either *flip* or *discard* the stack of pancakes above the spatula.

More concretely, let us represent a stack of pancakes by a sequence of distinct integers between 1 and n , representing the sizes of the pancakes, with each number marked to indicate the burned side(s) of the corresponding pancake. For example, $\underline{1}\bar{4}3\bar{2}$ represents a stack of four pancakes: a one-inch pancake burned on the bottom; a four-inch pancake burned on the top; an unburned three-inch pancake, and a two-inch pancake burned on both sides. We store this sequence in a data structure that supports the following operations:

- $\text{POSITION}(x)$: Return the position of integer x in the current sequence, or 0 if x is not in the sequence.
 - $\text{VALUE}(k)$: Return the k th integer in the current sequence, or 0 if the sequence has no k th element. VALUE is essentially the inverse of POSITION .
 - $\text{TOPBURNED}(k)$: Return TRUE if and only if the top side of the k th pancake in the current sequence is burned.
 - $\text{FLIP}(k)$: Reverse the order and the burn marks of the first k elements of the sequence.
 - $\text{DISCARD}(k)$: Discard the first k elements of the sequence.
- (a) Describe an algorithm to filter and sort any stack of n burned pancakes using $O(n)$ of the operations listed above. Try to make the big-Oh constant small.

$$\underline{1}\bar{4}3\bar{2} \xrightarrow{\text{FLIP}(4)} \bar{2}3\bar{4}\bar{1} \xrightarrow{\text{DISCARD}(1)} 3\bar{4}\bar{1} \xrightarrow{\text{FLIP}(2)} \bar{4}3\bar{1} \xrightarrow{\text{FLIP}(3)} \underline{1}3\bar{4}$$

- (b) Describe a data structure that supports each of the operations listed above in $O(\log n)$ amortized time. Together with part (a), such a data structure gives us an algorithm to filter and sort any stack of n burned pancakes in $O(n \log n)$ time.
- * (c) Prove or disprove: Sorting *unburned* pancakes requires at least $\Omega(\log n)$ amortized time per operation. More concretely: Any data structure that stores a permutation of the set $\{1, 2, \dots, n\}$ and supports the operations POSITION , VALUE , and FLIP requires at least $\Omega(\log n)$ amortized time per operation in the cell-probe model. [We'll talk about cell-probe lower bounds later in the semester. Meanwhile, the following paper may be relevant: Erik Demaine and Mihai Pătraşcu, *Lower Bounds for Dynamic Connectivity*, *STOC 2004*.]
- ★ (d) Describe an integer-RAM algorithm that computes, in $o(n \log n)$ time, a sequence of $O(n)$ FLIP operations that sorts a given stack of n unburned pancakes. Alternatively, prove that no such algorithm exists. [Hint: There are sorting algorithms that run in $o(n \log n)$ time on the integer RAM; that's not the hard part.]

- ★4. Link-cut trees (like most other dynamic tree data structures) are built on top of splay trees, but their analysis does not reflect the rich set of results known for splay trees. Can we prove stronger results about the efficiency of link-cut trees, or any other dynamic tree data structure, similar to the static optimality theorem, working set theorem, sequential access theorem, dynamic finger theorem, and the like for splay trees? Here are some concrete examples:

Suppose we perform an depth-first search of some rooted tree T , calling $\text{EXPOSE}(v)$ each time we visit any node v . For link-cut trees, the analysis in class implies a total running time of at most $O(n \log n)$. However, I believe the sequential access theorem implies that if T is a *path*, the actual running time is only $O(n)$. Is the total running time always $O(n)$, or does it depend on the tree T ?

Fix a tree T with n nodes and a sequence X of m EXPOSE operations. For each node v , let m_v denote the number of times that $\text{EXPOSE}(v)$ appears in X . For link-cut trees, the analysis in class implies a total running time of $O(m \log n)$, but is it really $O(\sum_v m_v \log(m/m_v))$, or something else? Does the actual running time also depend on T , or the given permutation of X ?

Are there interesting lower bounds for dynamic trees (or perhaps I should say self-adjusting representations of static trees), like Wilber's lower bounds and their geometric descendants for self-adjusting binary search trees?

These questions are probably very hard, so any partial results are potentially interesting.