A decorative vertical bar on the left side of the slide. It consists of a dark teal background with a white vertical stripe. To the right of the teal bar are several orange circles of varying sizes, arranged in a cluster. The title text is positioned to the right of this decorative bar.

# DATA STRUCTURES USING 'C'



UNIVERSITY of NORTH TEXAS

# *CSCE 3110*

## *Data Structures & Algorithm Analysis*

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Hashing

Reading: Chap.5, *Weiss*



# *How to Implement a Dictionary?*

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- Sequences
  - ordered
  - unordered
- Binary Search Trees
- Skip lists
- Hashtables



# *Hashing*

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- Another important and widely useful technique for implementing dictionaries
- Constant time per operation (on the average)
- Worst case time proportional to the size of the set for each operation (just like array and chain implementation)



## *Basic Idea*

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- Use *hash function* to map keys into positions in a *hash table*

### Ideally

- If element  $e$  has key  $k$  and  $h$  is hash function, then  $e$  is stored in position  $h(k)$  of table
- To search for  $e$ , compute  $h(k)$  to locate position. If no element, dictionary does not contain  $e$ .



# Example

## Dictionary Student Records

- Keys are ID numbers (951000 - 952000), no more than 100 students
- Hash function:  $h(k) = k - 951000$  maps ID into distinct table positions 0-1000
- array `table[1001]`





## *Analysis (Ideal Case)*

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- $O(b)$  time to initialize hash table ( $b$  number of positions or buckets in hash table)
- $O(1)$  time to perform *insert*, *remove*, *search*



## *Ideal Case is Unrealistic*

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- Works for implementing dictionaries, but many applications have key ranges that are too large to have 1-1 mapping between buckets and keys!

### Example:

- Suppose key can take on values from 0 .. 65,535 (2 byte unsigned int)
- Expect  $\approx 1,000$  records at any given time
- Impractical to use hash table with 65,536 slots!



# Hash Functions

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- If key range too large, use hash table with fewer buckets and a hash function which maps multiple keys to same bucket:

$h(k_1) = \beta = h(k_2)$ :  $k_1$  and  $k_2$  have **collision** at slot  $\beta$

- Popular hash functions: hashing by division  
 $h(k) = k\%D$ , where  $D$  number of buckets in hash table

- Example: hash table with 11 buckets

$$h(k) = k\%11$$

$$80 \rightarrow 3 \text{ (} 80\%11= 3\text{), } 40 \rightarrow 7, 65 \rightarrow 10$$

$$58 \rightarrow 3 \text{ collision!}$$



# Collision Resolution Policies

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- Two classes:
  - (1) Open hashing, a.k.a. separate chaining
  - (2) Closed hashing, a.k.a. open addressing
- Difference has to do with whether collisions are stored *outside the table* (open hashing) or whether collisions result in storing one of the records at *another slot in the table* (closed hashing)



# Closed Hashing

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- Associated with closed hashing is a *rehash strategy*:  
“If we try to place  $x$  in bucket  $h(x)$  and find it occupied, find alternative location  $h_1(x)$ ,  $h_2(x)$ , etc. Try each in order, if none empty table is full,”
- $h(x)$  is called *home bucket*
- Simplest rehash strategy is called *linear hashing*  
$$h_i(x) = (h(x) + i) \% D$$
- In general, our collision resolution strategy is to generate a sequence of hash table slots (probe sequence) that can hold the record; test each slot until find empty one (probing)



# Example Linear (Closed) Hashing

- $D=8$ , keys  $a, b, c, d$  have hash values  $h(a)=3$ ,  $h(b)=0$ ,  $h(c)=4$ ,  $h(d)=3$
- Where do we insert  $d$ ? 3 already filled
- Probe sequence using linear hashing:  
 $h_1(d) = (h(d)+1)\%8 = 4\%8 = 4$   
 $h_2(d) = (h(d)+2)\%8 = 5\%8 = \mathbf{5^*}$   
 $h_3(d) = (h(d)+3)\%8 = 6\%8 = 6$   
etc.  
7, 0, 1, 2
- Wraps around the beginning of the table!

0	b
1	
2	
3	a
4	c
5	<b>d</b>
6	
7	



# Operations Using Linear Hashing

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- Test for membership: *findItem*
- Examine  $h(k)$ ,  $h_1(k)$ ,  $h_2(k)$ , ..., until we find  $k$  or an empty bucket or home bucket
- If no deletions possible, strategy works!
- What if deletions?
- If we reach empty bucket, cannot be sure that  $k$  is not somewhere else and empty bucket was occupied when  $k$  was inserted
- Need special placeholder *deleted*, to distinguish bucket that was never used from one that once held a value
- May need to reorganize table after many deletions



# *Performance Analysis - Worst Case*

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- ⊕ Initialization:  $O(b)$ ,  $b$  # of buckets
- ⊕ Insert and search:  $O(n)$ ,  $n$  number of elements in table; all  $n$  key values have same home bucket
- ⊕ No better than linear list for maintaining dictionary!



