Online Algorithms

Online Algorithms Tutorial 4 — Self-Organizing Datastructures

Self-Organizing Datastructures

Consider a set- or dictionary-datastructure:

- C++ map, set, etc.
- Java HashMap, TreeMap, etc.
- Python dict, set
- . . .

Self-Organizing Datastructures

Consider a set- or dictionary-datastructure:

- C++ map, set, etc.
- Java HashMap, TreeMap, etc.
- Python dict, set
- . . .

Operations:

- Find given key
- Add key
- Delete key

Classic vs. Self-organizing

Classic:

- Insert, Delete change the set and internal representation
- Find does not change anything

Classic vs. Self-organizing

Classic:

- Insert, Delete change the set and internal representation
- Find does not change anything

Self-organizing:

- Insert, Delete change the set and internal representation
- Find may change internal representation

Classic vs. Self-organizing

Classic:

- Insert, Delete change the set and internal representation
- Find does not change anything

Self-organizing:

- Insert, Delete change the set and internal representation
- Find may change internal representation

Pro's & Con's:

- Possibly faster if some keys are requested more often than others
- Can be used in data compression
- More complicated
- Bad for multithreading



Homework: Linked lists

Linked List:

- Not a very practical set implementation
- Can be used for compression
- Instead of encoding bytes or byte pairs, encode list positions
- Length depends on position in list
- Common byte pairs become shorter, uncommon ones longer

Results:

- MoveToFront 2-competitive
- FrequencyCount, Transpose not competitive

Arrays? Double-ended queues?

If sets are small or don't change much, (sorted) arrays are practical. Iteration is often more important than queries.

Arrays? Double-ended queues?

If sets are small or don't change much, (sorted) arrays are practical. Iteration is often more important than queries.

MOVETOFRONT does not work well for arrays.

Arrays? Double-ended queues?

If sets are small or don't change much, (sorted) arrays are practical. Iteration is often more important than queries.

MOVETOFRONT does not work well for arrays.

Double-ended queues or ring-buffers can work.

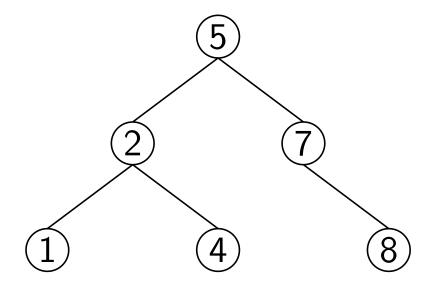
Hash tables

Have not been studied much; not much potential as self-organizing datastructures. Collisions should be rare.

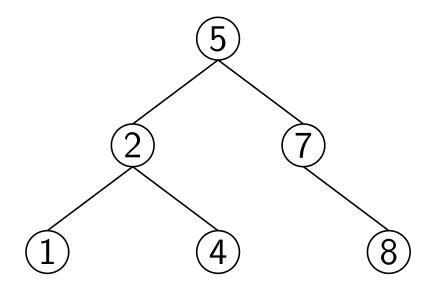


A lot of potential for self-organization: There are very many different binary trees that encode the same set.

A lot of potential for self-organization: There are very many different binary trees that encode the same set.



A lot of potential for self-organization: There are very many different binary trees that encode the same set.



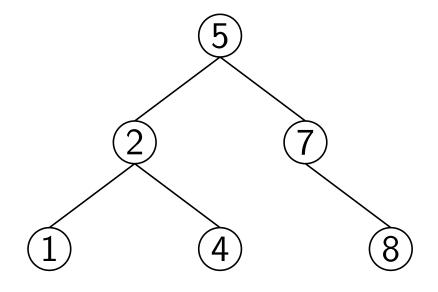
Search 5: Costs 1

Search 4: Costs 3

Search 7: Costs 2



A lot of potential for self-organization: There are very many different binary trees that encode the same set.





