CS 240 – Data Structures and Data Management

Module 5: Other Dictionary Implementations

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Outline

- Dictionaries with Lists Revisited
 - Dictionary ADT
 - implementations so far
 - Skip Lists
 - Re-ordering items



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- Dictionaries with Lists Revisited
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Dictionary ADT: Implementations thus far

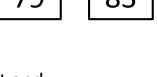
- A dictionary is a collection of key-value pairs (KVPs)
 - search, insert, and delete
- Realizations
 - Balanced search trees (AVL trees)
 - $\Theta(\log n)$ search, insert, and delete
 - complex code and not necessarily the fastest running time in practice
 - Binary search trees
 - $\Theta(height)$ search, insert and delete
 - simpler than AVL tree
 - randomization helps efficiency
 - Ordered array
 - simple implementation
 - $\Theta(\log n)$ search
 - $\Theta(n)$ insert and delete
 - Ordered linked list
 - simple implementation
 - $\Theta(n)$ search, insert and delete
 - search is the bottleneck, insert and delete would be $\Theta(1)$ if do search first and account for its running time separately

37

65

69

efficient search (like binary search) in ordered linked list?



Outline

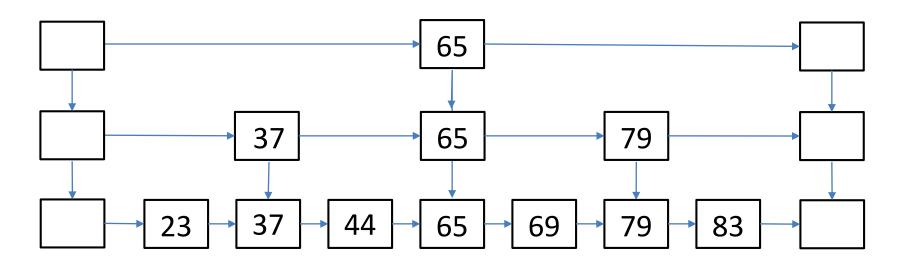
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Ordered array has efficient binary search

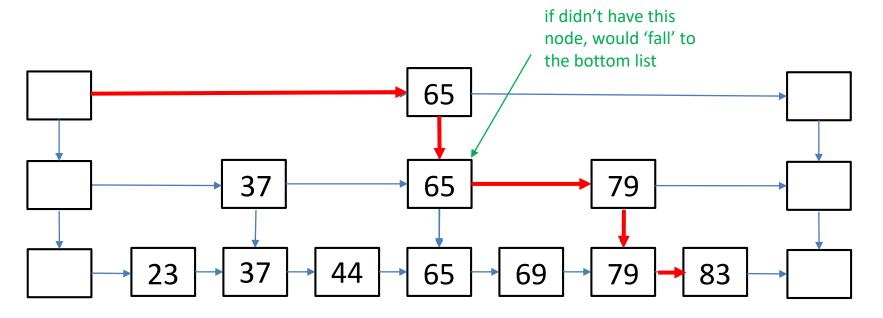
0	1	2	3	4	5	6
23	37	44	65	69	79	83

Can we imitate binary search in an ordered linked list?



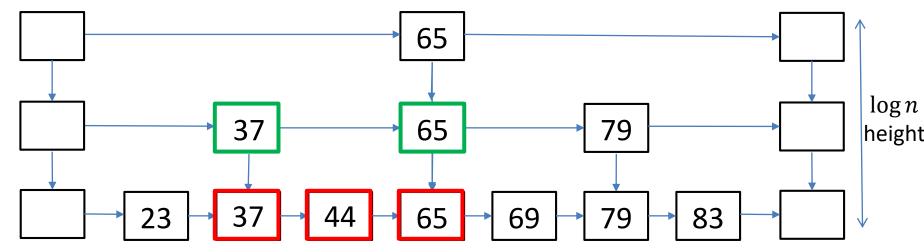


Search(83)





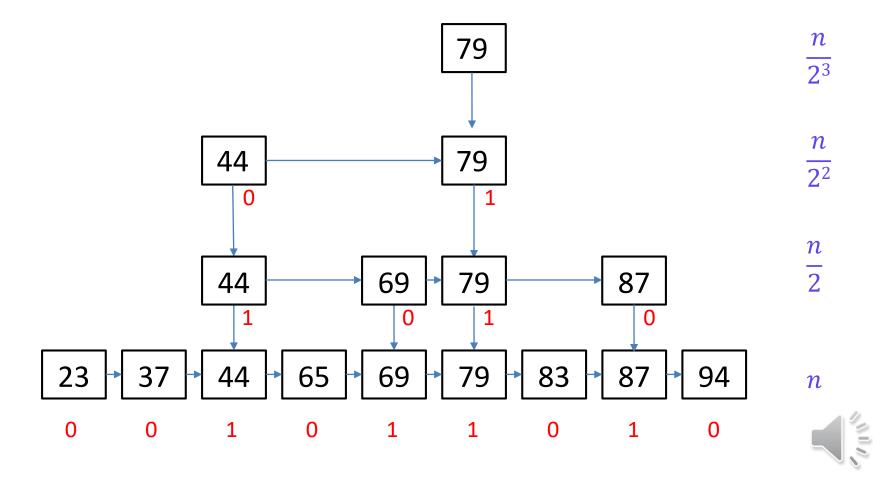
- Imitating binary search with a *hierarchy* of linked lists
 - build from bottom to top, each higher up list has 1/2 of previous list items
 - $\log n$ height (total number of linked lists needed)



