Algorithms

ROBERT SEDGEWICK | KEVIN WAYNE

3.4 HASH TABLES

hash functions

separate chaining

linear probing

context

Robert Sedgewick | Kevin Wayne

Algorithms

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Premature optimization

" Programmers waste enormous amounts of time thinking about, or worrying about, the speed of noncritical parts of their programs, and these attempts at efficiency actually have a strong negative impact when debugging and maintenance are considered.

We should forget about small efficiencies, say about 97% of the time: premature optimization is the root of all evil.

Yet we should not pass up our opportunities in that critical 3%. "

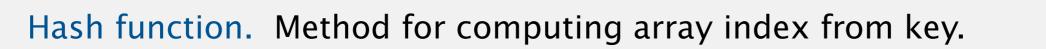


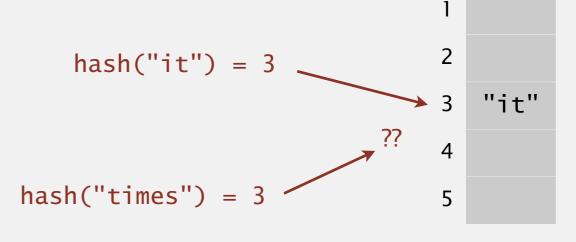
Symbol table implementations: summary

implementation		guarantee		a	iverage case	ordered	key	
	search	insert	delete	search hit	insert	delete	ops?	interface
sequential search (unordered list)	Ν	Ν	N	Ν	N	N		equals()
binary search (ordered array)	log N	Ν	Ν	log N	N	Ν	~	compareTo()
BST	Ν	Ν	Ν	log N	log N	\sqrt{N}	~	compareTo()
red-black BST	log N	log N	log N	log N	log N	log N	~	compareTo()

- Q. Can we do better?
- A. Yes, but with different access to the data.

Save items in a key-indexed table (index is a function of the key).





0

Issues.

- Computing the hash function.
- Equality test: Method for checking whether two keys are equal.
- Collision resolution: Algorithm and data structure to handle two keys that hash to the same array index.

Classic space-time tradeoff.

- No space limitation: trivial hash function with key as index.
- No time limitation: trivial collision resolution with sequential search.
- Space and time limitations: hashing (the real world).

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Computing the hash function

Idealistic goal. Scramble the keys uniformly to produce a table index.

- Efficiently computable.
- Each table index equally likely for each key.





Ex. Social Security numbers.

- 573 = California, 574 = Alaska • Bad: first three digits.
- Better: last three digits.

(assigned in chronological order within geographic region)

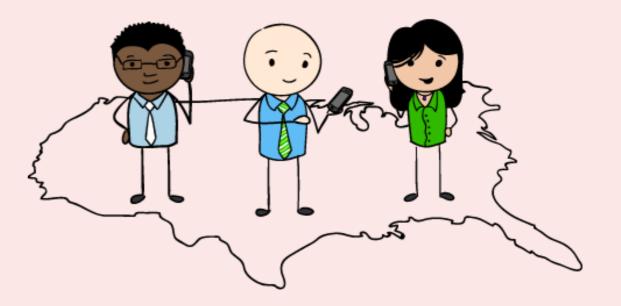
Practical challenge. Need different approach for each key type.

Hash tables: quiz 1

Which of the following would be a good hash function for U.S. phone numbers to integers between 0 and 999?

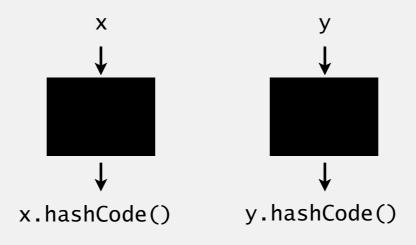
- A. First three digits.
- **B.** Second three digits.
- C. Last three digits.
- **D.** Either B or C.
- E. I don't know.

(609) 867-5309



All Java classes inherit a method hashCode(), which returns a 32-bit int.

Requirement. If x.equals(y), then (x.hashCode() == y.hashCode()).
Highly desirable. If !x.equals(y), then (x.hashCode() != y.hashCode()).



Default implementation. Memory address of x. Legal (but poor) implementation. Always return 17. Customized implementations. Integer, Double, String, File, URL, Date, ... User-defined types. Users are on their own.

Implementing hash code: integers, booleans, and doubles

Java library implementations

```
public final class Integer
{
    private final int value;
    ...
    public int hashCode()
    { return value; }
}
```

```
public final class Boolean
{
    private final boolean value;
    ...
    public int hashCode()
    {
        if (value) return 1231;
        else return 1237;
     }
}
```

```
public final class Double
{
   private final double value;
   . . .
   public int hashCode()
       long bits = doubleToLongBits(value);
       return (int) (bits ^ (bits >>> 32));
   }
}
            convert to IEEE 64-bit representation;
                xor most significant 32-bits
                with least significant 32-bits
```

Warning: -0.0 and +0.0 have different hash codes

5

Treat string of length *L* as *L*-digit, base-31 number:

 $h = s[0] \cdot 31^{L-1} + \ldots + s[L-3] \cdot 31^{2} + s[L-2] \cdot 31^{1} + s[L-1] \cdot 31^{0}$

public final class String	cha	r Unicode
{ private final char[] s;		
	'a'	97
<pre>public int hashCode()</pre>	'b'	98
{ int hash = 0;	'c'	99
for (int i = 0; i < length(); i++) hash = s[i] + (31 * hash);		
return hash;		
}		
<pre>Java library implementation</pre>		

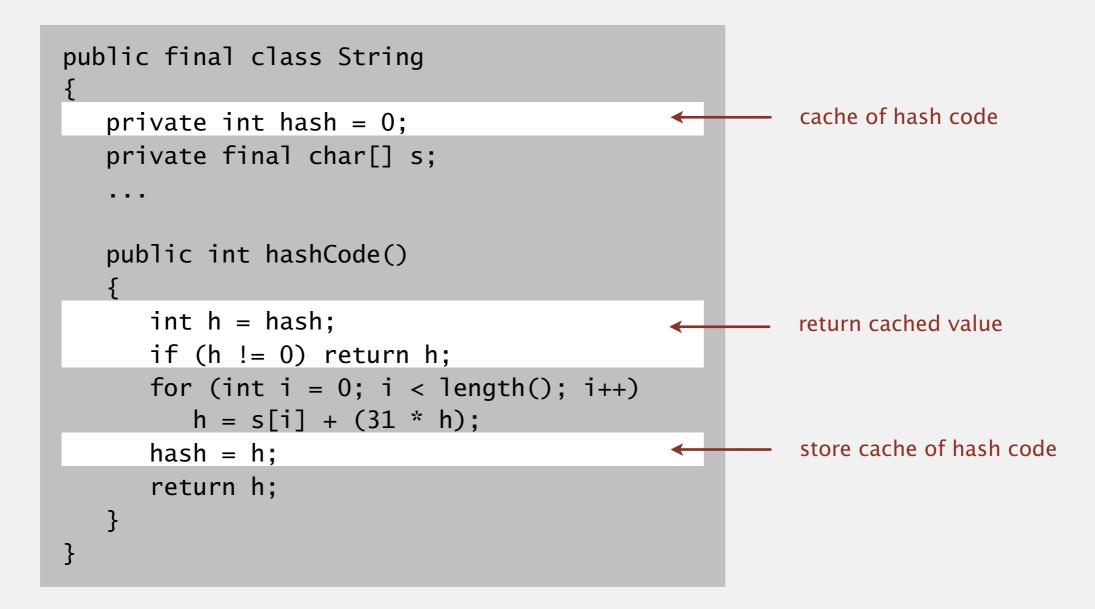
Horner's method: only *L* multiplies/adds to hash string of length *L*.

String s = "call";
s.hashCode();
$$\leftarrow$$
 3045982 = 99.31³ + 97.31² + 108.31¹ + 108.31⁰
= 108 + 31. (108 + 31. (97 + 31. (99)))

Implementing hash code: strings

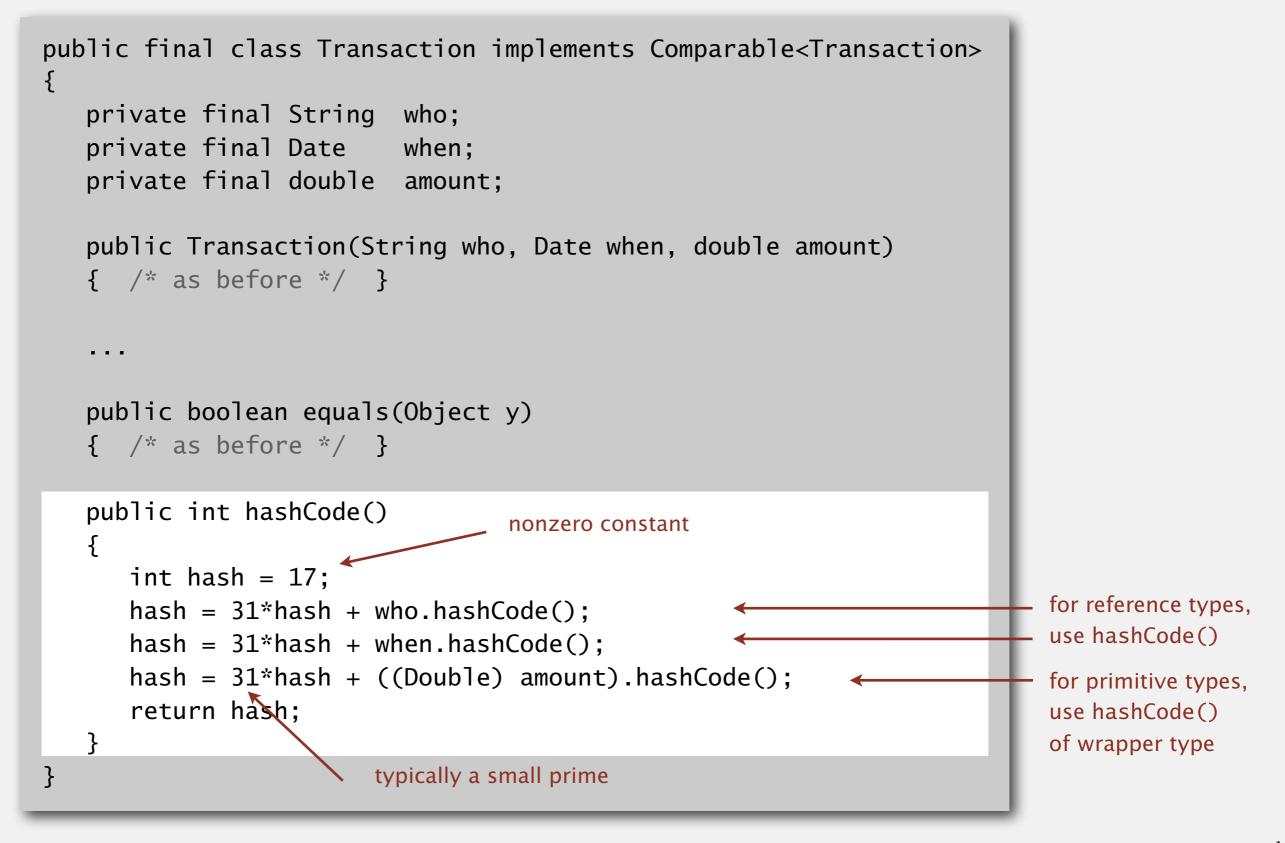
Performance optimization.

