

# 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

- Dictionaries with Lists Revisited
  - Dictionary ADT
    - implementations so far
  - Skip Lists
  - Biased Search Requests

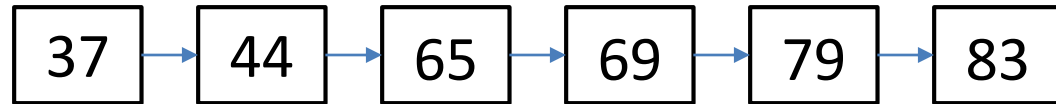
# Outline

- Dictionaries with Lists Revisited
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  - Skip Lists
  - Biased Search Requests

# 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**
    - $\Theta(\text{height})$  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**



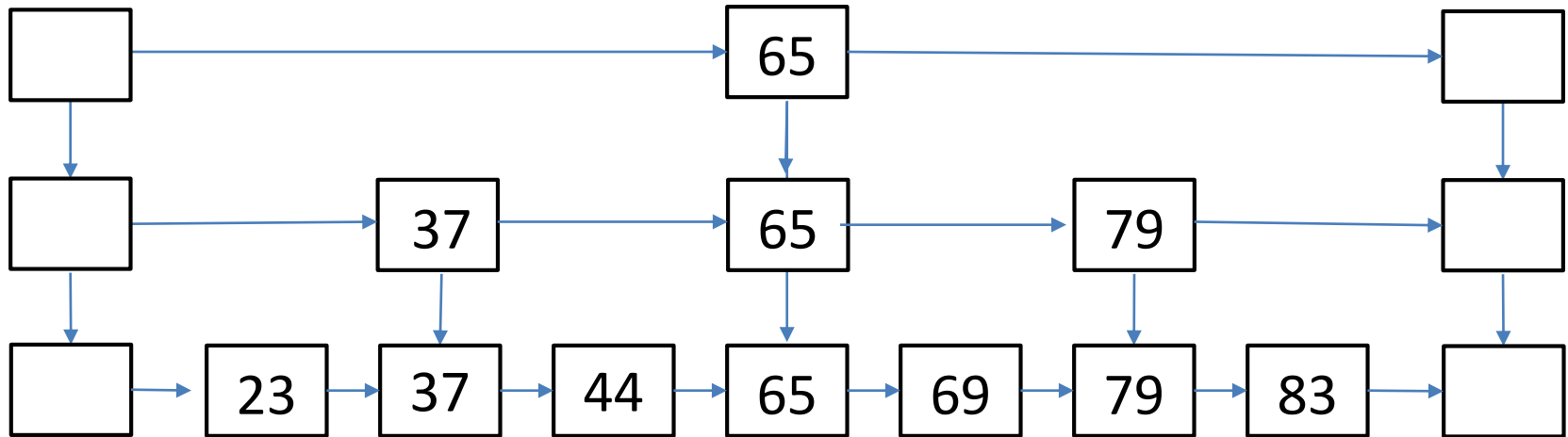
- simple implementation
- $\Theta(n)$  search, insert and delete
- search is the bottleneck, insert and delete would be  $\Theta(1)$  if do search first and account for its running time separately
- efficient search (like binary search) in ordered linked list?

# Outline

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

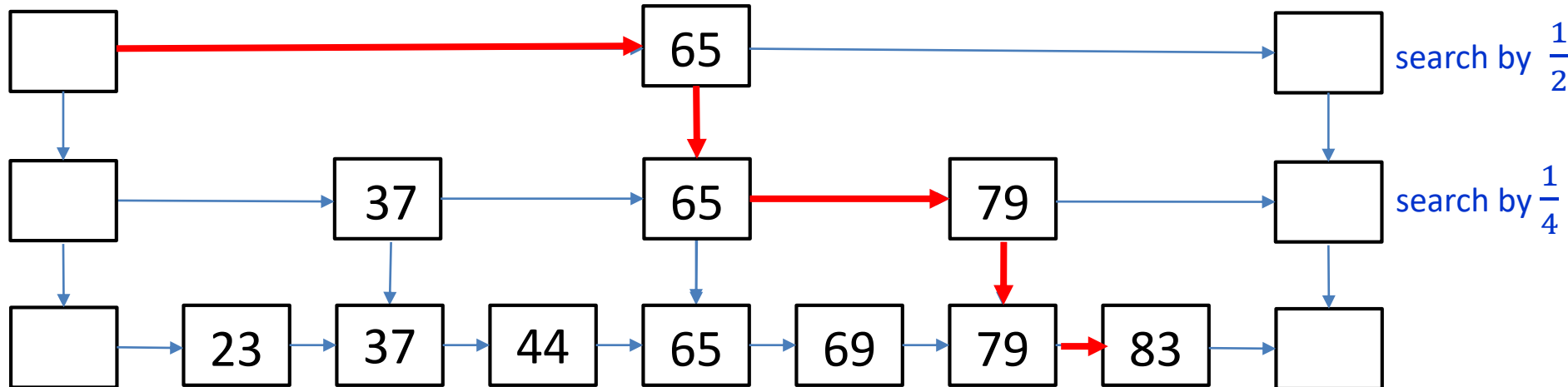
# Skip Lists: Motivation

- Build a *hierarchy* of linked lists to imitate binary search in ordered linked list with
  - start from the bottom list and take every second item in the list above
  - downward links are needed to navigate from list above to the list below



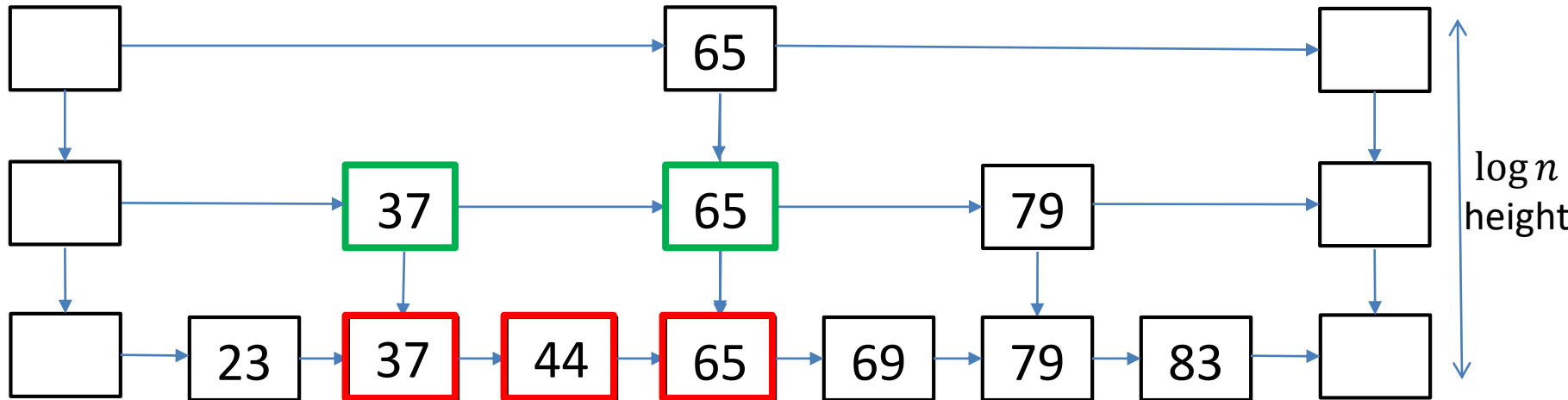
# Skip Lists: Motivation

- Search goes through the higher lists while possible, before dropping down to the list below
  - top list enables search by  $\frac{1}{2}$  of the list, next by  $\frac{1}{4}$  of the list, and so on
- Search(83)



# Skip Lists: Motivation

- Hierarchy of linked lists
  - each list has 1/2 of items from the list below
  - total number of linked lists (height) is  $\log n$
  - total number of nodes  $\leq 2n$

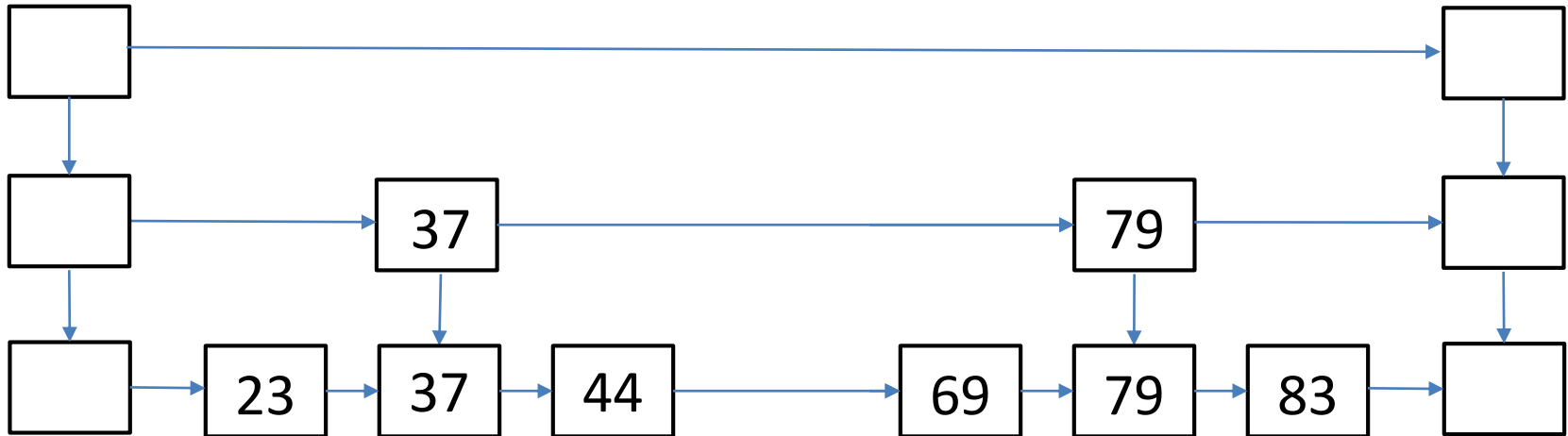


- When searching, go through the highest level possible
  - thus visit at most **two items** at each level, and total time to search  $\Theta(\log n)$