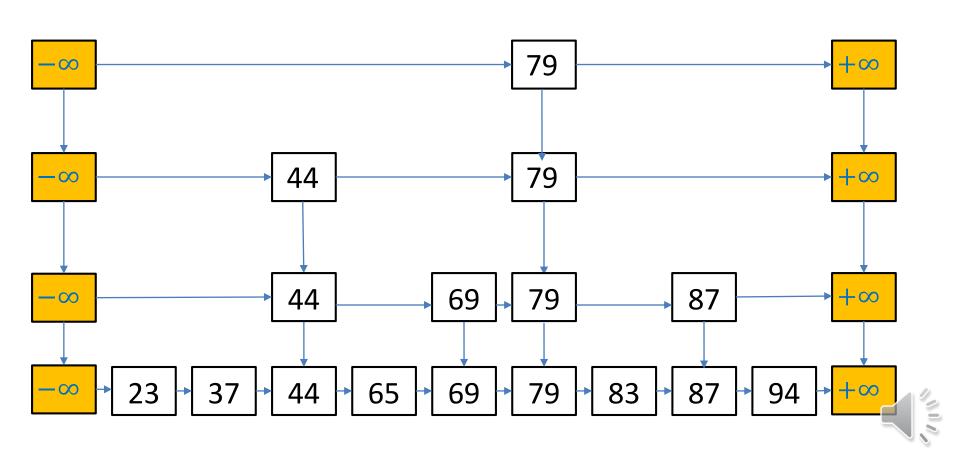
- When searching, go through the highest level possible
 - thus visit at most two items at each level
- Easy to implement if data structure is static
 - know all items beforehand, no need to insert or delete, but in static case an ordered array will work, and is more efficient (no links)
- To enable insert and delete, use randomization

- For next level, choose each item from previous level with probability ½ (coin toss)
- *i*th list is expected to have $n/2^i$ nodes
- Expect about log(n) lists in total

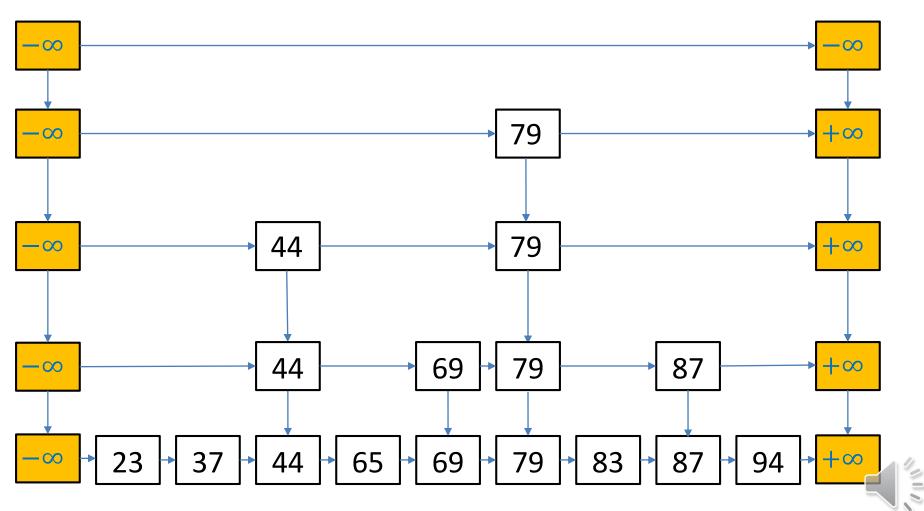
expected number of nodes



- Insert 'boundary' nodes with special sentinel symbols $-\infty$ and $+\infty$
 - to simplify code for searching

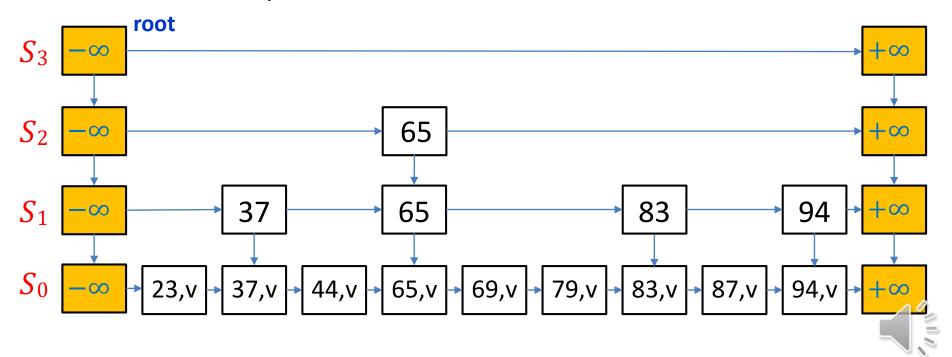


- Insert sentinel only level, with only $-\infty$ and $+\infty$
 - to simplify code for searching



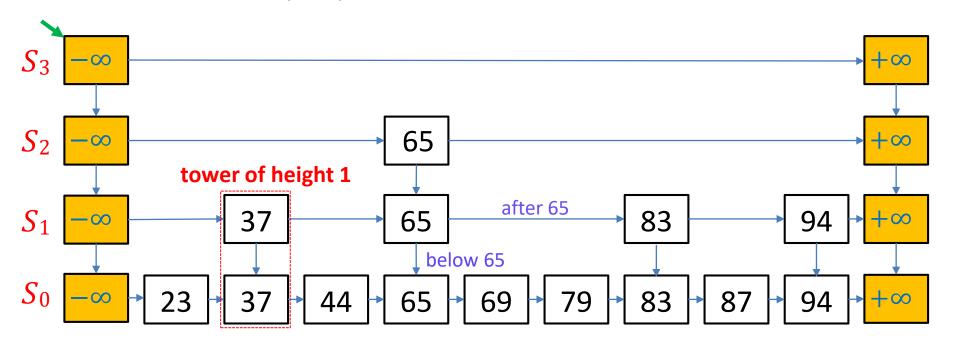
Skip Lists [Pugh'1989]

- A hierarchy S of ordered linked lists (*levels*) $S_0, S_1, ..., S_h$
 - S_0 contains the KVPs of S in non-decreasing order
 - other lists store only keys, or links to nodes in S_0
 - each S_i contains special keys (sentinels) $-\infty$ and $+\infty$
 - each S_i is randomly generated subsequence of S_{i-1} i.e., $S_0 \supseteq S_1 \supseteq ... \supseteq S_h$
 - S_h contains only sentinels, the left sentinel is the root



