

# 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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Fall 2022

References: Sedgewick 9.1-9.4

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# Outline

## 1 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:**  $\Theta(\text{height})$  search, insert and delete
- **Balanced BST** (AVL trees):  
 $\Theta(\log n)$  search, insert, and delete

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Improvements/Simplifications?

- **Can show:** The average-case height of binary search trees (over all possible insertion sequences) is  $O(\log n)$ .
- How can we shift the average-case to expected height via randomization?

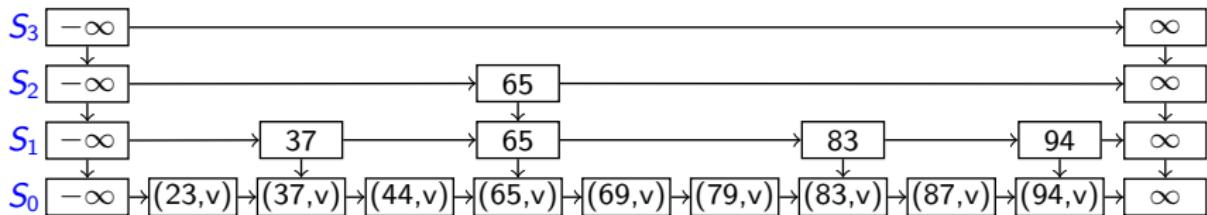
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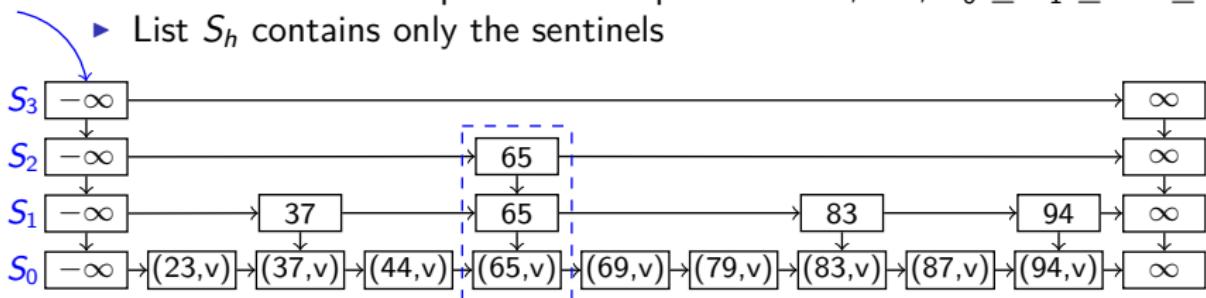
# Skip Lists

- 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.,  $S_0 \supseteq S_1 \supseteq \dots \supseteq S_h$
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- 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  $p.\text{after}$  and  $p.\text{below}$

# Search in Skip Lists

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

*getPredecessors* ( $k$ )

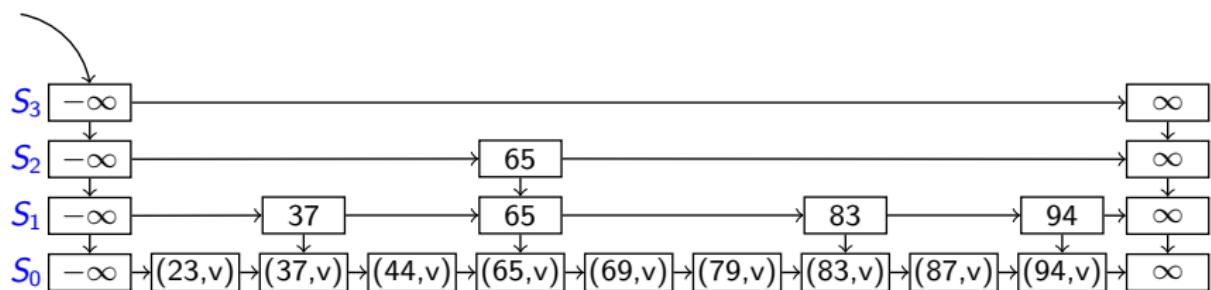
1.  $p \leftarrow$  topmost left sentinel
2.  $P \leftarrow$  stack of nodes, initially containing  $p$
3. **while**  $p.below \neq \text{NIL}$  **do**
4.      $p \leftarrow p.below$
5.     **while**  $p.after.key < k$  **do**  $p \leftarrow p.after$
6.      $P.push(p)$
7. **return**  $P$

