# Algorithms



## Symbol table implementations: summary

implementation		guarantee			average cas	e	ordered	key
	search	insert	delete	search	insert	delete	ops?	interface
sequential search (unordered list)	n	n	n	n	n	n		equals()
binary search (ordered array)	log n	n	n	log n	n	n		compareTo()
BST	n	n	n	log n	log n	$\sqrt{n}$		compareTo()
red-black BST	$\log n$	$\log n$	$\log n$	$\log n$	$\log n$	$\log n$	<b>✓</b>	compareTo()
hashing	n	n	n		1			equals() hashCode()

<sup>†</sup> subject to certain technical assumptions

- Q. Can we do better?
- A. Yes, but only with different access to the symbol table keys.

### Hashing: basic plan

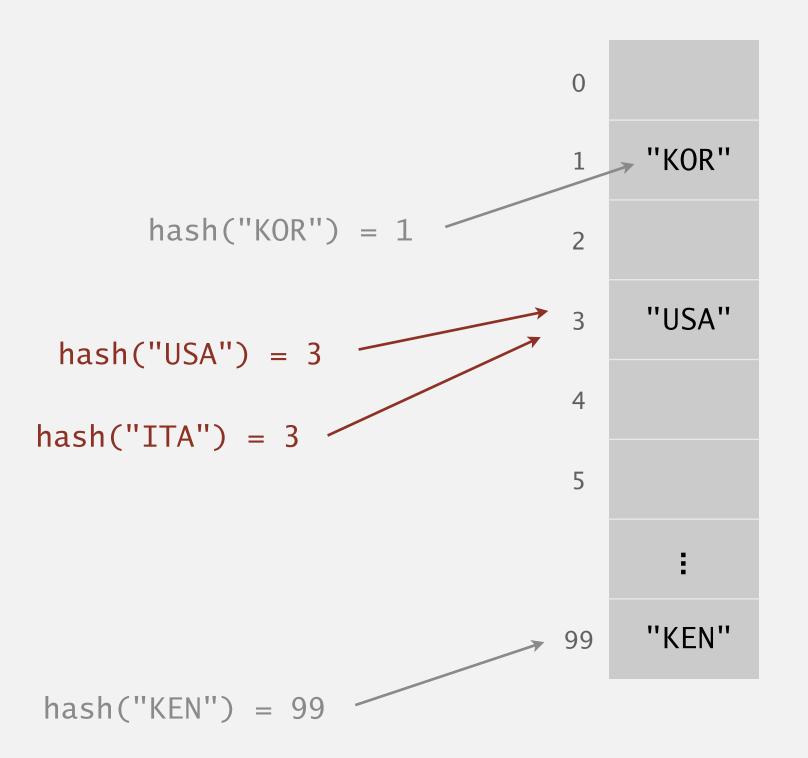
Save key-value pairs in a key-indexed table, where the index is a function of the key.

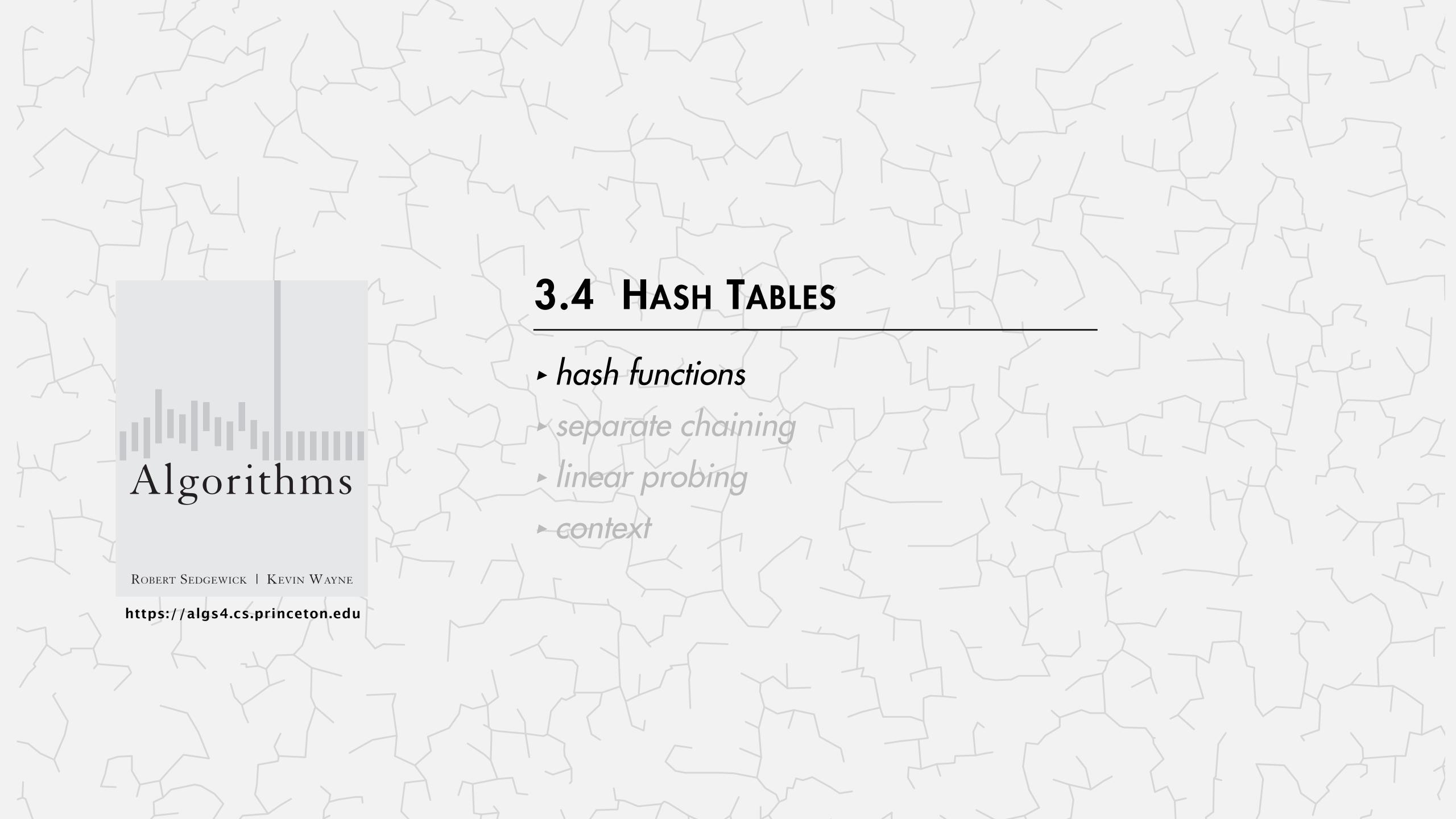
Hash function: Mathematical function that maps (hashes) a key to an array index.

Collision: Two distinct keys that hash to same index.

Issue. Collisions are unavoidable.

- How to limit collisions?
   [good hash functions]
- How to accommodate collisions?
   [novel algorithms and data structures]





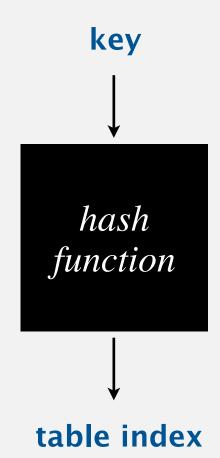
### Designing a hash function

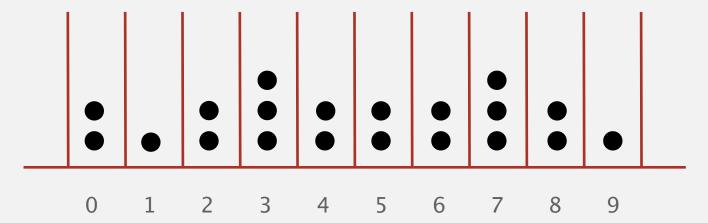
#### Required properties. [for correctness]

- Deterministic.
- Each key hashes to a table index between 0 and m-1.