# *Performance Analysis - Avg Case*

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- Distinguish between successful and unsuccessful searches
  - Delete = successful search for record to be deleted
  - Insert = unsuccessful search along its probe sequence
- Expected cost of hashing is a function of how full the table is: load factor  $\alpha = n/b$
- It has been shown that average costs under linear hashing (probing) are:
  - Insertion:  $1/2(1 + 1/(1 - \alpha)^2)$
  - Deletion:  $1/2(1 + 1/(1 - \alpha))$



# Improved Collision Resolution

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- Linear probing:  $h_i(x) = (h(x) + i) \% D$ 
  - all buckets in table will be candidates for inserting a new record before the probe sequence returns to home position
  - clustering of records, leads to long probing sequences
- Linear probing with skipping:  $h_i(x) = (h(x) + ic) \% D$ 
  - $c$  constant other than 1
  - records with adjacent home buckets will not follow same probe sequence
- (Pseudo)Random probing:  $h_i(x) = (h(x) + r_i) \% D$ 
  - $r_i$  is the  $i^{\text{th}}$  value in a random permutation of numbers from 1 to  $D-1$
  - insertions and searches use the *same* sequence of “random” numbers



# Example

II

insert 1052 (h.b. 7)

0	1001
1	9537
2	3016
3	
4	
5	
6	
7	9874
8	2009
9	9875
10	

1. What if next element has home bucket 0?

$$h(k) = k \% 11$$

→ go to bucket 3

Same for elements with home bucket 1 or 2!

Only a record with home position 3 will stay.

⇒  $p = 4/11$  that next record will go to bucket 3

2. Similarly, records hashing to 7,8,9 will end up in 10

3. Only records hashing to 4 will end up in 4 ( $p=1/11$ ); same for 5 and 6

0	1001
1	9537
2	3016
3	
4	
5	
6	
7	9874
8	2009
9	9875
10	1052

next element in bucket 3 with  $p = 8/11$



# Hash Functions - Numerical Values

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- ❊ Consider:  $h(x) = x \% 16$ 
  - ❑ poor distribution, not very random
  - ❑ depends solely on least significant four bits of key
- ❊ Better, *mid-square* method
  - ❑ if keys are integers in range  $0, 1, \dots, K$ , pick integer  $C$  such that  $DC^2$  about equal to  $K^2$ , then
$$h(x) = \lfloor x^2 / C \rfloor \% D$$
extracts middle  $r$  bits of  $x^2$ , where  $2^r = D$  (a base- $D$  digit)
  - ❑ better, because most or all of bits of key contribute to result



# Hash Function – Strings of Characters

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## ❁ Folding Method:

```
int h(String x, int D) {  
    int i, sum;  
    for (sum=0, i=0; i<x.length(); i++)  
        sum+= (int)x.charAt(i);  
    return (sum%D);  
}
```

- ❁ sums the ASCII values of the letters in the string
  - ASCII value for "A" =65; sum will be in range 650-900 for 10 upper-case letters; good when D around 100, for example
- ❁ order of chars in string has no effect



# Hash Function – Strings of Characters

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## ● Much better: Cyclic Shift

```
static long hashCode(String key, int D) {  
    int h=0;  
    for (int i=0, i<key.length(); i++){  
        h = (h << 4) | ( h >> 27);  
        h += (int) key.charAt(i);  
    }  
    return h%D;  
}
```