# Skip Lists: Motivation

- Deleted 65, no longer every second item is in the list above

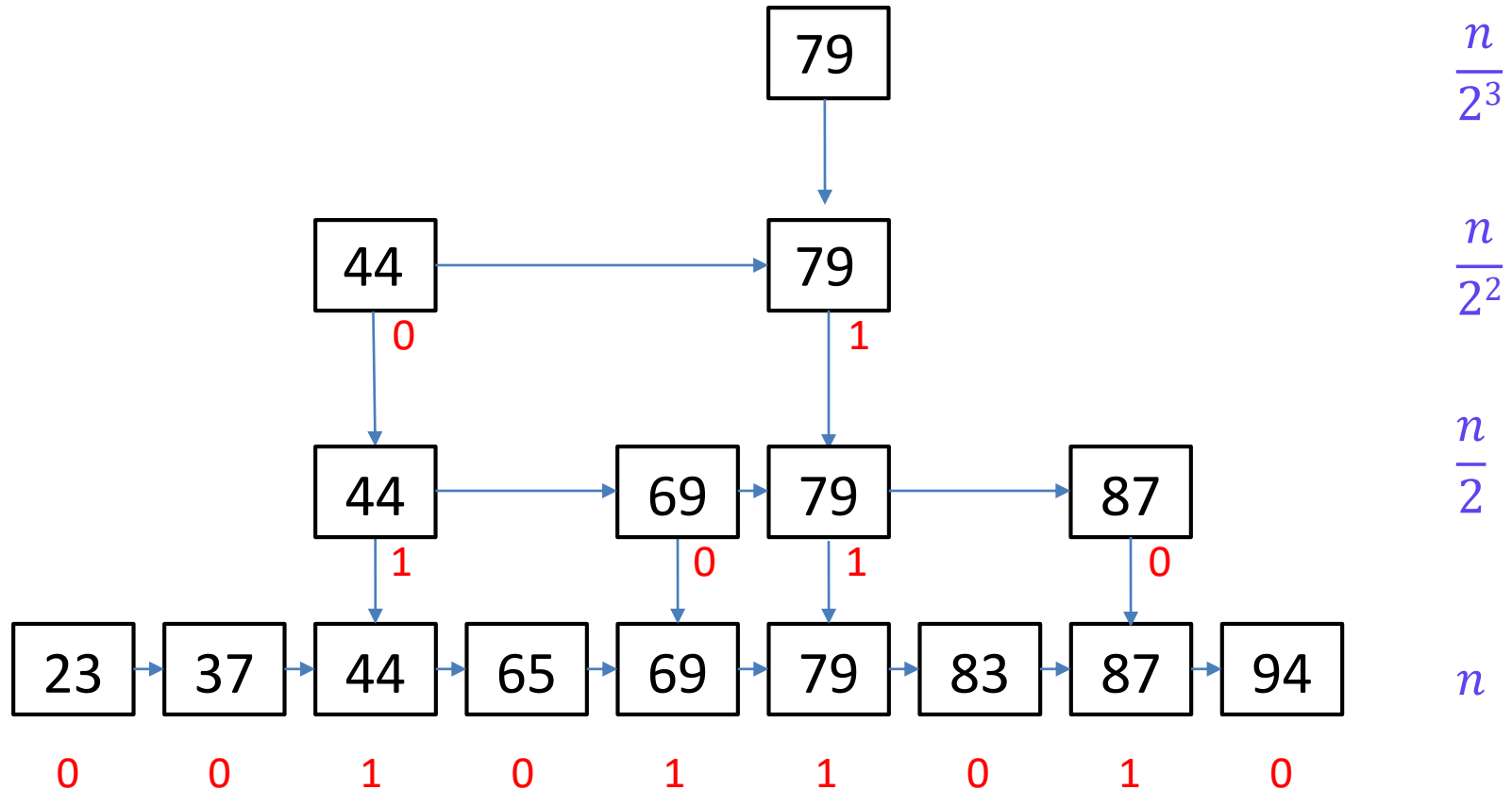


- Big problem:** deletion or insertion of items ruins ‘every second item is in the list above’ property
  - crucial property for efficiency
- Thus the hierarchy of linked lists works only for *static* dictionary
  - know all items beforehand, and **do not** insert or delete
    - but in static case an ordered array is more efficient in practice (no links)
- Randomization* enables hierarchical linked list **with** efficient insert and delete
  - instead of requiring a deterministic subset of items in list above, *randomly* chose a subset of the items in the list above

# Skip Lists: Motivation

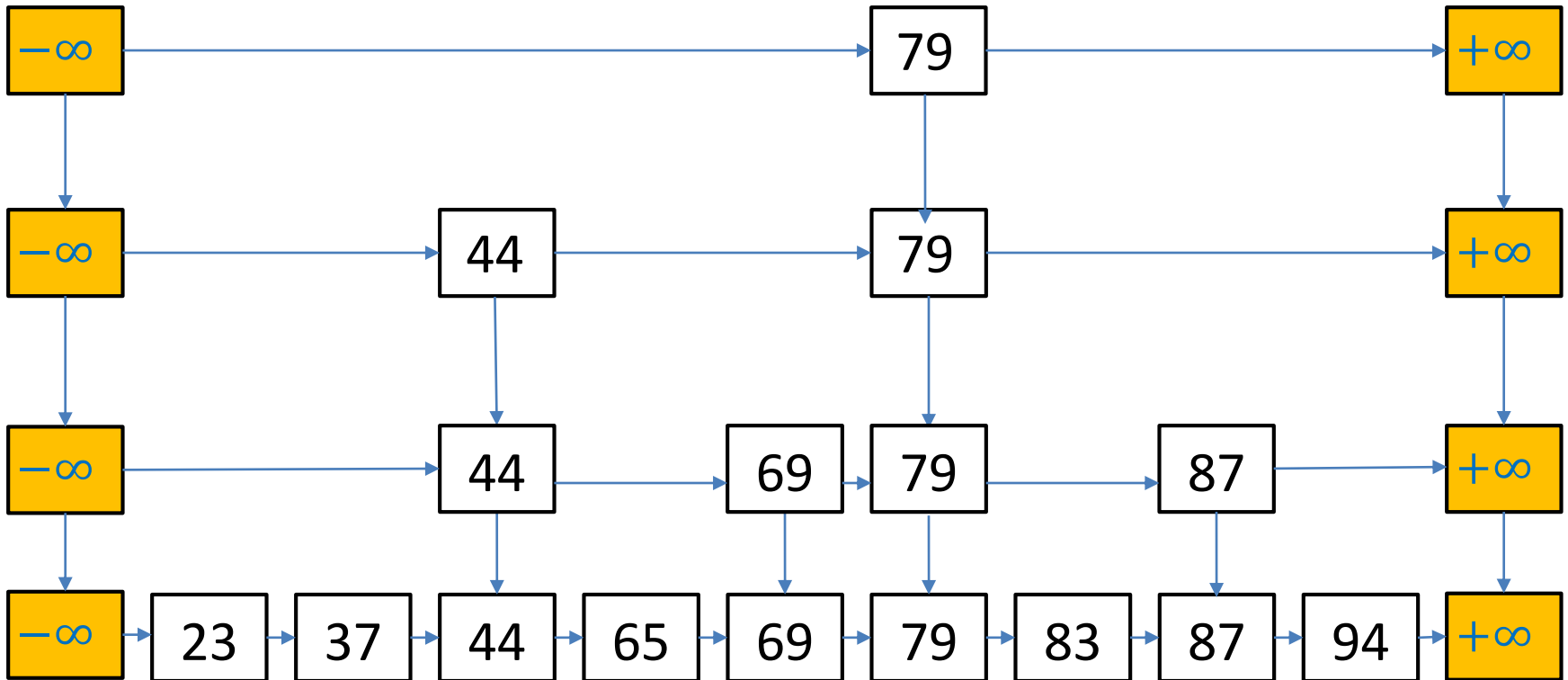
- For next level, choose each item from previous level with probability  $\frac{1}{2}$  (coin toss)
- $i$ th list is expected to have  $n/2^i$  nodes
- Expect about  $\log(n)$  lists in total

