CS 240 - Data Structures and Data Management

Module 5: Other Dictionary Implementations

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Based on lecture notes by many previous cs240 instructors

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Outline

- 5 Dictionaries with Lists revisited
 - Dictionary ADT: Implementations thus far
 - Skip Lists
 - Re-ordering Items

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Dictionary ADT: Implementations thus far

A dictionary is a collection of key-value pairs (KVPs), supporting operations search, insert, and delete.

Realizations we have seen so far:

- Unordered array or linked list: $\Theta(1)$ insert, $\Theta(n)$ search and delete
- Ordered array: $\Theta(\log n)$ search, $\Theta(n)$ insert and delete
- Binary search trees: $\Theta(height)$ search, insert and delete
- Balanced BST (AVL trees):
 Θ(log n) search, insert, and delete

Improvements/Simplifications?

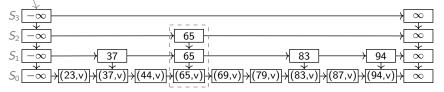
- Can show: If the KVPs were inserted in random order, then the expected height of the binary search tree would be $O(\log n)$.
- How can we use randomization within the data structure to mirror what would happen on random input?

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Skip Lists

- A hierarchy S of ordered linked lists (levels) S_0, S_1, \dots, S_h :
 - ▶ Each list S_i contains the special keys $-\infty$ and $+\infty$ (sentinels)
 - ▶ List S_0 contains the KVPs of S in non-decreasing order. (The other lists store only keys, or links to nodes in S_0 .)
 - ▶ Each list is a subsequence of the previous one, i.e., $S_0 \supseteq S_1 \supseteq \cdots \supseteq S_h$
 - ▶ List S_h contains only the sentinels; the left sentinel is the *root*



- Each KVP belongs to a tower of nodes
- There are (usually) more nodes than keys
- The skip list consists of a reference to the topmost left node.
- Each node p has references p.after and p.below

Search in Skip Lists

For each level, find **predecessor** (node before where k would be). This will also be useful for *insert*/*delete*.

```
getPredecessors (k)

1. p \leftarrow \text{root}

2. P \leftarrow \text{stack of nodes, initially containing } p

3. while p.below \neq \text{NIL do}

4. p \leftarrow p.below

5. while p.after.key < k \text{ do } p \leftarrow p.after

6. P.push(p)

7. return P
```

```
skipList::search (k)

1. P \leftarrow getPredecessors(k)

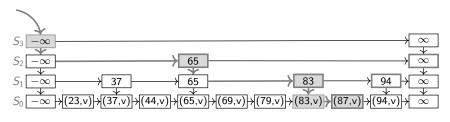
2. p_0 \leftarrow P.top() // predecessor of k in S_0

3. if p_0.after.key = k return p_0.after

4. else return "not found, but would be after p_0"
```

Example: Search in Skip Lists

Example: search(87)



- key compared with k
- added to P

Insert in Skip Lists

```
skipList::insert(k, v)
```

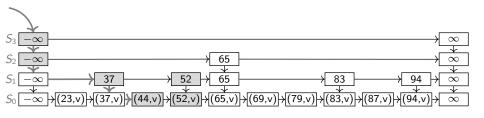
- Randomly repeatedly toss a coin until you get tails
 - ► Negative binomial distribution
- Let i the number of times the coin came up heads
 - we want k to be in lists S_0, \ldots, S_i .
 - ▶ $i \rightarrow height$ of tower of k
 - $P(\text{tower of key } k \text{ has height } \geq i) = \left(\frac{1}{2}\right)^i$
- Increase height h of skip list, if needed, to have h > i levels.
- Use getPredecessors(k) to get stack P. The top i items of P are the predecessors p_0, p_1, \dots, p_i of where k should be in each list S_0, S_1, \dots, S_i
- Insert (k, v) after p_0 in S_0 , and k after p_j in S_j for $1 \le j \le i$

Example: Insert in Skip Lists

Example: skipList::insert(52, v)

Coin tosses: $H,T \Rightarrow i = 1$

getPredecessors(52)



Example 2: Insert in Skip Lists

Example: skipList::insert(100, v)

Insert in Skip Lists

```
skipList::insert(k, v)
   P \leftarrow getPredecessors(k)
    for (i \leftarrow 0; random(2) = 1; i \leftarrow i+1) {} // random tower height
     while i > P.size()
                                                     // increase skip-list height?
3.
4.
             root ← new sentinel-only list, linked in appropriately
             add left sentinel of root at bottom of stack P
5.
                                                        // insert (k, v) in S_0
6. p \leftarrow P.pop()
7.
     z_{below} \leftarrow new node with (k, v), inserted after p
     while i > 0
                                                        // insert k in S_1, \ldots, S_i
8.
9.
             p \leftarrow P.pop()
10.
            z \leftarrow new node with k added after p
11. z.below \leftarrow z_{below}; z_{below} \leftarrow z
            i \leftarrow i - 1
12.
```