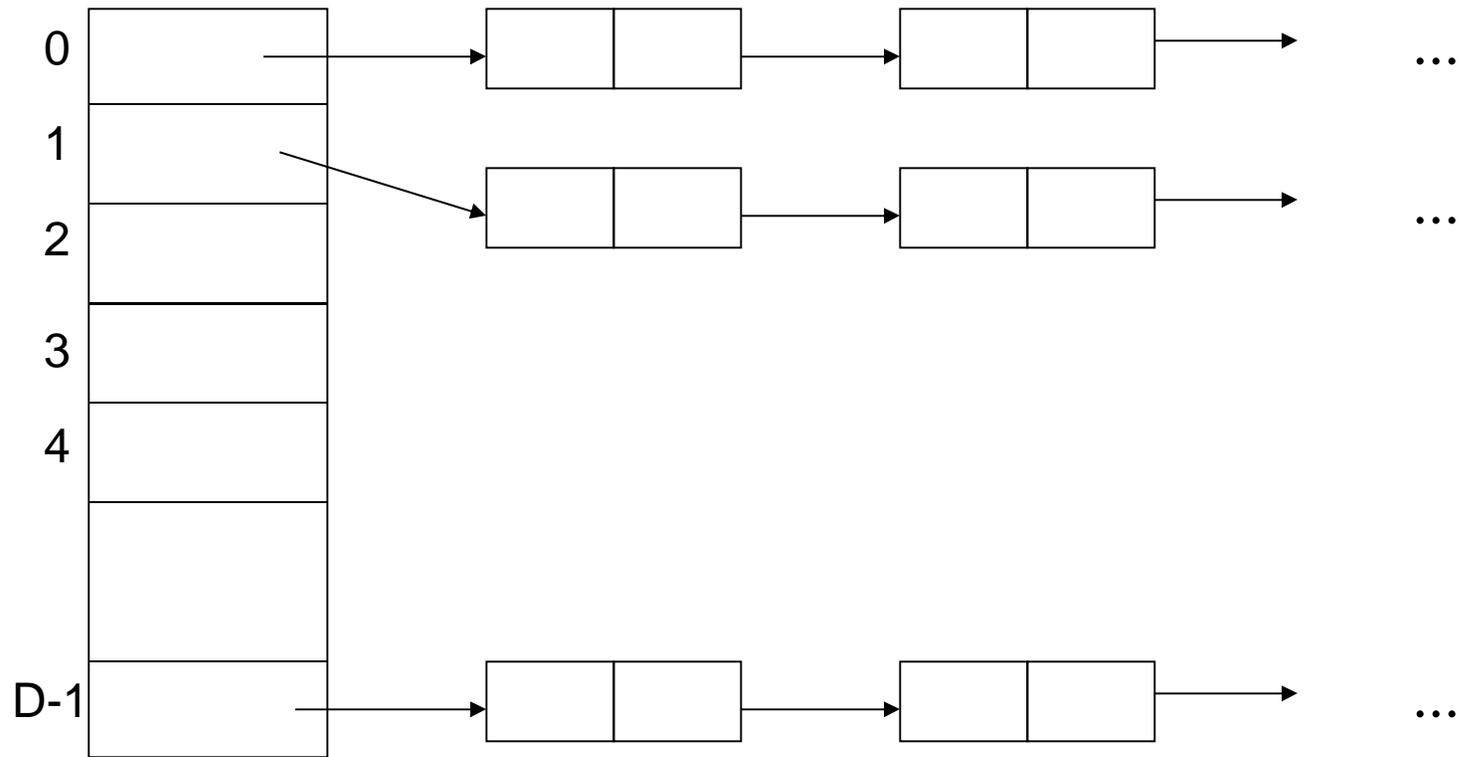
# *Open Hashing*

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- Each bucket in the hash table is the head of a linked list
- All elements that hash to a particular bucket are placed on that bucket's linked list
- Records within a bucket can be ordered in several ways
  - by order of insertion, by key value order, or by frequency of access order



# Open Hashing Data Organization





# Analysis

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- Open hashing is most appropriate when the hash table is kept in main memory, implemented with a standard in-memory linked list
- We hope that number of elements per bucket roughly equal in size, so that the lists will be short
- If there are  $n$  elements in set, then each bucket will have roughly  $n/D$
- If we can estimate  $n$  and choose  $D$  to be roughly as large, then the average bucket will have only one or two members



## *Analysis Cont'd*

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Average time per dictionary operation:

- $D$  buckets,  $n$  elements in dictionary  $\Rightarrow$  average  $n/D$  elements per bucket
- *insert, search, remove* operation take  $O(1+n/D)$  time each
- If we can choose  $D$  to be about  $n$ , constant time
- Assuming each element is likely to be hashed to any bucket, running time constant, independent of  $n$



## *Comparison with Closed Hashing*

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- ❁ Worst case performance is  $O(n)$  for both
- ❁ Number of operations for hashing
  - ❁ 23 6 8 10 23 5 12 4 9 19
  - ❁  $D=9$
  - ❁  $h(x) = x \% D$



## *Hashing Problem*

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- Draw the 11 entry hashtable for hashing the keys 12, 44, 13, 88, 23, 94, 11, 39, 20 using the function  $(2i+5) \bmod 11$ , closed hashing, linear probing
- Pseudo-code for listing all identifiers in a hashtable in lexicographic order, using open hashing, the hash function  $h(x) = \text{first character of } x$ . What is the running time?