expected  
number of nodes



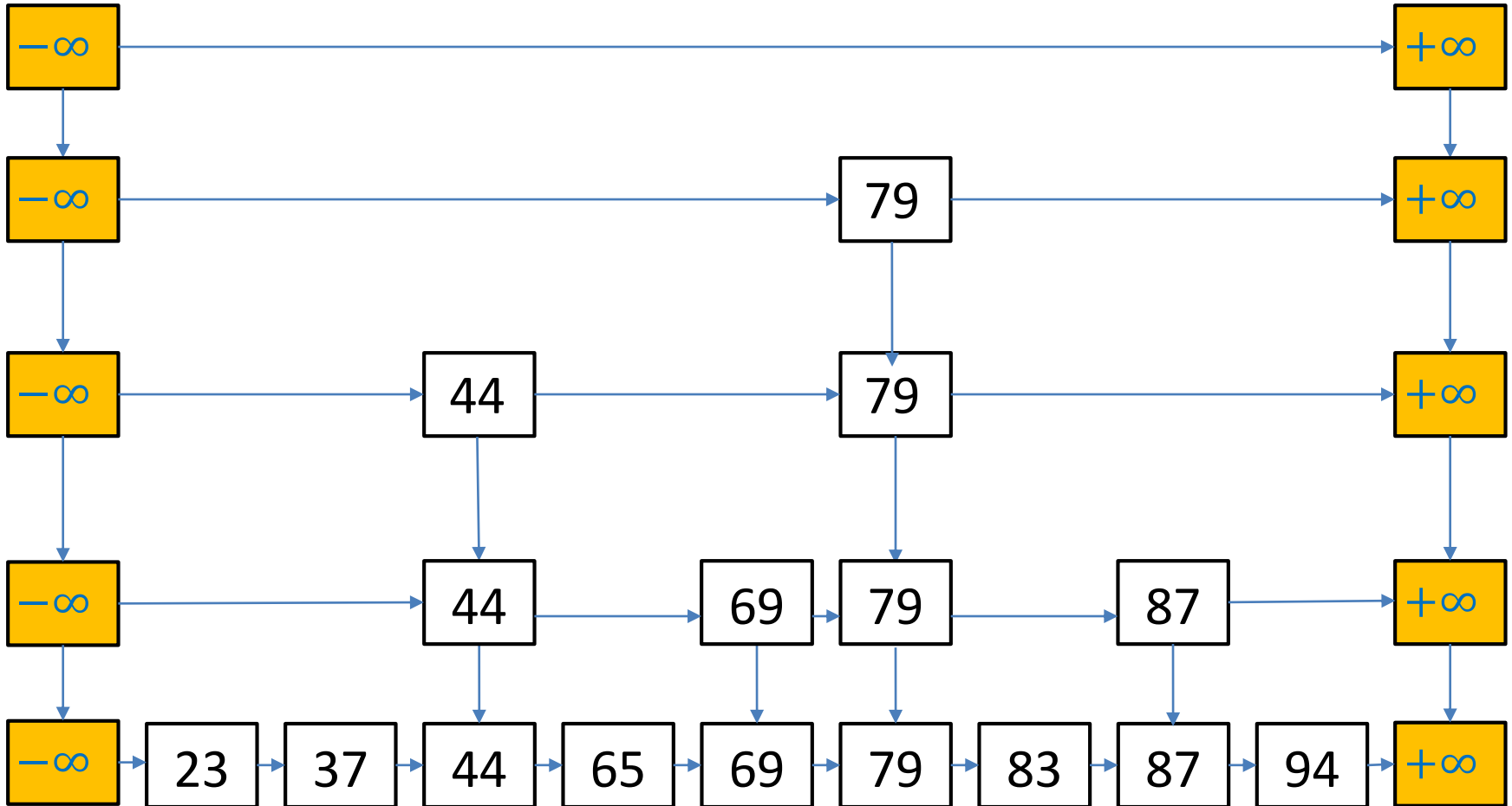
# Skip Lists: Motivation

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



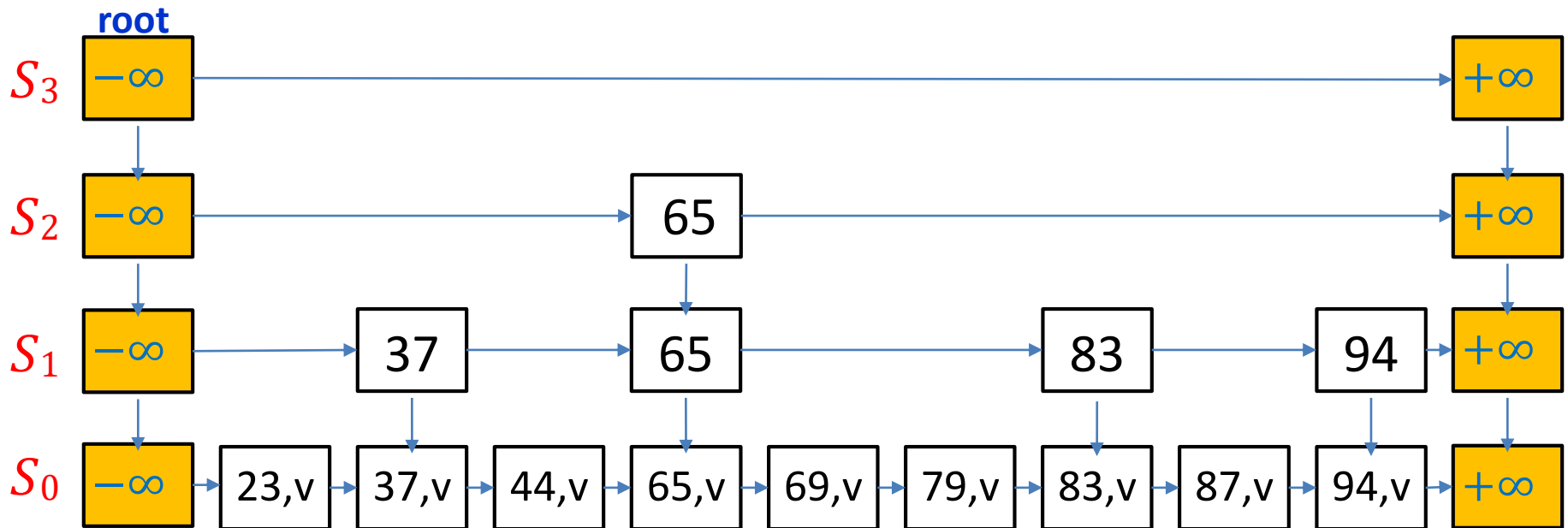
# Skip Lists: Motivation

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



# Skip Lists [Pugh'1989]

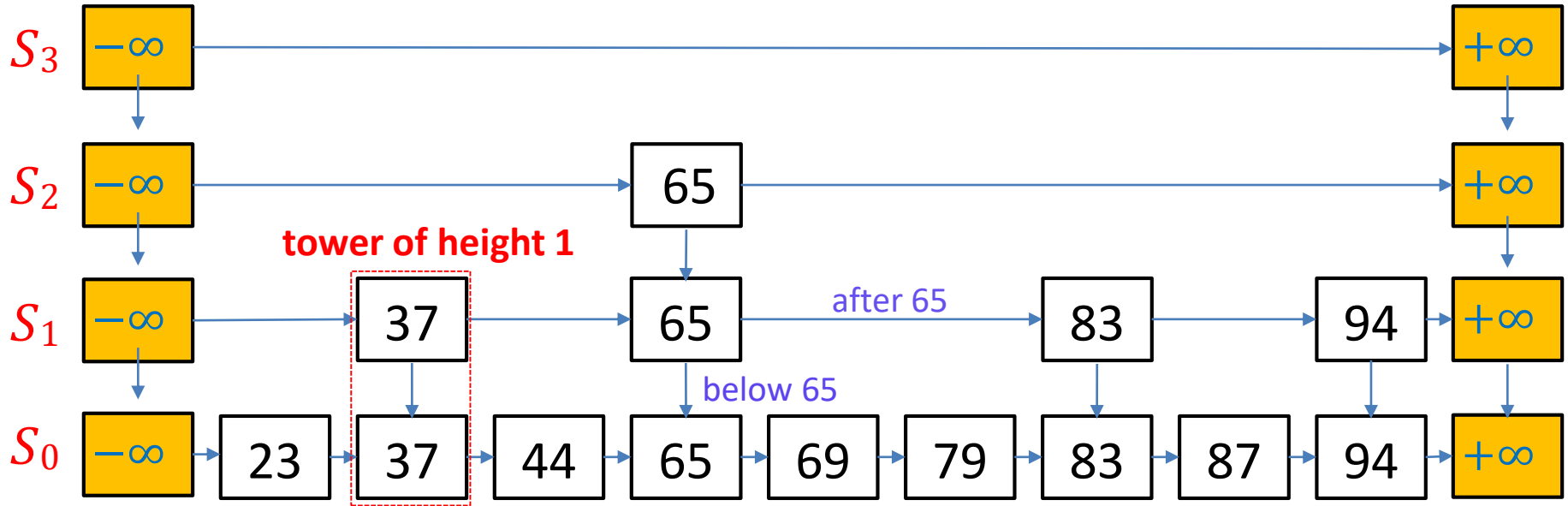
- A hierarchy  $S$  of ordered linked lists (*levels*)  $S_0, S_1, \dots, S_h$ 
  - $S_0$  contains the KVPs of  $S$  in non-decreasing order
  - other lists store only keys
  - each  $S_i$  contains special keys (sentinels)  $-\infty$  and  $+\infty$
  - each  $S_i$  is randomly generated subsequence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



- $n$  is number of KVP (number of items stored); in this example,  $n = 9$

# Skip Lists

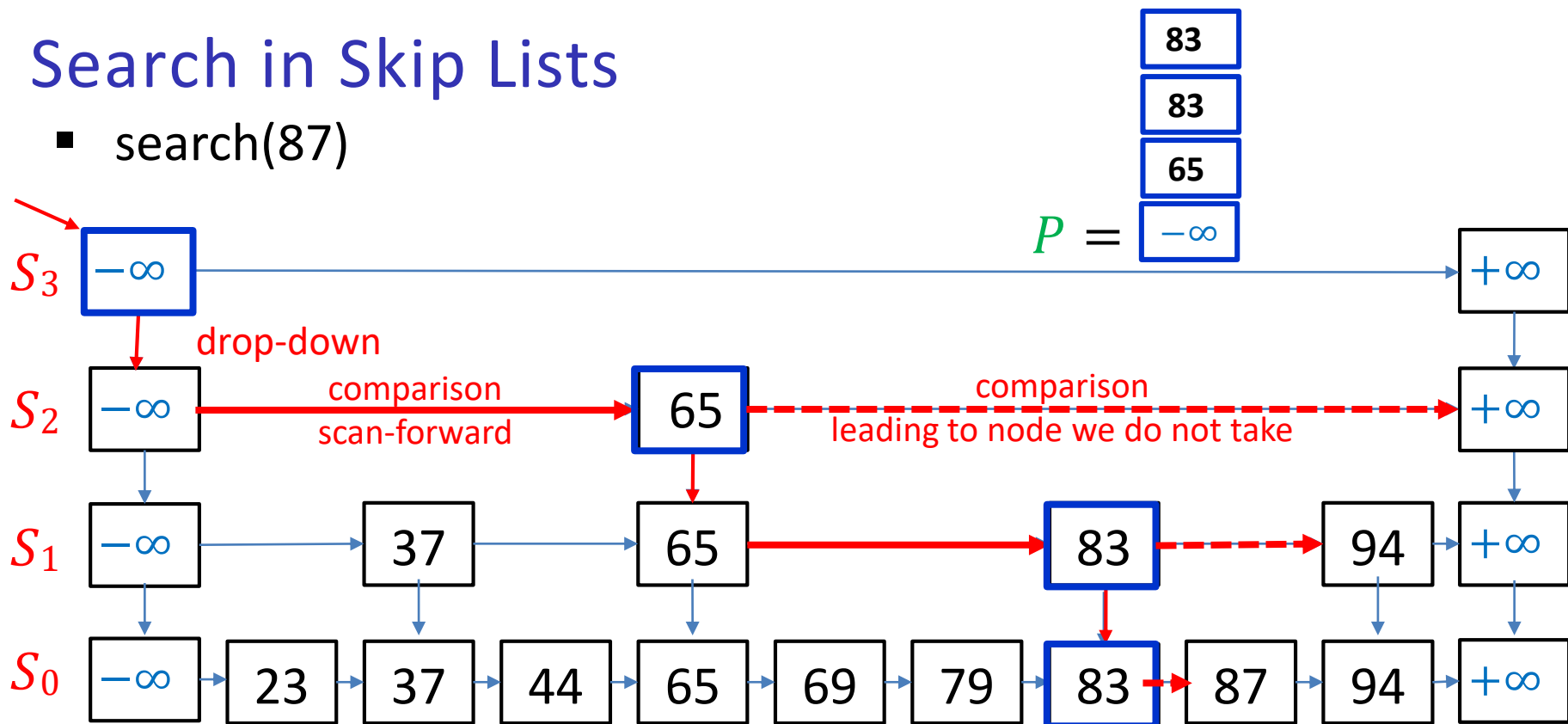
- Show only keys from now on



- Each KVP belongs to a *tower* of nodes
  - tower height is number of nodes – 1
- Height of the skip list is the maximum height of any tower
  - height is 3 in this example
- Each node  $p$  has references to  $\text{after}(p)$  and  $\text{below}(p)$
- There are (usually) more nodes than keys

# Search in Skip Lists

- search(87)



- For each level, **predecessor** of key  $k$  is
  - If key  $k$  is present at the level: node before node with key  $k$
  - if key  $k$  is not present at the level: node before node where  $k$  would have been
- $P$  collects predecessors of key  $k$  for all levels
  - nodes where we drop down and the rightmost node in  $S_0$  with key  $< k$
  - 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 ← root
```