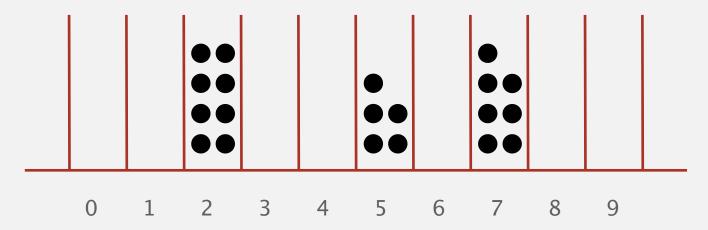
#### Desirable properties. [for performance]

- Very fast to compute.
- For any subset of n input keys, each table index gets approximately  $n \mid m$  keys.





leads to good hash-table performance (m = 10, n = 20)



leads to poor hash-table performance (m = 10, n = 20)

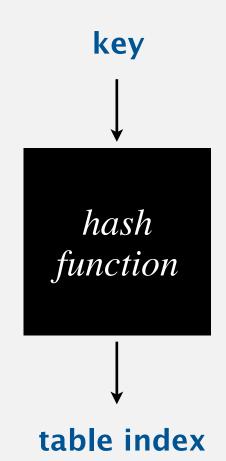
### Designing a hash function

#### Required properties. [for correctness]

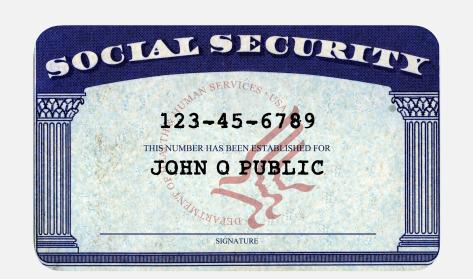
- Deterministic.
- Each key hashes to a table index between 0 and m-1.

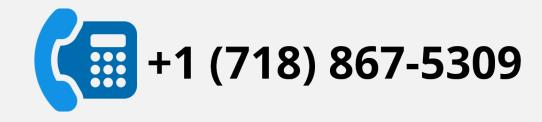
#### Desirable properties. [for performance]

- Very fast to compute.
- For any subset of n input keys, each table index gets approximately  $n \mid m$  keys.



- Ex 1. [m = 10,000] Last 4 digits of U.S. Social Security number.
- Ex 2. [m = 10,000] Last 4 digits of phone number.





## Hash tables: quiz 1



#### Which is the last digit of your day of birth?

- **A.** 0 or 1
- **B.** 2 or 3
- **C.** 4 or 5
- **D.** 6 or 7
- **E.** 8 or 9



## Hash tables: quiz 2



#### Which is the last digit of your year of birth?

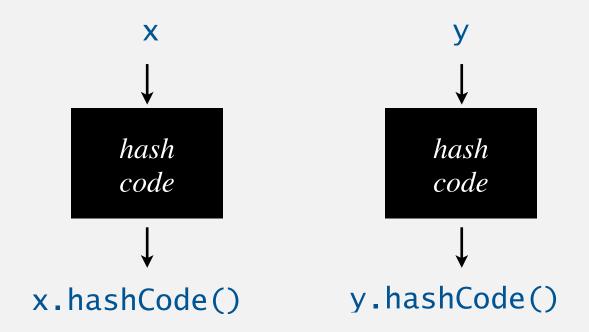
- **A.** 0 or 1
- **B.** 2 or 3
- **C.** 4 or 5
- **D.** 6 or 7
- **E.** 8 or 9



#### Java's hashCode() conventions

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

```
Required. [for correctness] If x.equals(y), then x.hashCode() == y.hashCode(). Highly desirable. [for efficiency] If !x.equals(y), then x.hashCode() != y.hashCode().
```



Customized implementations. Integer, Double, String, java.net.URL, ... Legal (but highly undesirable) implementation. Always return 17. User-defined types. Users are on their own.

### Implementing hashCode(): integers and doubles

#### Java library implementations

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

```
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
if used only least significant 32 bits,
    all integers between -2<sup>21</sup> and 2<sup>21</sup>
       would have same hash code (0)
```

with least significant 32-bits

### Implementing hashCode(): user-defined types

#### 31x + y rule.

- Initialize hash to 1.
- Repeatedly multiply hash by 31 and add hash of each significant field.

```
public final class Transaction
   private final String who;
   private final Date when;
   private final double amount;
   public int hashCode()
      int hash = 1;
                                                                            for reference types,
      hash = 31*hash + who.hashCode();
                                                                            use hashCode()
      hash = 31*hash + when.hashCode();
      hash = 31*hash + ((Double) amount).hashCode();
                                                                            for primitive types,
      return hash;
                                                                            use hashCode()
                                                                            of wrapper type
```

### Implementing hashCode(): user-defined types

#### 31x + y rule.

- Initialize hash to 1.
- Repeatedly multiply hash by 31 and add hash of each significant field.

```
public final class Transaction
   private final String who;
   private final Date when;
   private final double amount;
   public int hashCode()
      return Objects.hash(who, when, amount);
              a varargs method that applies
               31x + y rule to its arguments
```

### Implementing hashCode()

#### "Standard" recipe for user-defined types.

- Combine each significant field using the 31x + y rule.
- Shortcut 1: use Objects.hash() for all fields (except arrays).
- Shortcut 2: use Arrays.hashCode() for array of primitives.
- Shortcut 3: use Arrays.deepHashCode() for array of objects.



Principle. Every significant field contributes to hash.

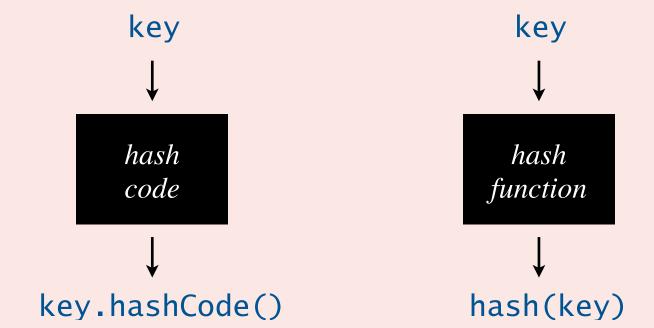
In practice. Recipe above works reasonably well; used in Java libraries.

### Hash tables: quiz 3



#### Which Java function maps hashable keys to integers between 0 and m-1?

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



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

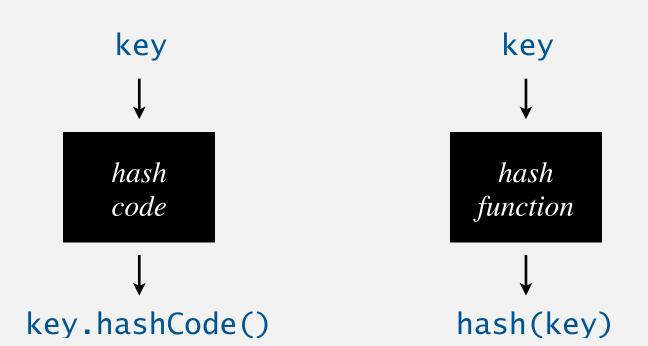
C. Both A and B.