*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  $p_0$ "

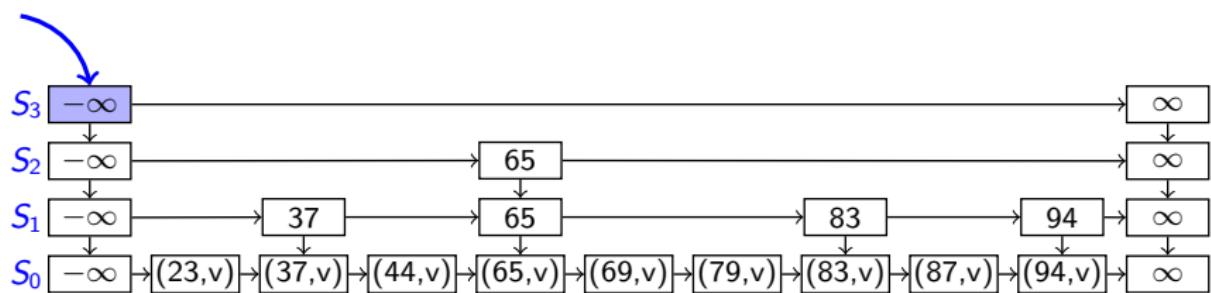
# Example: Search in Skip Lists

**Example:** *search(87)*



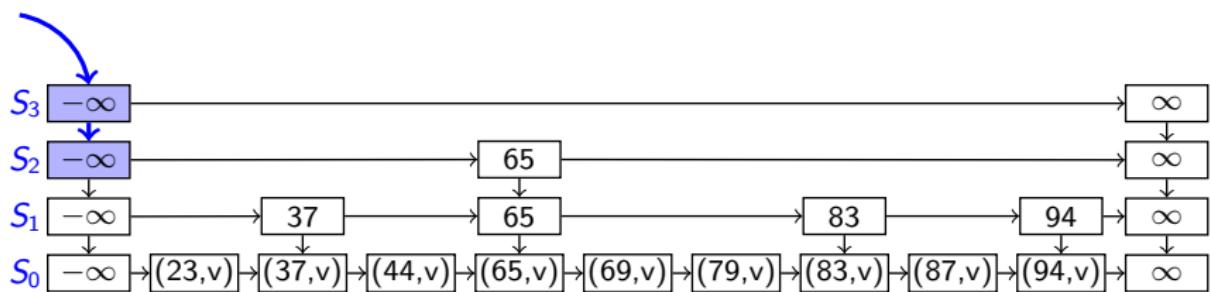
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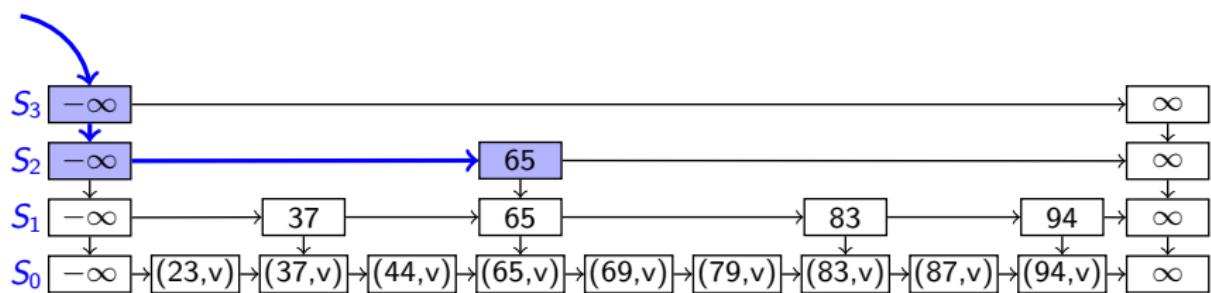
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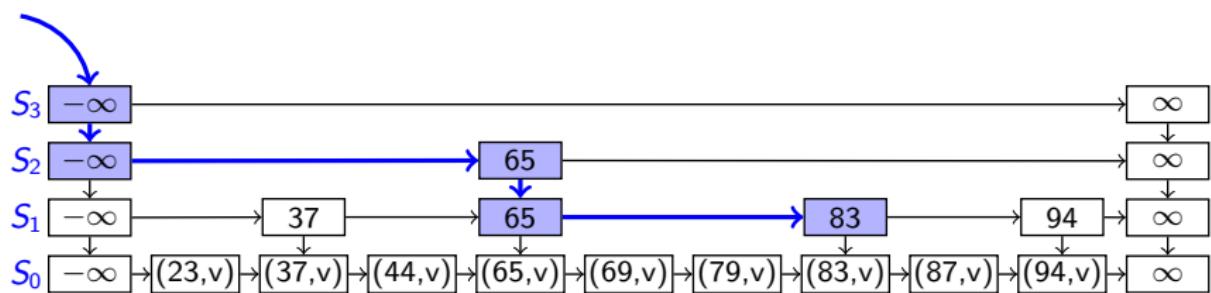