```
  P ← stack of nodes, initially containing p
```

```
  while p.below ≠ NIL do // keep dropping down until reach  $S_0$ 
```

```
    p ← p.below
```

```
    while p.after.key < k do
```

```
      p ← p.after // move to the right
```

```
      P.push(p) // this is next predecessor
```

```
  return P
```

```
skipList::search(k)
```

```
  P ← getPredecessors(k)
```

```
  q ← P.top() // predecessor of k in  $S_0$ 
```

```
  if q.after.key = k return q.after
```

```
  else return 'not found, but would be after q'
```



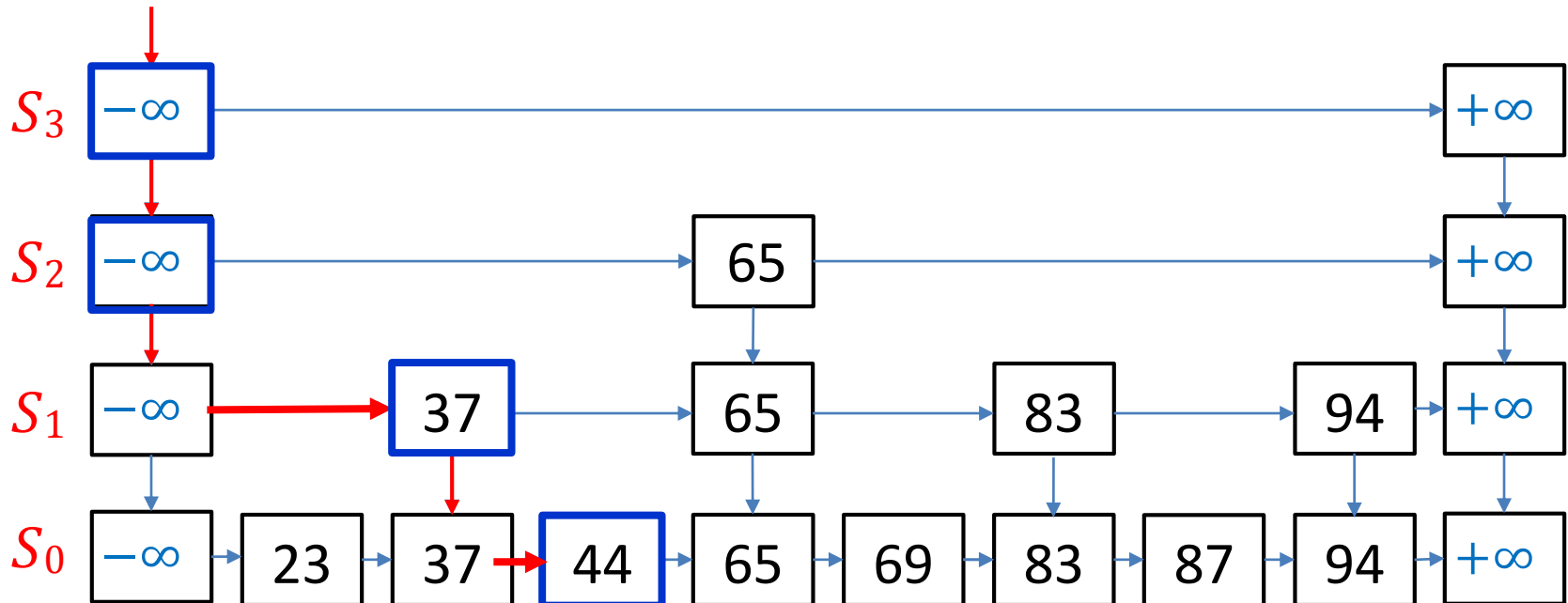
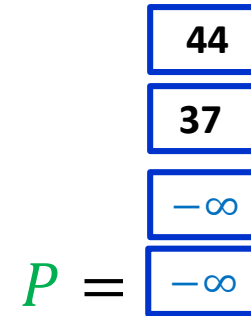
# Insert in Skip Lists

$S_3$  ← if in  $S_2$ , then insert new item with probability  $\frac{1}{2}$   
 $S_2$  ← if in  $S_1$ , then insert new item with probability  $\frac{1}{2}$   
 $S_1$  ← insert new item with probability  $\frac{1}{2}$   
 $S_0$  ← 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  $i = 0$

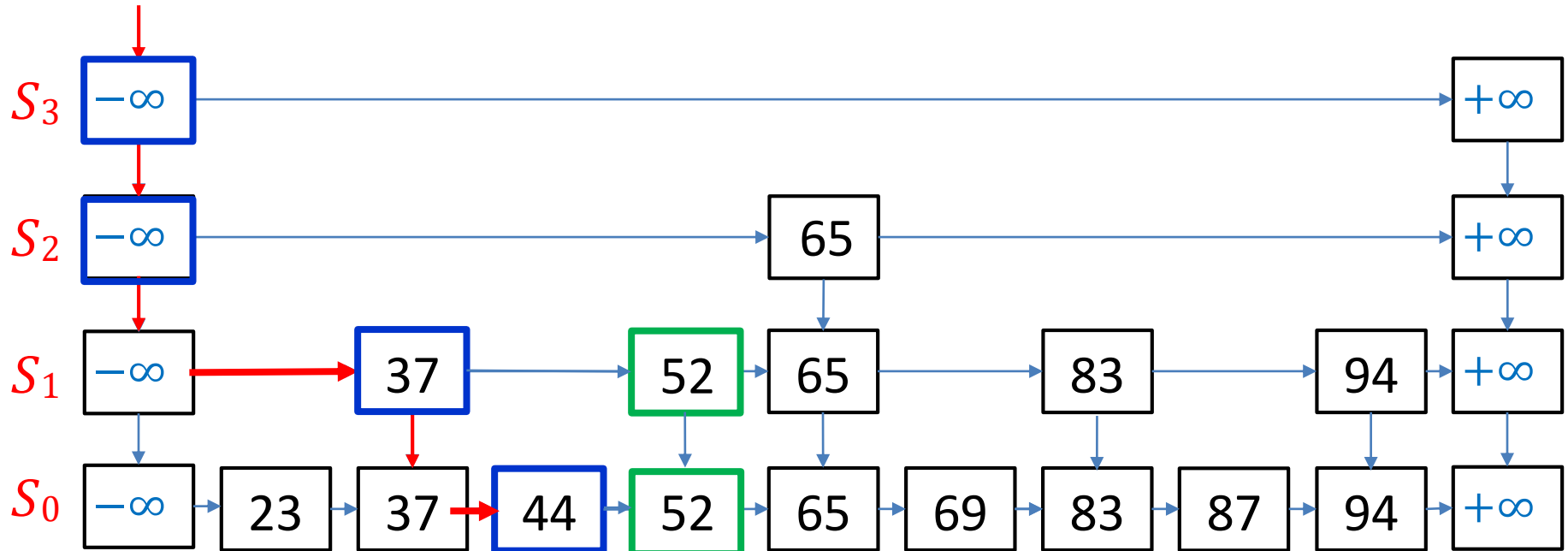