D. Neither A nor B.

### Modular hashing

correct

Hash code. An int between  $-2^{31}$  and  $2^{31}-1$ . Hash function. An int between 0 and m-1 (for use as array index). *m* typically a prime or a power of 2 private int hash(Key key) return key.hashCode() % m; } bug the remainder operator can evaluate to a negative integer private int hash(Key key) return Math.abs(key.hashCode()) % m; } 1-in-a-billion bug hashCode() of "polygenelubricants" and new Double(-0.0) is  $-2^{31}$ private int hash(Key key) { return Math.abs(key.hashCode() % m); }

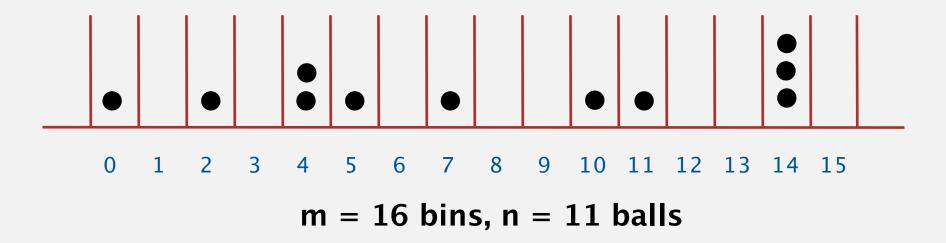


### Uniform hashing assumption

Uniform hashing assumption. Each key is equally likely to hash to any of m possible indices.

and independently of other keys

Bins and balls. Toss n balls uniformly at random into m bins.



#### Bad news. [birthday problem]

- In a random group of 23 people, more likely than not that two people share the same birthday.
- Expect two balls in the same bin after  $\sim \sqrt{\pi\,m\,/\,2}$  tosses.

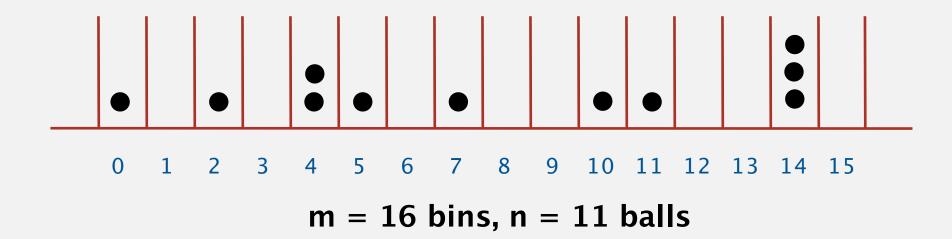
23.9 when m = 365

### Uniform hashing assumption

Uniform hashing assumption. Each key is equally likely to hash to any of m possible indices.

and independently of other keys

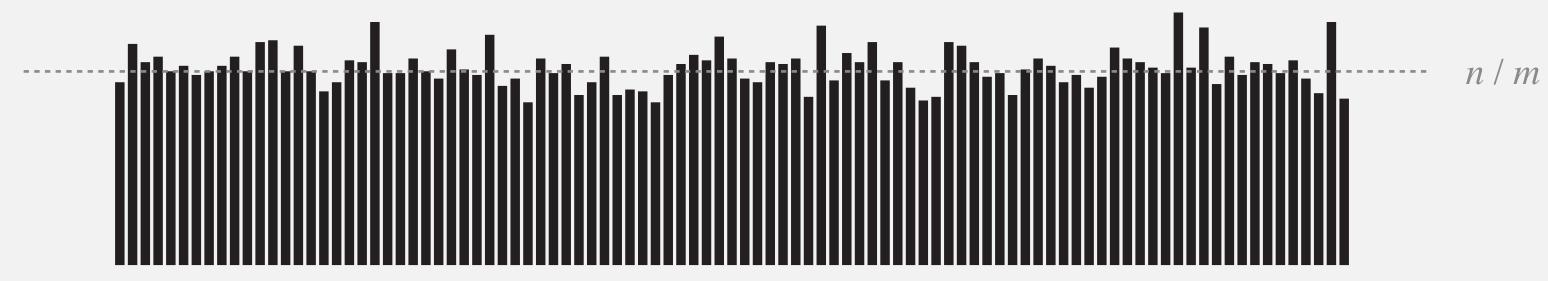
Bins and balls. Toss n balls uniformly at random into m bins.



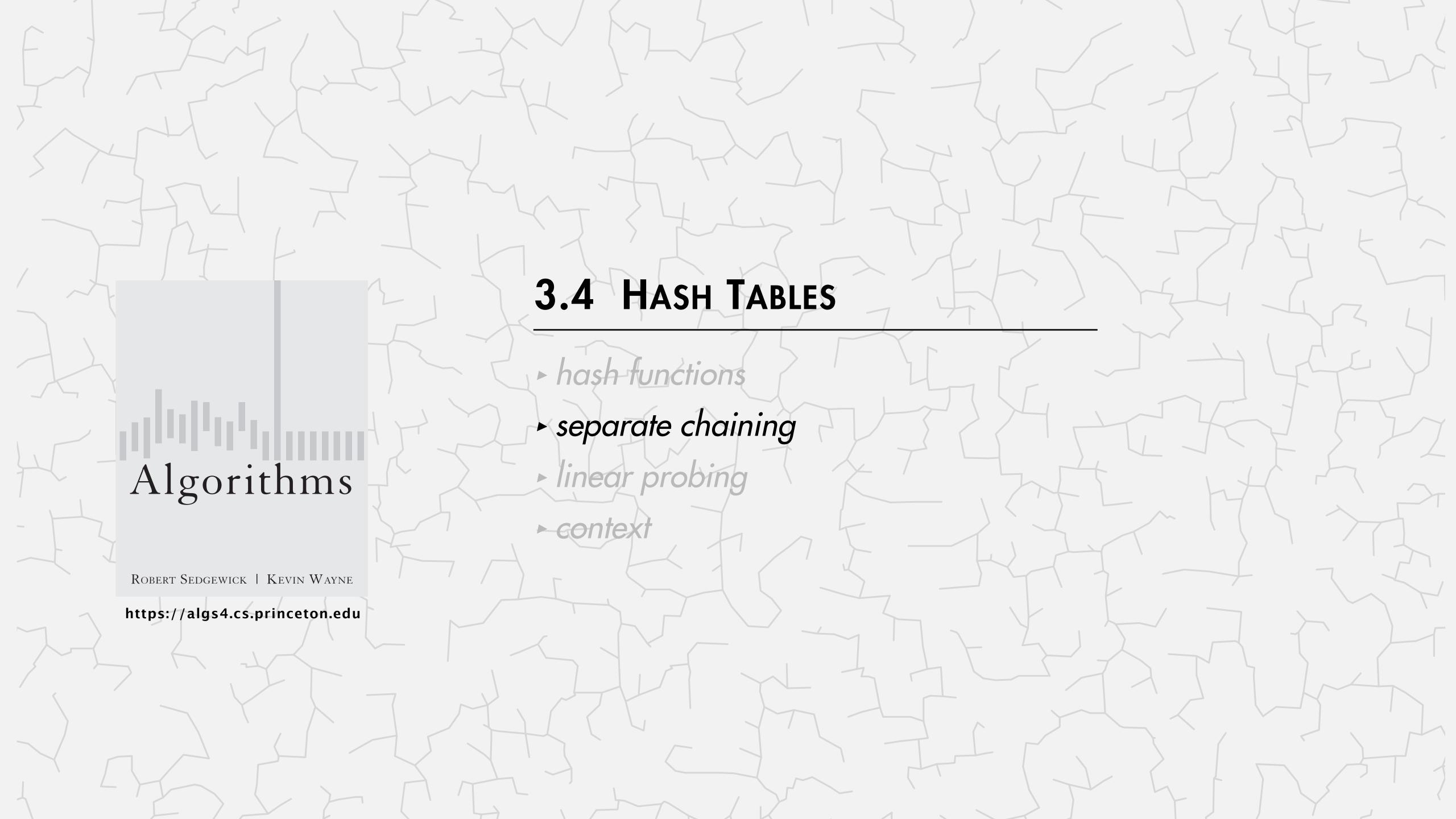
Good news. [load balancing]

Binomial(n, 1 / m)

- When n >> m, expect most bins to have approximately n / m balls.
- When n = m, expect most loaded bin has  $\sim \ln n / \ln \ln n$  balls.