Search 5: Costs 1

Search 4: Costs 3

Search 7: Costs 2

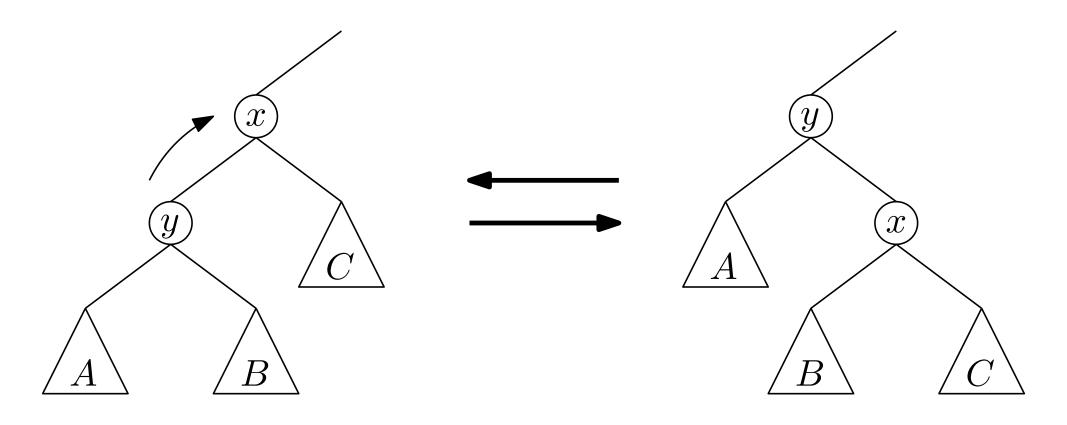
Search 4: Costs 1

Search 7: Costs 2



Operations on trees

Rotations: Transform one binary tree into the other.



Cost: 1 unit. Can also be used to realize AVL-trees. Can be applied anywhere in the tree.



Can we move from any tree to any other by rotations?

Can we move from any tree to any other by rotations?

Yes! Reaching a right path needs at most n-1 rotations:

- We can always rotate a vertex into the right path
- Increases number of edges on the right path by 1
- ullet In total, we have n-1 edges

Can we move from any tree to any other by rotations?

Yes! Reaching a right path needs at most n-1 rotations:

- We can always rotate a vertex into the right path
- Increases number of edges on the right path by 1
- ullet In total, we have n-1 edges

To transform from one tree to another: Transform both to right path, apply rotations from target to right path in reverse.

 \Rightarrow at most 2n-2 rotations.

Can we move from any tree to any other by rotations?

Yes! Reaching a right path needs at most n-1 rotations:

- We can always rotate a vertex into the right path
- Increases number of edges on the right path by 1
- \bullet In total, we have n-1 edges

To transform from one tree to another: Transform both to right path, apply rotations from target to right path in reverse.

 \Rightarrow at most 2n-2 rotations.

Actually, 2n-6 are sufficient.



Relaxed competitive ratio f(n) (similar to asymptotic):

$$A(\sigma, T_0) \leq f(n) \cdot \text{OPT}(\sigma, T_0) + \mathcal{O}(n).$$

Relaxed competitive ratio f(n) (similar to asymptotic):

$$A(\sigma, T_0) \leq f(n) \cdot \text{OPT}(\sigma, T_0) + \mathcal{O}(n).$$

Allows initial move to any initial tree!

Relaxed competitive ratio f(n) (similar to asymptotic):

$$A(\sigma, T_0) \leq f(n) \cdot \text{OPT}(\sigma, T_0) + \mathcal{O}(n).$$

Allows initial move to any initial tree!

Related problem: Optimal static binary search tree.

- ullet Given request probabilities for n keys
- Compute the tree minimizing the average depth
- NP-hard!

Relaxed competitive ratio f(n) (similar to asymptotic):

$$A(\sigma, T_0) \leq f(n) \cdot \text{OPT}(\sigma, T_0) + \mathcal{O}(n).$$

Allows initial move to any initial tree!

