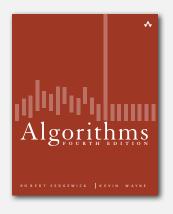
# 3.4 Hash Tables



- hash functions
- → separate chaining
- ▶ linear probing
- applications

# Optimize judiciously

"More computing sins are committed in the name of efficiency (without necessarily achieving it) than for any other single reason including blind stupidity. " — William A. Wulf

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

"We follow two rules in the matter of optimization: Rule 1: Don't do it.
Rule 2 (for experts only). Don't do it yet - that is, not until you have a perfectly clear and unoptimized solution." — M. A. Jackson

**Reference: Effective Java by Joshua Bloch** 



2

0

2

3 "it"

#### Algorithms, 4th Edition · Robert Sedgewick and Kevin Wayne · Copyright © 2002–2011 · August 2, 2011 12:18:41 PM

#### ST implementations: summary

implementation	guarantee			average case			ordered	operations
implementation	search	insert	delete	search hit	insert	delete	iteration?	on keys
sequential search (linked list)	Ν	N	N	N/2	N	N/2	no	equals()
binary search (ordered array)	lg N	N	N	lg N	N/2	N/2	yes	compareTo()
BST	N	N	N	1.38 lg N	1.38 lg N	?	yes	compareTo()
red-black tree	2 lg N	2 lg N	2 lg N	1.00 lg N	1.00 lg N	1.00 lg N	yes	compareTo()

Hashing: basic plan

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

Hash function. Method for computing array index from key.

hash("it") = 3

#### Issues.

- Computing the hash function.
- Equality test: Method for checking whether two keys are equal.

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).

Hash function. Method for computing array index from key.

hash("it") = 3 ?? 4 hash("times") = 3 5

# 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).

# hash functions

- ▶ separate chaining ▶ linear probing
- ▶ applications

# 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.

thoroughly researched problem, still problematic in practical applications

# Ex 1. Phone numbers.

- Bad: first three digits.
- Better: last three digits.

# Ex 2. Social Security numbers. -

573 = California, 574 = Alaska (assigned in chronological order within geographic region)

- Bad: first three digits.
- Better: last three digits.

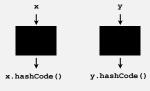
# Practical challenge. Need different approach for each key type.

# Java's hash code conventions

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. Trivial (but poor) implementation. Always return 17. Customized implementations. Integer, Double, String, File, URL, Date, ... User-defined types. Users are on their own.

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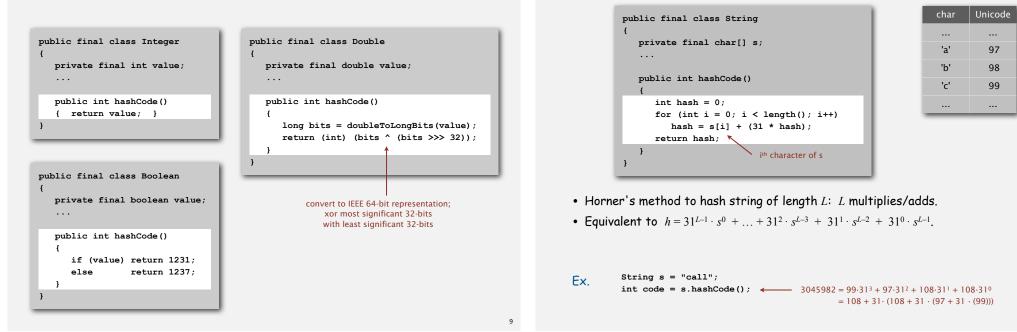
key

table

index

# Implementing hash code: integers, booleans, and doubles

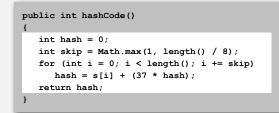
# Implementing hash code: strings



# War story: String hashing in Java

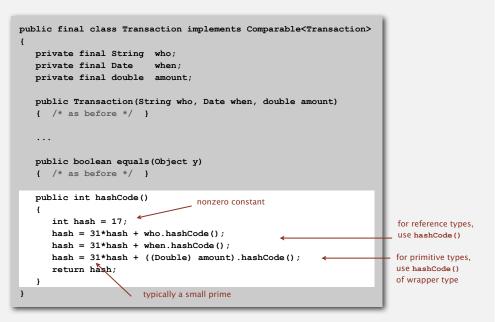
#### String hashCode() in Java 1.1.