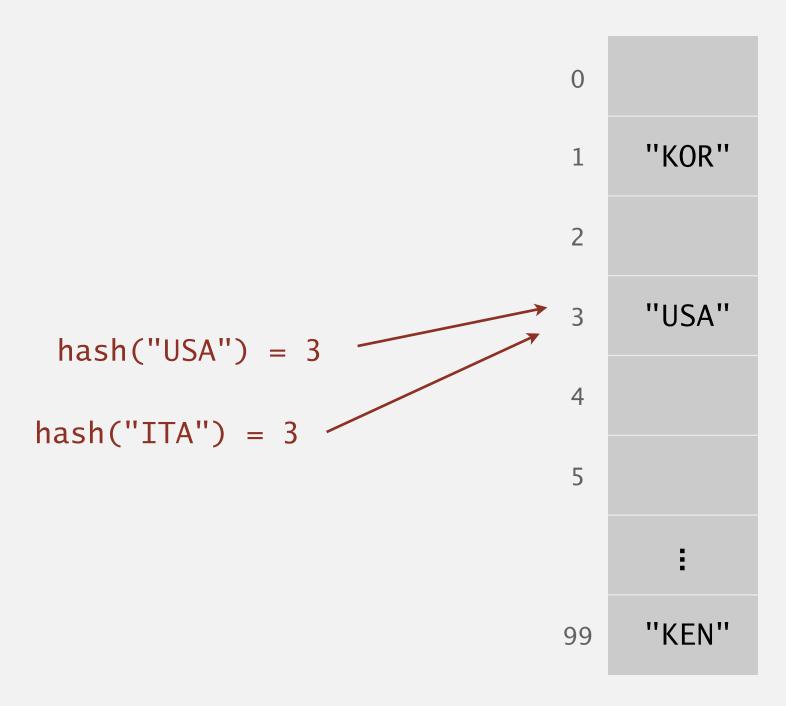
hash value frequencies for words in Tale of Two Cities (m = 97)



#### Collisions

Collision. Two distinct keys that hash to the same index.

- Birthday problem ⇒ can't avoid collisions. ←— unless you have a ridiculous (quadratic) amount of memory
- Load balancing  $\Rightarrow$  no index gets too many collisions.
  - ⇒ ok to scan through all colliding keys.

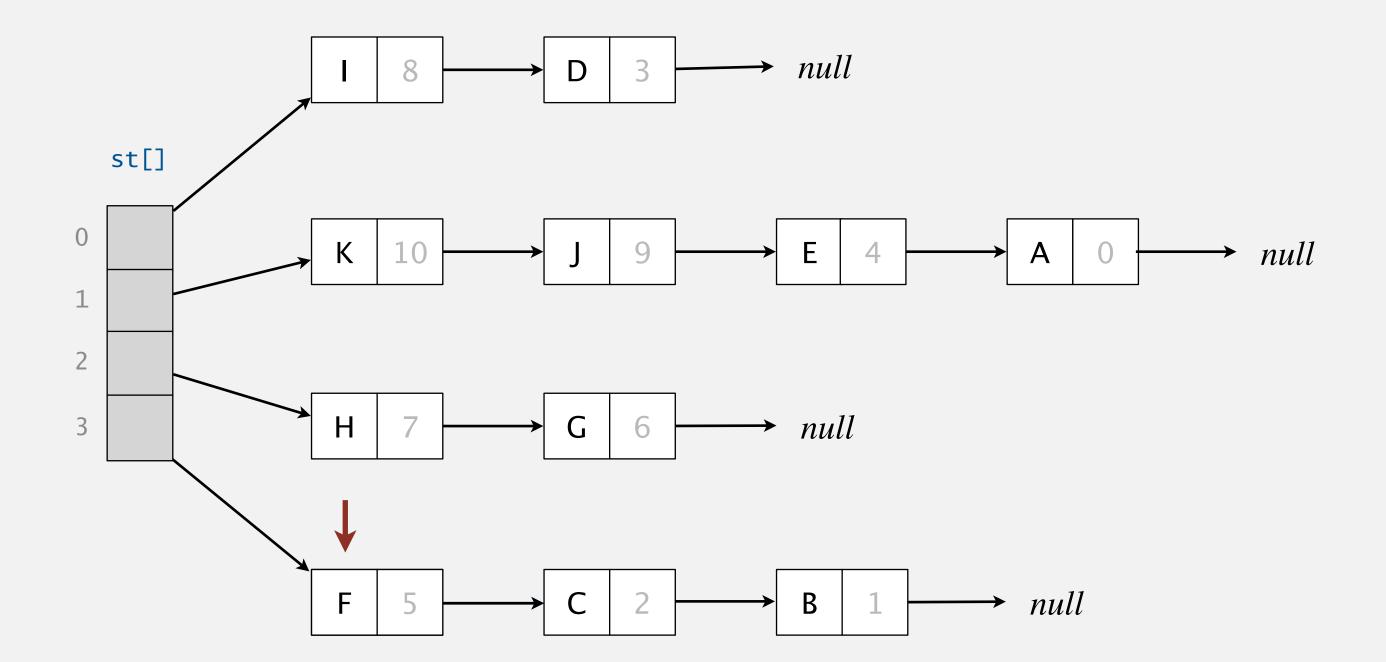


### Separate-chaining hash table

#### Use an array of *m* linked lists.

- Hash: map key to table index i between 0 and m 1.
- Insert: add key-value pair at front of chain i (if not already in chain).

put(L, 11) hash(L) = 3 separate-chaining hash table (m = 4)

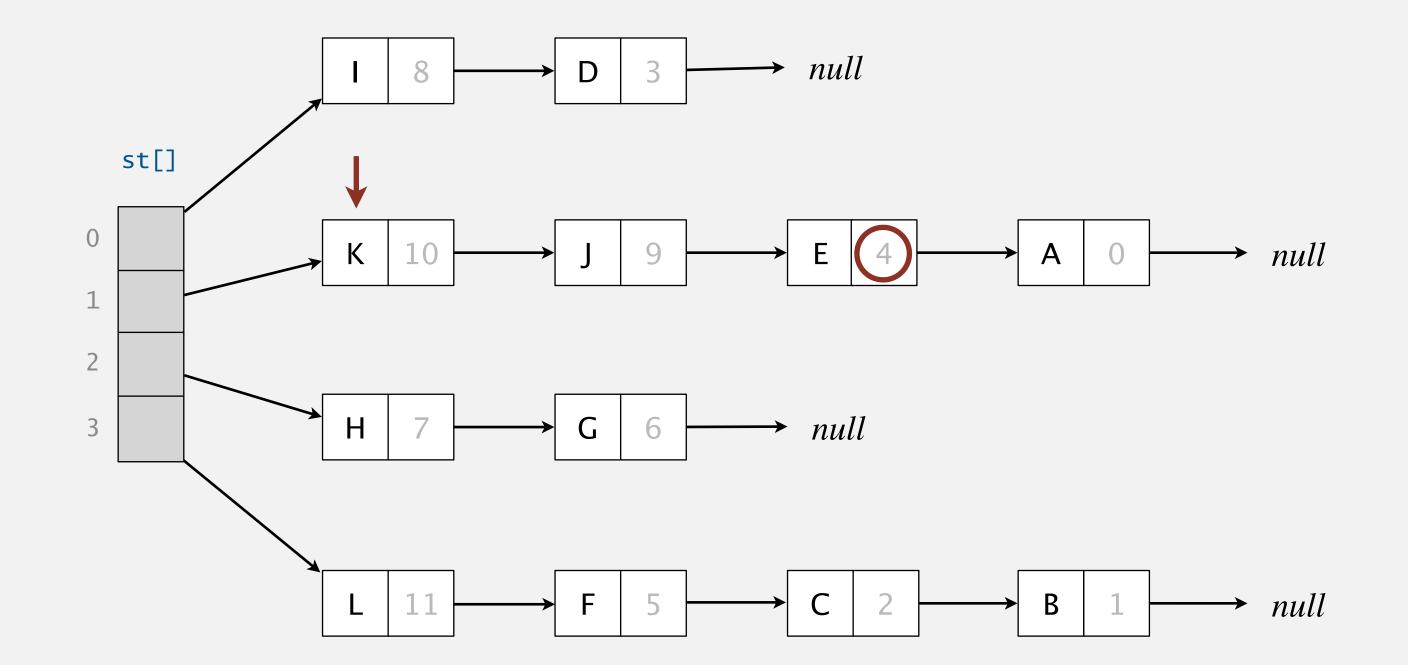


### Separate-chaining hash table

#### Use an array of *m* linked lists.

- Hash: map key to table index i between 0 and m-1.
- Insert: add key-value pair at front of chain i (if not already in chain).
- Search: perform sequential search in chain i.