Related problem: Optimal static binary search tree.

- ullet Given request probabilities for n keys
- Compute the tree minimizing the average depth
- NP-hard!

f(n)-static-competitive:

Compare A not to OPT, but to optimal static tree.

How bad can we be? What is the worst possible competitive ratio?

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

What is the best we can hope for?

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

What is the best we can hope for?

 $\mathcal{O}(1)$; constant competitive ratio.

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

What is the best we can hope for?

 $\mathcal{O}(1)$; constant competitive ratio.

Is there is an obvious $O(\log n)$ -competitive algorithm?

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

What is the best we can hope for?

 $\mathcal{O}(1)$; constant competitive ratio.

Is there is an obvious $O(\log n)$ -competitive algorithm?

Yes! Balanced static search trees (AVL, red-black, ...)!

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

What is the best we can hope for?

 $\mathcal{O}(1)$; constant competitive ratio.

Is there is an obvious $O(\log n)$ -competitive algorithm?

Yes! Balanced static search trees (AVL, red-black, ...)!

Simple strategies (as for linked lists)?

How bad can we be? What is the worst possible competitive ratio?

 $\mathcal{O}(n)$; corresponds to requesting last element of path.

What is the best we can hope for?

 $\mathcal{O}(1)$; constant competitive ratio.

Is there is an obvious $O(\log n)$ -competitive algorithm?

Yes! Balanced static search trees (AVL, red-black, ...)!

Simple strategies (as for linked lists)?

ROTATEONCE, ROTATETOTOP



ROTATEONCE is bad (like TRANSPOSE): Start with path, request last two elements alternatingly.

ROTATEONCE is bad (like TRANSPOSE): Start with path, request last two elements alternatingly.

ROTATETOTOP:

- Any initial tree
- Sequence $(1,\ldots,n)$
- How does the tree look now?

ROTATEONCE is bad (like TRANSPOSE): Start with path, request last two elements alternatingly.

ROTATETOTOP:

- Any initial tree
- Sequence $(1,\ldots,n)$
- How does the tree look now?

After 1: 1 is at the root! After 2: Left path is 2-1!

How good are simple strategies?

ROTATEONCE is bad (like TRANSPOSE): Start with path, request last two elements alternatingly.

ROTATETOTOP:

- Any initial tree
- Sequence $(1,\ldots,n)$
- How does the tree look now?

After 1: 1 is at the root! After 2: Left path is 2-1!

After n: Left path n-...-1.

How good are simple strategies?

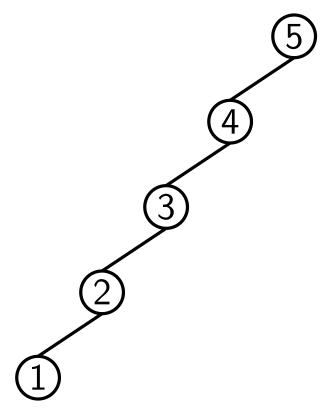
ROTATEONCE is bad (like TRANSPOSE): Start with path, request last two elements alternatingly.

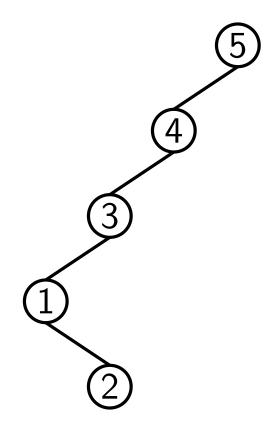
ROTATETOTOP:

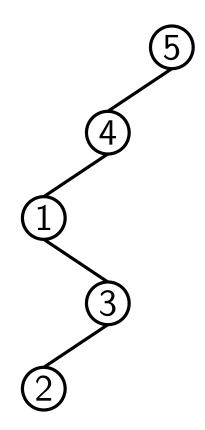
- Any initial tree
- Sequence $(1,\ldots,n)$
- How does the tree look now?