- For long strings: only examine 8-9 evenly spaced characters.
- Benefit: saves time in performing arithmetic.



• Downside: great potential for bad collision patterns.

#### 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 a reference type, use hashCode (). \_\_\_\_\_ applies rule recursively

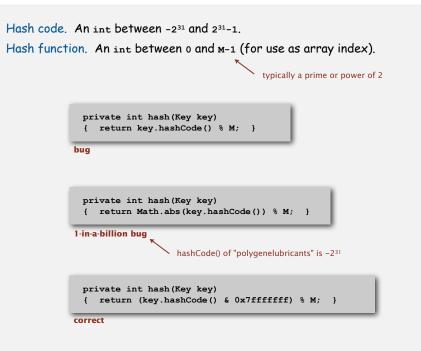
In practice. Recipe works reasonably well; used in Java libraries.

In theory. Need a theorem for each type to ensure reliability.

Basic rule. Need to use the whole key to compute hash code;

consult an expert for state-of-the-art hash codes.

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 ~  $\sqrt{\pi M/2}$  tosses.

Coupon collector. Expect every bin has  $\ge 1$  ball after ~  $M \ln M$  tosses.

Load balancing. After M tosses, expect most loaded bin has  $\Theta(\log M / \log \log M)$  balls.

# Uniform hashing assumption

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

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





Hash value frequencies for words in Tale of Two Cities (M = 97)

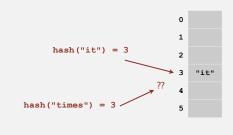
Java's String data uniformly distribute the keys of Tale of Two Cities

#### Collisions

Collision. Two distinct keys hashing to same index.

- Birthday problem ⇒ can't avoid collisions unless you have a ridiculous (quadratic) amount of memory.
- Coupon collector + load balancing  $\Rightarrow$  collisions will be evenly distributed.

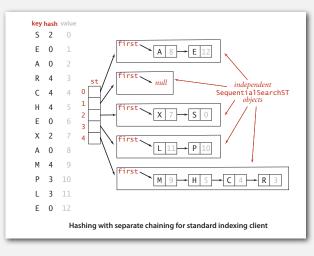
Challenge. Deal with collisions efficiently.



# Separate chaining ST

# 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*<sup>th</sup> chain (if not already there).
- Search: only need to search *i*<sup>th</sup> chain.



▶ separate chaining

# Separate chaining ST: Java implementation

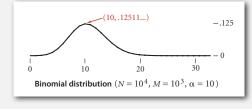
1	<pre>private int N; // number of key-value pairs</pre>
1	private int M; // hash table size
1	<pre>private SequentialSearchST<key, value=""> [] st; // array of STs</key,></pre>
	<pre>public SeparateChainingHashST()</pre>
1	<pre>public SeparateChainingHashST(int M) {</pre>
	<pre>this.M = M; st = (SequentialSearchST<key, value="">[]) new SequentialSearchST[M]; for (int i = 0; i &lt; M; i++) st[i] = new SequentialSearchST<key, value="">();</key,></key,></pre>
	<pre>st = (SequentialSearchST<key, value="">[]) new SequentialSearchST[M]; for (int i = 0; i &lt; M; i++)</key,></pre>
	<pre>st = (SequentialSearchST<key, value="">[]) new SequentialSearchST[M]; for (int i = 0; i &lt; M; i++)     st[i] = new SequentialSearchST<key, value="">(); } private int hash(Key key)</key,></key,></pre>
	<pre>st = (SequentialSearchST<key, value="">[]) new SequentialSearchST[M]; for (int i = 0; i &lt; M; i++)     st[i] = new SequentialSearchST<key, value="">(); }</key,></key,></pre>
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:	<pre>st = (SequentialSearchST<key, value="">[]) new SequentialSearchST[M]; for (int i = 0; i &lt; M; i++)    st[i] = new SequentialSearchST<key, value="">(); } private int hash(Key key) { return (key.hashCode() &amp; 0x7fffffff) % M; } public Value get(Key key) { return st[hash(key)].get(key); }</key,></key,></pre>
1	<pre>st = (SequentialSearchST<key, value="">[]) new SequentialSearchST[M]; for (int i = 0; i &lt; M; i++) st[i] = new SequentialSearchST<key, value="">(); } private int hash(Key key) { return (key.hashCode() &amp; 0x7fffffff) % M; } public Value get(Key key)</key,></key,></pre>

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# Analysis of separate chaining

Proposition. Under uniform hashing assumption, probability 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.