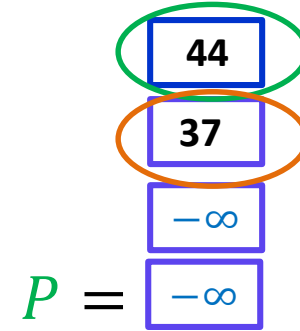
# Insert in Skip Lists: Example 1

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



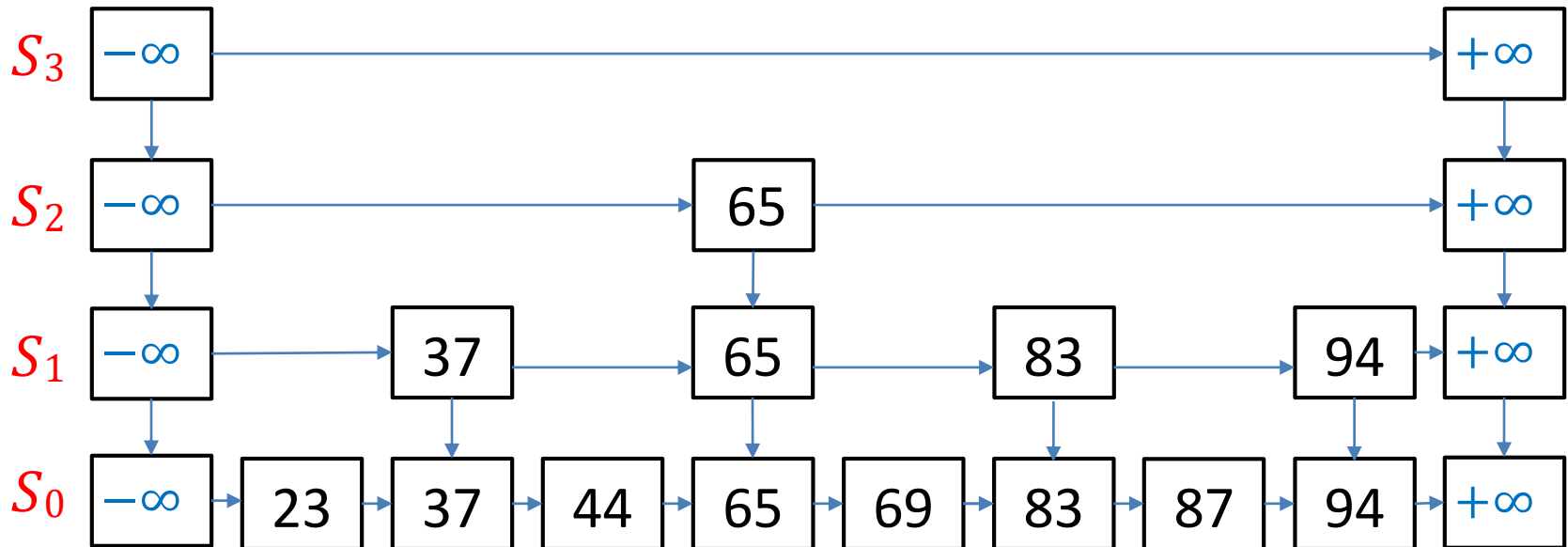
# Insert in Skip Lists: Example 1

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



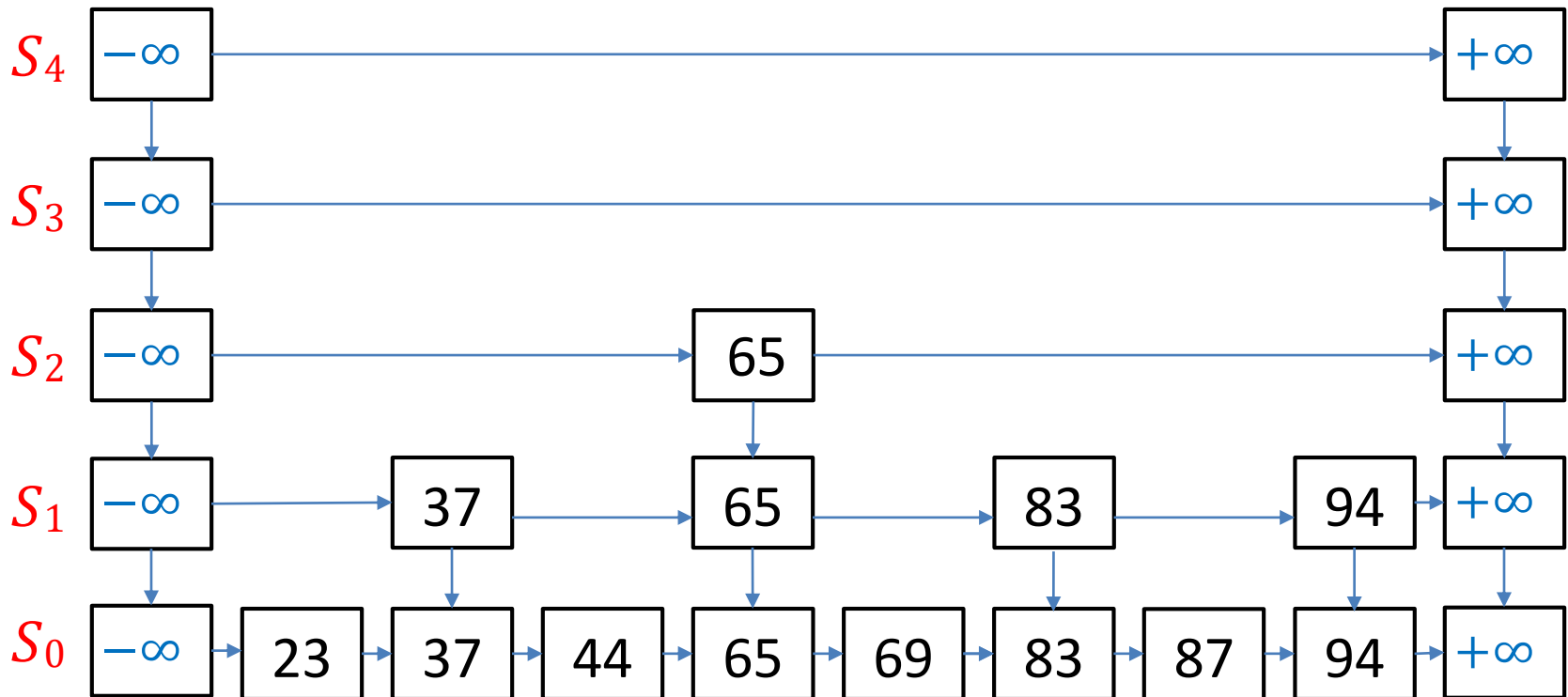
# Insert in Skip Lists: Example 2

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



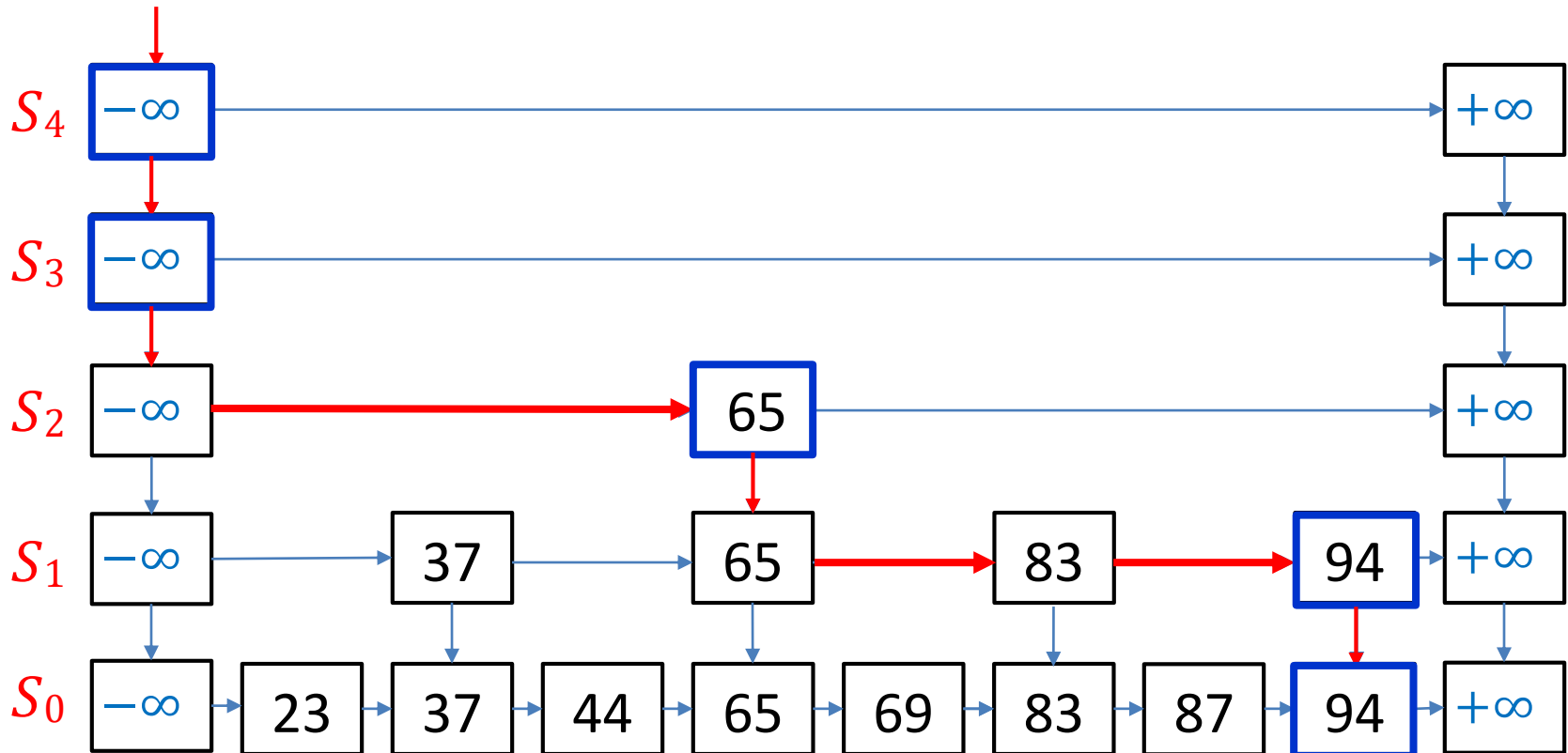
# Insert in Skip Lists: Example 2

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



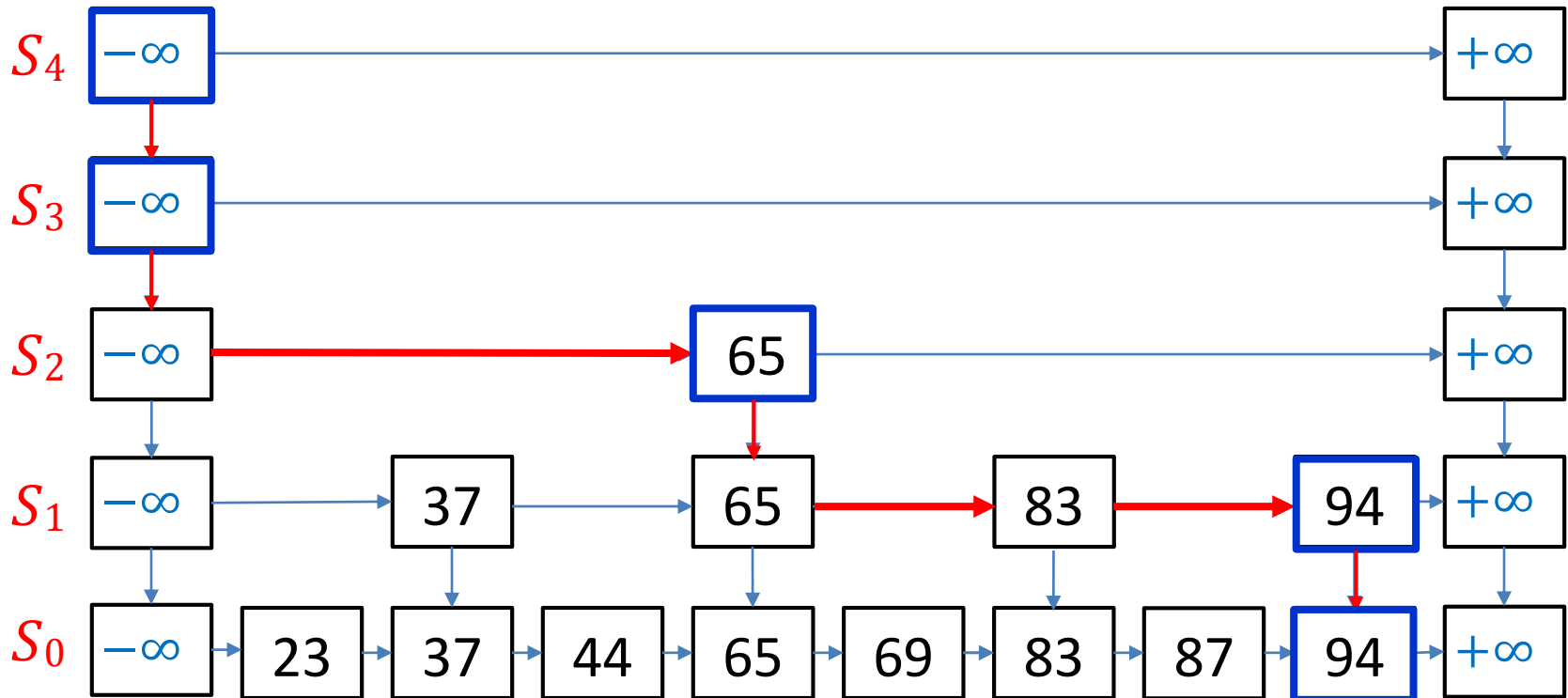
# Insert in Skip Lists: Example 2

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



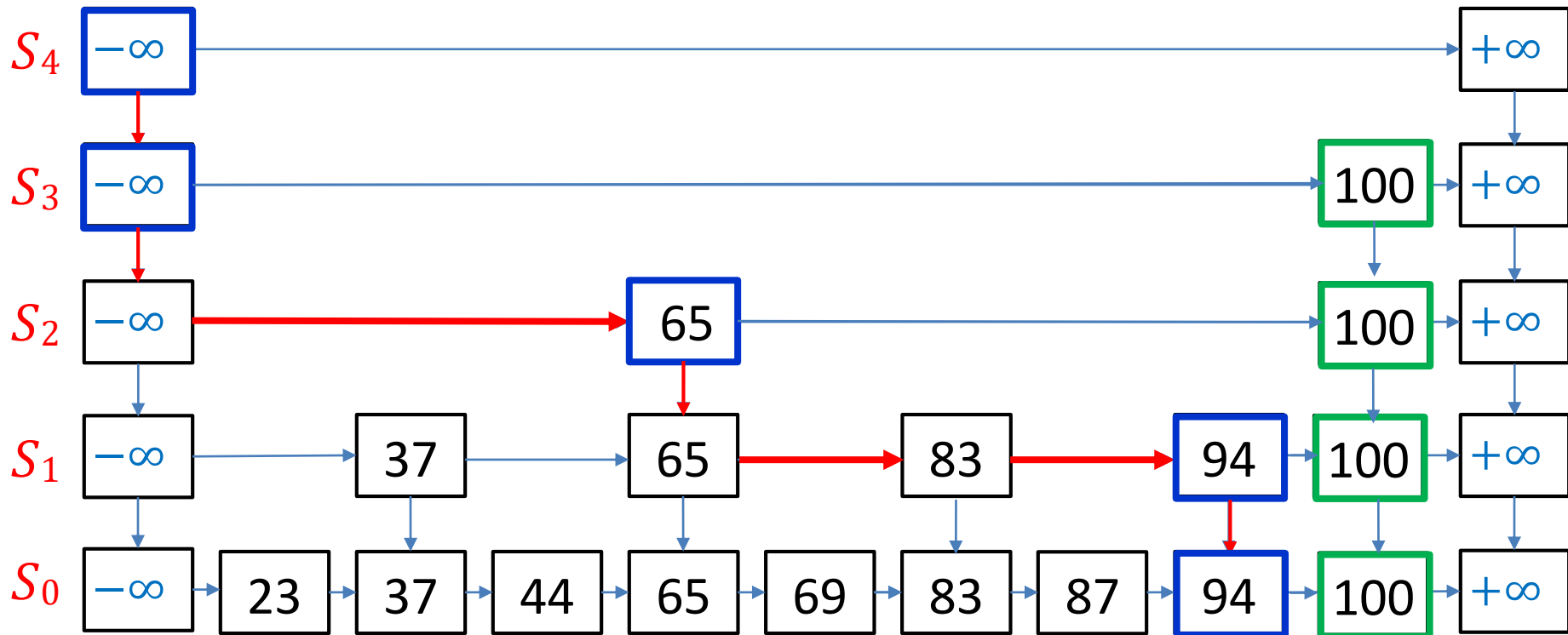
# Insert in Skip Lists: Example 2

- `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: Example 2

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



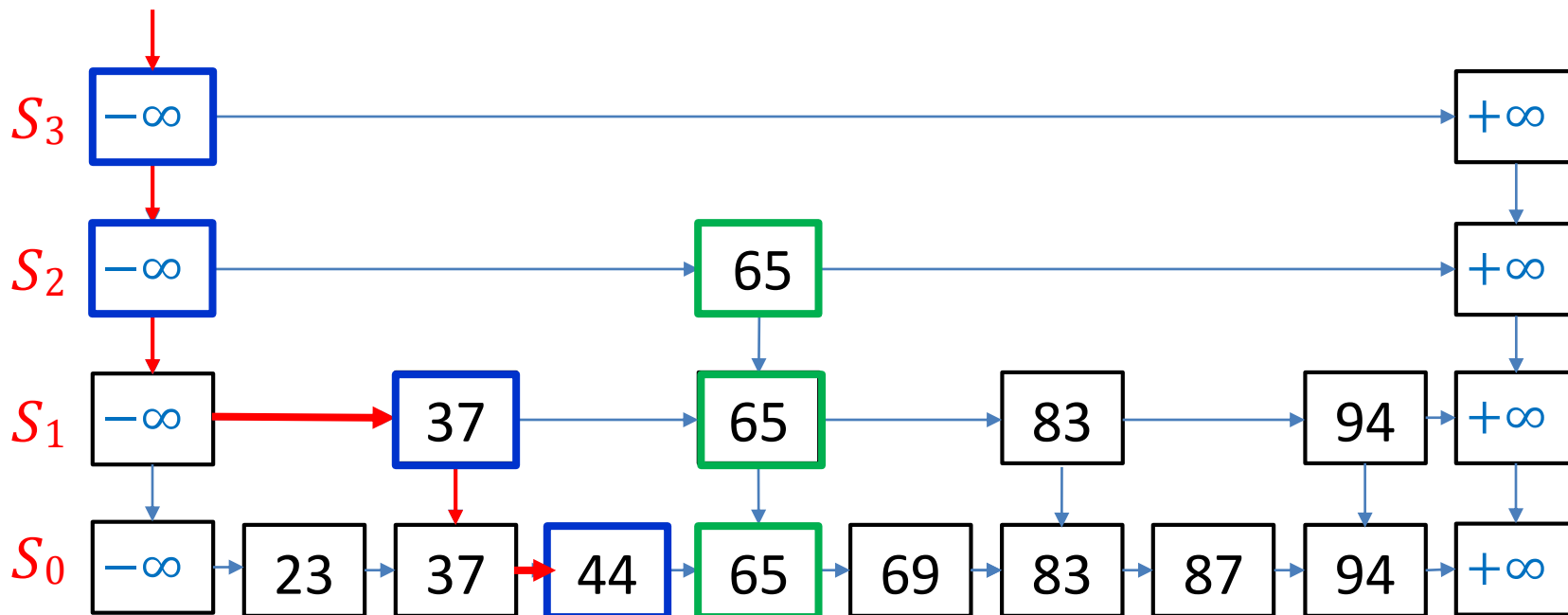
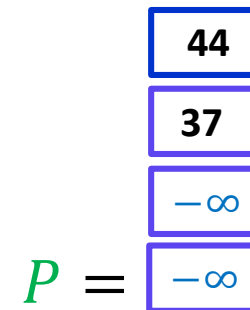


# Insert in Skip Lists