- Cache the hash value in an instance variable.
- Return cached value.



Q. What if hashCode() of string is $0? \leftarrow hashCode()$ of "pollinating sandboxes" is 0

Implementing hash code: user-defined types



Hash code design

"Standard" recipe for user-defined types.

- Combine each significant field using the 31x + y rule.
- If field is a primitive type, use wrapper type hashCode().
- If field is null, use 0.
- If field is a reference type, use hashCode(). ← _____ applies rule recursively

In practice. Recipe above works reasonably well; used in Java libraries. In theory. Keys are bitstring; "universal" family of hash functions exist.

> awkward in Java since only one (deterministic) hashCode()

Basic rule. Need to use the whole key to compute hash code; consult an expert for state-of-the-art hash codes.

Hash tables: quiz 1

Which of the following is an effective way to map a hashable key to an integer between 0 and M-1 ?

Α.

B.

private int hash(Key key)
{ return key.hashCode() % M; }

private int hash(Key key)
{ return Math.abs(key.hashCode()) % M; }

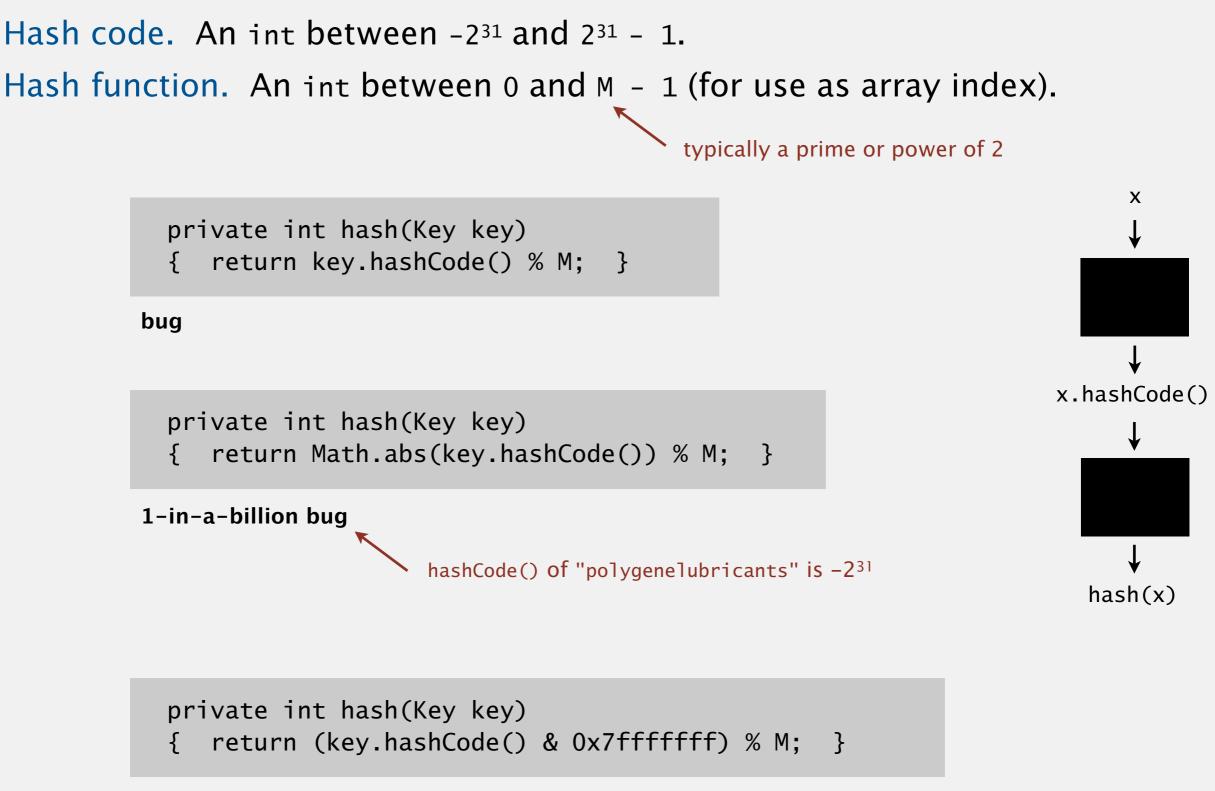
- C. Both A and B.
- **D.** Neither A nor B.
- **E.** *I don't know.*

Х

x.hashCode()

hash(x)

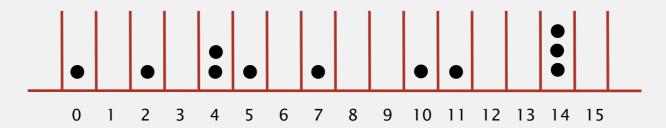
Modular hashing



Uniform hashing assumption

Uniform hashing assumption. Each key is equally likely to hash to an integer between 0 and M - 1.

Bins and balls. Throw balls uniformly at random into *M* bins.



Birthday problem. Expect two balls in the same bin after $\sim \sqrt{\pi M/2}$ tosses.

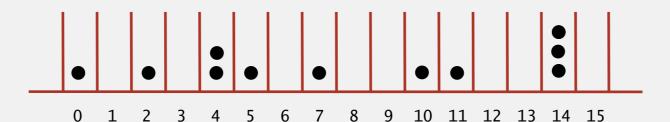
Coupon collector. Expect every bin has ≥ 1 ball after $\sim M \ln M$ tosses.

Load balancing. After *M* tosses, expect most loaded bin has $\sim \ln M / \ln \ln M$ balls.

Uniform hashing assumption

Uniform hashing assumption. Each key is equally likely to hash to an integer between 0 and M - 1.

Bins and balls. Throw balls uniformly at random into *M* bins.





Java's String data uniformly distribute the keys of Tale of Two Cities

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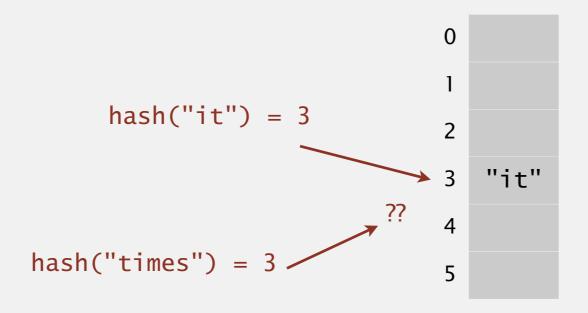
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Collisions

Collision. Two distinct keys hashing to same index.

- Birthday problem \Rightarrow can't avoid collisions. \leftarrow
- Coupon collector \Rightarrow not too much wasted space.
- Load balancing \Rightarrow no index gets too many collisions.



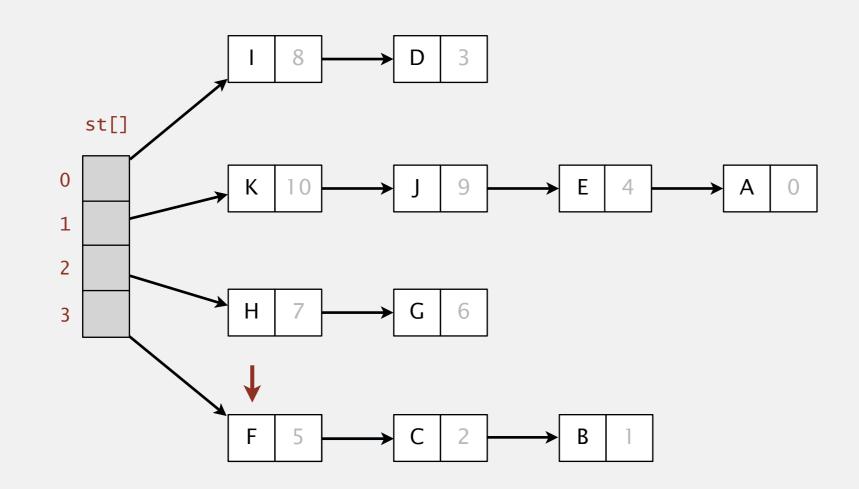
Challenge. Deal with collisions efficiently.

unless you have a ridiculous (quadratic) amount of memory Use an array of *M* < *N* linked lists. [H. P. Luhn, IBM 1953]

- Hash: map key to integer *i* between 0 and M 1.
- Insert: put at front of *i*th chain (if not already in chain).
- Search: sequential search in *i*th chain.

put(L, 11) hash(L) = 3

separate-chaining hash table (M = 4)

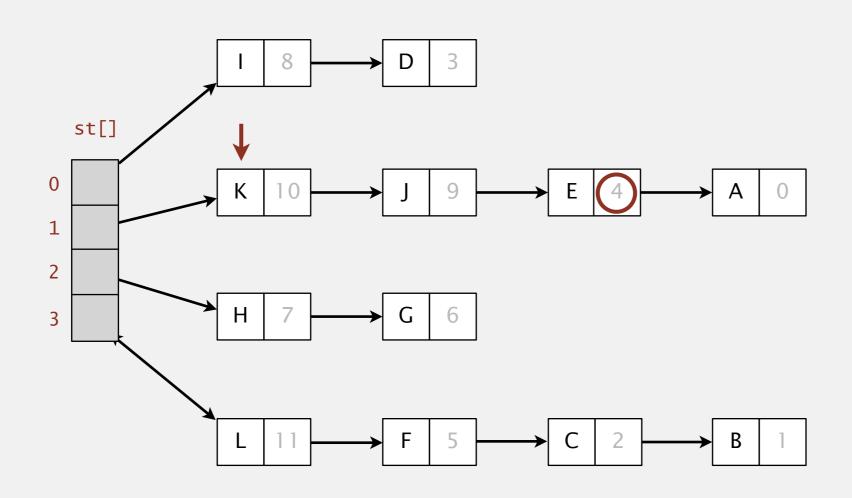


Use an array of *M* < *N* linked lists. [H. P. Luhn, IBM 1953]