 $\label{eq:get(E)} get(E)$  separate-chaining hash table (m = 4)



### Separate-chaining hash table: Java implementation

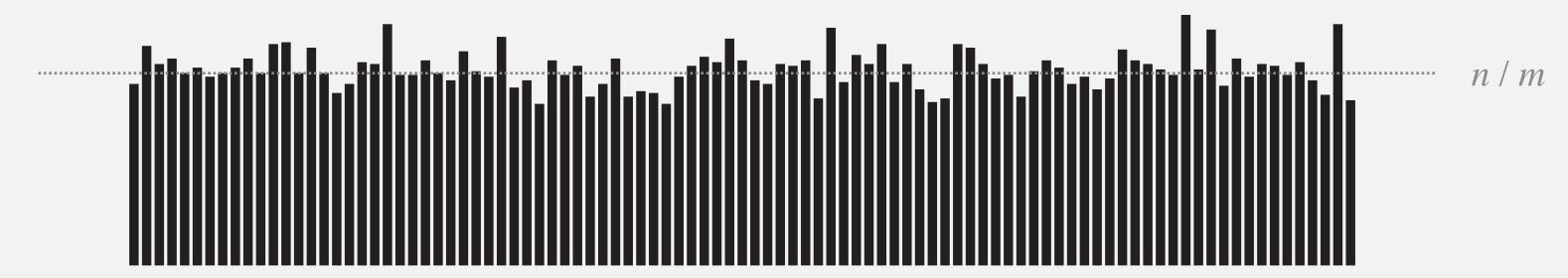
```
public class SeparateChainingHashST<Key, Value>
   private int m = 128; // number of chains
                                                                          array resizing
   private Node[] st = new Node[m]; // array of chains
                                                                           code omitted
   private static class Node
      private Object key;
                                    no generic array creation
      private Object val;
                                    (declare key and value of type Object)
      private Node next;
   private int hash(Key key)
   { /* as before */ }
   public Value get(Key key) {
      int i = hash(key);
      for (Node x = st[i]; x != null; x = x.next)
         if (key.equals(x.key)) return (Value) x.val;
      return null;
```

### Separate-chaining hash table: Java implementation

```
public class SeparateChainingHashST<Key, Value>
  private int m = 128; // number of chains
  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)
  { /* as before */ }
   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

Recall load balancing: Under the uniform hashing assumption, the length of each chain is tightly concentrated around mean = n/m.



hash value frequencies for words in Tale of Two Cities (m = 97)

calls to either
equals() or hashCode()

Consequence. Expected number of probes for search/insert is  $\Theta(n \mid m)$ .

- m too small  $\Rightarrow$  chains too long.
- m too large  $\Rightarrow$  too many empty chains.
- Typical choice:  $m \sim \frac{1}{4} n \Rightarrow \Theta(1)$  time for search/insert.

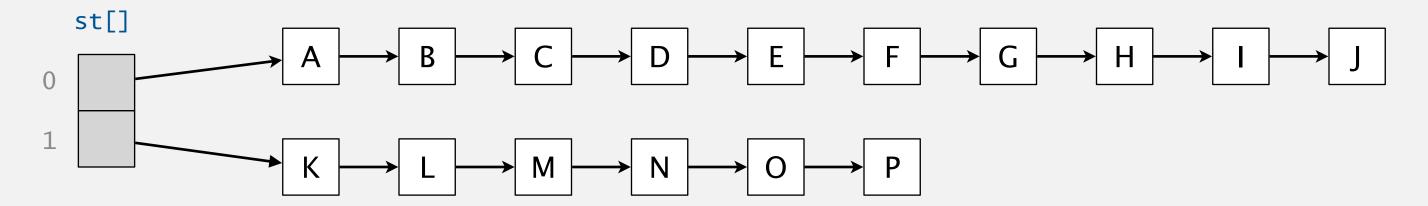
m times faster than sequential search

### Resizing in a separate-chaining hash table

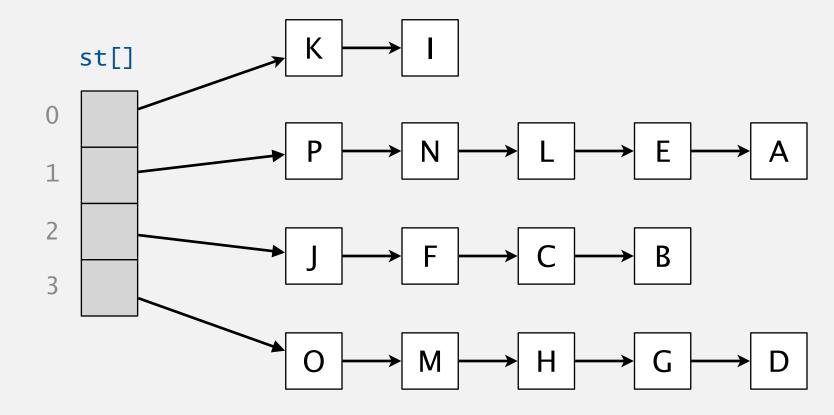
Goal. Average length of chain n/m is  $\Theta(1)$ .

- Double length m of array when  $n/m \ge 8$ .
- Halve length m of array when  $n/m \le 2$ .

#### before resizing (n/m = 8)



#### after resizing (n/m = 4)

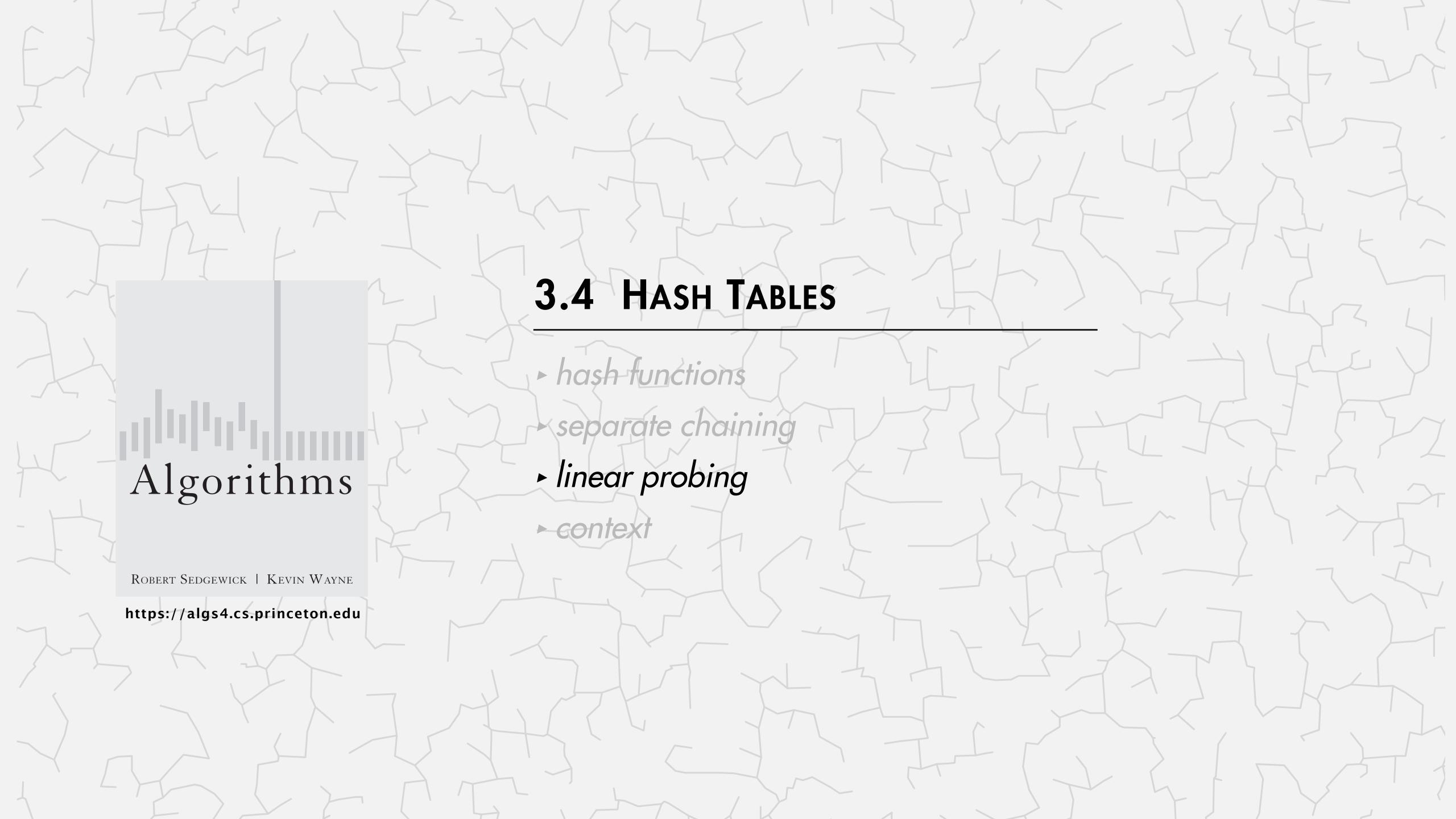


## Symbol table implementations: summary

implementation		guarantee		ā	average cas	e	ordered	key
	search	insert	delete	search	insert	delete	ops?	interface
sequential search (unordered list)	n	n	n	n	n	n		equals()
binary search (ordered array)	log n	n	n	log n	n	n	<b>~</b>	compareTo()
BST	n	n	n	log n	$\log n$	$\sqrt{n}$	<b>~</b>	compareTo()
red-black BST	$\log n$	$\log n$	$\log n$	log n	$\log n$	$\log n$	<b>~</b>	compareTo()
separate chaining	n	n †	n	1	1	1		equals() hashCode()

† under uniform hashing assumption

can achieve  $\Theta(1)$  probabilistic, amortized guarantee by choosing a hash function at random (see "universal hashing")



### Linear-probing hash table: insert

- Maintain key-value pairs in two parallel arrays, with one key per cell.
- Resolve collisions by linear probing: search successive cells until either finding the key or an unused cell.

Inserting into a linear-probing hash table.

linear-probing hash table



## Linear-probing hash table: search

- Maintain key-value pairs in two parallel arrays, with one key per cell.
- Resolve collisions by linear probing:
   search successive cells until either finding the key or an unused cell.

Searching in a linear-probing hash table.

linear-probing hash table



## Linear-probing hash table demo



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, ....

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

Note. Array length m must be greater than number of key-value pairs n.

	0	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
keys[]	Р	М			Α	С	S	Н	L		Ε				R	X

m = 16

### Linear-probing symbol table: Java implementation