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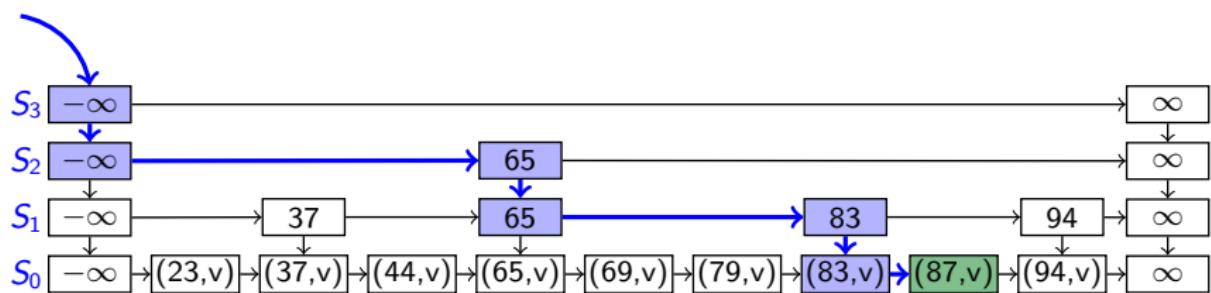
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# Insert in Skip Lists

*skipList::insert(k, v)*

- Randomly repeatedly toss a coin until you get tails
- Let  $i$  the number of times the coin came up heads; this will be the height of the tower of  $k$

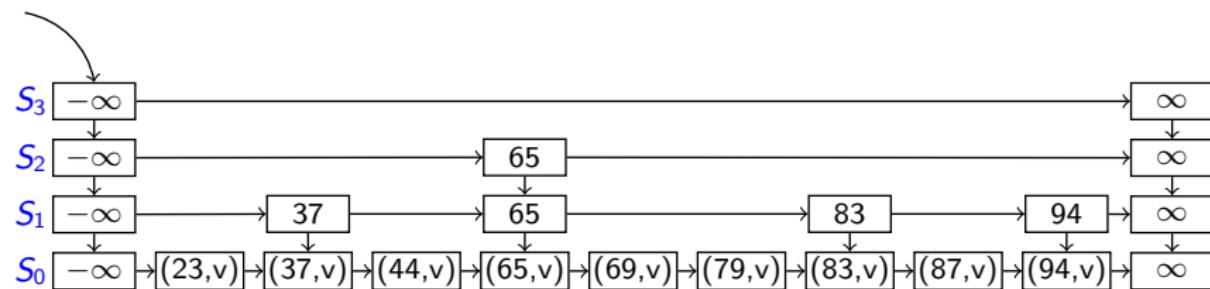
$$P(\text{tower of key } k \text{ has height } \geq i) = \left(\frac{1}{2}\right)^i$$