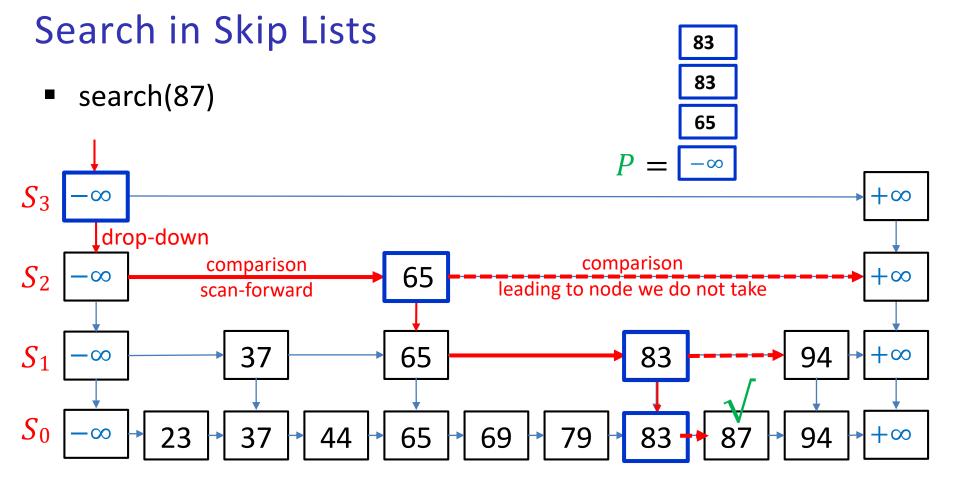
Skip Lists [Pugh'1989]

Will show only keys from now on



- Each KVP belongs to a tower of nodes
- Height of the skip list is the maximum height of any tower
- Each node p has references to after(p) and below(p)
- There are (usually) more nodes than keys





- For each level, **predecessor** of key k is the node before node with key k, or, if key k is not present at that level, the node before where k would be
- P collects predecessors of key k at level S_0 , $S_{1,...}$
 - these are needed for insert/delete
- k is in skip list if and only if P. top(). after has key k



Search in Skip Lists

```
getPredecessors(k)
         p \leftarrow root
         P \leftarrow stack of nodes, initially containing p
         while p. below \neq NIL do // keep dropping down until reach S_0
             p \leftarrow p. below
             while p, after, key < k do
                     p \leftarrow p. after //move to the right
              P.push(p)
                             // this is next predecessor
         return P
```

```
skipList::search(k) \\ P \leftarrow getPredecessors(k) \\ top \leftarrow P.top() \quad //predecessor of $k$ in $S_0$ \\ if $top. after. key = k$ return $top. after$ \\ else return 'not found, but would be after $top'$
```

Insert in Skip Lists

```
S_3 if in S_2, then insert new item with probability ½ S_2 if in S_1, then insert new item with probability ½ S_1 insert new item with probability ½ insert new item
```

- Keep "tossing a coin" until T appears
- Insert into S_0 and as many other S_i as there are heads
- Examples
 - H, H, T (insert into S_0, S_1, S_2) \Rightarrow will say i = 2
 - H,T (insert into S_0, S_1) \Rightarrow will say i=1
 - T (insert into S_0) \Rightarrow will say i = 0



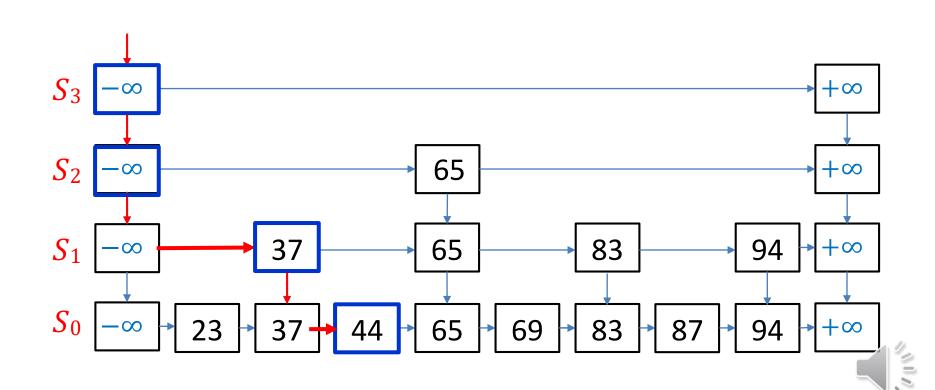
- skipList::insert(52, v)
- coin tosses: $H, T \Rightarrow i = 1$
- getPredecessors(52)

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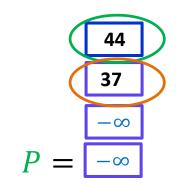
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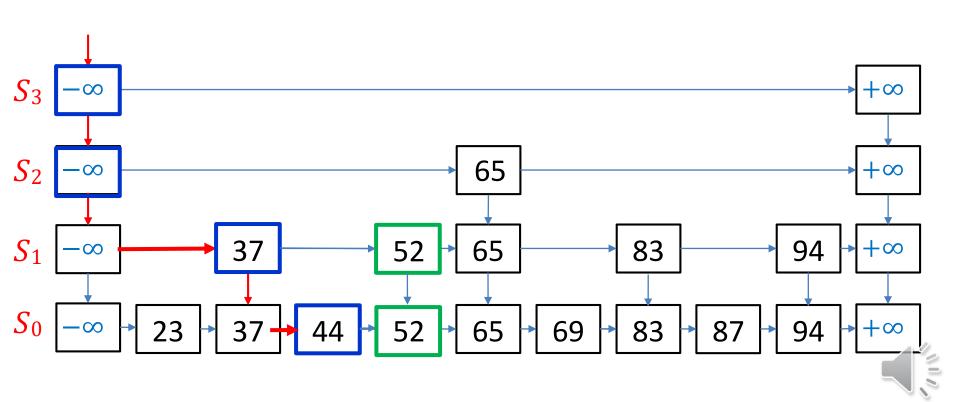
 $-\infty$

