

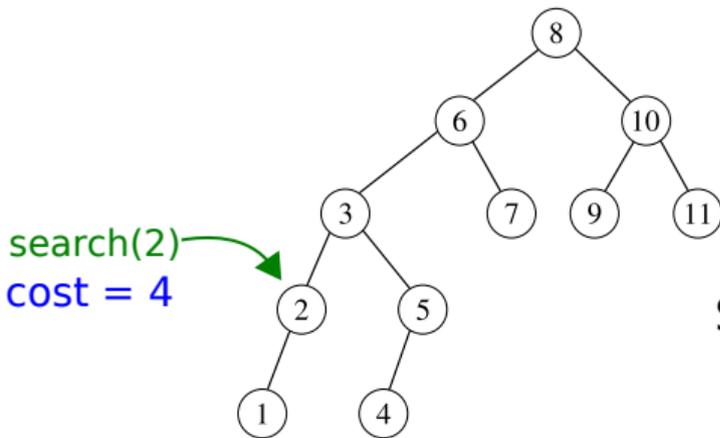
Multi-finger Binary Search Trees

Parinya Chalermsook Mayank Goswami

László Kozma Kurt Mehlhorn Thatchaphol Saranurak

Binary Search Tree (BST)

n keys



search cost = depth + 1

Search sequence:

x_1, x_2, x_3, \dots

Tree can be adapted using rotations, to prepare for next search.

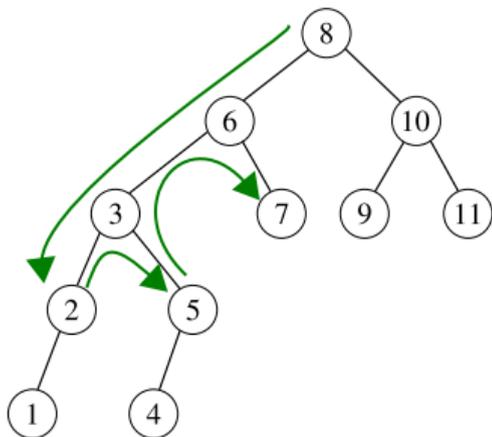
cost = pointer moves + rotations

- offline cost: **OPT**

- online algorithms: e.g. **Splay tree**

[Sleator, Tarjan, 1983]

Finger Search



search(2)

search(5)

search(7)

...

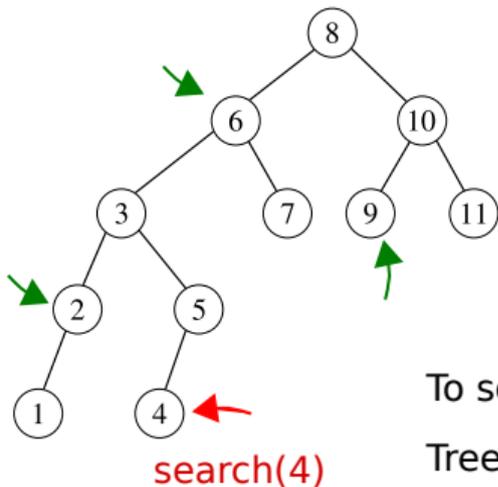
Search starts from previous location

Tree is static, chosen optimally

cost = amount of movement by finger

optimal finger search cost: **OPT_1**

Multiple fingers



k fingers, stationed at nodes
search starts from any of the fingers

To serve request, must move some finger there
Tree is static, chosen optimally

cost = amount of movement by **all** fingers

optimal k -finger search cost: OPT_k

Multiple fingers

$$OPT_1 \geq OPT_2 \geq \dots \geq OPT_k \geq \dots \geq OPT_n$$

matched by online
BST with rotations

[Cole et al., 2000]
[Iacono-Langerman, 2016]

$$OPT \leq O(OPT_1)$$

Our result

matched by online
BST with rotations

(with small overhead)

$$OPT \leq O(OPT_k) \cdot \log k$$

Our results

1. BSTs (with rotations) can simulate the k-finger optimum with small overhead:

$$OPT \leq O(OPT_k) \cdot \log k$$

$\log(k)$ factor is optimal.