#### \_ equals() and hashCode()

Consequence. Number of probes for search/insert is proportional to N/M.

- M too large  $\Rightarrow$  too many empty chains.
- M too small  $\Rightarrow$  chains too long.
- Typical choice:  $M \sim N/5 \Rightarrow$  constant-time ops.

# ST implementations: summary

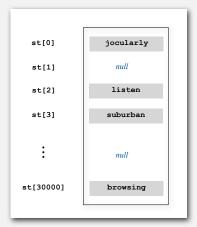
implementation	guarantee			average case			ordered	operations
mplementation	search	insert	delete	search hit	insert	delete	iteration?	on keys
sequential search (linked list)	Ν	Ν	Ν	N/2	N	N/2	no	equals()
binary search (ordered array)	lg N	Ν	Ν	lg N	N/2	N/2	yes	compareTo()
BST	Ν	Ν	Ν	1.38 lg N	1.38 lg N	?	yes	compareTo()
red-black tree	2 lg N	2 lg N	2 lg N	1.00 lg N	1.00 lg N	1.00 lg N	yes	compareTo()
separate chaining	lg N *	lg N *	lg N *	3-5 *	3-5 *	3-5 *	no	equals()

\* under uniform hashing assumption

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# Collision resolution: open addressing

Open addressing. [Amdahl-Boehme-Rocherster-Samuel, IBM 1953] When a new key collides, find next empty slot, and put it there.



linear probing (M = 30001, N = 15000)

#### hash functions

separate chaining

# Inear probing

▶ applications

M times faster than

sequential search

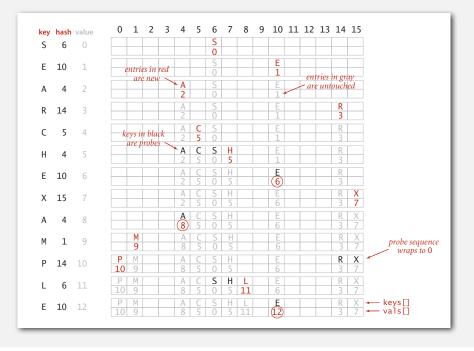
# Linear probing

# Use an array of size M > N.

- 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.

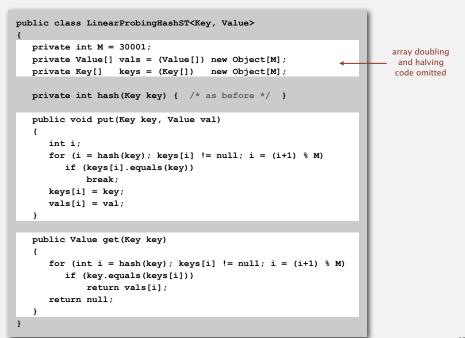


# Linear probing: trace of standard indexing client



#### 25

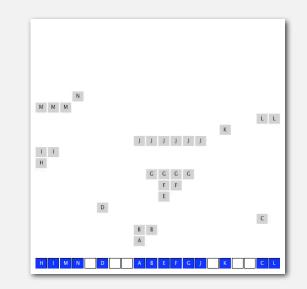
# Linear probing ST implementation



#### Clustering

Cluster. A contiguous block of items.

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



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# 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 ~ 3/2.

Full. With *M* cars, mean displacement is ~  $\sqrt{\pi M/8}$ 

# Analysis of linear probing

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

$$\sim \frac{1}{2} \left( 1 + \frac{1}{1 - \alpha} \right) \qquad \sim \frac{1}{2} \left( 1 + \frac{1}{(1 - \alpha)^2} \right)$$
  
search hit search miss / insert

Pf. [Knuth 1962] A landmark in analysis of algorithms.