P = | -∞

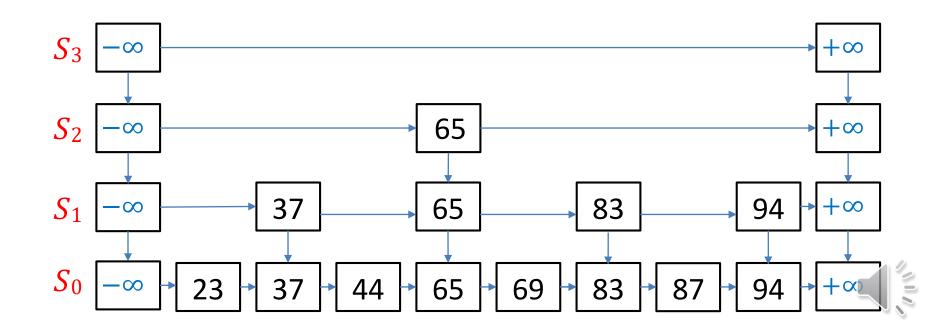


- skipList::insert(52, v)
- coin tosses: $H, T \Rightarrow i = 1$
- getPredecessors(52)
- now insert into S_0 and S_1

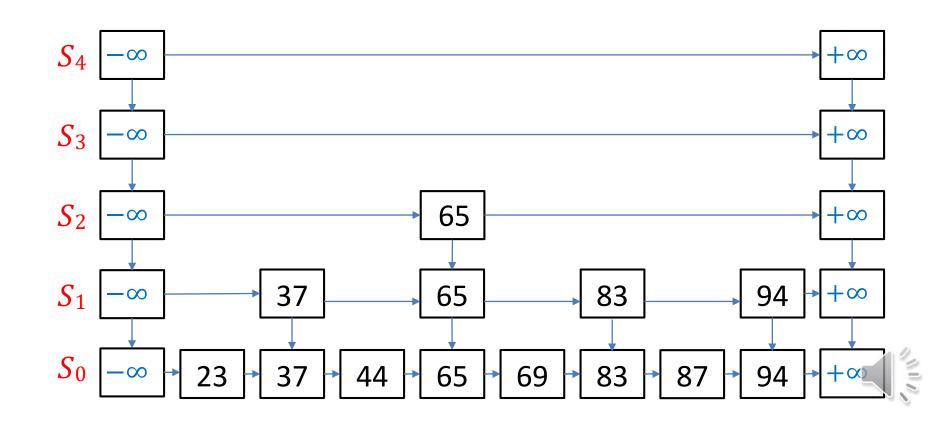




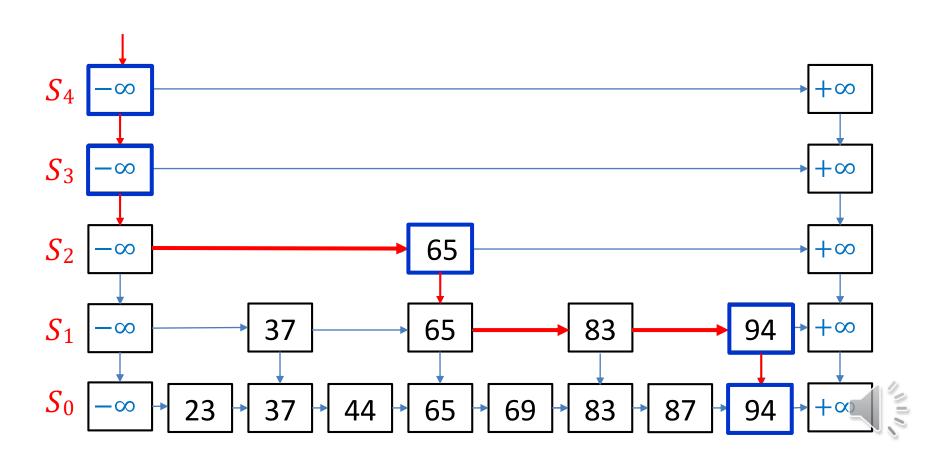
- skipList::insert(100, v)
- coin tosses: $H, H, H, T \Rightarrow i = 3$
- first increase height



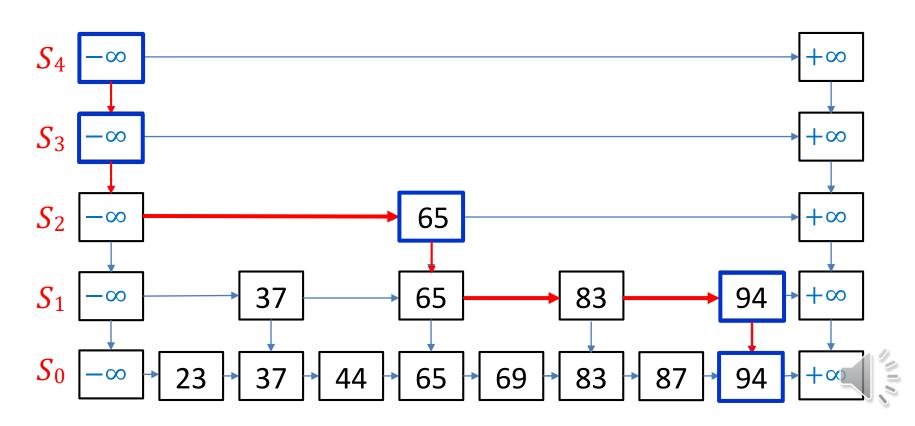
- skipList::insert(100, v)
- coin tosses: $H, H, H, T \Rightarrow i = 3$
- first increase height
- next *getPredecessors* (100)