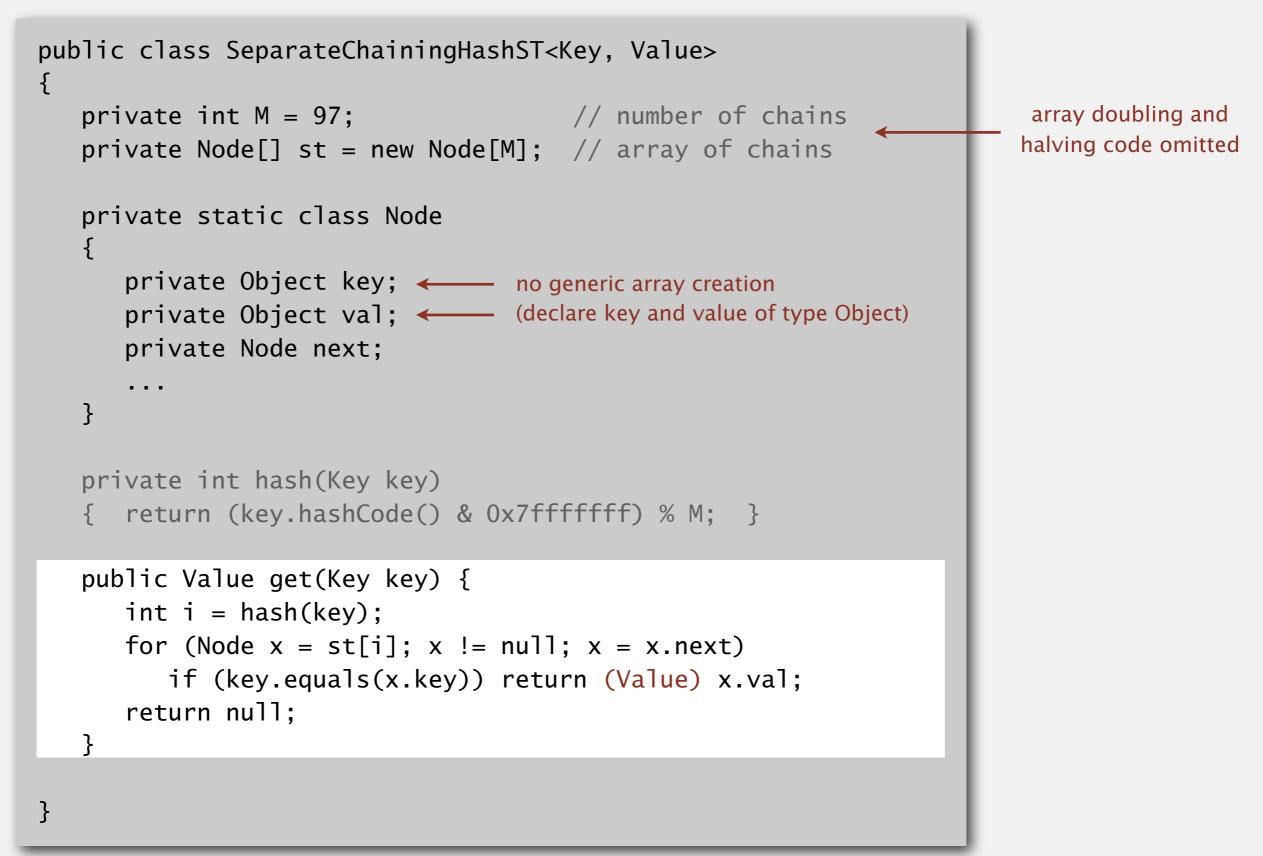
- Hash: map key to integer *i* between 0 and M 1.
- Insert: put at front of *i*th chain (if not already in chain).
- Search: sequential search in *i*th chain.

get(E) hash(E) = 1

separate-chaining hash table (M = 4)



Separate-chaining symbol table: Java implementation



Separate-chaining symbol table: Java implementation

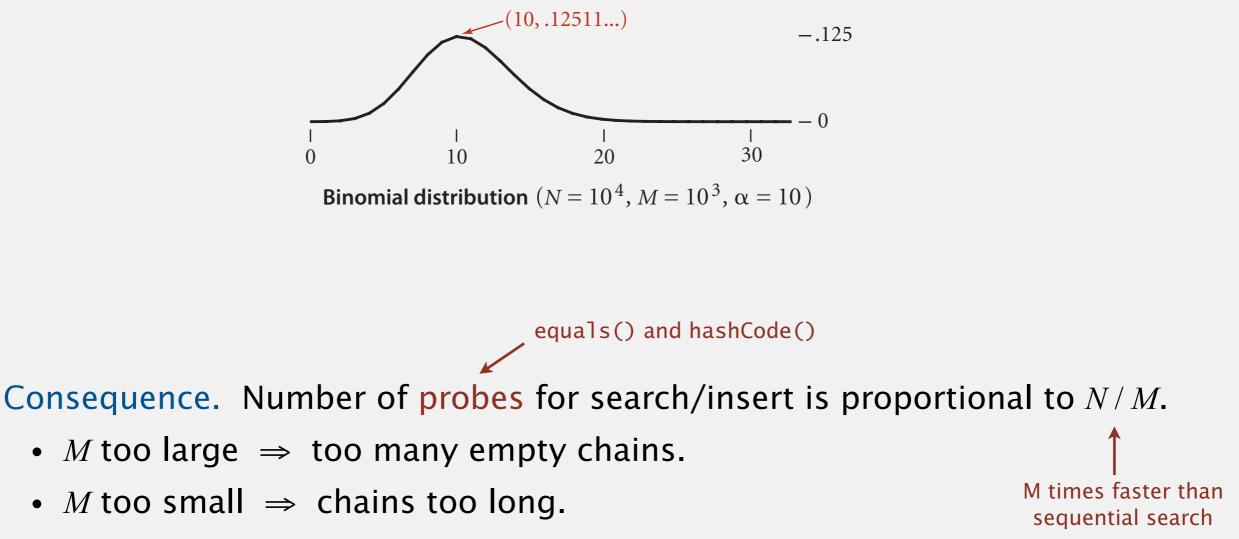
```
public class SeparateChainingHashST<Key, Value>
{
                        // number of chains
   private int M = 97;
   private Node[] st = new Node[M]; // array of chains
   private static class Node
      private Object key;
      private Object val;
      private Node next;
      . . .
   }
   private int hash(Key key)
   { return (key.hashCode() & 0x7ffffff) % M; }
   public void put(Key key, Value val) {
      int i = hash(key);
      for (Node x = st[i]; x != null; x = x.next)
        if (key.equals(x.key)) { x.val = val; return; }
      st[i] = new Node(key, val, st[i]);
   }
```

}

Analysis of separate chaining

Proposition. Under uniform hashing assumption, prob. that the number of keys in a list is within a constant factor of N/M is extremely close to 1.

Pf sketch. Distribution of list size obeys a binomial distribution.



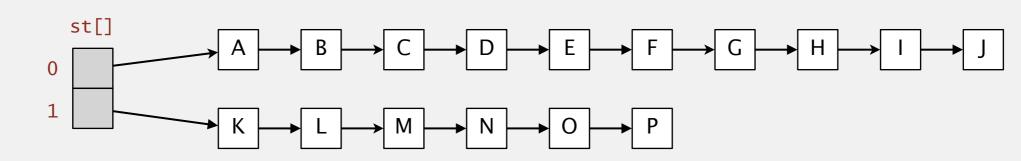
• Typical choice: $M \sim \frac{1}{4} N \Rightarrow$ constant-time ops.

Resizing in a separate-chaining hash table

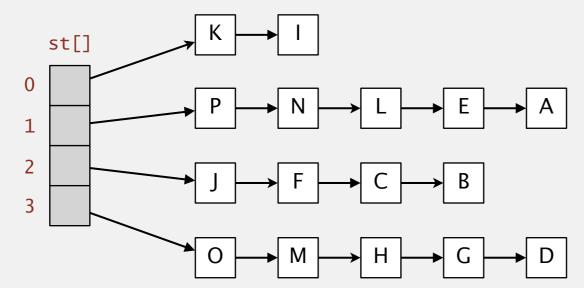
Goal. Average length of list N / M = constant.

- Double size of array *M* when $N/M \ge 8$; halve size of array *M* when $N/M \le 2$.
- Note: need to rehash all keys when resizing.

 x.hashCode() does not change;
 but hash(x) can change



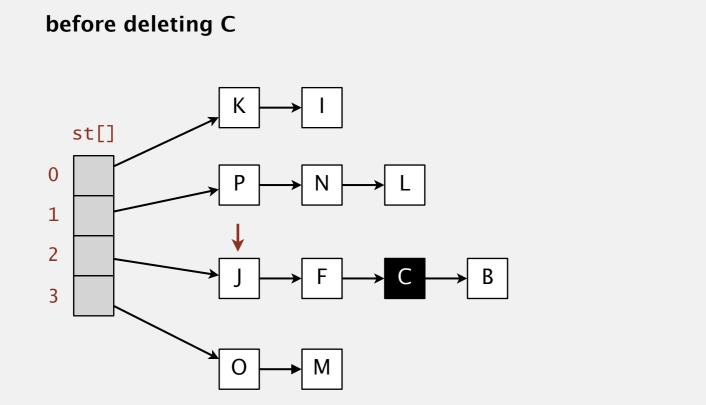
after resizing (N/M = 4)



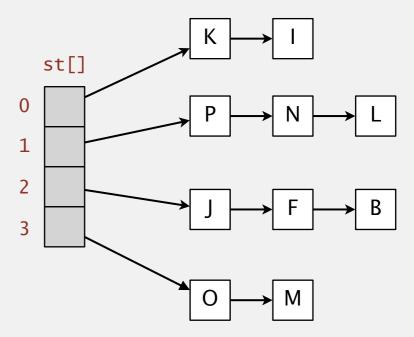
before resizing (N/M = 8)

Deletion in a separate-chaining hash table

- Q. How to delete a key (and its associated value)?
- A. Easy: need to consider only chain containing key.



after deleting C



Symbol table implementations: summary

implementation		guarantee		a	iverage case	ordered	key	
	search	insert	delete	search hit	insert	delete	ops?	interface
sequential search (unordered list)	Ν	Ν	Ν	Ν	Ν	Ν		equals()
binary search (ordered array)	log N	Ν	Ν	log N	N	Ν	~	compareTo()
BST	N	Ν	Ν	log N	log N	\sqrt{N}	~	compareTo()
red-black BST	log N	log N	log N	log N	log N	log N	~	compareTo()
separate chaining	N	N	N	1 *	1 *	1 *		equals() hashCode()

* under uniform hashing assumption

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Collision resolution: open addressing

Open addressing. [Amdahl-Boehme-Rocherster-Samuel, IBM 1953]

- Maintain keys and values in two parallel arrays.
- When a new key collides, find next empty slot, and put it there.

linear-probing hash table (M = 16, N = 10)

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
keys[]	Р	М			А	С		Н	L		Е				R	Х
	put(k	(, 14)						K								
	hash(K) = 7															
vals[]	11	10			9	5		6	12		13				4	8

Hash. Map key to integer *i* between 0 and M - 1.