- Increase height 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  $1 \leq j \leq i$

## Example: Insert in Skip Lists

Example: `skipList::insert(52, v)`

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

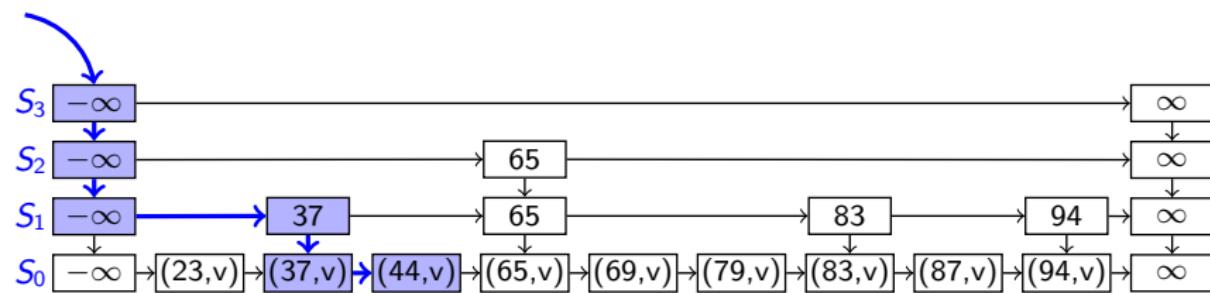


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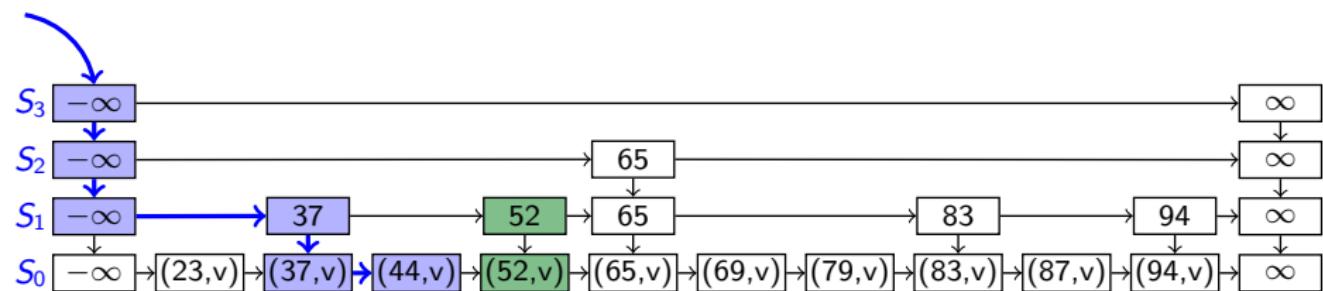