```
public class LinearProbingHashST<Key, Value>
  private int m = 32768;
                                                                           array resizing
  private Value[] vals = (Value[]) new Object[m];
                                                                           code omitted
  private Key[] keys = (Key[]) new Object[m];
  private int hash(Key key)
   { /* as before */ }
  private void put(Key key, Value val) { /* next slide */ }
  public Value get(Key key)
     for (int i = hash(key); keys[i] != null; i = (i+1) % m)
        if (key.equals(keys[i]))
           return vals[i];
      return null;
```

### Linear-probing symbol table: Java implementation

```
public class LinearProbingHashST<Key, Value>
  private int m = 32768;
  private Value[] vals = (Value[]) new Object[m];
  private Key[] keys = (Key[]) new Object[m];
  private int hash(Key key)
  { /* as before */ }
  public Value get(Key key) { /* previous slide */ }
  public void put(Key key, Value val)
     int i;
     for (i = hash(key); keys[i] != null; i = (i+1) % m)
        if (keys[i].equals(key))
            break;
     keys[i] = key;
     vals[i] = val;
```

array resizing

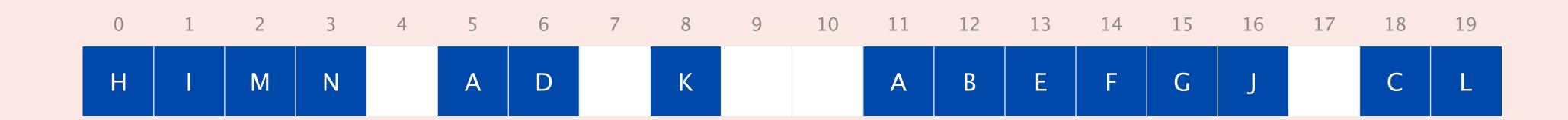
code omitted

32

### Hash tables: quiz 4



Under the uniform hashing assumption, where is the next key most likely to be added in this linear-probing hash table (no resizing)?



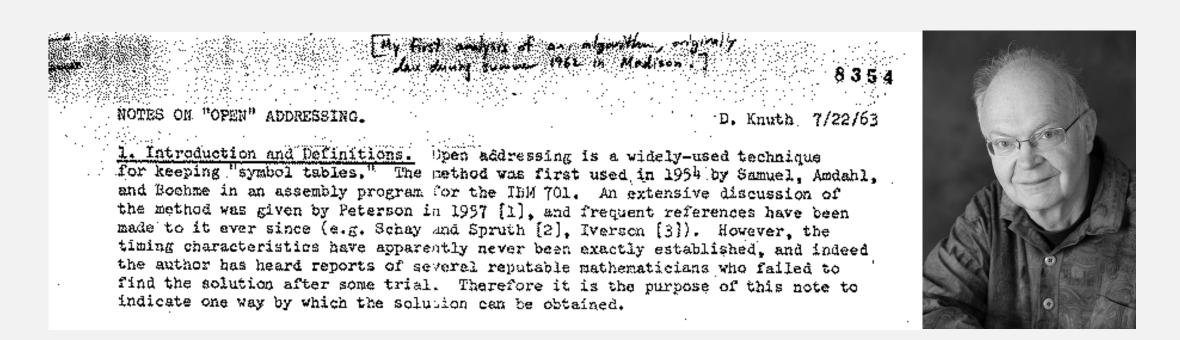
- A. Index 4.
- **B.** Index 17.
- C. Either index 4 or 17.
- D. All open indices are equally likely.

### 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 at most

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

#### Pf. [beyond course scope]



#### 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

## ST implementations: summary

implementation		guarantee			average cas	e	ordered	key	