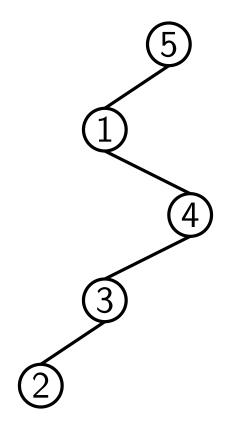
After 1: 1 is at the root! After 2: Left path is 2-1!

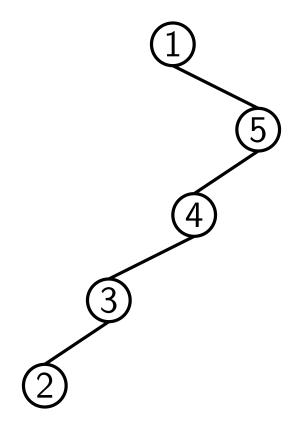
After n: Left path n-...-1.

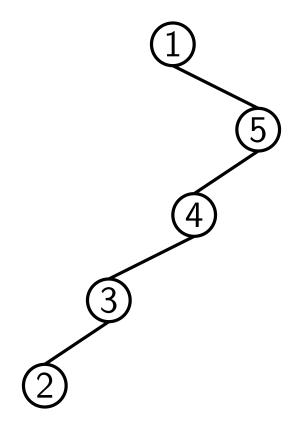




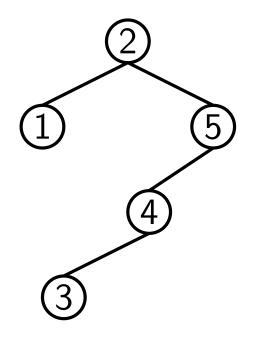








After request for 2:



Generally, the tree stays a path.

 $\Rightarrow \Omega(n^2)$ cost per repetition of $(1,\ldots,n)$.

Generally, the tree stays a path.

 $\Rightarrow \Omega(n^2)$ cost per repetition of $(1,\ldots,n)$.

We can achieve $\mathcal{O}(n)$ per repetition:

First rotate into right path; costs O(n).

Then $\mathcal{O}(1)$ per request.

 \Rightarrow ROTATETOTOP is $\Omega(n)$ -competitive.

Better idea?

Can we do better?

- No better (proven) algorithm than $\mathcal{O}(\log n)$
- But a (conjectured) candidate for $\mathcal{O}(1)$: Splay trees

Better idea?

Can we do better?

- No better (proven) algorithm than $\mathcal{O}(\log n)$
- But a (conjectured) candidate for $\mathcal{O}(1)$: Splay trees

Basic idea: ROTATETOTOP with changed rotations

- Consider up to two rotations at once
- If they are in the same direction, rotate upper edge first
- Operations: Zig (last rotation), Zig-Zag, Zig-Zig
- Zig and Zig-Zag are normal bottom-up rotations
- Zig-Zig: Does not destroy balance

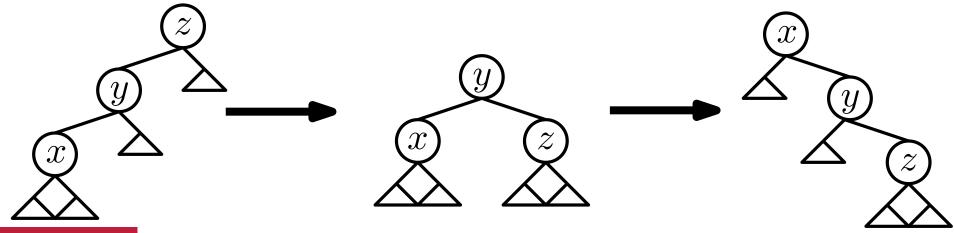
Better idea?

Can we do better?