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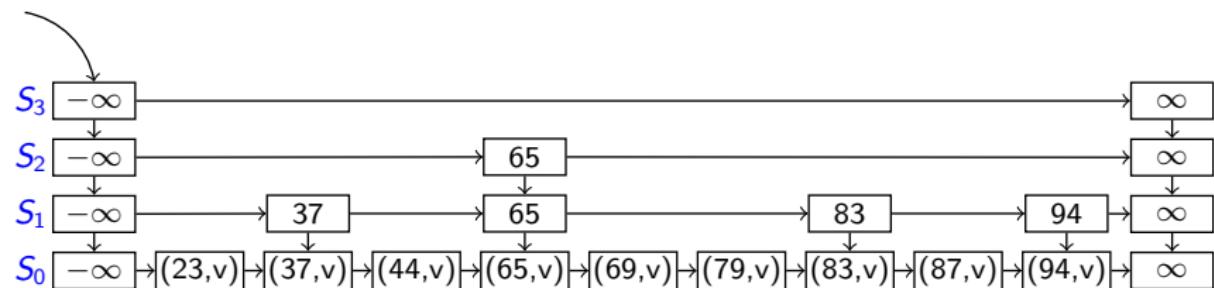
`getPredecessors(52)`



## Example 2: Insert in Skip Lists

Example: `skipList::insert(100, v)`

Coin tosses: H,H,H,T  $\Rightarrow i = 3$

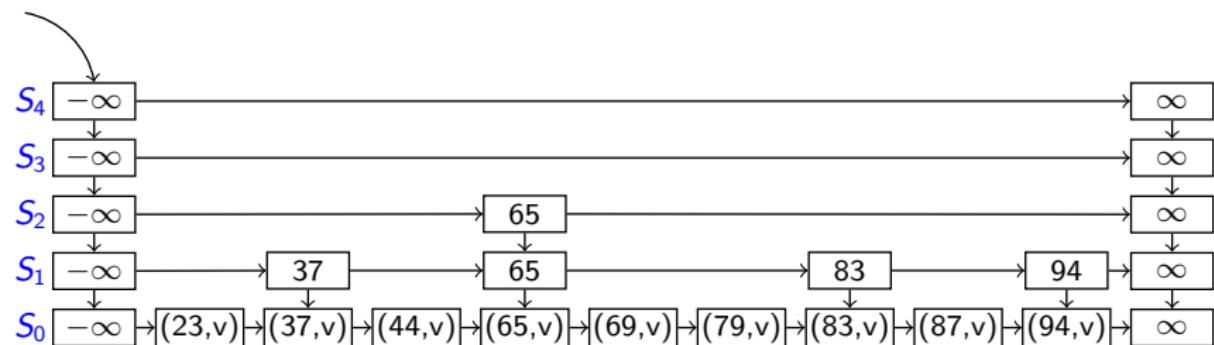


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Example: `skipList::insert(100, v)`

Coin tosses: H,H,H,T  $\Rightarrow i = 3$

*Height increase*



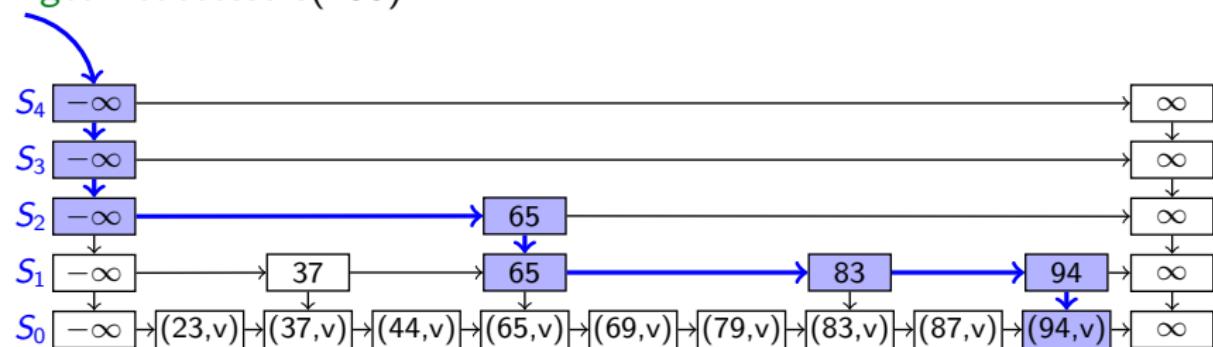
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Example: `skipList::insert(100, v)`