```
skipList::insert(k, v)  
  for (i ← 0; random(2) = 1; i ← i + 1) {}           // random tower height  
  for (h ← 0, p ← root.below; p ≠ NIL; p ← p.below) do h ++  
  while i ≥ h                                         // increase skip-list height if needed  
    root ← new sentinel-only list linked in appropriately  
    h ++  
  P ← getPredecessors(k)  
  p ← P.pop()  
  zBellow ← new node with (k, v) inserted after p // insert (k, v) in  $S_0$   
  while i > 0                                         // insert k in  $S_1 S_2, \dots, S_i$   
    p ← P.pop()  
    z ← new node with k added after p  
    z.below ← zBellow  
    zBellow ← z  
    i ← i - 1
```

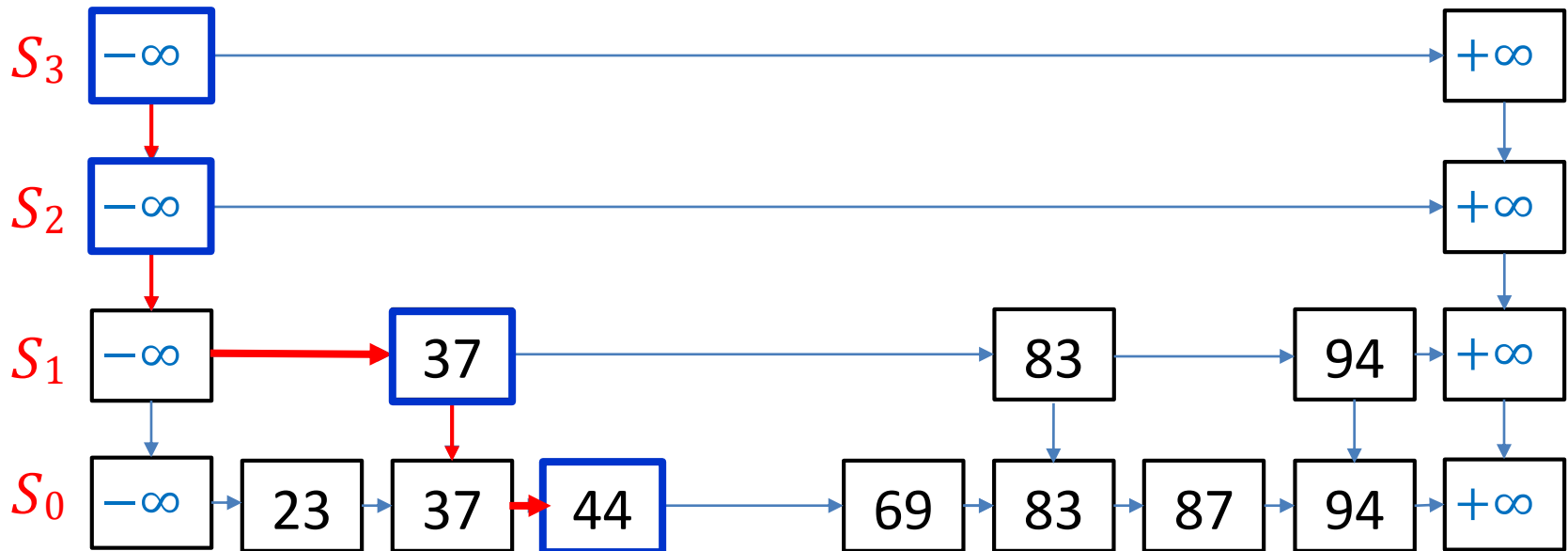
# Example: Delete in Skip Lists

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



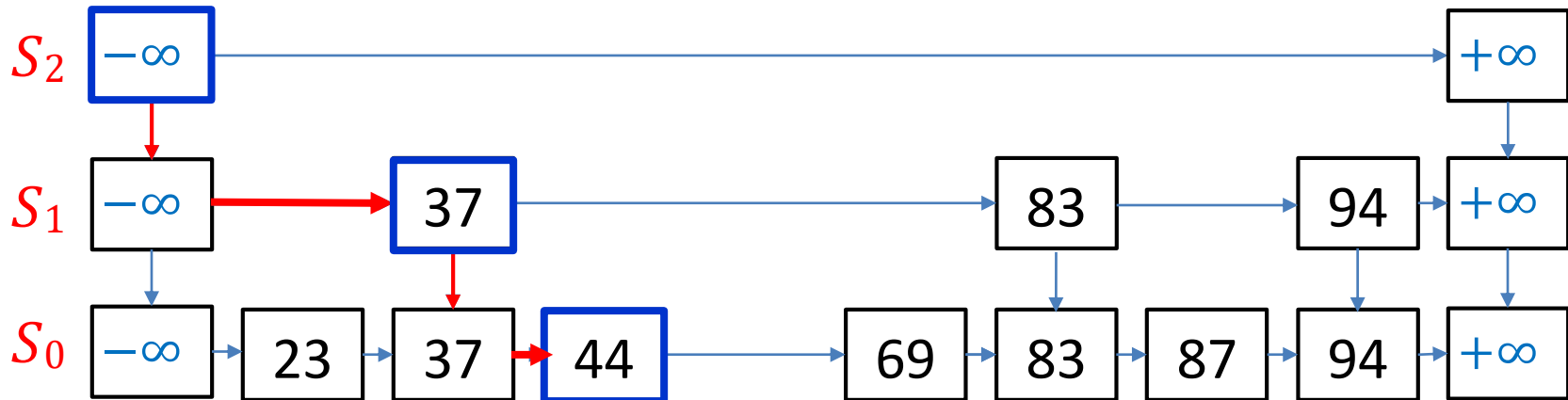
# Example: Delete in Skip Lists

- *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 ← getPredecessors(k)
```

```
while P is non-empty
```

```
    p ← P.pop() // predecessor of k in some layer
```

```
    if p.after.key = k
```

```
        p.after ← p.after.after
```

```
    else break // no more copies of k
```

```
p ← left sentinel of the root-list
```

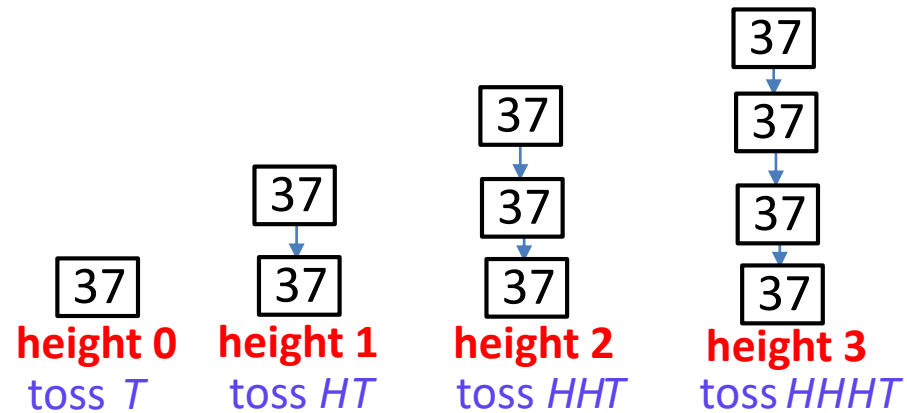
```
while p.below.after is the  $\infty$  sentinel
```

```
    // the two top lists are both only sentinels, remove one
```

```
    p.below ← p.below.below // removes the second empty list
```

```
    p.after.below ← p.after.below.below
```

# Skip List Analysis



- Let  $X_k$  be the height of tower for key  $k$

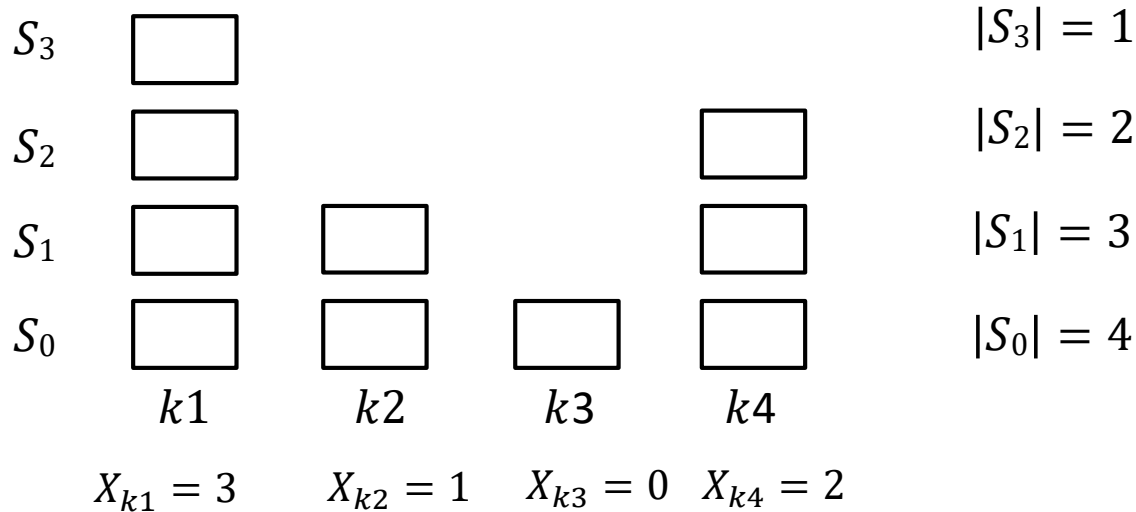
$$P(X_k \geq 1) = \frac{1}{2} \quad P(X_k \geq 2) = \frac{1}{2} \cdot \frac{1}{2} \quad P(X_k \geq 3) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2}$$

toss H...                      toss HH...                      toss HHH...