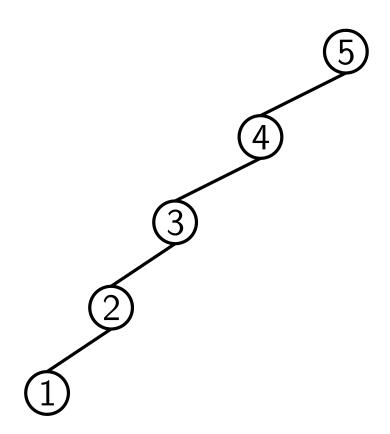
- No better (proven) algorithm than $\mathcal{O}(\log n)$
- But a (conjectured) candidate for $\mathcal{O}(1)$: Splay trees

Basic idea: ROTATETOTOP with changed rotations

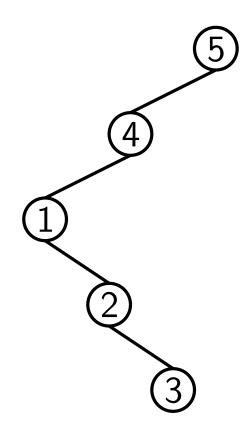
- Consider up to two rotations at once
- If they are in the same direction, rotate upper edge first
- Operations: Zig (last rotation), Zig-Zag, Zig-Zig
- Zig and Zig-Zag are normal bottom-up rotations
- Zig-Zig: Does not destroy balance



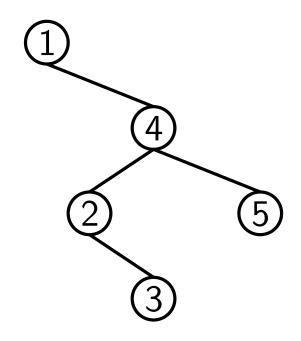
Splaying a path



Splaying a path



Splaying a path



Static optimality theorem:

For any access probabilities, splay trees are within a constant factor of optimal static binary trees.

Static optimality theorem:

For any access probabilities, splay trees are within a constant factor of optimal static binary trees.

Proof is technical, would take too long, but understandable.

https://www.cs.cmu.edu/~sleator/papers/self-adjusting.pdf

Static optimality theorem:

For any access probabilities, splay trees are within a constant factor of optimal static binary trees.

Proof is technical, would take too long, but understandable.

https://www.cs.cmu.edu/~sleator/papers/self-adjusting.pdf

Key lemma: Splay Access Lemma

Static optimality theorem:

For any access probabilities, splay trees are within a constant factor of optimal static binary trees.

Proof is technical, would take too long, but understandable.

https://www.cs.cmu.edu/~sleator/papers/self-adjusting.pdf

Key lemma: Splay Access Lemma

Number of nodes below node: n(v)

Rank of v: $r(v) = \log_2(n(v))$

Tree potential $\Phi(T) = \sum_{v} r(v)$

Static optimality theorem:

For any access probabilities, splay trees are within a constant factor of optimal static binary trees.

Proof is technical, would take too long, but understandable.

https://www.cs.cmu.edu/~sleator/papers/self-adjusting.pdf

Key lemma: Splay Access Lemma

Number of nodes below node: n(v)

Rank of v: $r(v) = \log_2(n(v))$

Tree potential $\Phi(T) = \sum_{v} r(v)$

Cost of access/splaying with k rotations:

$$k + \Phi(U) - \Phi(T) \in \mathcal{O}(\log n).$$



Static optimality theorem:

For any access probabilities, splay trees are within a constant factor of optimal static binary trees.

Proof is technical, would take too long, but understandable.

https://www.cs.cmu.edu/~sleator/papers/self-adjusting.pdf

Key lemma: Splay Access Lemma

Number of nodes below node: n(v)

Rank of v: $r(v) = \log_2(n(v))$

Tree potential $\Phi(T) = \sum_{v} r(v)$ Careful: Initial potential > 0!

