lly Efficient Use of Keys: the Trie

much about cost of comparisons. worst case is length of string. nould throw extra factor of key length, L, into costs: parisons really means $\Theta(ML)$ operations. for key X, keep looking at same chars of X M times. tter? Can we get search cost to be O(L)?

multi-way decision tree, with one decision per character

9:39 2018 CS61B: Lecture #31 2

CS61B Lecture #31

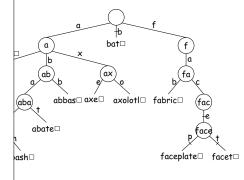
ed search structures (DS(IJ), Chapter 9

bm Numbers (DS(IJ), Chapter 11)

Adding Item to a Trie

ding bat and faceplate.

icked.



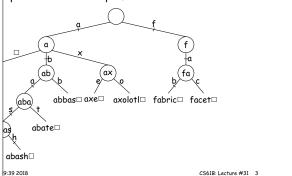
9:39 2018 CS61B: Lecture #31 4

The Trie: Example

show paths followed for "abash" and "fabric"

I node corresponds to a possible prefix.

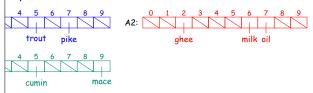
n path to node = that prefix.



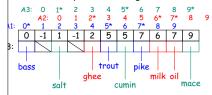
Scrunching Example

(unrelated to Tries on preceding slides)

rrays, each indexed 0..9



them, but keep track of original index of each item:



9:39 2018 CS61B: Lecture #31 6

A Side-Trip: Scrunching

bvious implementation for internal nodes is array inaracter.

erformance, L length of search key.

independent of N, number of keys. Is there a depen-

ays are sparsely populated by non-null values—waste of

arrays on top of each other!

mpty) entries of one array to hold non-null elements of

arkers to tell which entries belong to which array.

9:39 2018 CS61B: Lecture #31 5

9:39 2018 C561B: Lecture #31 1

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

9:39 2018 CS61B: Lecture #31 8

Practicum

ina idea is cute, but

od if we want to expand our trie.

plicated.

9:39 2018

nore useful for representing large, sparse, fixed tables rows and columns.

, number of children in trie tends to drop drastically s a few levels down from the root.

ce, might as well use linked lists to represent set of

rays for the first few levels, which are likely to have

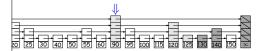
CS61B: Lecture #31 7

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

9:39 2018 CS61B: Lecture #31 10

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

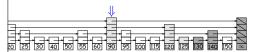
9:39 2018 C561B: Lecture #31 9

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

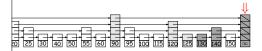
9:39 2018 CS61B: Lecture #31 12

robabilistic Balancing: Skip Lists

an be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ıple:



tart at top layer on left, search until next step would hen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

9:39 2018 CS61B: Lecture #31 11

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are > k high as there are that are k high.

hes fast with high probability.

9:39 2018 CS61B: Lecture #31 14

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

hought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

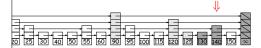
9:39 2018 CS61B: Lecture #31 13

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

hought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

9:39 2018 CS61B: Lecture #31 16

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

hought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

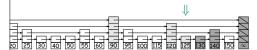
9:39 2018 CS61B: Lecture #31 15

robabilistic Balancing: Skip Lists

in be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

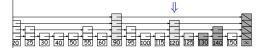
9:39 2018 CS61B: Lecture #31 18

robabilistic Balancing: Skip Lists

n be thought of as a kind of n-ary search tree in which put the keys at "random" heights.

thought of as an ordered list in which one can skip large

ple:



tart at top layer on left, search until next step would nen go down one layer and repeat.

, we search for 125 and 127. Gray nodes are looked at; nodes are overshoots.

he nodes were chosen randomly so that there are about nodes that are >k high as there are that are k high.

hes fast with high probability.

9:39 2018 CS61B: Lecture #31 17

Summary

arch trees allows us to realize $\Theta(\lg N)$ performance.

-black trees:

 $\left|V\right>$ performance for searches, insertions, deletions. ood for external storage. Large nodes minimize # of ations

performance for searches, insertions, and deletions, s length of key being processed.

to manage space efficiently.

idea: scrunched arrays share space.

able $\Theta(\lg N)$ performace for searches, insertions, dele-

plement.

for interesting ideas: probabilistic balance, randomstructures.

9:39 2018 CS61B: Lecture #31 20

Example: Adding and deleting

m initial list:

9:39 2018



r, we add 126 and 127 (choosing random heights for emove 20 and 40:

CS61B: Lecture #31 19



s here have been modified.

tructures that Implement Abstractions

linked lists, circular buffers

le.

Queue: heaps

Set: binary search trees, red-black trees, B-trees,

arrays or linked lists

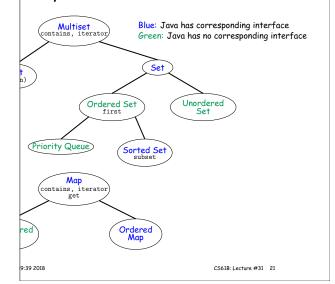
d Set: hash table

lap: hash table

: red-black trees, B-trees, sorted arrays or linked lists

9:39 2018 CS61B: Lecture #31 22

immary of Collection Abstractions



Corresponding Classes in Java

ction)

ist, LinkedList, Stack, ArrayBlockingQueue,

et

Queue: PriorityQueue Set (SortedSet): TreeSet

d Set: HashSet

Nap: HashMap

o (SortedMap): TreeMap

9:39 2018 CS61B: Lecture #31 23