#### 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

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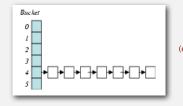
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BST	N	N	N	1.38 lg N	1.38 lg N	?	yes	compareTo()
red-black tree	2 lg N	2 lg N	2 lg N	1.00 lg N	1.00 lg N	1.00 lg N	yes	compareTo()
separate chaining	lg N *	lg N *	lg N *	3-5 *	3-5 *	3-5 *	no	equals()
linear probing	lg N *	lg N *	lg N *	3-5 *	3-5 *	3-5 *	no	equals()

\* under uniform hashing assumption

# War story: algorithmic complexity attacks

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



malicious adversary learns your hash function (e.g., by reading Java API) and causes a big pile-up in single slot that grinds performance to a halt

# 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.

# Algorithmic complexity attack on Java

Goal. Find family of strings with the same hash code. Solution. The base-31 hash code is part of Java's string API.

key	hashCode()		key	hashCode()	key	hashCode()
"Aa"	2112	"2	AaAaAaAa"	-540425984	'BBAaAaAa"	-540425984
"BB"	2112	"2	AaAaAaBB"	-540425984	'BBAaAaBB"	-540425984
_		"2	AaAaBBAa"	-540425984	'BBAaBBAa"	-540425984
		"2	AaAaBBBBB"	-540425984	'BBAaBBBB"	-540425984
		"2	AaBBAaAa"	-540425984	'BBBBAaAa"	-540425984
		"2	AaBBAaBB"	-540425984	'BBBBAaBB"	-540425984
		"2	AaBBBBBAa"	-540425984	'BBBBBBAa"	-540425984
		"2	AaBBBBBBB"	-540425984	BBBBBBBB"	-540425984

2<sup>N</sup> 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 shal = MessageDigest.getInstance("SHA1"); byte[] bytes = shal.digest(password);

/\* prints bytes as hex string \*/

Applications. Digital fingerprint, message digest, storing passwords. Caveat. Too expensive for use in ST implementations.

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# Separate chaining vs. linear probing

#### Separate chaining.

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

# Linear probing.

- Less wasted space.
- Better cache performance.

#### Hashing: variations on the theme

Many improved versions have been studied.

#### Two-probe hashing. (separate-chaining variant)

- Hash to two positions, put key in shorter of the two chains.
- Reduces expected length of the longest chain to  $\log \log N$ .

# 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.
- Difficult to implement delete.

# Hashing vs. balanced search trees

#### Hashing.

- 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 strings (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 trees: java.util.TreeMap, java.util.TreeSet.
- Hashing: java.util.HashMap, java.util.IdentityHashMap.

# hash functions

separate chaining

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40

linear probing

# ▶ applications

#### 37

#### Set API

# Mathematical set. A collection of distinct keys.

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

# Exception filter

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



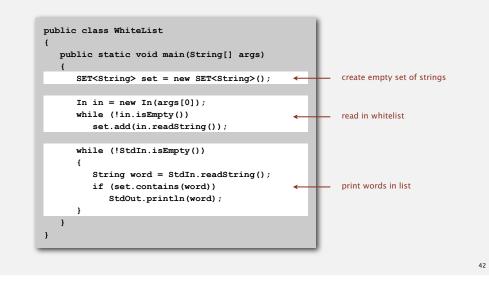
# Exception filter applications

- 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 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 check for stolen cards stolen cards credit cards number

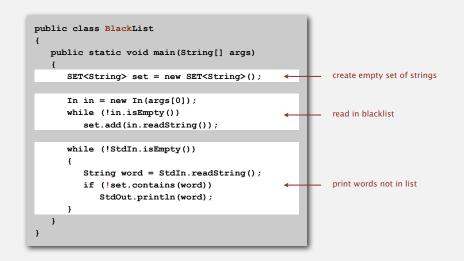
# Exception filter: Java implementation

- Read in a list of words from one file.
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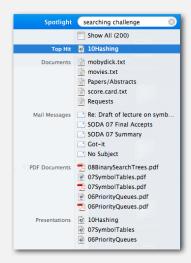
# Exception filter: Java implementation

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



# File indexing

# Goal. Index a PC (or the web).



# File indexing

Goal. Given a list of files specified as command-line arguments, create an index so that 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

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

import FileIndex.java SET.java ST.java

Comparator null

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

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#### Book index

Goal. Index for an e-book.

	stack of int (intStack), 140 symbol table (ST), 503 text index (TI), 525 union-find (UF), 159 Abstract in-place merging, 351-	and linked lists, 92, 94-95 merging, 349-350 multidimensional, 117-118 references, 86-87, 89 sorting, 265-267, 273-276
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# File indexing