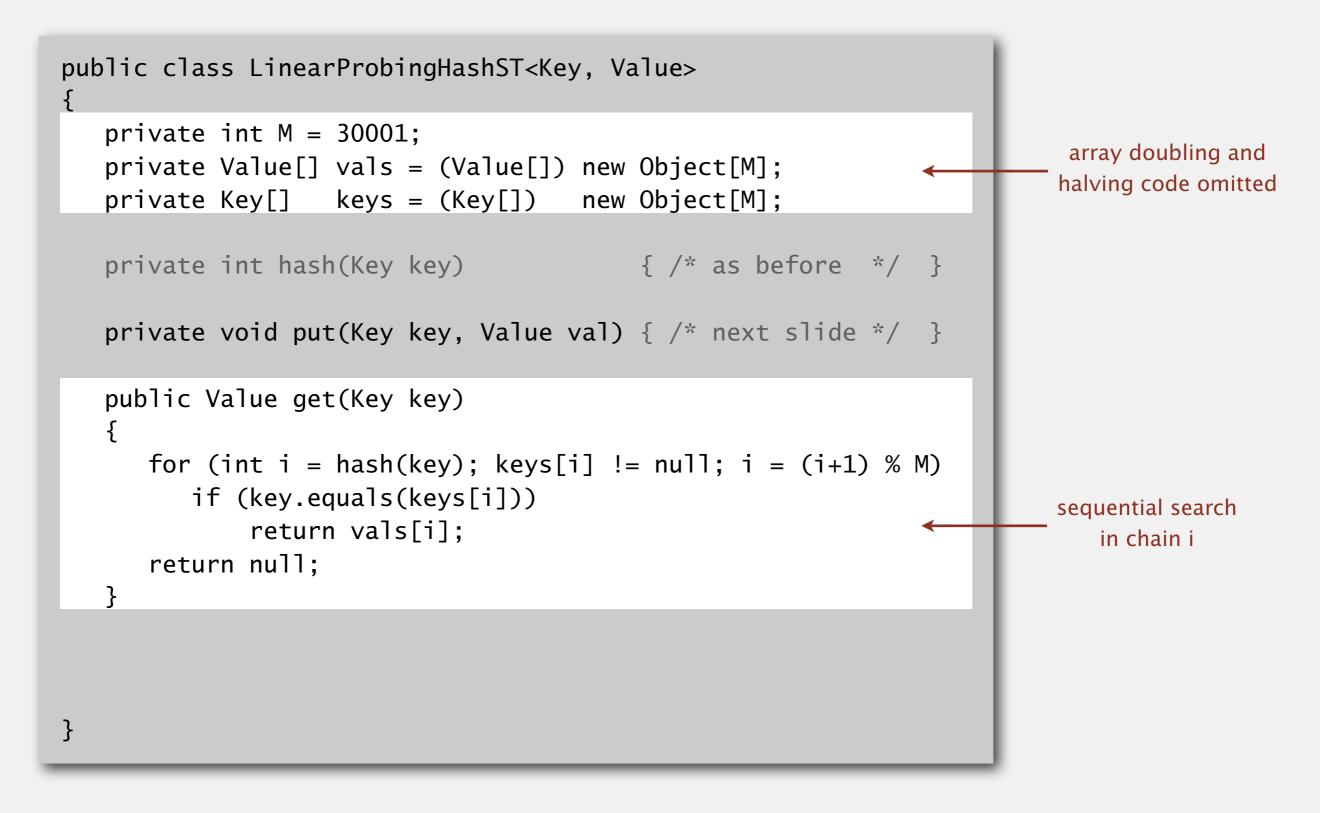
Insert. Put at table index *i* if free; if not try i + 1, i + 2, etc.

Search. Search table index *i*; if occupied but no match, try i + 1, i + 2, etc.

Note. Array size *M* must be greater than number of key-value pairs *N*.



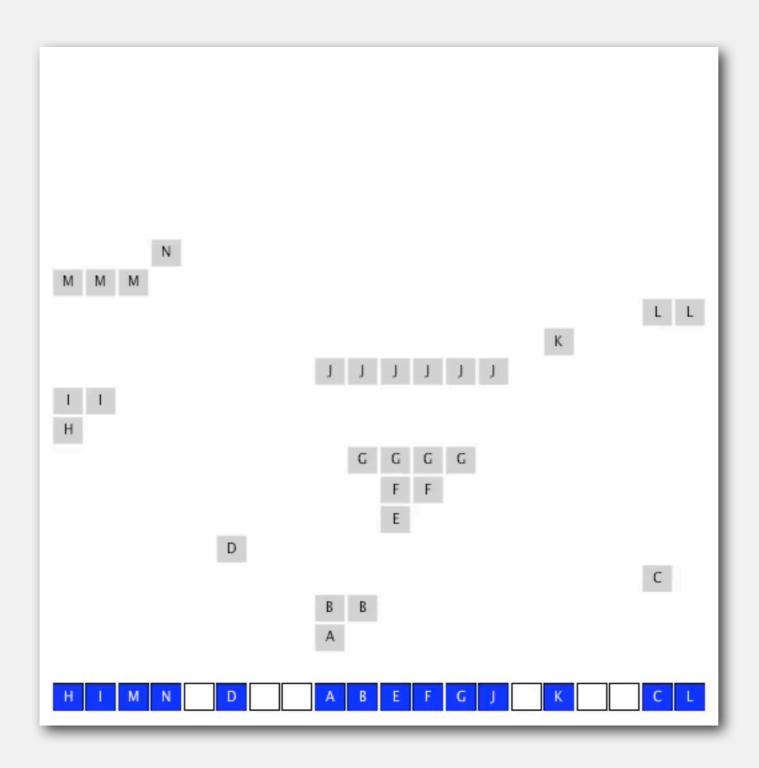




```
public class LinearProbingHashST<Key, Value>
{
  private int M = 30001;
  private Value[] vals = (Value[]) new Object[M];
  private Key[] keys = (Key[]) new Object[M];
  private int hash(Key key) { /* as before */ }
  private Value get(Key key) { /* prev slide */ }
  public void put(Key key, Value val)
   {
     int i;
     for (i = hash(key); keys[i] != null; i = (i+1) % M)
                                                                   sequential search
        if (keys[i].equals(key))
                                                                      in chain i
            break;
     keys[i] = key;
     vals[i] = val;
   }
}
```

Cluster. A contiguous block of items.

Observation. New keys likely to hash into middle of big clusters.



Knuth's parking problem

Model. Cars arrive at one-way street with *M* parking spaces. Each desires a random space *i* : if space *i* is taken, try i + 1, i + 2, etc.

Q. What is mean displacement of a car?



Half-full.With M/2 cars, mean displacement is ~ 5/2.Full.With M cars, mean displacement is ~ $\sqrt{\pi M/8}$.

Key insight. Cannot afford to let linear-probing hash table get too full.

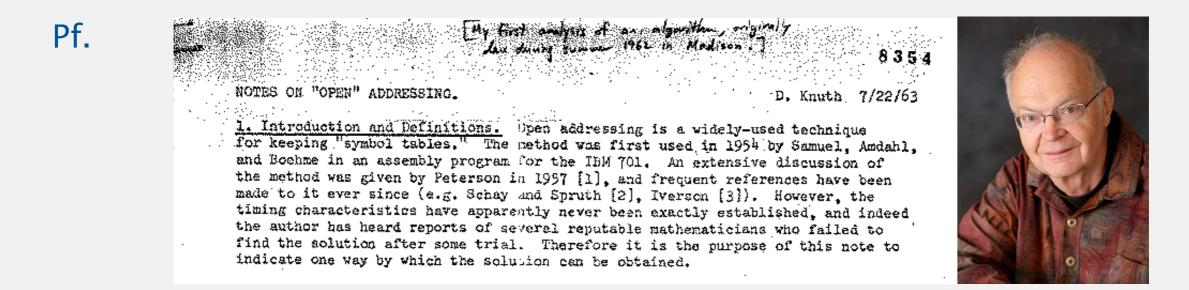
Analysis of linear probing

Proposition. Under uniform hashing assumption, the average # of probes in a linear probing hash table of size *M* that contains $N = \alpha M$ keys is:



search hit

search miss / insert



Parameters.