Coin tosses: H,H,H,T  $\Rightarrow i = 3$

*Height increase*

`getPredecessors(100)`



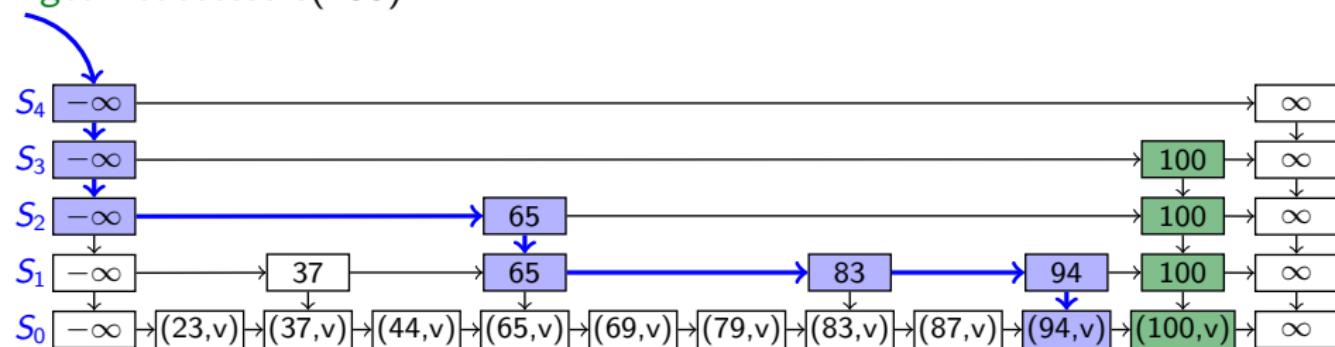
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Example: `skipList::insert(100, v)`

Coin tosses: H,H,H,T  $\Rightarrow i = 3$

*Height increase*

`getPredecessors(100)`



## Delete in Skip Lists

It is easy to remove a key since we can find all predecessors.

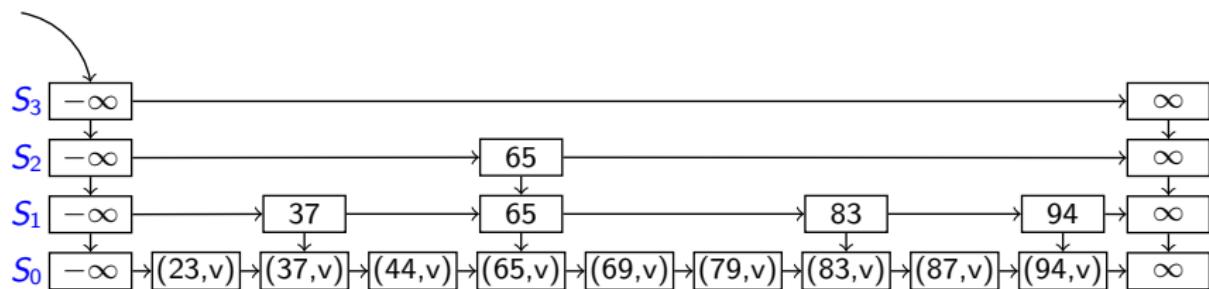
Then eliminate layers if there are multiple ones with only sentinels.