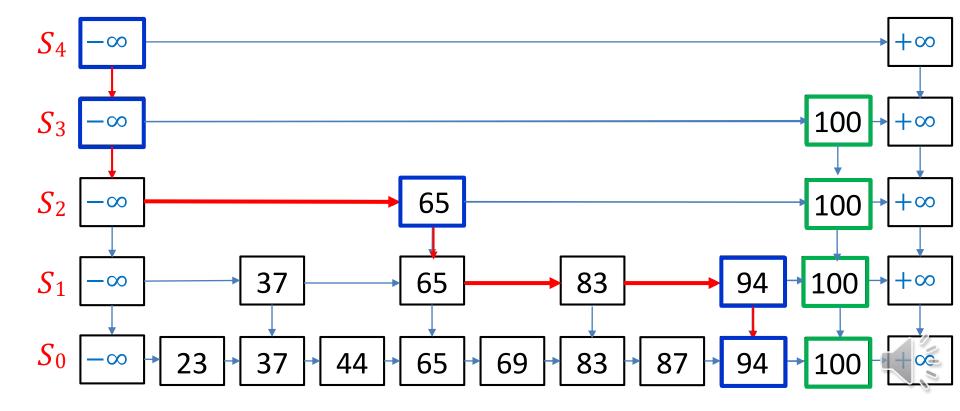
- skipList::insert(100, v)
- coin tosses: $H, H, H, T \Rightarrow i = 3$
- first increase height
- next *getPredecessors* (100)



- skipList::insert(100, v)
- coin tosses: $H, H, H, T \Rightarrow i = 3$
- first increase height
- next *getPredecessors* (100)
- insert new key



- skipList::insert(100, v)
- coin tosses: $H, H, H, T \Rightarrow i = 3$
- first increase height
- next getPredecessors (100)
- insert new key

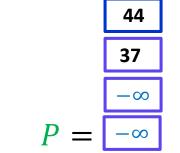


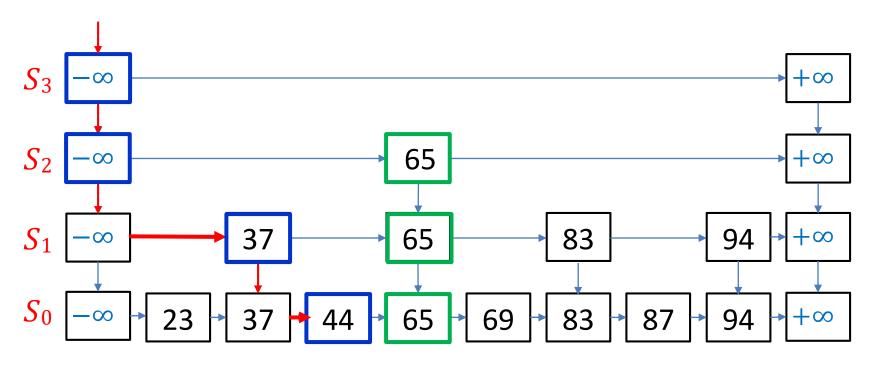
Insert in Skip Lists

```
skipList::insert(k, v)
          P \leftarrow getPredecessors(k)
          for (i \leftarrow 0; random(2) = 1; i \leftarrow i + 1) {} // random tower height
                                                                   // increase skip-list height?
          while i \geq P. size()
               root \leftarrow new sentinel-only list linked in appropriately
               P.append(left sentinel of root)
           p \leftarrow P.pop()
                                                                    // insert (k, v) in S_0
           zBellow \leftarrow \text{new node with } (k, v) \text{ inserted after } p
           while i > 0
                                                                   // insert k in S_1 S_2,..., S_k
                p \leftarrow P.pop()
                z \leftarrow new node with k added after p
                z.below \leftarrow zBellow
                zBellow \leftarrow z
               i \leftarrow i - 1
```

Example: Delete in Skip Lists

- skipList::delete(65)
 - first *getPredecessors*(*S*, 65)
 - then delete key 65 from all S_i
 - P has predecessor of each node to be deleted

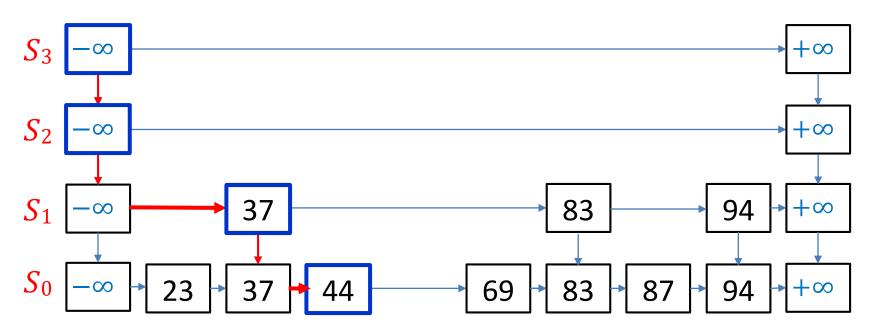






Example: Delete in Skip Lists

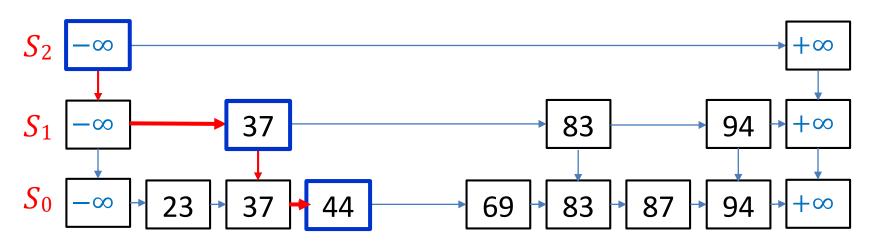
- skipList::delete(65)
 - first *getPredecessors*(*S*, 65)
 - then delete key 65 from all S_i
 - P has predecessor of each node to be deleted
 - height decrease: delete all unnecessary S_i , if any





Example: Delete in Skip Lists

- skipList::delete(65)
 - first *getPredecessors*(*S*, 65)
 - then delete key 65 from all S_i
 - P has predecessor of each node to be deleted
 - height decrease: delete all unnecessary S_i , if any