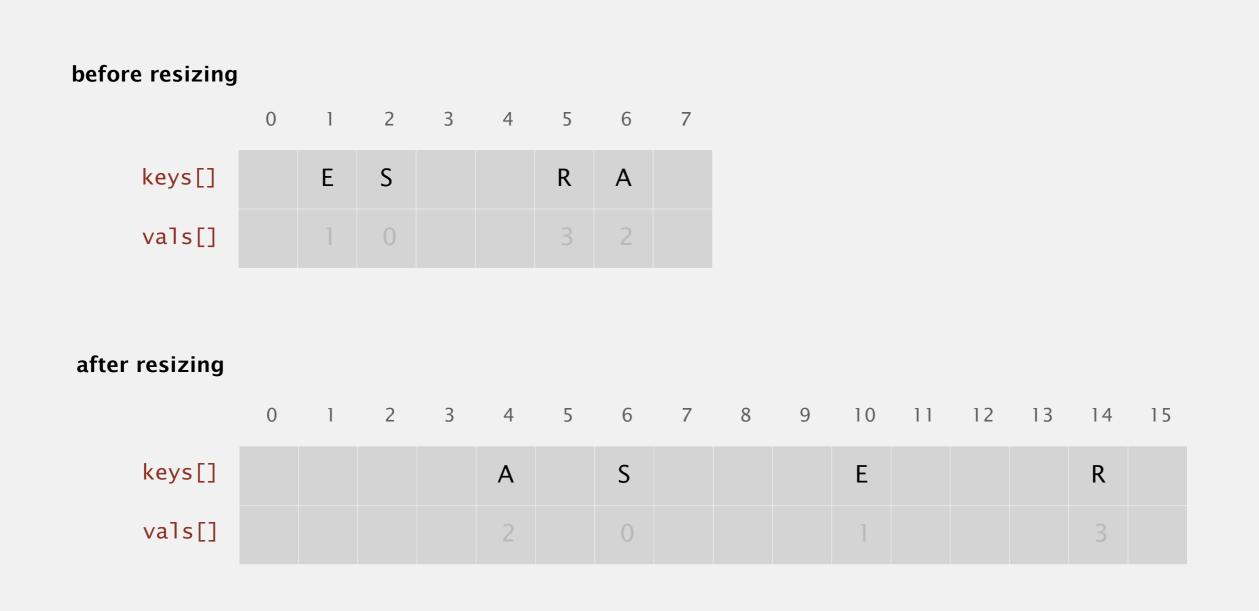
- *M* too large \Rightarrow too many empty array entries.
- M too small \Rightarrow search time blows up.
- Typical choice: $\alpha = N/M \sim \frac{1}{2}$.

probes for search hit is about 3/2# probes for search miss is about 5/2

Resizing in a linear-probing hash table

Goal. Average length of list $N/M \le \frac{1}{2}$.

- Double size of array *M* when $N/M \ge \frac{1}{2}$.
- Halve size of array M when $N/M \le \frac{1}{8}$.
- Need to rehash all keys when resizing.



Deletion in a linear-probing hash table

- Q. How to delete a key (and its associated value)?
- A. Requires some care: can't just delete array entries.



ST implementations: summary

(ordered array) $\log N$ N N $\log N$ $\log N$ \sqrt{N} \checkmark \sim \sim BST N N N $\log N$ $\log N$ $\log N$ $\log N$ $\log N$ \sqrt{N} \checkmark \sim \sim red-black BST $\log N$ <th< th=""><th>key</th></th<>	key							
implementation	search	insert	delete	search hit	insert	delete	ops?	interface
-	Ν	Ν	Ν	Ν	Ν	Ν		equals()
	log N	Ν	Ν	log N	Ν	Ν	~	compareTo()
BST	Ν	Ν	Ν	log N	log N	\sqrt{N}	~	compareTo()
red-black BST	log N	log N	log N	log N	log N	log N	~	compareTo()
separate chaining	N	Ν	Ν	1 *	1 *	1 *		equals() hashCode()
linear probing	Ν	Ν	Ν	1 *	1 *	1 *		equals() hashCode()

* under uniform hashing assumption

3-SUM (REVISITED)

3-SUM. Given N distinct integers, find three such that a + b + c = 0. Goal. N² expected time case, N extra space.

3.4 HASH TABLES

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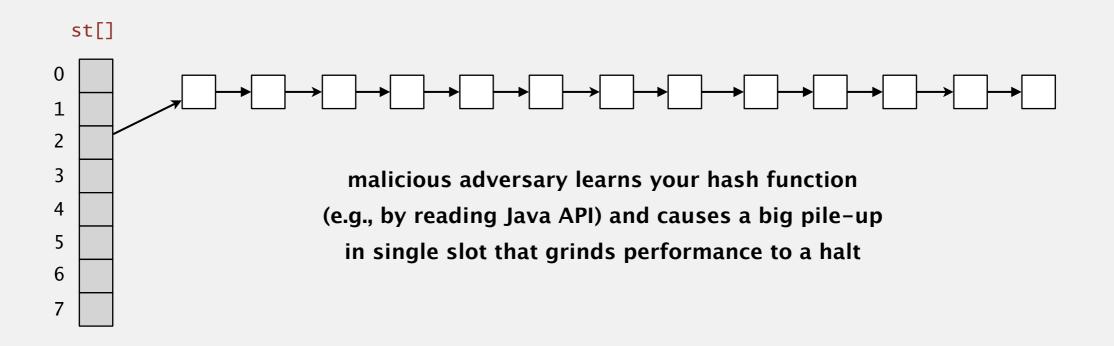
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War story: algorithmic complexity attacks

Q. Is the uniform hashing assumption important in practice?
A. Obvious situations: aircraft control, nuclear reactor, pacemaker, HFT, ...
A. Surprising situations: denial-of-service attacks.



Real-world exploits. [Crosby-Wallach 2003]

- Bro server: send carefully chosen packets to DOS the server, using less bandwidth than a dial-up modem.
- Perl 5.8.0: insert carefully chosen strings into associative array.
- Linux 2.4.20 kernel: save files with carefully chosen names.

A Java bug report.

Jan Lieskovsky 2011-11-01 10:13:47 EDT

Description

Julian Wälde and Alexander Klink reported that the String.hashCode() hash function is not sufficiently collision resistant. hashCode() value is used in the implementations of HashMap and Hashtable classes:

http://docs.oracle.com/javase/6/docs/api/java/util/HashMap.html
http://docs.oracle.com/javase/6/docs/api/java/util/Hashtable.html

A specially-crafted set of keys could trigger hash function collisions, which can degrade performance of HashMap or Hashtable by changing hash table operations complexity from an expected/average O(1) to the worst case O(n). Reporters were able to find colliding strings efficiently using equivalent substrings and meet in the middle techniques.

This problem can be used to start a denial of service attack against Java applications that use untrusted inputs as HashMap or Hashtable keys. An example of such application is web application server (such as tomcat, see bug #750521) that may fill hash tables with data from HTTP request (such as GET or POST parameters). A remote attack could use that to make JVM use excessive amount of CPU time by sending a POST request with large amount of parameters which hash to the same value.

This problem is similar to the issue that was previously reported for and fixed in e.g. perl: http://www.cs.rice.edu/~scrosby/hash/CrosbyWallach UsenixSec2003.pdf Goal. Find family of strings with the same hashCode(). Solution. The base-31 hash code is part of Java's String API.

key	hashCode()
"Aa"	2112
"BB"	2112

key	hashCode()	key	hashCode()
"AaAaAaAa"	-540425984	"BBAaAaAa"	-540425984
"AaAaAaBB"	-540425984	"BBAaAaBB"	-540425984
"AaAaBBAa"	-540425984	"BBAaBBAa"	-540425984
"AaAaBBBB"	-540425984	"BBAaBBBB"	-540425984
"AaBBAaAa"	-540425984	"BBBBAaAa"	-540425984
"AaBBAaBB"	-540425984	"BBBBAaBB"	-540425984
"AaBBBBAa"	-540425984	"BBBBBBAa"	-540425984
"AaBBBBBB"	-540425984	"BBBBBBBBB"	-540425984

2^N strings of length 2N that hash to same value!

Diversion: one-way hash functions

One-way hash function. "Hard" to find a key that will hash to a desired value (or two keys that hash to same value).

```
Ex. MD4, MD5, SHA-0, SHA-1, SHA-2, WHIRLPOOL, RIPEMD-160, ....
known to be insecure
```

```
String password = args[0];
MessageDigest sha1 = MessageDigest.getInstance("SHA1");
byte[] bytes = sha1.digest(password);
/* prints bytes as hex string */
```

Applications. Crypto, message digests, passwords, Bitcoin, Caveat. Too expensive for use in ST implementations.

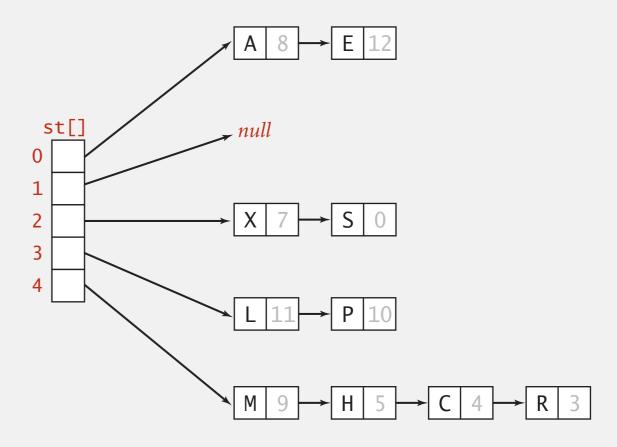
Separate chaining vs. linear probing

Separate chaining.

- Performance degrades gracefully.
- Clustering less sensitive to poorly-designed hash function.

Linear probing.

- Less wasted space.
- Better cache performance.



	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
keys[]	Р	М			А	С	S	Н	L		E				R	Х
vals[]	10	9			8	4	0	5	11		12				3	7

Many improved versions have been studied.

Two-probe hashing. [separate-chaining variant]

- Hash to two positions, insert key in shorter of the two chains.
- Reduces expected length of the longest chain to $\sim \lg \ln N$.

Double hashing. [linear-probing variant]

- Use linear probing, but skip a variable amount, not just 1 each time.
- Effectively eliminates clustering.
- Can allow table to become nearly full.
- More difficult to implement delete.

Cuckoo hashing. [linear-probing variant]

- Hash key to two positions; insert key into either position; if occupied, reinsert displaced key into its alternative position (and recur).
- Constant worst-case time for search.



Hash tables.

- Simpler to code.
- No effective alternative for unordered keys.
- Faster for simple keys (a few arithmetic ops versus log *N* compares).
- Better system support in Java for String (e.g., cached hash code).

Balanced search trees.

- Stronger performance guarantee.
- Support for ordered ST operations.
- Easier to implement compareTo() correctly than equals() and hashCode().

Java system includes both.

- Red-black BSTs: java.util.TreeMap, java.util.TreeSet.
- Hash tables: java.util.HashMap, java.util.IdentityHashMap.