*skipList::delete(k)*

1.  $P \leftarrow \text{getPredecessors}(k)$
2. **while**  $P$  is non-empty
3.      $p \leftarrow P.pop()$      // predecessor of  $k$  in some layer
4.     **if**  $p.after.key = k$
5.          $p.after \leftarrow p.after.after$
6.     **else break**     // no more copies of  $k$
7.      $p \leftarrow$  topmost left sentinel
8.     **while**  $p.below.after$  is the  $\infty$ -sentinel  
        // the two top lists are both only sentinels, remove one
9.          $p.below \leftarrow p.below.below$
10.          $p.after.below \leftarrow p.after.below.below$

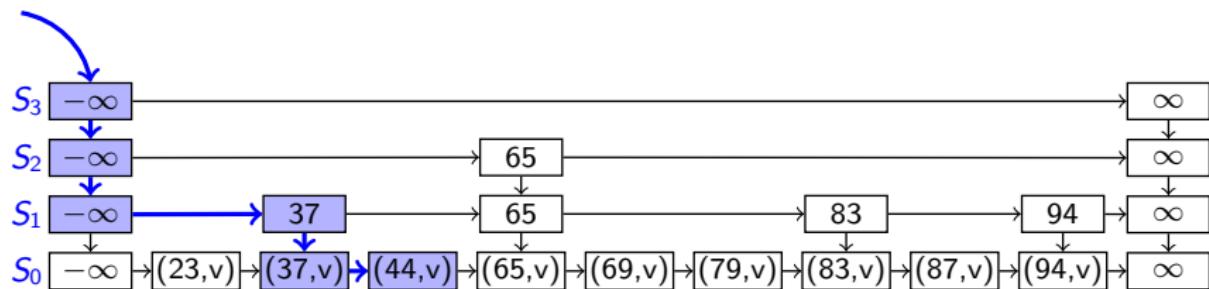
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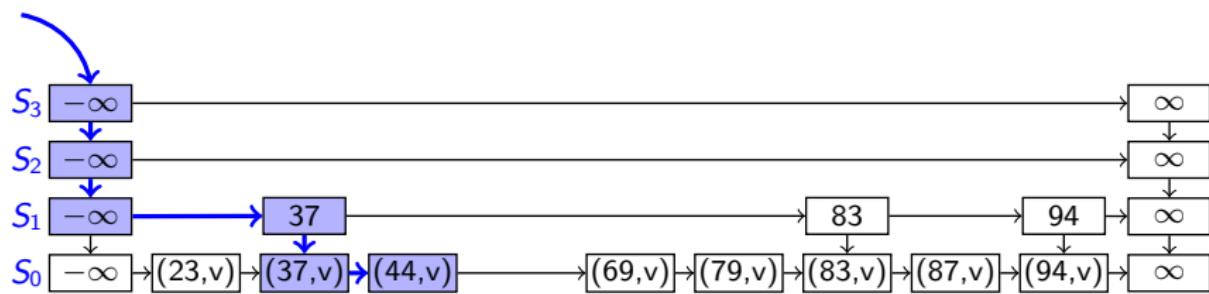
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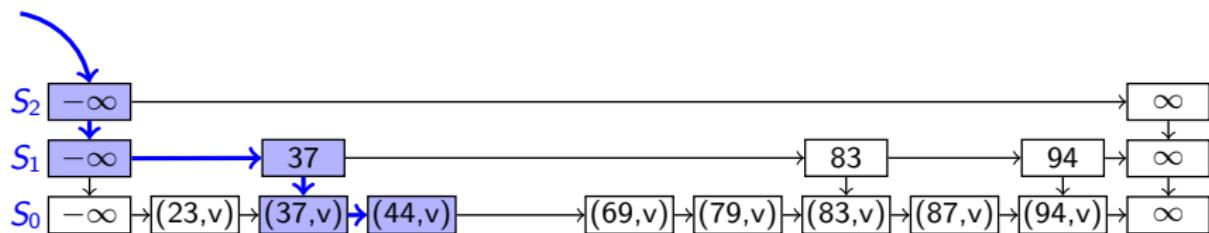