Delete in Skip Lists

```
skipList::delete(k)
         P \leftarrow getPredecessors(k)
         while P is non-empty
                                                     // predecessor of k in some layer
                 p \leftarrow P.pop()
                 if p. after. key = k
                      p.after \leftarrow p.after.after
                                                     // no more copies of k
                 else break
          p \leftarrow \text{left sentinel of the root-list}
         while p. below. after is the \infty sentinel
            // the two top lists are both only sentinels, remove one
                                              // removes the second empty list
            p.below \leftarrow p.below.below
            p.after.below \leftarrow p.after.below.below
```



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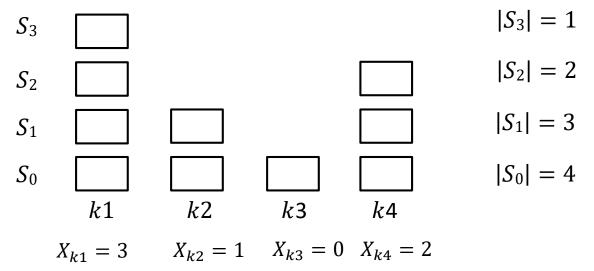
• Let X_k be the height of tower for key k

$$P(X_k \ge 1) = \frac{1}{2}, \ P(X_k \ge 2) = \frac{1}{2} \cdot \frac{1}{2}, \ P(X_k \ge 3) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2}$$

In general
$$P(X_k \ge i) = P(H \ H \ \dots \ H) = \left(\frac{1}{2}\right)^l$$
 $i \text{ times}$

- In the worst case, the height of a tower could be arbitrary large
 - no bound on height in terms of n
- Therefore operations could be arbitrarily slow, and space requirements arbitrarily large
- But this is exceedingly unlikely
- Therefore we analyse expected run-time and space-usage





- Let X_k be the height of tower for key k, we know $P(X_k \ge i) = \frac{1}{2^i}$
- If $X_k \ge i$ then list S_i includes key k
- Let $|S_i|$ be the number of keys in list S_i
 - sentinels do not count towards the length
 - S_0 always contains all n keys



$$S_3$$
 $I_{3,k1} = 1$ $I_{3,k2} = 0$ $I_{3,k3} = 0$ $I_{3,k4} = 0$
 S_2 $I_{2,k1} = 1$ $I_{2,k2} = 0$ $I_{2,k3} = 0$ $I_{2,k4} = 1$
 S_1 $I_{1,k1} = 1$ $I_{1,k2} = 1$ $I_{1,k3} = 0$ $I_{1,k4} = 1$
 S_0 $I_{1,k4} = 1$

- Let X_k be the height of tower for key k, we know $P(X_k \ge i) = \frac{1}{2^i}$
- If $X_k \ge i$ then list S_i includes key k
- Let $|S_i|$ be the number of keys in list S_i
 - sentinels do not count towards the length

$$\blacksquare \quad \text{Let} \quad I_{i,k} = \begin{cases} 0 & \text{if} \quad X_k < i \\ 1 & \text{if} \quad X_k \ge i \end{cases}$$

$$\bullet |S_i| = \sum_{k \in \mathcal{Y}} I_{i,k}$$



$$S_3$$
 $I_{3,k1} = 1$ $I_{3,k2} = 0$ $I_{3,k3} = 0$ $I_{3,k4} = 0$
 S_2 $I_{2,k1} = 1$ $I_{2,k2} = 0$ $I_{2,k3} = 0$ $I_{2,k4} = 1$
 S_1 $I_{1,k1} = 1$ $I_{1,k2} = 1$ $I_{1,k3} = 0$ $I_{1,k4} = 1$
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- Let X_k be the height of tower for key k, we know $P(X_k \ge i) = \frac{1}{2^i}$
- Let $|S_i|$ be the number of keys in list S_i

$$\blacksquare \quad \text{Let} \quad I_{i, k} = \begin{cases} 0 & \text{if} \quad X_k < i \\ 1 & \text{if} \quad X_k \ge i \end{cases}$$

- $|S_i| = \sum_{k \in \mathcal{Y}} |S_i| |S_i| = \sum_{k \in \mathcal{Y}} |S_i| |S$
- $E[|S_i|] = E\left[\sum_{key \ k} I_{i,k}\right] = \sum_{key \ k} E[I_{i,k}] = \sum_{key \ k} P(I_{i,k} = 1) = \sum_{key \ k} P(X_k \ge i) = \frac{n}{2^i}$
- The expected length of list S_i is $\frac{n}{2^i}$

 $|S_i|$ is number of keys in list S_i

$$\bullet \quad E[|S_i|] = \frac{n}{2^i}$$

Let
$$I_i = \begin{cases} 0 & \text{if } |S_i| = 0 \\ 1 & \text{if } |S_i| \ge 1 \end{cases}$$

$$S_4$$
 has only sentinels

$$S_3$$

 S_2

 S_1

 S_0

k2

$$I_2 = 1$$

 $I_4 = 0$

 $I_3 = 1$

*k*3

$$I_1 = 1$$

- $h = 1 + \sum_{i>1} I_i$ (here +1 is for the sentinel-only level)
- Since $I_i \leq 1$ we have that $E[I_i] \leq 1$
- Since $I_i \leq |S_i|$ we have that $E[I_i] \leq E[|S_i|] = \frac{\pi}{2i}$
- For ease of derivation, assume n is a power of 2

■
$$E[h] = E\left[1 + \sum_{i \ge 1} I_i\right] = 1 + \sum_{i \ge 1}^{\infty} E[I_i] = 1 + \sum_{i = 1}^{\log n} E[I_i] + \sum_{i = 1 + \log n}^{\infty} E[I_i]$$

$$\leq 1 + \sum_{i = 1}^{\log n} 1 + \sum_{i = 1 + \log n}^{\infty} \frac{n}{2^i}$$

$$\leq 1 + \log n + \sum_{i = 0}^{\infty} \frac{n}{2^{i+1 + \log n}}$$



 S_4 has only sentinels

$$I_4 = 0$$

 S_3

 $I_3 = 1$

• $|S_i|$ is number of keys in Liet S.