Algorithms

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3.5 SYMBOL TABLE APPLICATIONS

sets

dictionary clients

indexing clients

sparse vectors

Algorithms FOURTH EDITION

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Last updated on Mar 4, 2015, 10:04 AM

3.5 SYMBOL TABLE APPLICATIONS

▶ sets

dictionary clients

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Mathematical set. A collection of distinct keys.

public	class SET <key extend<="" th=""><th>ds Comparable<key>></key></th></key>	ds Comparable <key>></key>
	SET()	create an empty set
void	add(Key key)	add the key to the set
boolean	<pre>contains(Key key)</pre>	is the key in the set?
void	remove(Key key)	remove the key from the set
int	size()	number of keys in the set
Iterator <key></key>	iterator()	all keys in the set

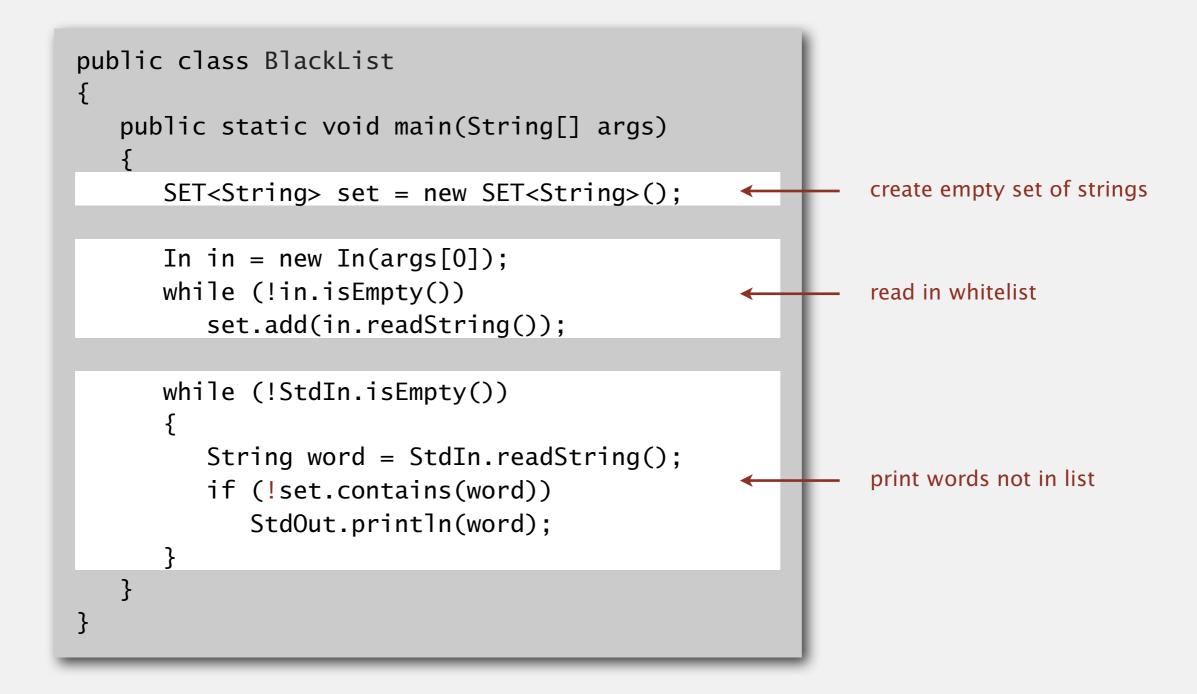
Q. How to implement efficiently?

- Read in a list of words from one file.
- Print out all words from standard input that are { in, not in } the list.

application	purpose	key	in list
spell checker	identify misspelled words	word	dictionary words
browser	mark visited pages	URL	visited pages
parental controls	block sites	URL	bad sites
chess	detect draw	board	positions
spam filter	eliminate spam	IP address	spam addresses
credit cards	check for stolen cards	number	stolen cards

Exception filter: Java implementation

- Read in a list of words from one file.
- Print out all words from standard input that are not in the list.



3.5 SYMBOL TABLE APPLICATIONS

dictionary clients

indexing clients

sparse vectors

sets

Algorithms

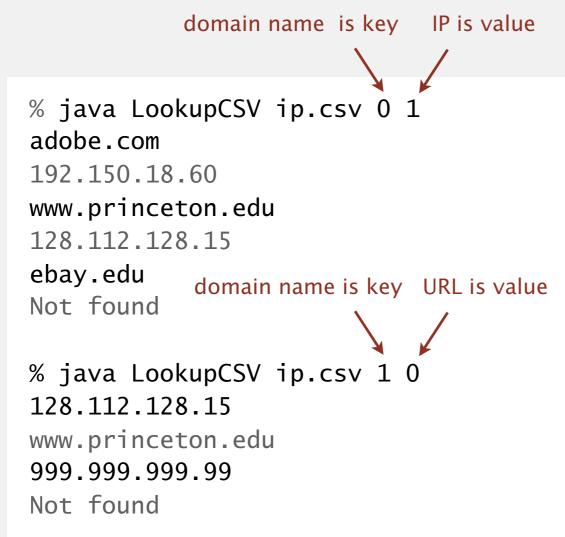
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Dictionary lookup

Command-line arguments.

- A comma-separated value (CSV) file.
- Key field.
- Value field.
- Ex 1. DNS lookup.



% more ip.csv

www.princeton.edu,128.112.128.15 www.cs.princeton.edu, 128.112.136.35 www.math.princeton.edu, 128.112.18.11 www.cs.harvard.edu,140.247.50.127 www.harvard.edu,128.103.60.24 www.yale.edu,130.132.51.8 www.econ.yale.edu,128.36.236.74 www.cs.yale.edu, 128.36.229.30 espn.com,199.181.135.201 yahoo.com,66.94.234.13 msn.com,207.68.172.246 google.com, 64.233.167.99 baidu.com,202.108.22.33 yahoo.co.jp,202.93.91.141 sina.com.cn,202.108.33.32 ebay.com,66.135.192.87 adobe.com, 192.150.18.60 163.com,220.181.29.154 passport.net,65.54.179.226 tom.com, 61.135.158.237 nate.com,203.226.253.11 cnn.com, 64.236.16.20 daum.net,211.115.77.211 blogger.com,66.102.15.100 fastclick.com,205.180.86.4 wikipedia.org,66.230.200.100 rakuten.co.jp,202.72.51.22

Dictionary lookup

Command-line arguments.

- A comma-separated value (CSV) file.
- Key field.
- Value field.
- Ex 2. Amino acids.

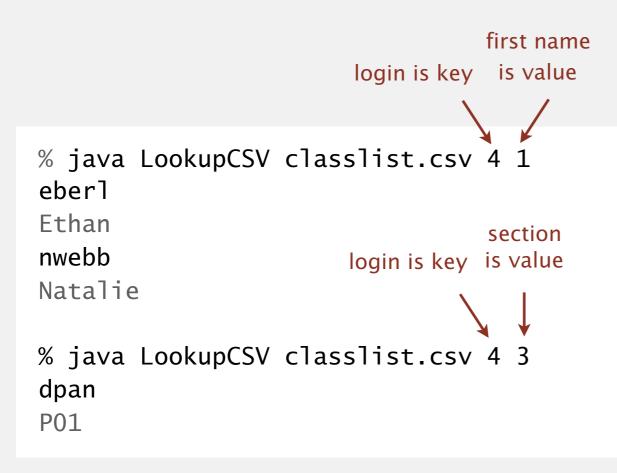
codon is key name is value
% java LookupCSV amino.csv 0 3
ACT
Threonine
TAG
Stop
CAT
Histidine

% more amino.csv TTT, Phe, F, Phenylalanine TTC, Phe, F, Phenylalanine TTA, Leu, L, Leucine TTG,Leu,L,Leucine TCT, Ser, S, Serine TCC, Ser, S, Serine TCA, Ser, S, Serine TCG, Ser, S, Serine TAT, Tyr, Y, Tyrosine TAC, Tyr, Y, Tyrosine TAA, Stop, Stop, Stop TAG, Stop, Stop, Stop TGT,Cys,C,Cysteine TGC,Cys,C,Cysteine TGA, Stop, Stop, Stop TGG, Trp, W, Tryptophan CTT,Leu,L,Leucine CTC, Leu, L, Leucine CTA, Leu, L, Leucine CTG, Leu, L, Leucine CCT, Pro, P, Proline CCC, Pro, P, Proline CCA, Pro, P, Proline CCG, Pro, P, Proline CAT, His, H, Histidine CAC, His, H, Histidine CAA,Gln,Q,Glutamine CAG,Gln,Q,Glutamine CGT, Arg, R, Arginine CGC, Arg, R, Arginine . . .

Command-line arguments.

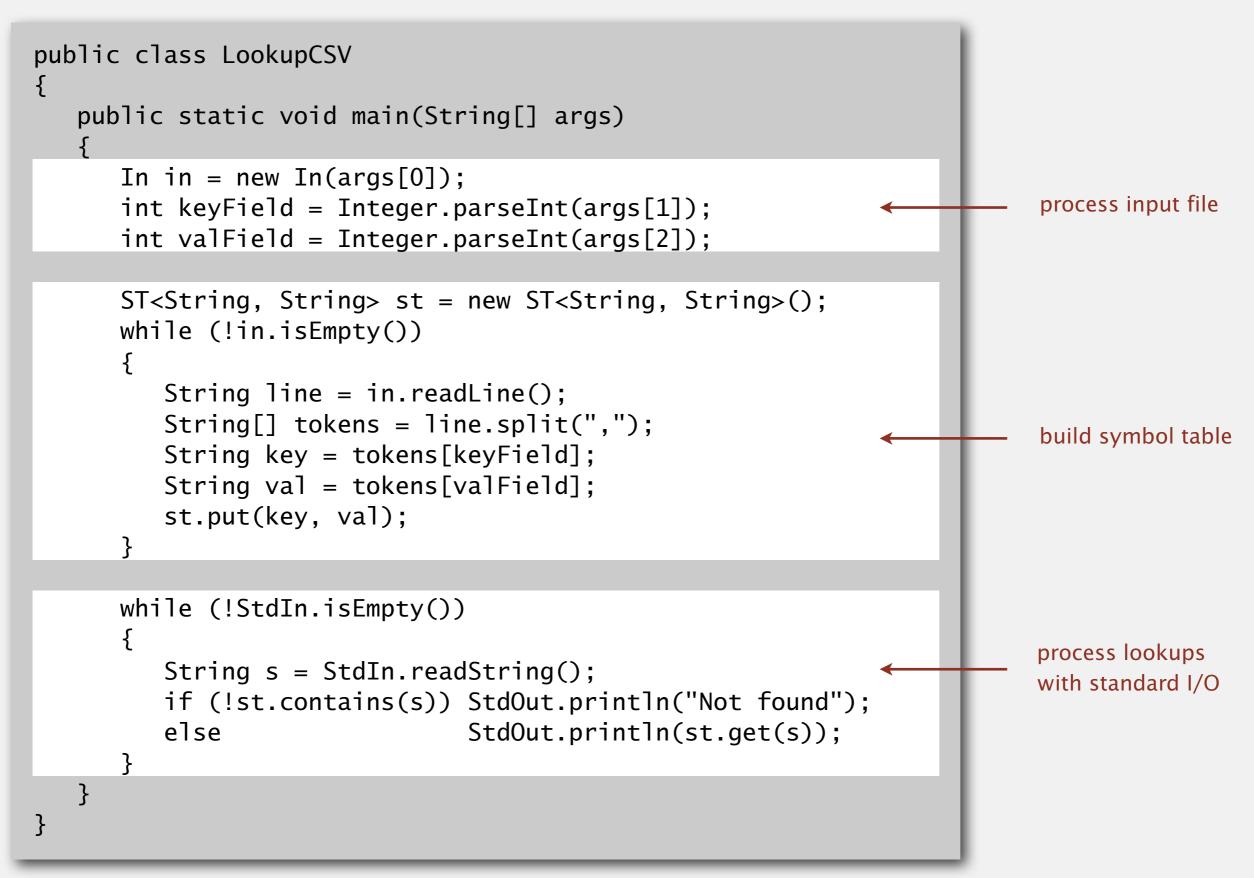
- A comma-separated value (CSV) file.
- Key field.
- Value field.