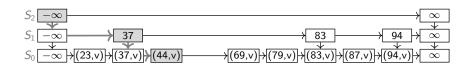
Delete in Skip Lists

It is easy to remove a key since we can find all predecessors. Then eliminate layers if there are multiple ones with only sentinels.

```
skipList::delete(k)
   P \leftarrow getPredecessors(k)
2. while P is non-empty
3
            p \leftarrow P.pop() // predecessor of k in some layer
            if p.after.kev = k
4.
5
                 p.after \leftarrow p.after.after
            else break // no more copies of k
6
    p \leftarrow left sentinel of the root-list
       while p.below.after is the \infty-sentinel
            // the two top lists are both only sentinels, remove one
            p.below \leftarrow p.below.below
9.
            p.after.below \leftarrow p.after.below.below
10.
```

Example: Delete in Skip Lists

Example: skipList::delete(65)

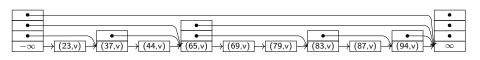


Analysis of Skip Lists

- Expected **space** usage: O(n)
- Expected height: O(log n)
- Crucial for all operations:
 - ▶ How often do we *drop down* (execute $p \leftarrow p.below$)?
 - ▶ How often do we *step forward* (execute $p \leftarrow p.after$)?
- skipList::search: O(log n) expected time
 - ▶ # drop-downs = height
 - expected # forward-steps is ≤ 1 in each level
 - ► expected total # forward-steps is in $O(\log n)$
- skipList::insert: O(log n) expected time
- skipList::delete: O(log n) expected time

Summary of Skip Lists

- O(n) expected space, all operations take $O(\log n)$ expected time.
- As described they are no faster than randomized binary search trees.
- Can show: A biased coin-flip to determine tower-height gives smaller expected run-times.
- Can save links (hence space) by implementing towers as array.



 Then skip lists are simple to implement. They are fast with good cache locality but can still suffer from cache misses.

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Re-ordering Items

- Recall: Unordered list/array implementation of ADT Dictionary search: $\Theta(n)$, insert: $\Theta(1)$, delete: $\Theta(1)$ (after a search)
- Lists/arrays are a very simple and popular implementation. Can we do something to make search more effective in practice?
- No: if items are accessed equally likely
- Yes: otherwise (we have a probability distribution of the items)
 - ▶ Intuition: Frequently accessed items should be in the front.
 - ► Two cases: Do we know the access distribution beforehand or not?
 - ► For short lists or extremely unbalanced distributions this may be faster than AVL trees or Skip Lists, and much easier to implement.

Optimal Static Ordering

Example:

| key | A | В | C | D | E |
|---------------------|----------------|----------------|----------------|-----------------|----------------|
| frequency of access | 2 | 8 | 1 | 10 | 5 |
| access-probability | $\frac{2}{26}$ | $\frac{8}{26}$ | $\frac{1}{26}$ | $\frac{10}{26}$ | $\frac{5}{26}$ |

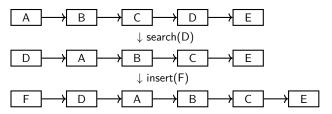
- We count cost *i* for accessing the key in the *i*th position.
- Order A, B, C, D, E has expected access cost $\frac{2}{26} \cdot 1 + \frac{8}{26} \cdot 2 + \frac{1}{26} \cdot 3 + \frac{10}{26} \cdot 4 + \frac{5}{26} \cdot 5 = \frac{86}{26} \approx 3.31$
- Order D, B, E, A, C has expected access cost

$$\frac{10}{26} \cdot 1 + \frac{8}{26} \cdot 2 + \frac{5}{26} \cdot 3 + \frac{2}{26} \cdot 4 + \frac{1}{26} \cdot 5 = \frac{66}{26} \approx 2.54$$

- Claim: Over all possible static orderings, the one that sorts items by non-increasing access-probability minimizes the expected access cost.
- Proof Idea: For any other ordering, exchanging two items that are out-of-order according to their access probabilities makes the total cost decrease.

Dynamic Ordering: MTF

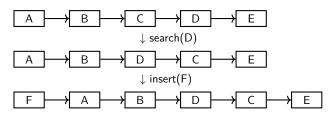
- What if we do not know the access probabilities ahead of time?
- Rule of thumb (temporal locality): A recently accessed item is likely to be used soon again.
- In list: Always insert at the front
- Move-To-Front heuristic (MTF): Upon a successful search, move the accessed item to the front of the list



 We can also do MTF on an array, but should then insert and search from the back so that we have room to grow.

Dynamic Ordering: Transpose

Transpose heuristic: Upon a successful search, swap the accessed item with the item immediately preceding it



Performance of dynamic ordering:

- Transpose does not adapt quickly to changing access patterns.
- MTF works well in practice.
- Can show: MTF is "2-competitive":
 No more than twice as bad as the optimal static ordering.