	search	insert	delete	search	insert	delete	ops?	interface	
sequential search (unordered list)	n	n	n	n	n	n		equals()	
binary search (ordered array)	log n	n	n	log n	n	n	<b>~</b>	compareTo()	
BST	n	n	n	log n	log n	$\sqrt{n}$	<b>~</b>	compareTo()	
red-black BST	$\log n$	log n	log n	$\log n$	$\log n$	$\log n$	<b>✓</b>	compareTo()	
separate chaining	n	n	n	1	1	1		equals() hashCode()	
linear probing	n	n	n	1 †	1	1		equals() hashCode()	

<sup>†</sup> under uniform hashing assumption

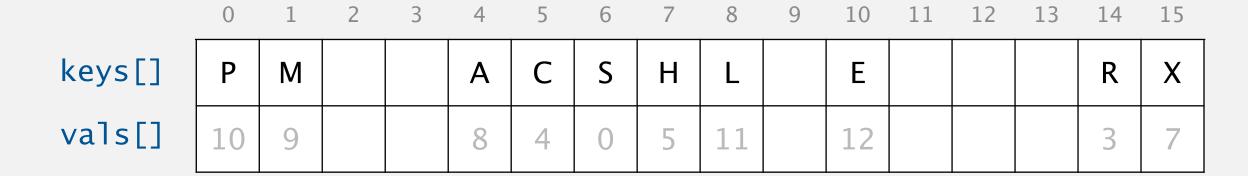
## Separate chaining vs. linear probing

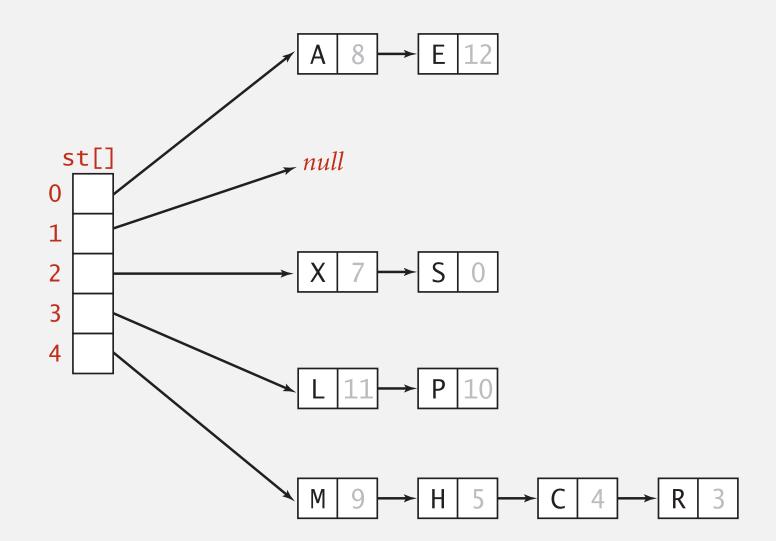
#### Separate chaining.

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

#### Linear probing.

- Unrivaled data locality.
- More probes because of clustering.



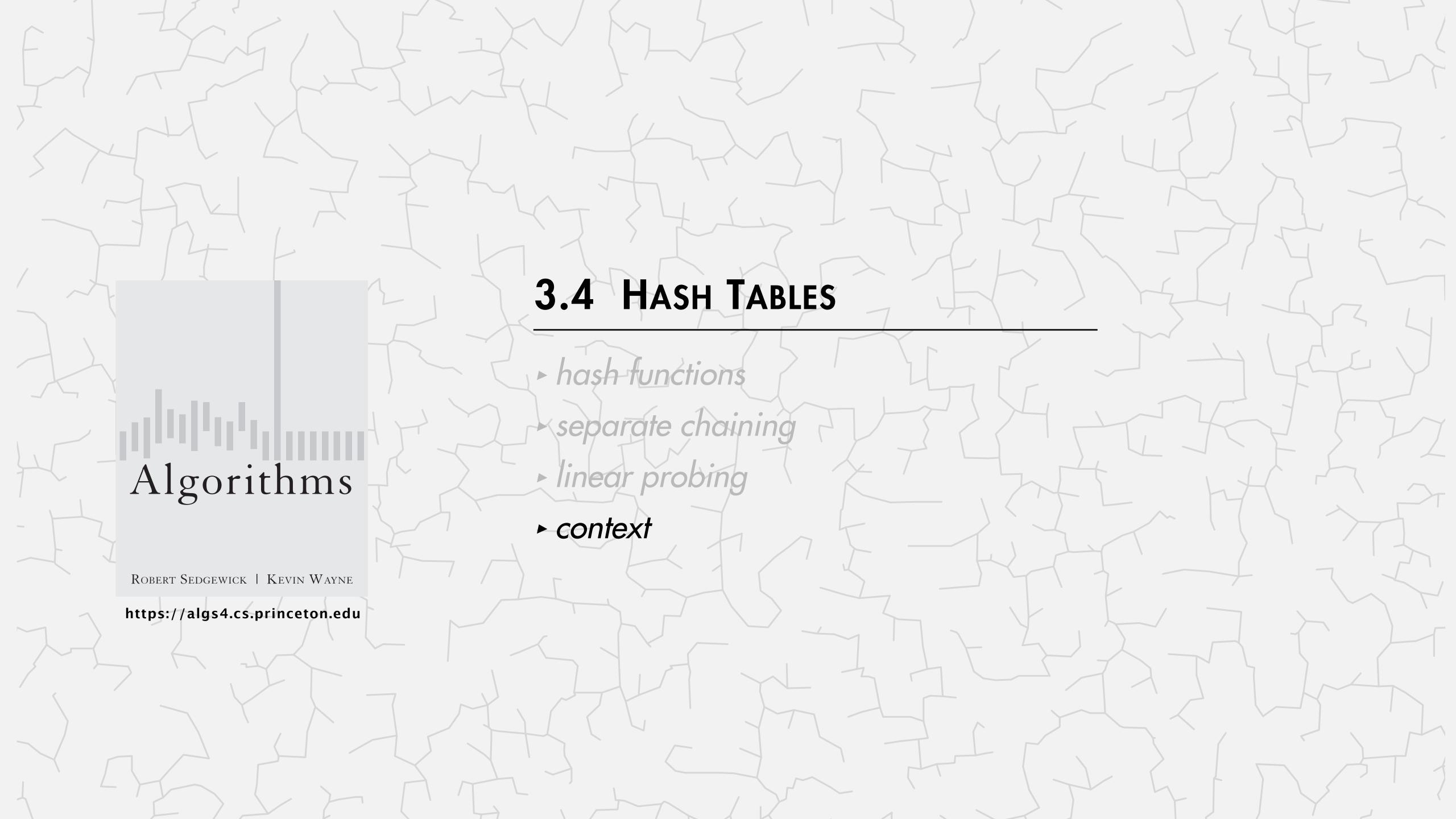


# 3-SUM (REVISITED)



3–Sum. Given n distinct integers, find three such that a + b + c = 0.

Goal.  $\Theta(n^2)$  expected time;  $\Theta(n)$  extra space.



#### Hashing: variations on the theme

Many many improved versions have been studied.

Use different probe sequence, i.e., not h(k), h(k) + 1, h(k) + 2, ...

[ quadratic probing, double hashing, pseudo-random probing, ... ]

Google Swiss Table

Facebook F14

Python 3

Python 3

Python 3

Python 3

Python 3

Python 3

Facebook F14

Python 3

Facebook F14

Python 3

Python 3

Python 3

Insert tombstones prophylactically, to avoid primary clustering. 

eliminates primary clustering; 
maintains data locality

[ graveyard hashing ]







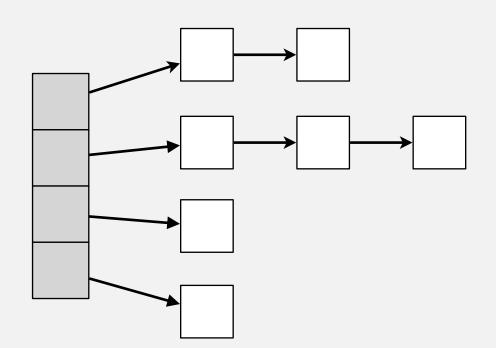
#### Hash tables vs. balanced search trees

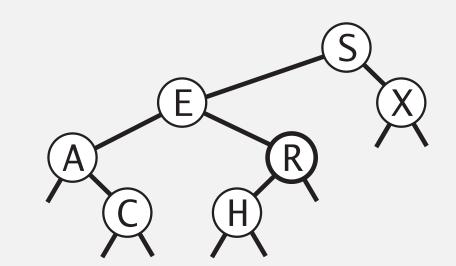
#### Hash tables.

- Simpler to code.
- Typically faster in practice.
- No effective alternative for unordered keys.