Ex 3. Class list.



% more classlist.csv 13, Berl, Ethan Michael, P01, eberl 12, Cao, Phillips Minghua, P01, pcao 11, Chehoud, Christel, P01, cchehoud 10, Douglas, Malia Morioka, P01, malia 12, Haddock, Sara Lynn, P01, shaddock 12, Hantman, Nicole Samantha, P01, nhantman 11, Hesterberg, Adam Classen, P01, ahesterb 13, Hwang, Roland Lee, P01, rhwang 13, Hyde, Gregory Thomas, P01, ghyde 13,Kim,Hyunmoon,P01,hktwo 12,Korac,Damjan,P01,dkorac 11, MacDonald, Graham David, P01, gmacdona 10, Michal, Brian Thomas, P01, bmichal 12, Nam, Seung Hyeon, P01, seungnam 11, Nastasescu, Maria Monica, P01, mnastase 11, Pan, Di, P01, dpan 12, Partridge, Brenton Alan, P01, bpartrid 13, Rilee, Alexander, P01, arilee 13, Roopakalu, Ajay, P01, aroopaka 11, Sheng, Ben C, P01, bsheng 12, Webb, Natalie Sue, P01, nwebb •

Dictionary lookup: Java implementation



3.5 SYMBOL TABLE APPLICATIONS

dictionary clients

sets

Algorithms

indexing clients

sparse vectors

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Goal. Index a PC (or the web).

Spotlight	searching challenge 🛛 🛞
	Show All (200)
Top Hit	🙀 10Hashing
Documents	 mobydick.txt movies.txt Papers/Abstracts score.card.txt Requests
Mail Messages	 Re: Draft of lecture on symb SODA 07 Final Accepts SODA 07 Summary Got-it No Subject
PDF Documents	 08BinarySearchTrees.pdf 07SymbolTables.pdf 07SymbolTables.pdf 06PriorityQueues.pdf 06PriorityQueues.pdf
Presentations	 06PriorityQueues.pdf 10Hashing 07SymbolTables 06PriorityQueues

File indexing

Goal. Given a list of text files, create an index so that you can efficiently find all files containing a given query string.

% ls *.txt
aesop.txt magna.txt moby.txt
sawyer.txt tale.txt

% java FileIndex *.txt

freedom
magna.txt moby.txt tale.txt

whale
moby.txt

lamb
sawyer.txt aesop.txt

% ls *.java BlackList.java Concordance.java DeDup.java FileIndex.java ST.java SET.java WhiteList.java

% java FileIndex *.java

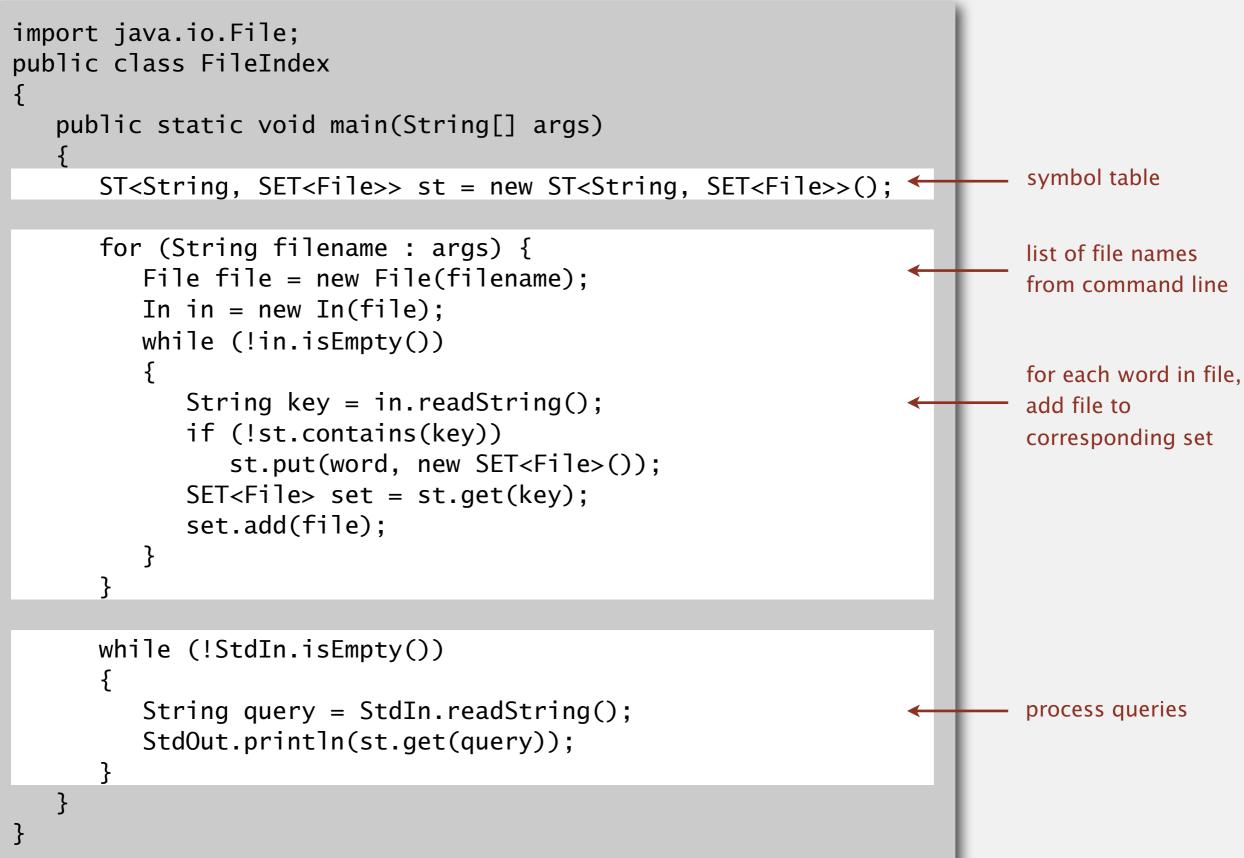
import
FileIndex.java SET.java ST.java

Comparator null

Which data type below would be the best choice to represent the file index?

- A. SET<ST<File, String>>
- B. SET<ST<String, File>>
- C. ST<File, SET<String>>
- D. ST<String, SET<File>>
- E. I don't know.

File indexing



Book index

Goal. Index for an e-book.

Index

Abstract data type (ADT), 127-195 abstract classes, 163 classes, 129-136 collections of items, 137-139 creating, 157-164 defined, 128 duplicate items, 173-176 equivalence-relations, 159-162 FIFO queues, 165-171 first-class, 177-186 generic operations, 273 index items, 177 insert/remove operations, 138-139 modular programming, 135 polynomial, 188-192 priority queues, 375-376 pushdown stack, 138-156 stubs, 135 symbol table, 497-506 ADT interfaces array (myArray), 274 complex number (Complex), 181 existence table (ET), 663 full priority queue (PQfull), 397 indirect priority queue (PQi), 403 itcm (myItem), 273, 498 key (myKey), 498 polynomial (Poly), 189 point (Point), 134 priority queue (PQ), 375 queue of int (intQueue), 166