# Example: Delete in Skip Lists

Example: `skipList::delete(65)`

`getPredecessors(65)`

*Height decrease*



# Summary of Skip Lists

- Expected **space** usage:  $O(n)$
- Expected **height**:  $O(\log n)$   
A skip list with  $n$  items has height at most  $3 \log n$  with probability at least  $1 - 1/n^2$
- Crucial for all operations:
  - ▶ How often do we *drop down* (execute  $p \leftarrow p.\text{below}$ )?
  - ▶ How often do we *scan forward* (execute  $p \leftarrow p.\text{after}$ )?
- *skipList::search*:  $O(\log n)$  expected time
  - ▶ # drop-downs = height
  - ▶ expected # scan-forwards is  $\leq 2$  in each level
- *skipList::insert*:  $O(\log n)$  expected time
- *skipList::delete*:  $O(\log n)$  expected time
- Skip lists are fast and simple to implement in practice

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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?

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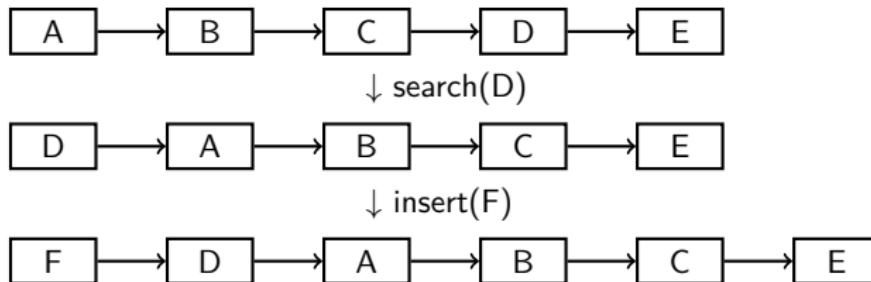
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- **Claim:** Over all possible static orderings, the one that sorts items by non-increasing access-probability minimizes the expected access cost.
- **Proof Idea:** For any other ordering, exchanging two items that are out-of-order according to their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

# Dynamic Ordering: MTF

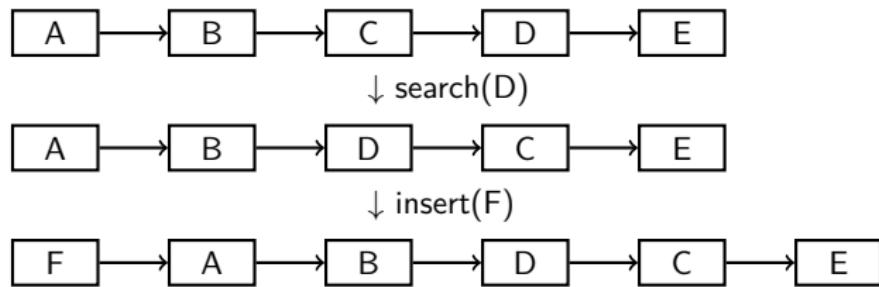
- What if we do *not know the access probabilities* ahead of time?
- Rule of thumb (**temporal locality**): A recently accessed item is likely to be used soon again.
- In list: Always insert at the front
- **Move-To-Front heuristic** (MTF): Upon a successful search, move the accessed item to the front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

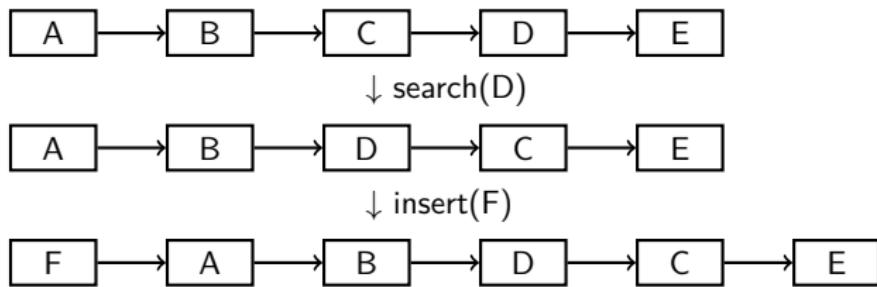
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## Performance of dynamic ordering:

- 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.