- In general 
$$P(X_k \geq i) = P(\underbrace{H H \dots H}_{i \text{ times}}) = \left(\frac{1}{2}\right)^i$$

- In the worst case, the height of a tower could be arbitrary large
  - no bound on height in terms of  $n$
- Operations could be arbitrarily slow, and space requirements arbitrarily large
  - but this is exceedingly unlikely
- Let us analyse *expected* run-time and space-usage (randomized data structure)

# Skip List Analysis



- Let  $X_k$  be the height of tower for key  $k$ , we know  $P(X_k \geq i) = \frac{1}{2^i}$
- If  $X_k \geq i$  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

# Skip List Analysis

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

- Let  $X_k$  be the height of tower for key  $k$ , we know  $P(X_k \geq i) = \frac{1}{2^i}$
- If  $X_k \geq i$  then list  $S_i$  includes key  $k$
- Let  $|S_i|$  be the number of keys in list  $S_i$
- Let  $I_{i,k} = \begin{cases} 0 & \text{if } X_k < i \\ 1 & \text{if } X_k \geq i \end{cases} = \begin{cases} 0 & \text{if list } S_i \text{ does not include key } k \\ 1 & \text{if list } S_i \text{ includes key } k \end{cases}$
- $|S_i| = \sum_{\text{key } k} I_{i,k}$
- $E[|S_i|] = E\left[\sum_{\text{key } k} I_{i,k}\right] = \sum_{\text{key } k} E[I_{i,k}] = \sum_{\text{key } k} P(I_{i,k} = 1) = \sum_{\text{key } k} P(X_k \geq i) = \sum_{\text{key } k} \frac{1}{2^i} = \frac{n}{2^i}$
- The expected length of list  $S_i$  is  $\frac{n}{2^i}$



# Skip List Analysis

- $|S_i|$  is number of keys in list  $S_i$ 
  - $E[|S_i|] = \frac{n}{2^i}$

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

- $h = 1 + \sum_{i \geq 1} I_i$  (here +1 is for the sentinel-only level)

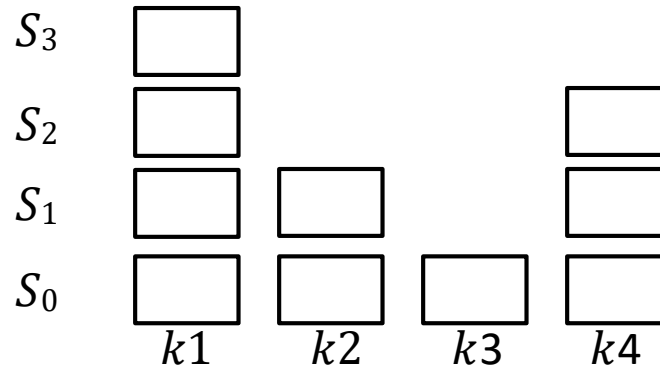
- Since  $I_i \leq 1$  we have that  $E[I_i] \leq 1$

- Since  $I_i \leq |S_i|$  we have that  $E[I_i] \leq E[|S_i|] = \frac{n}{2^i}$

- For ease of derivation, assume  $n$  is a power of 2

$$\begin{aligned}
 E[h] &= E\left[1 + \sum_{i \geq 1} I_i\right] = 1 + \sum_{i \geq 1} E[I_i] = 1 + \sum_{i=1}^{\log n} E[I_i] + \sum_{i=1+\log n}^{\infty} E[I_i] \\
 &\leq 1 + \sum_{i=1}^{\log n} 1 + \sum_{i=1+\log n}^{\infty} \frac{n}{2^i} \\
 &\leq 1 + \log n + \sum_{i=0}^{\infty} \frac{n}{2^{i+1+\log n}}
 \end{aligned}$$

$S_4$  has only sentinels



$$I_4 = 0$$

$$I_3 = 1$$

$$I_2 = 1$$

$$I_1 = 1$$

# Skip List Analysis

$S_4$  has only sentinels

$$I_4 = 0$$



$$I_3 = 1$$

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

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

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- $h = 1 + \sum_{i \geq 1} I_i$  (here  $h$  is height)

- Since  $I_i \leq 1$  we have that  $E[I_i] \leq 1$

- Since  $I_i \leq |S_i|$  we have that  $E[I_i] \leq E[|S_i|]$

- For ease of derivation, assume  $E[I_i] = E[|S_i|]$

- $E[h] = E\left[1 + \sum_{i \geq 1} I_i\right] = 1 + \sum_{i \geq 1} E[I_i]$

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

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

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

$$S = \sum_{i=0}^{\infty} \frac{1}{2^i} = \cancel{1} + \cancel{\frac{1}{2}} + \frac{1}{2^2} + \frac{1}{2^3} + \dots$$

$$2S = \sum_{i=0}^{\infty} \frac{2}{2^i} = 2 + \cancel{1} + \cancel{\frac{1}{2}} + \frac{1}{2^2} + \dots$$

$$2S - S = 2$$

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

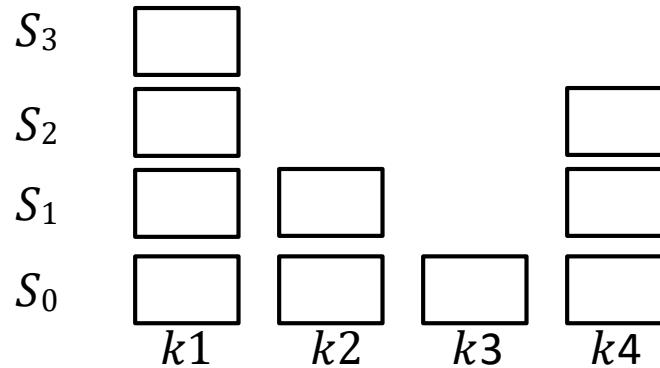
# Skip List Analysis

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

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$S_4$  has only sentinels



$$I_4 = 0$$

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$$I_2 = 1$$

$$I_1 = 1$$

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- For ease of derivation, assume  $n$  is a power of 2

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

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

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

$$= 1 + \log n + 1$$

- Expected skip list height  $\leq 2 + \log n$

# 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

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

## 2. Space for nodes storing keys

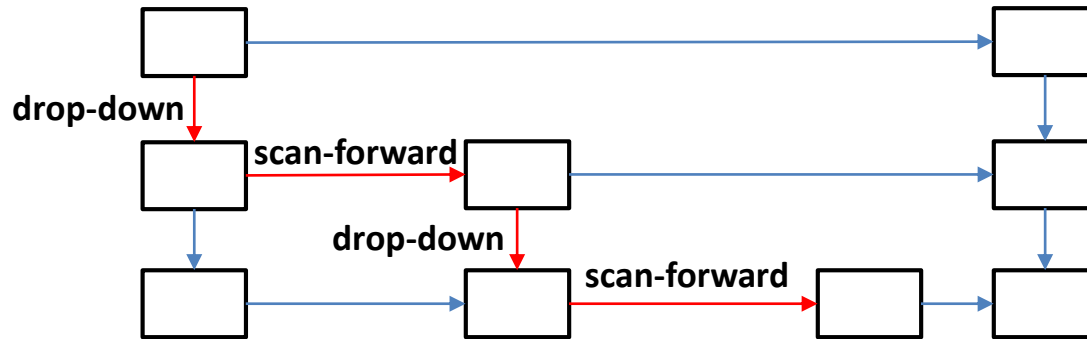
- Let  $|S_i|$  be the number of keys in list  $S_i$

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

- expected space for keys is  $E\left[\sum_{i \geq 0} |S_i|\right] = \sum_{i \geq 0} E[|S_i|] = \sum_{i \geq 0} \frac{n}{2^i} = 2n$

- Total expected space is  $\Theta(n)$

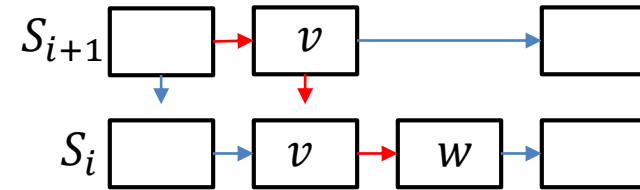
# Skip List Analysis: Expected Running Time



- *search*, *insert*, and *delete* are dominated by the runtime of *getPredecessors*
- So we analyze the expected time of *getPredecessors*
  - runtime is proportional to number of 'drop-down' and 'scan-forward'
- We 'drop-down'  $h$  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 next that expected number of 'scan-forward' is also  $O(\log n)$
- So total expected running time is  $O(\log n)$

# Expected Number of Scan-Forward Operations

- Number 'scan-forward' at level  $i$ 
  - 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  $S_{i+1}$
  - Thus if we do scan forward from  $v$  to  $w$ , then the tower of  $w$  has height  $i$ 
    - $P(\text{tower of } w \text{ has height } i \mid \text{tower of } w \text{ has height at least } i) = 1/2$
    - scan forward (i.e. at least one scan) from  $v$  to  $w$  with probability at most  $1/2$ 
      - 'at most' because we could scan-down down if search  $key < w$
    - repeating argument, probability of scan-forward at least  $l$  times is at most  $(1/2)^l$



$$E[\# \text{ scan-forward at level } i] = \sum_{l \geq 1} l \cdot P(\text{scans} = l) \stackrel{\substack{\text{theorem in probability} \\ \text{theory}}}{=} \sum_{l \geq 1} P(\text{scans} \geq l) \leq \sum_{l \geq 1} \frac{1}{2^l} = 1$$