$$\bullet \quad E[|S_i|] = \frac{n}{2^i}$$

Let
$$I_i = \begin{cases} 0 & \text{if } |S_i| = 0 \\ 1 & \text{if } |S_i| \ge 0 \end{cases}$$

•
$$h = 1 + \sum_{i>1} I_i$$
 (here +

- Since $I_i \leq 1$ we have that
- Since $I_i \leq |S_i|$ we have

$$E[h] = E\left[1 + \sum_{i \ge 1} I_i\right] = \begin{bmatrix} S = \sum_{i=0}^{\infty} \frac{1}{2^i} \\ S = 2S - S = 2 \end{bmatrix}$$

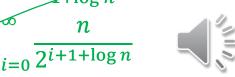
$$\sum_{i=0}^{\infty} \frac{n}{2^{i+1+\log n}} = \frac{1}{2} \sum_{i=0}^{\infty} \frac{n}{2^{i} 2^{\log n}}$$

$$=\frac{1}{2}\sum_{i=0}^{\infty}\frac{n}{2^{i}n}$$

$$=\frac{1}{2}\sum_{i=0}^{\infty}\frac{1}{2^i}=1$$

$$S = \sum_{i=0}^{\infty} \frac{1}{2^{i}} \qquad 2S = \sum_{i=0}^{\infty} \frac{1}{2^{i-1}} = 2 + \sum_{i=0}^{\infty} \frac{1}{2^{i}}$$

$$S = 2S - S = 2$$



 $|S_i|$ is number of keys in list S_i

$$\bullet \quad E[|S_i|] = \frac{n}{2^i}$$

Let $I_i = \begin{cases} 0 & \text{if } |S_i| = 0 \\ 1 & \text{if } |S_i| \ge 1 \end{cases}$

 S_4 has only sentinels

$$S_4$$
 has only sentinels

 S_3

 S_2

 S_1

 S_0

k2

*k*3

$$I_3 = 1$$

 $I_4 = 0$

$$I_2 = 1$$

$$I_1=1$$

- $h = 1 + \sum_{i \ge 1} I_i$ (here +1 is for the sentinel-only level)
- Since $I_i \leq 1$ we have that $E[I_i] \leq 1$
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- For ease of derivation, assume n is a power of 2

$$E[h] = E\left[1 + \sum_{i \ge 1} I_i\right] = 1 + \sum_{i \ge 1}^{\infty} E[I_i] = 1 + \sum_{i = 1}^{\log n} E[I_i] + \sum_{i = 1 + \log n}^{\infty} E[I_i]$$

$$\le 1 + \sum_{i = 1}^{\log n} 1 + \sum_{i = 1 + \log n}^{\infty} \frac{n}{2^i}$$



 $\leq 1 + \log n + 1$

Skip List Analysis: Expected Space

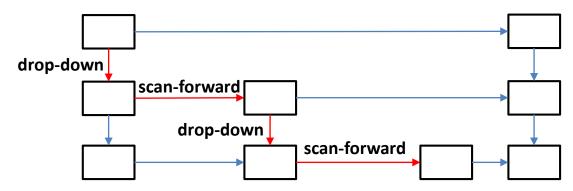
- We need space for nodes storing sentinels and nodes storing keys
- 1. Space for nodes storing sentinels
 - there are 2h + 2 sentinels, where h be the skip list height
 - $E[h] \leq 2 + \log n$
 - expected space for sentinels is at most

$$E[2h + 2] = 2E[h] + 2 \le 6 + 2\log n$$

- 2. Space for nodes storing keys
 - Let $|S_i|$ be the number of keys in list S_i
 - $\bullet \quad E[|S_i|] = \frac{n}{2^i}$
 - expected space for keys is $E\left|\sum_{i>0}|S_i|\right| = \sum_{i\geq0}\frac{n}{2^i} = 2n$
- Total expected space is $\Theta(n)$



Skip List Analysis: Expected Running Time



- search, insert, and delete are dominated by the running time of getPredecessors
- So let us analyze the expected time of getPredecessors
- In getPredecessors, running time is proportional to the number of 'drop-down' and 'scan-forward'
- We can 'drop-down' at most h times, where h is skip list height
 - expected height h is $O(\log n)$
 - total expected time spent on 'drop-down' operations is $O(\log n)$
- Will show on the next slide that the expected number of 'scan-forward' is also $O(\log n)$
- So the expected running time is $O(\log n)$

Skip List Analysis: Expected Running Time

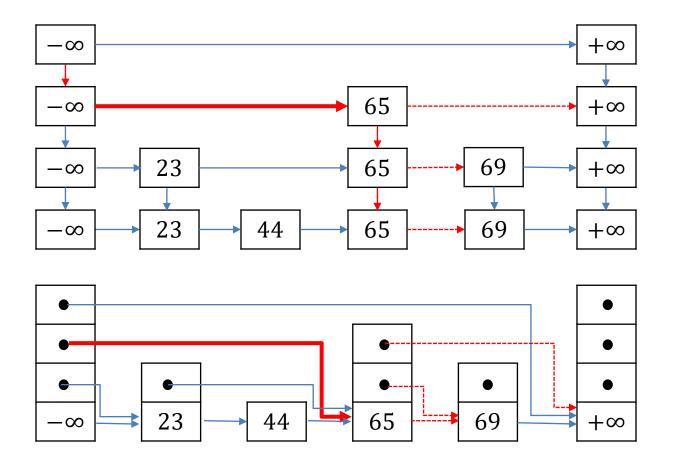
- What about 'scan-forward'?
 - assume i < h (if i = h, then we are at the top list and do not scan forward at all)
 - let v be leftmost key in S_i we visit during search
 - we v reached by dropping down from S_{i+1}
 - let w be the key right after v
 - height of tower of w in this case is at least i
 - What is the probability of scanning from v to w?
 - If we do scan forward from v to w, then w did not exist in S_{i+1}
 - otherwise, we would scan forward from v to w in S_{i+1}
 - in other words, we always enter the tower of any node 'at the top'
 - Thus if we do scan forward from v to w, then the tower of w has height i
 - $P(\text{tower of } w \text{ has height } i | \text{tower of } w \text{ has height at least } i) = \frac{1}{2}$
 - we scan forward from v to w with probability at most $\frac{1}{2}$
 - 'at most' because we could scan-down down if key < w
 - repeating the argument, the probability of scan-forward l times is at most $(1/2)^l$

$$E[\text{ number of scans}] = \sum_{l \ge 1} l \cdot P(\text{scans} = l) = \sum_{l \ge 1} P(\text{scans} \ge l) \le \sum_{l \ge 1} \frac{1}{2^l} = 1$$