#### Balanced search trees.

- Stronger performance guarantees.
- Support for ordered ST operations.
- Easier to implement compareTo() than hashCode().





#### Java includes both.

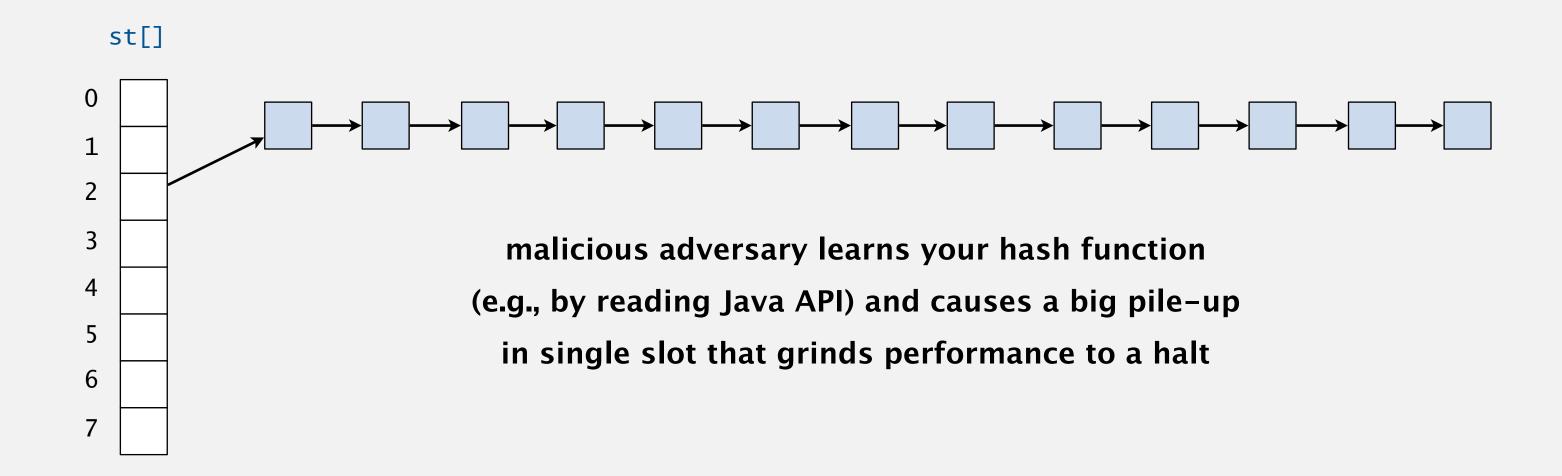
- BSTs: java.util.TreeMap. ← red-black BST
- Hash tables: java.util.HashMap, java.util.IdentityHashMap.

separate chaining
(Java 8: if chain gets too long, use red-black BST for chain)



### Algorithmic complexity attacks

- Q. Is the uniform hashing assumption important in practice?
- A1. Yes: aircraft control, nuclear reactor, pacemaker, HFT, missile-defense system, ...
- A2. Yes: denial-of-service (DoS) attacks.



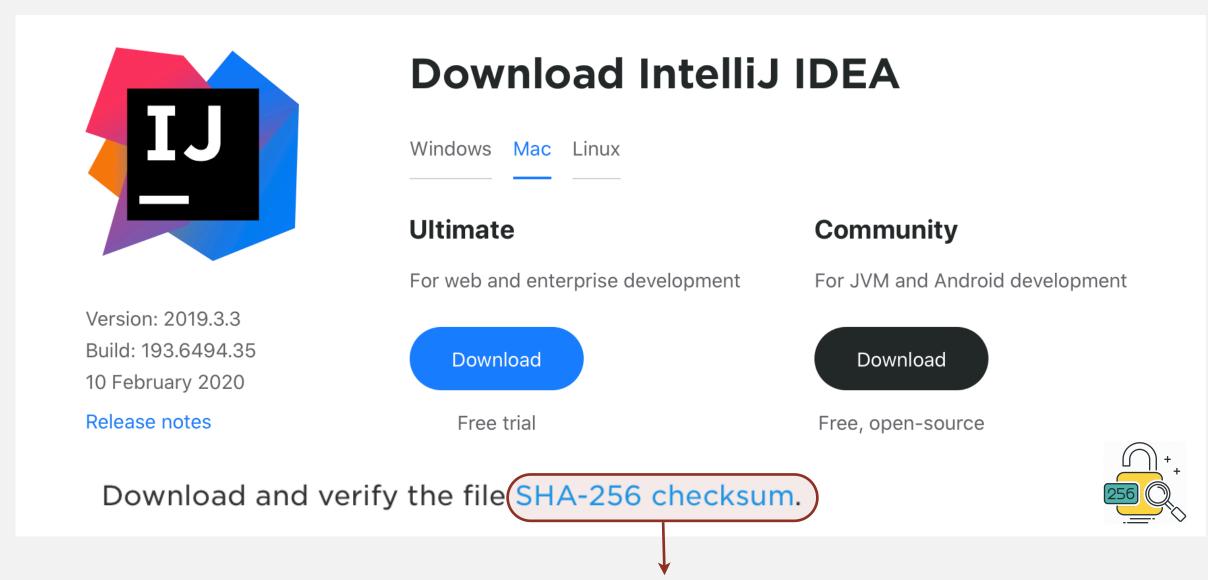
#### Real-world exploits. [Crosby-Wallach 2003]

- Linux 2.4.20 kernel: save files with carefully chosen names.
- Bro server: send carefully chosen packets to DoS the server, using less bandwidth than a dial-up modem.

### Hashing: beyond symbol tables

File verification. When downloading a file from the web:

- Vendor publishes hash of file.
- Client checks whether hash of downloaded file matches.
- If mismatch, file corrupted. ← (e.g., error in transmission or infected by virus)



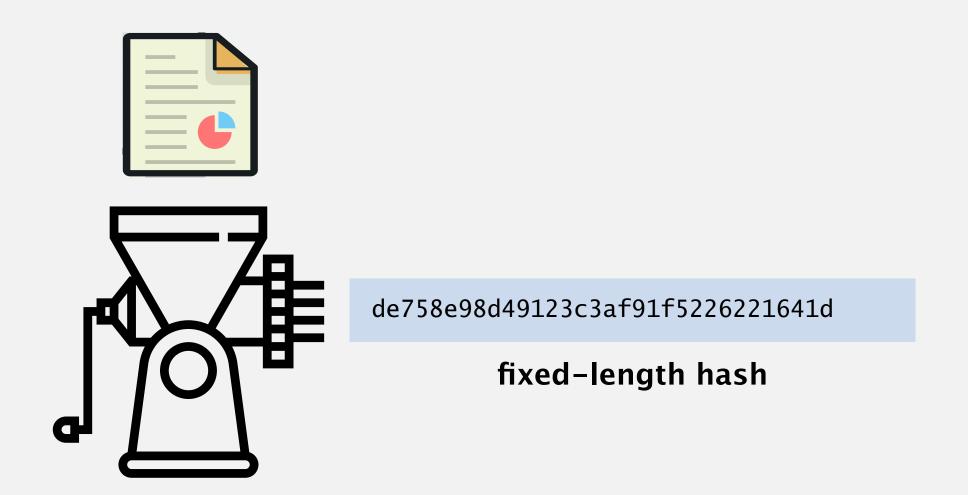
c62ed2df891ccbb40d890e8a0074781801f086a3091a4a2a592a96afaba31270

~/Desktop> sha256sum ideaIC-2022.3.3.dmg
c62ed2df891ccbb40d890e8a0074781801f086a3091a4a2a592a96afaba31270

### Hashing: cryptographic applications

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

Ex. MD5, SHA-1, SHA-256, SHA-512, SHA3-512, Whirlpool, BLAKE3, ....



Applications. File verification, digital signatures, cryptocurrencies, password authentication, blockchain, non-fungible tokens, Git commit identifiers, ....



ALGORITHM (NOUN)
WORD USED BY
PROGRAMMERS WHEN
THEY DO NOT WANT TO
EXPLAIN WHAT THEY DID.