

Data Structures 1

NTIN066

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- Run a program available on gitlab to obtain results of experiments.
- Determine the asymptotic behavior of the measured data. Note that certain results may depend on multiple parameters; therefore, the dependence on all relevant parameters should be analyzed.
- State the mathematical theorems that theoretically justify the observed empirical behavior. Cite all your sources.
- Discuss the correspondence between theory and experiment, explaining how the theoretical results account for, or deviate from, the empirical measurements.
- It may be helpful to apply regression to substantiate the asymptotic estimates. It is advisable to compare the outcomes of regression across different candidate functions. Any suitable mathematical software may be used to perform the regression analysis.
- Submit your report in PDF format.

Problem Setting

- We are given a set of keys $x_1 < x_2 < \dots < x_n$.
- Each key x_i is accessed with frequency $w_i > 0$.
- Goal: construct a binary search tree T minimizing the total search cost

$$\sum_{i=1}^n w_i \cdot \text{depth}_T(x_i),$$

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Observation

If T is SOBST, then the subtree of every node in T is SOBST.

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Recurrence

If r is chosen as the root ($i \leq r \leq j$), then

$$C(i, j) = \min_{r=i, \dots, j} (C(i, r-1) + C(r+1, j)) + \sum_{k=i}^j w_k.$$

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Complexity

- There are $\mathcal{O}(n^2)$ subproblems.
- Each requires $\mathcal{O}(n)$ choices of r .
- Total running time: $\mathcal{O}(n^3)$.
- With prefix sums and optimization: $\mathcal{O}(n^2)$.

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- Where are all elements in the interval $[a, b]$ stored?
- What is the time complexity of this interval query?

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- What is the time complexity of this operation?
- How can this additional information be updated during the operations `SPLAY`, `INSERT`, and `DELETE`?

Procedure

- 1 Insert/delete the given node as in a BST
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Amortized complexity

- Finding the required nodes has the same complexity as SPLAY: $\mathcal{O}(\log n)$ amortized
- The actual deletion of a node takes $\mathcal{O}(1)$ time and the potential of the tree decreases
- When adding a leaf, the potential of the nodes u_1, \dots, u_h on the path to the root increases by $\sum_{i=1}^h \log(s(u_i) + 1) - \log(s(u_i)) \leq \log(n)$
- The amortized complexity of the operations INSERT and DELETE is $\mathcal{O}(\log n)$

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Theorem (working set)

Let x_1, \dots, x_k be a sequence of searched elements, and let z_i denote the number of distinct elements searched for since the previous search of the element x_i . The total cost of searching for the elements x_1, \dots, x_k is $\mathcal{O}(n \log n + k + \sum_i \log(1 + z_i))$.

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Example

Search sequence	a	b	c	a	c	b	b	a	b
Values z_i	0	1	2	2	1	2	0	2	1

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Theorem (searching elements in increasing order)

If the search sequence S contains elements in increasing order, then the total time required to search for S in a splay tree is $\mathcal{O}(n)$.

Questions

- Does there exist a search sequence on which a splay tree is asymptotically faster than a statically optimal tree constructed for that sequence?
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Theorem (Static Optimality)

Let x_1, \dots, x_k be a sequence of searches for elements from a set X , where every element of X is searched for at least once. Let T be a static tree on X , and suppose the total cost of searching for the elements in the sequence in T is f . Then the total cost of performing the searches in a splay tree is $\mathcal{O}(f)$.

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Dynamic Optimality Conjecture

Let A be any binary search tree algorithm that accesses an element x by traversing the path from the root to x at a cost of $d_x + 1$, and that between accesses can make any rotations in the tree at a cost of 1 per rotation where d_x is the depth of x . Let $A(S)$ be the cost for A to perform the sequence S of accesses. Then the cost for a splay tree to perform the same accesses is $\mathcal{O}(n + A(S))$.