2. There is an online BST with cost:

$$O(OPT_k) \cdot \log^7 k$$

The proof has three ingredients:

1. **Simulate** k-finger strategy by BST with single root-pointer (and rotations)
2. Find **online** k-finger strategy
3. **Learn** optimal underlying tree

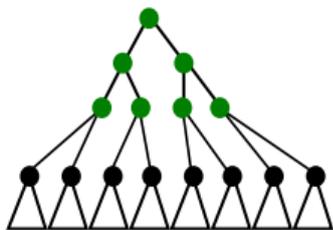
1. Simulate k-finger strategy by BST

with optimal $O(\log k)$ overhead,

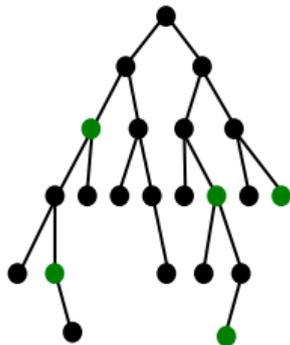
refining previous approach with overhead $O(k)$

[Demaine, Iacono, Langerman, Özkan, 2013]

Idea: store nodes with fingers as a balanced subtree

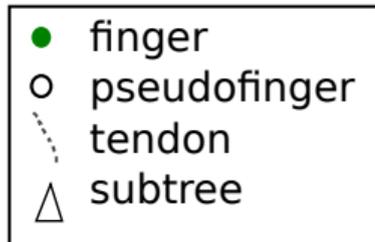
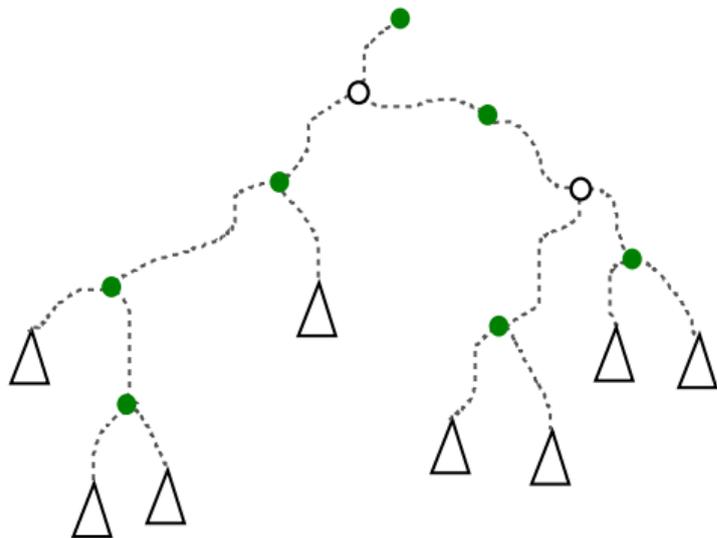