Cost of access/splaying with k rotations:

$$k + \Phi(U) - \Phi(T) \in \mathcal{O}(\log n).$$



Dynamic Optimality Conjecture

Splay trees are $\mathcal{O}(1)$ -competitive in our relaxed sense:

A sequence σ of accesses costs $\mathcal{O}(n + \mathrm{OPT}(\sigma))$.

Dynamic Optimality Conjecture

Splay trees are $\mathcal{O}(1)$ -competitive in our relaxed sense: A sequence σ of accesses costs $\mathcal{O}(n+\mathrm{OPT}(\sigma))$.

Status: open. Works for all special sequence families studied so far. Works experimentally in practice.

Dynamic Optimality Conjecture

Splay trees are $\mathcal{O}(1)$ -competitive in our relaxed sense: A sequence σ of accesses costs $\mathcal{O}(n+\mathrm{OPT}(\sigma))$.

Status: open. Works for all special sequence families studied so far. Works experimentally in practice.

Scanning Theorem

Access sequences like $1, \ldots, n$ or $n, \ldots, 1$ cost $\mathcal{O}(n)$.

Dynamic Optimality Conjecture

Splay trees are $\mathcal{O}(1)$ -competitive in our relaxed sense: A sequence σ of accesses costs $\mathcal{O}(n+\mathrm{OPT}(\sigma))$.

Status: open. Works for all special sequence families studied so far. Works experimentally in practice.

Scanning Theorem

Access sequences like $1, \ldots, n$ or $n, \ldots, 1$ cost $\mathcal{O}(n)$.

Static Finger Theorem

In a sequence σ with many accesses close to a *finger* f:

$$\mathcal{O}\left(n\log n + \sum_{x \in \sigma} \log(|x - f| + 2)\right).$$

Further Properties

Dynamic Finger Theorem

Let x, y be consecutive elements in the sequence σ . Total cost:

$$\mathcal{O}\left(n+\sum_{x,y}\log(|x-y|+2)\right).$$

Further Properties

Dynamic Finger Theorem

Let x, y be consecutive elements in the sequence σ . Total cost:

$$\mathcal{O}\left(n+\sum_{x,y}\log(|x-y|+2)\right).$$

Working Set Theorem

Let t(x) be the number of distinct elements accessed since last access to x. Total cost:

$$\mathcal{O}\left(n\log n + \sum_{x} \log(t(x) + 2)\right).$$

Possible application: Sorting

Possible application: Sorting

Insert:

As for binary trees, but then splay the inserted element.

Possible application: Sorting

Insert:

As for binary trees, but then splay the inserted element.

SPLAY SORT:

Insert n elements into splay tree, then walk the tree in-order.

Possible application: Sorting

Insert:

As for binary trees, but then splay the inserted element.

SPLAY SORT:

Insert n elements into splay tree, then walk the tree in-order.

In-order traversal: O(n), insertion: $O(n \log n)$ worst-case

Possible application: Sorting

Insert:

As for binary trees, but then splay the inserted element.

SPLAY SORT:

Insert n elements into splay tree, then walk the tree in-order.

In-order traversal: O(n), insertion: $O(n \log n)$ worst-case

Practice: Slower than QuickSort, MergeSort, etc. for random sequences

Possible application: Sorting

Insert:

As for binary trees, but then splay the inserted element.

SPLAY SORT:

Insert n elements into splay tree, then walk the tree in-order.

In-order traversal: O(n), insertion: $O(n \log n)$ worst-case

Practice: Slower than QuickSort, MergeSort, etc. for random sequences

But: Adaptive! Works better with partially-sorted inputs!

Possible application: Sorting

Insert:

As for binary trees, but then splay the inserted element.

SPLAY SORT:

Insert n elements into splay tree, then walk the tree in-order.

In-order traversal: O(n), insertion: $O(n \log n)$ worst-case

Practice: Slower than QuickSort, MergeSort, etc. for random sequences

But: Adaptive! Works better with partially-sorted inputs!

There are specialized, better algorithms for that case though.