Expected number of scan-forwards at any level is 1, over all levels h, which is $O(\log n)$

Arrays Instead of Linked Lists

- As described now, they are no faster than randomized binary search trees
- Can save links by implementing each tower as an array
 - this not only saves space, but gives better running time in practice
 - when 'scan-forward', we know the correct array location to look at (level i)
- Search(67)





Summary of Skip Lists

- For a skip list with n items
 - expected space usage is O(n)
 - expected running time for search, insert, delete is $O(\log n)$
- Two efficiency improvements
 - use arrays for key towers for more efficient implementation
 - can show: a biased coin-flip to determine tower-height gives smaller expected run-times
 - with arrays and biased coin-flip skip lists are fast in practice and easy to implement



Outline

- Dictionaries with Lists Revisited
 - Dictionary ADT
 - implementations so far
 - Skip Lists
 - Re-ordering items



Re-ordering Items

- Unordered arrays (or lists) are among simplest data structures to implement
- But for Dictionary ADT
 - search: $\Theta(n)$, insert: $\Theta(1)$, delete: $\Theta(1)$ (after a search)
- Lists/arrays are a very simple a popular implementation
- Can we make search in unordered arrays (or lists) more effective in practice?
 - No: if items are accessed equally likely
 - Yes: otherwise
 - intuition: frequently accessed items should be in the front
 - Two cases
 - know the access distribution beforehand
 - do not know access distribution beforehand
 - For short lists or extremely unbalanced distributions this may be faster than
 AVL trees or Skip Lists, and easier to implement



Optimal Static Ordering

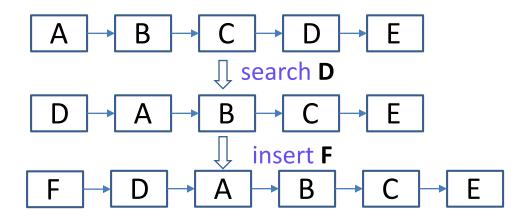
key	А	В	С	D	E
frequency of access	2	8	1	10	5
access probability	2 26	8 26	$\frac{1}{26}$	10 26	5 26

■ Order
$$C$$
 A B D E has expected cost $\frac{1}{26} \cdot 1 + \frac{2}{26} \cdot 2 + \frac{8}{26} \cdot 3 + \frac{10}{26} \cdot 4 + \frac{5}{26} \cdot 5 \approx 3.61$
■ Order D B E A C has expected cost $\frac{10}{26} \cdot 1 + \frac{8}{26} \cdot 2 + \frac{5}{26} \cdot 3 + \frac{2}{26} \cdot 4 + \frac{1}{26} \cdot 5 \approx 2.54$

- Claim: ordering items by non-increasing access-probability minimizes expected access cost, i.e. best static ordering
- Proof Idea: for any other ordering, exchanging two items that are out-of-order according to access probabilities makes total cost decrease

Dynamic Ordering

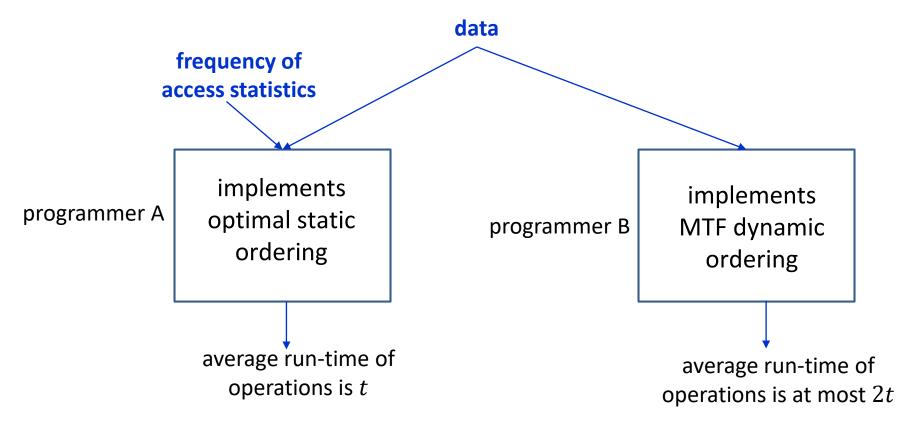
- What if we do not know the access probabilities ahead of time?
- Rule of thumb (temporal locality)
 - recently accessed item is likely to be accessed soon again
- In list: always insert at the front
- Move-To-Front heuristic (MTF): after search, move the accessed item to the front



- 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: MTF

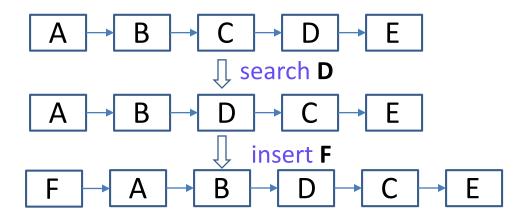
- Can show: MTF is "2-competitive"
 - no more than twice as bad as the optimal "offline" ordering





Dynamic Ordering: Transpose

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



- Avoids drastic changes MTF might do, while still adapting to access patterns
- Worst case is $\Theta(n)$ for both transpose and MTF

