

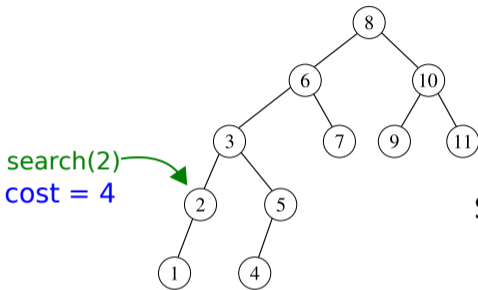
# Multi-finger Binary Search Trees

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# 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.

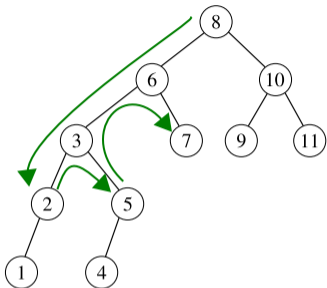
cost = pointer moves + rotations

- offline cost: **OPT**

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

[Sleator, Tarjan, 1983]

# Finger Search



search(2)

search(5)

search(7)

...

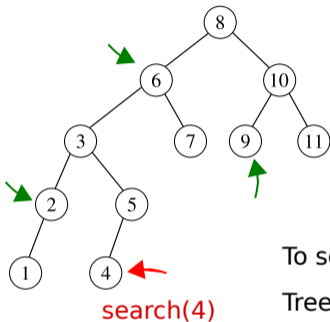
Search starts from previous location

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

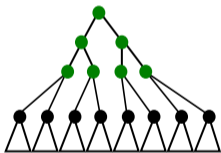
# 1. Simulate k-finger strategy by BST

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

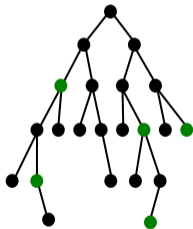
refining previous approach with overhead  $O(k)$

[Demaine, Iacono, Langerman, Özkan, 2013]

**Idea:** store nodes with fingers as a balanced subtree



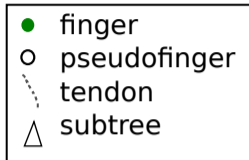
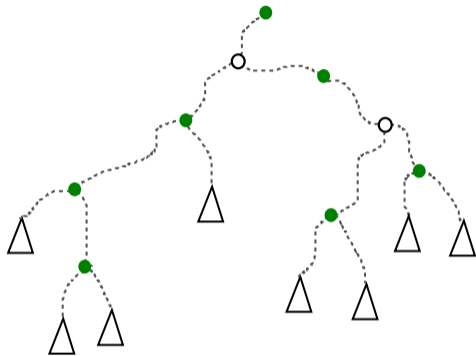
binary search tree



finger tree



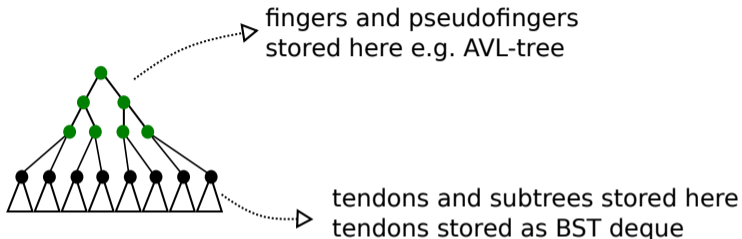
# finger tree



$O(k)$  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



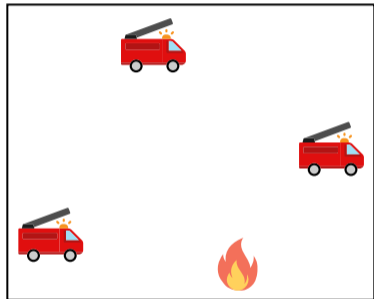
access finger --> search AVL tree ...  $O(\log k)$

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:

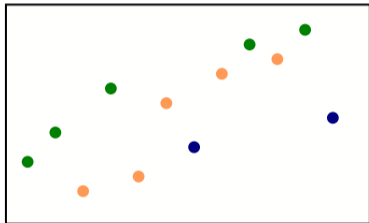
then it must be  $O(OPT_k) \log k$

(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)