binary search tree



finger tree

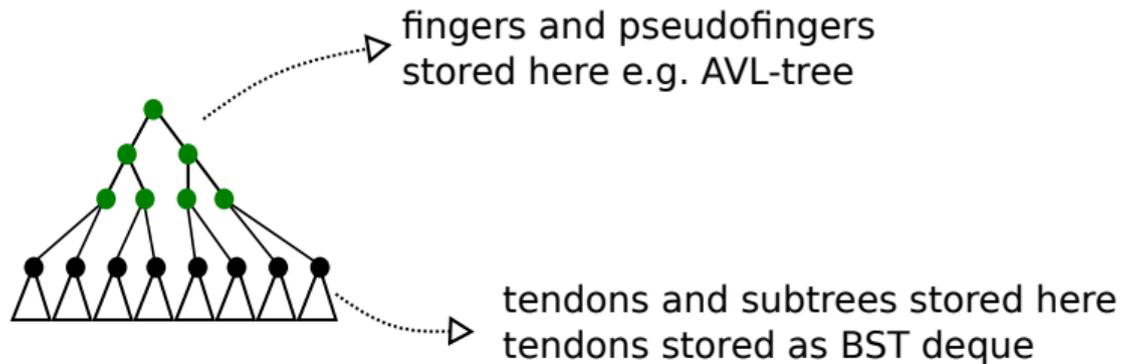
finger tree



$O(k)$ fingers and pseudofingers,
can store as e.g. AVL-tree

tendons: almost sorted,
can update in $O(1)$ amortized time

BST simulation of finger tree



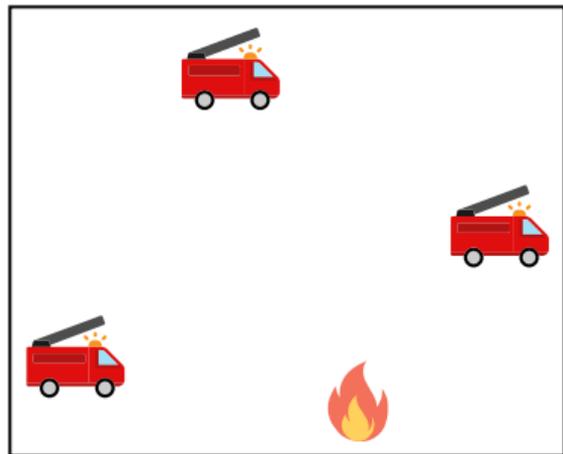
access finger --> search AVL tree ... $O(\log k)$

move finger --> update connected tendons
insert/delete in deque ... $O(1)$

update node role as (pseudo)finger;
may trigger AVL re-balancing ... $O(\log k)$

2. Find online k-finger strategy

metric space



k-server problem

Goal: serve requests,
minimize total movement

Our task is the special case when the metric is a BST

Relevant result: $O(\log^6 k)$ competitive online algorithm
[Lee; Bubeck et al. 2018]

3. Learn best underlying tree

Search sequence:

x_1, x_2, x_3, \dots

- **Idea:** Process in epochs of length $O(n \log n)$
- Maintain distribution over all $\sim 4^n$ trees

When epoch starts:

pick a tree from distribution, rotate to it
(rotation cost amortized over epoch)

When epoch ends:

evaluate all trees, update distribution using
multiplicative-weights-update (MWU)

Loss of tree T = cost during epoch if we used T

MWU guarantee

nr. of possible
trees $\sim 4^n$

max possible cost
within epoch $\sim n^2$

$$\text{overall cost} < (1 + \varepsilon)T^* + \frac{\ln N \cdot T^{max}}{\varepsilon}$$

total cost with best tree
in hindsight

poly(n) additive term

Summary of main result

1. **Simulate** k-finger strategy by BST
with single root-pointer (and rotations)
 $O(\log k)$ factor loss
refines technique of [Demaine et al.]
2. Find **online** k-finger strategy
 $O(\log^6 k)$ factor loss
first connection between k-server and BST?
3. **Learn** optimal underlying tree
 $O(1+\epsilon)$ factor, additive term
MWU technique, randomized

Applications

1. Better understand limitations of BST model

[Sleator, Tarjan, 1983]:

Splay tree cost is $O(OPT)$

Dynamic Optimality Conjecture

From our results:

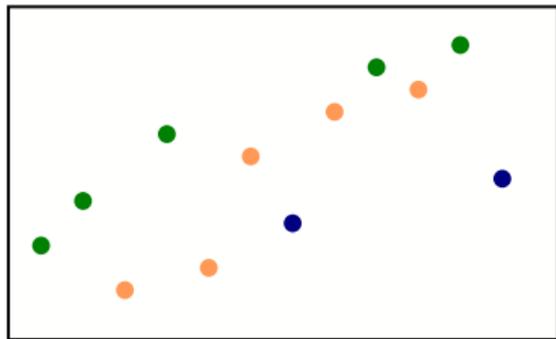
then it must be $O(OPT_k) \log k$

(even $k=1$ case is hard to prove)

barrier to Dynamic Optimality

Applications

2. Understand which search sequences are easy



Example:

k-monotone sequence
(shuffle of k runs)

k-finger strategy:

each finger serves one run
($O(n)$ traversal of tree)
total cost: $O(nk)$

\Rightarrow BST cost: $O(n \log k)$

Applications

2. Understand which search sequences are easy

Example:

weak unified bound

(search is close to a recent search)

amortized cost for search(x_k):

$$\min_t \{ \log |x_k - x_{k-t}| \} \cdot f(t)$$

rank-diff. from
recent search

penalty factor for
looking back too far

we give a k-finger strategy (not so easy)

OPEN QUESTIONS

1. De-randomize our online BST

(randomness comes from picking tree in MWU strategy)

Possible randomized/deterministic separation?

2. Towards dynamic optimality

Show that Splay tree matches $O(\text{OPT}_k) f(k)$

Show that some online BST matches $O(\text{OPT}_k) \log k$

(another $\log(k)$ factor unavoidable if online k -server is used)