# Expected Number of Scan-Forward Operations

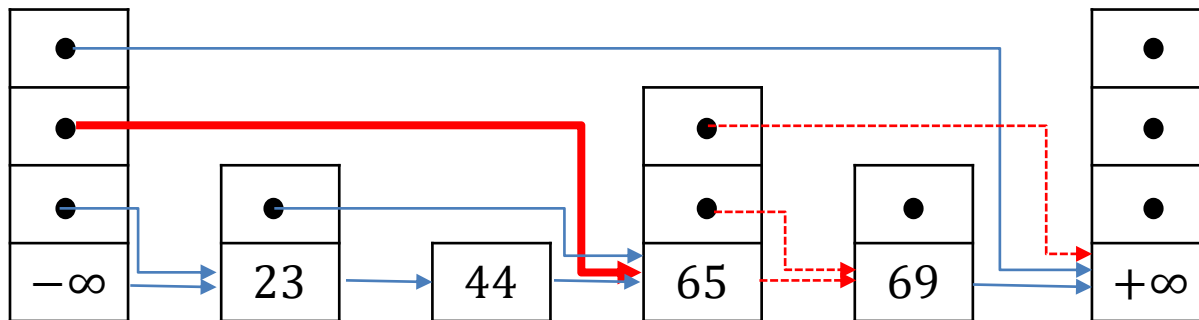
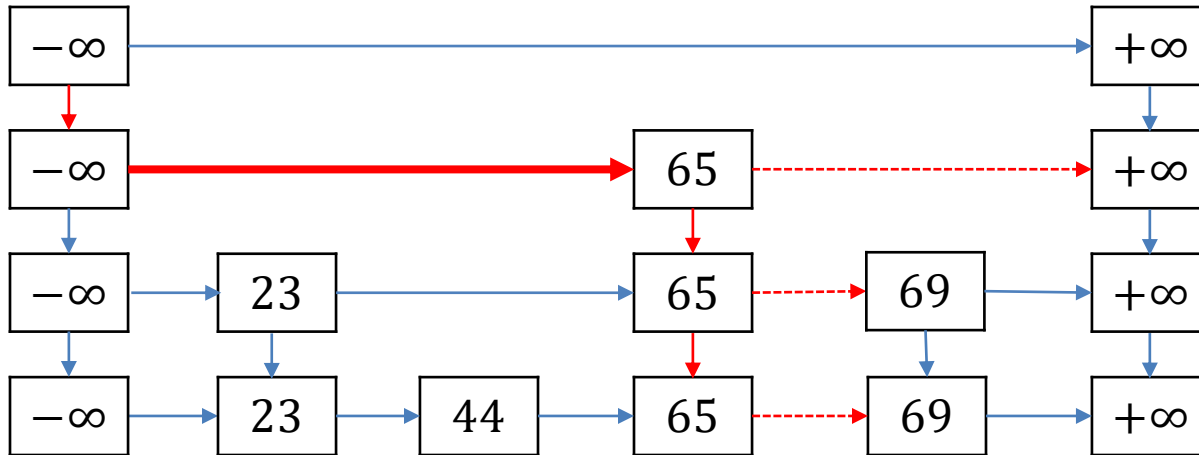
- At level  $i < h$ :  $E[\text{number of scan-forward}] \leq 1$
- Also, expected number of scan-forward at level  $i < \text{number of keys at level } S_i$ 
  - $|S_i|$  is the number of keys in list on level  $i$ , and  $E[|S_i|] = \frac{n}{2^i}$
- For ease of derivation, assume  $n$  is a power of 2
- Expected number of scan-forward over all levels

$$\begin{aligned} \sum_{i \geq 0} E[\# \text{ of scan-forward at level } i] &= \\ &= \sum_{i=1}^{\log n} E[\# \text{ of scan-for at level } i] + \sum_{i=1+\log n}^{\infty} E[\# \text{ of scan-for at level } i] \\ &\leq \sum_{i=1}^{\log n} 1 + \sum_{i=1+\log n}^{\infty} \frac{n}{2^i} \\ &\leq \log n + 1 \end{aligned}$$

- Expected number of scan-forwards 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
  - **Biased Search Requests**

# Improving Unsorted Lists/Arrays

- Unordered lists/arrays are among simplest data structures to implement
- But for Dictionary ADT
  - inefficient *search*:  $\Theta(n)$
- Can we make search in unordered lists/arrays more effective in practice?
  - No if items are accessed equally likely
    - can show average-case search is  $\Theta(n)$
  - Yes if the search requests are biased
    - some items are accessed much more frequently than others
    - 80/20 rule: 80% of outcomes result from 20% of causes
    - access = insertion or successful search
    - frequently accessed items should be in the front
      - two cases
        - know the access distribution beforehand
          - optimal static ordering
        - do not know access distribution beforehand
          - dynamic ordering

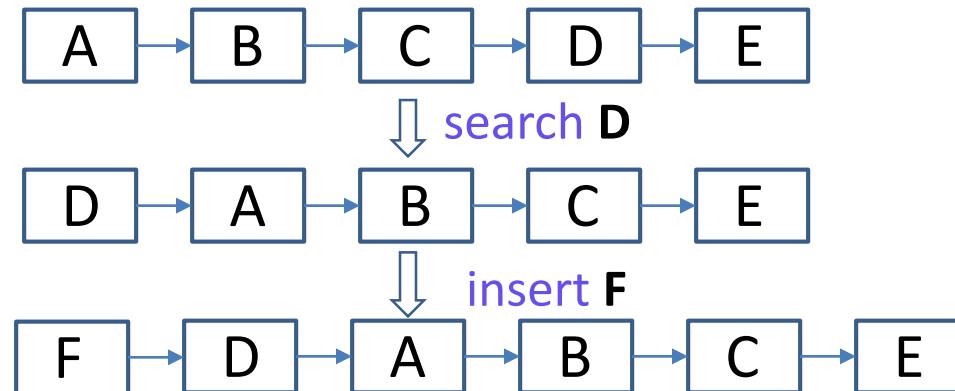
# Optimal Static Ordering

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

- Order  $A \quad B \quad C \quad D \quad E$  has expected cost
 
$$\frac{2}{26} \cdot 1 + \frac{8}{26} \cdot 2 + \frac{1}{26} \cdot 3 + \frac{10}{26} \cdot 4 + \frac{5}{26} \cdot 5 \approx 3.31$$
- Order  $D \quad B \quad E \quad A \quad C$  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
  - static ordering*: order of items does not change
- Proof Idea: for any other ordering, exchanging two items that are out-of-order according to access probabilities makes total cost increase

# Dynamic Ordering

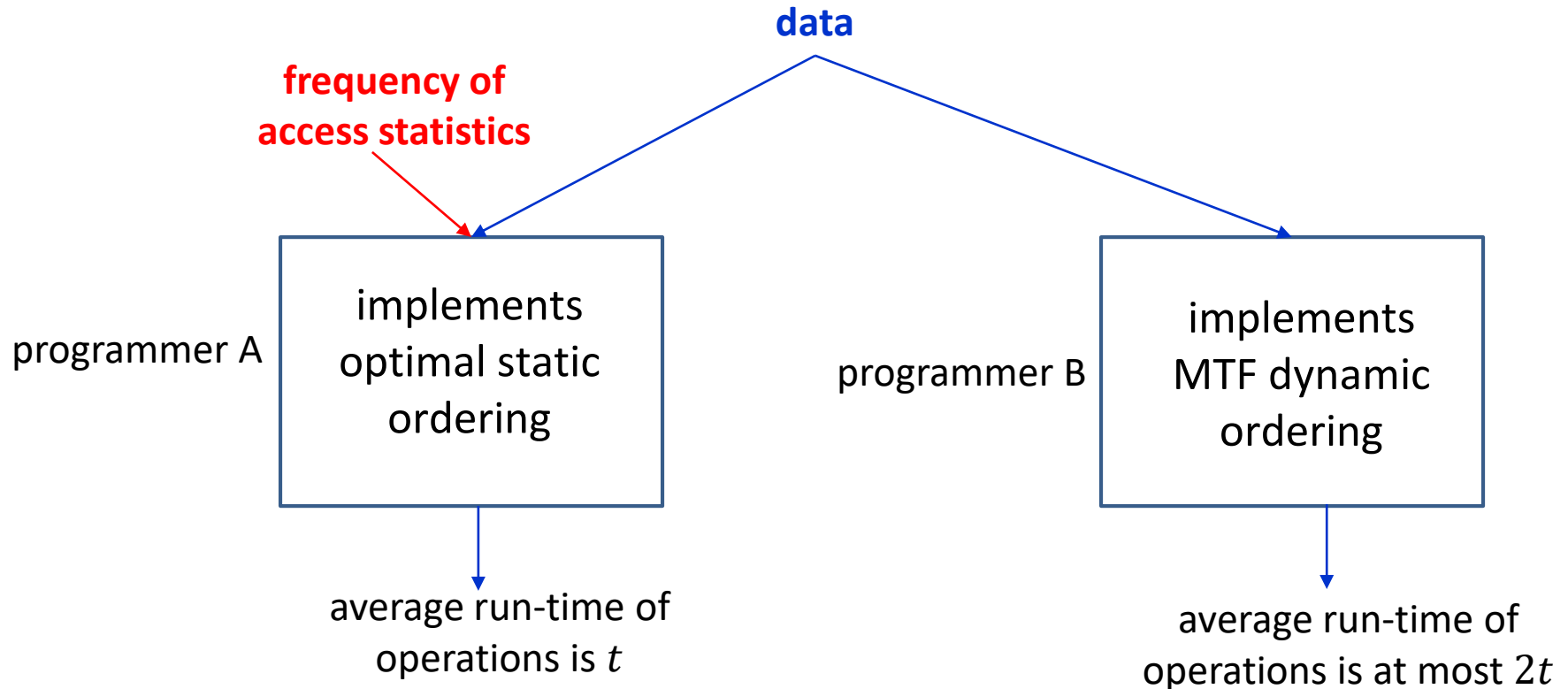
- *Dynamic ordering*: order of items is allowed to change
- What if we do *not know the access probabilities* ahead of time?
- Rule of thumb: recently accessed item is likely to be accessed soon again
- **Move-To-Front heuristic** (MTF): after search, move the accessed item to the front
  - additionally, in list: always insert at the front



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

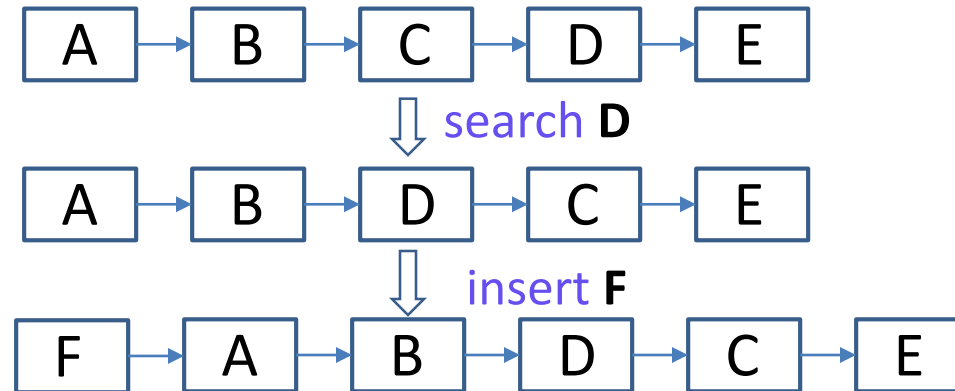
# Dynamic Ordering: MTF

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



# Dynamic Ordering: Other Heuristics

- **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
- **Frequency-count heuristic:** Keep counters how often items were accessed, and sort in non-decreasing order
  - works well in practice, but requires extra space

# Summary of Biased Search Requests

- We are unlikely to know the access-probabilities of items, so optimal static order is mostly of theoretical interest
- For any dynamic reordering heuristic, some sequence will defeat it
  - have  $\Theta(n)$  access cost for each item
- MTF and Frequency-Count work well in practice
- For MTF can prove theoretical guarantees
- There is very little overhead for MTF and other strategies, they should be applied whenever unordered arrays or lists are used