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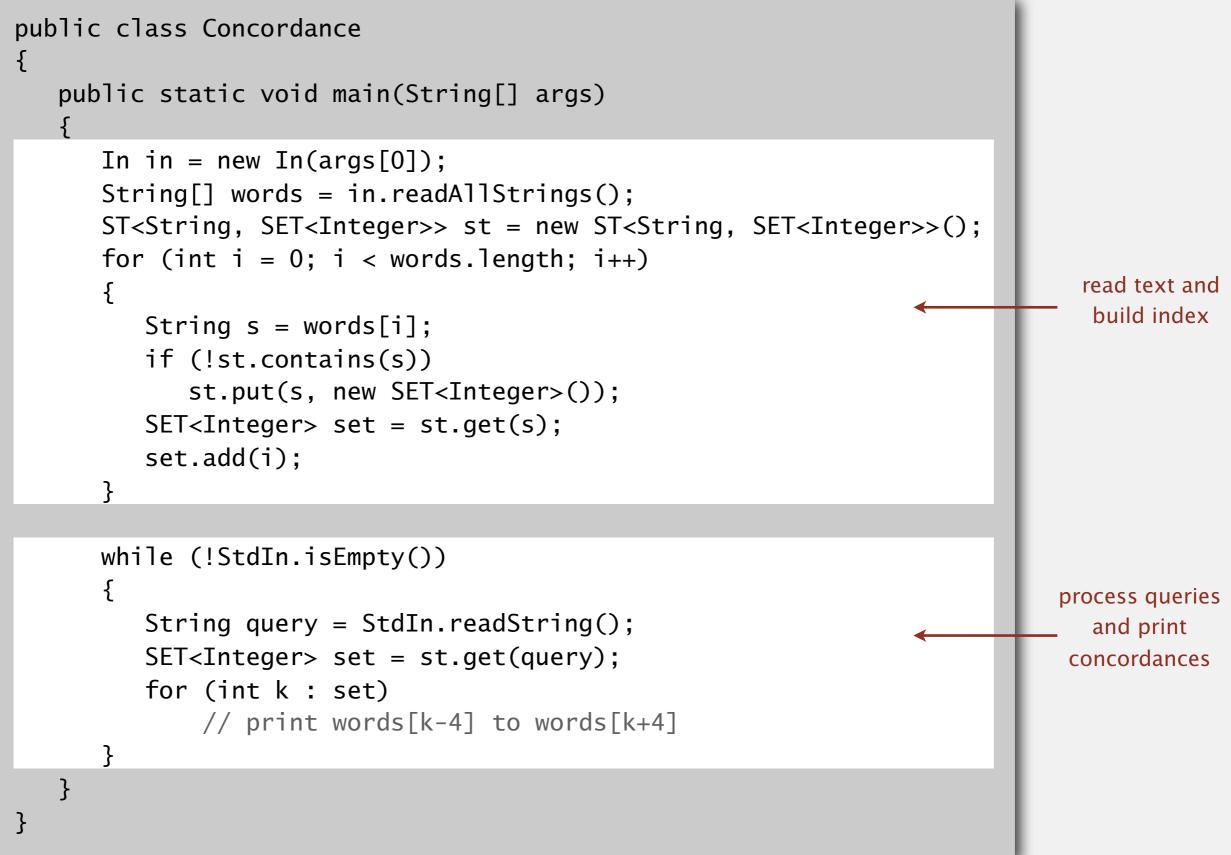
Concordance

Goal. Preprocess a text corpus to support concordance queries: given a word, find all occurrences with their immediate contexts.

```
% java Concordance tale.txt
cities
tongues of the two *cities* that were blended in
majesty
their turnkeys and the *majesty* of the law fired
me treason against the *majesty* of the people in
of his most gracious *majesty* king george the third
princeton
no matches
```

Solution. Key = query string; value = set of indices containing that string.

Concordance



3.5 SYMBOL TABLE APPLICATIONS

Algorithms

sparse vectors

indexing clients

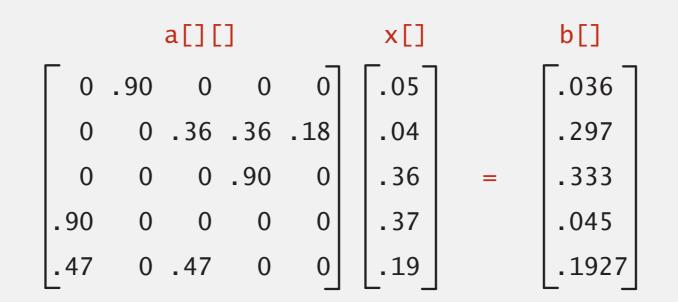
dictionary clients

sets

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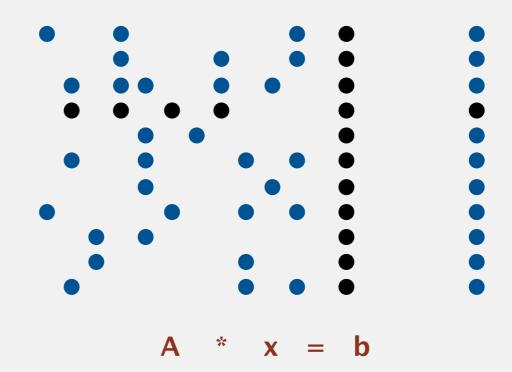
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Matrix-vector multiplication (standard implementation)



Problem. Sparse matrix-vector multiplication.

Assumptions. Matrix dimension is 10,000; average nonzeros per row ~ 10.



Vector representations

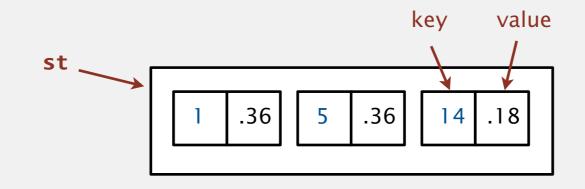
1d array (standard) representation.

- Constant time access to elements.
- Space proportional to *N*.

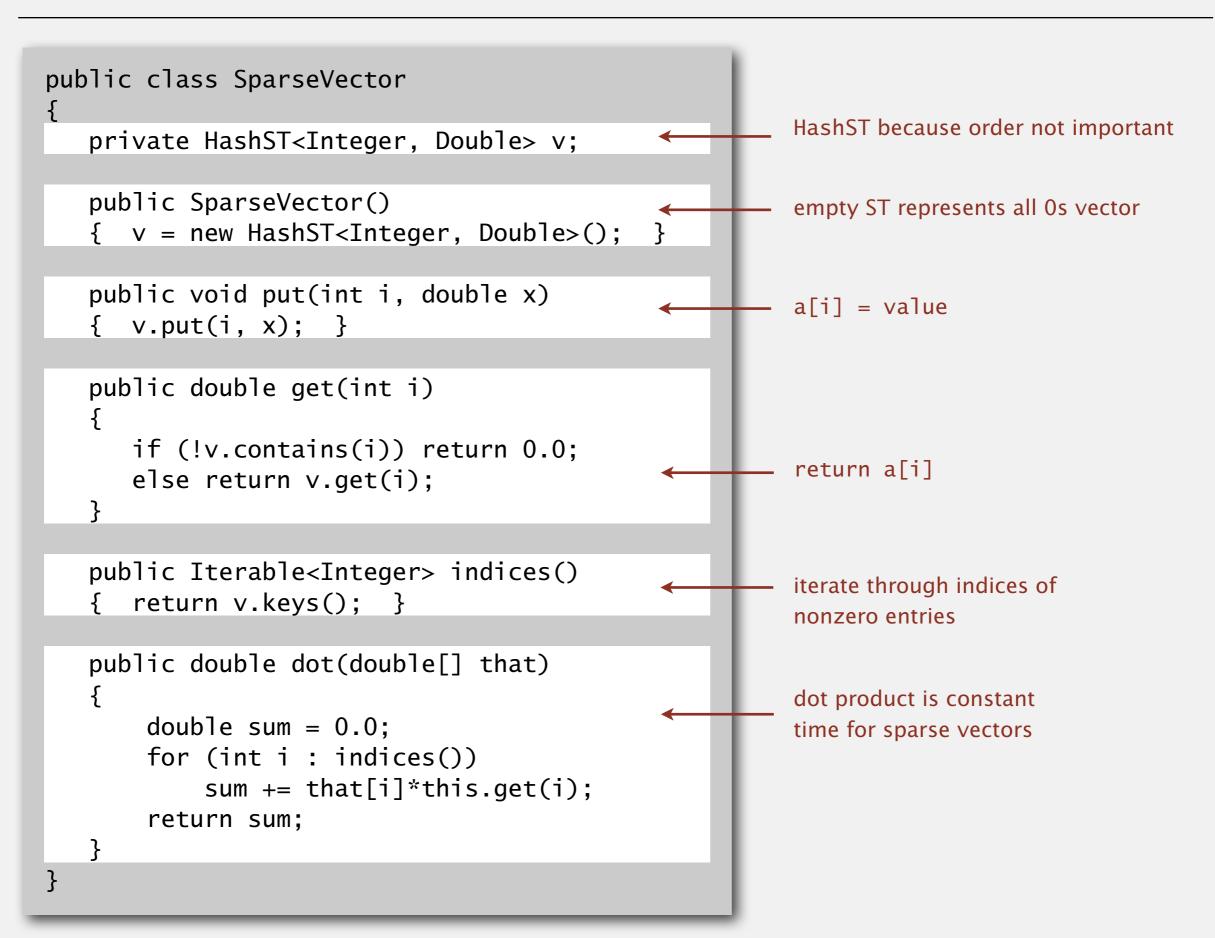
0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19
0	.36	0	0	0	.36	0	0	0	0	0	0	0	0	.18	0	0	0	0	0

Symbol table representation.

- Key = index, value = entry.
- Efficient iterator.
- Space proportional to number of nonzeros.



Sparse vector data type



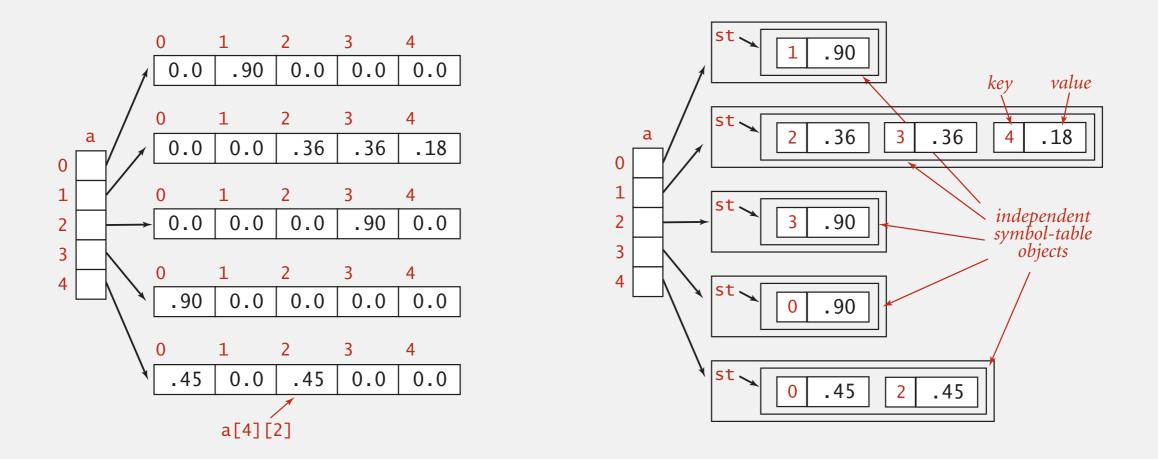
Matrix representations

2D array (standard) matrix representation: Each row of matrix is an array.

- Constant time access to elements.
- Space proportional to N².

Sparse matrix representation: Each row of matrix is a sparse vector.

- Efficient access to elements.
- Space proportional to number of nonzeros (plus *N*).



Sparse matrix-vector multiplication

