

**CSCI 136**  
**Data Structures &**  
**Advanced Programming**

**Spring 2018**

**Lecture 34**

**Profs 2070567 and 74655**

# Administrative Details

## Reminders

- No lab this week
  - Many TAs will be holding normal hours to answer questions about labs and practice exams
- Final exam
  - Monday, May 21 at 9:30am in Chemistry 123
  - Covers everything, with strong emphasis on post-midterm
  - Study guide, sample exam will be posted on handouts page
- Review session
  - Friday May 18
    - Time?

# Last Time

- Hash tables implement the Map interface
  - `[obj.hashCode() % array.length]` assigns objects to bins
  - **Collisions** occur when multiple objects map to the same bin
  - We can resolve collisions using:
    - **Linear probing** (aka open addressing)
    - **External chaining**

# Today's Outline

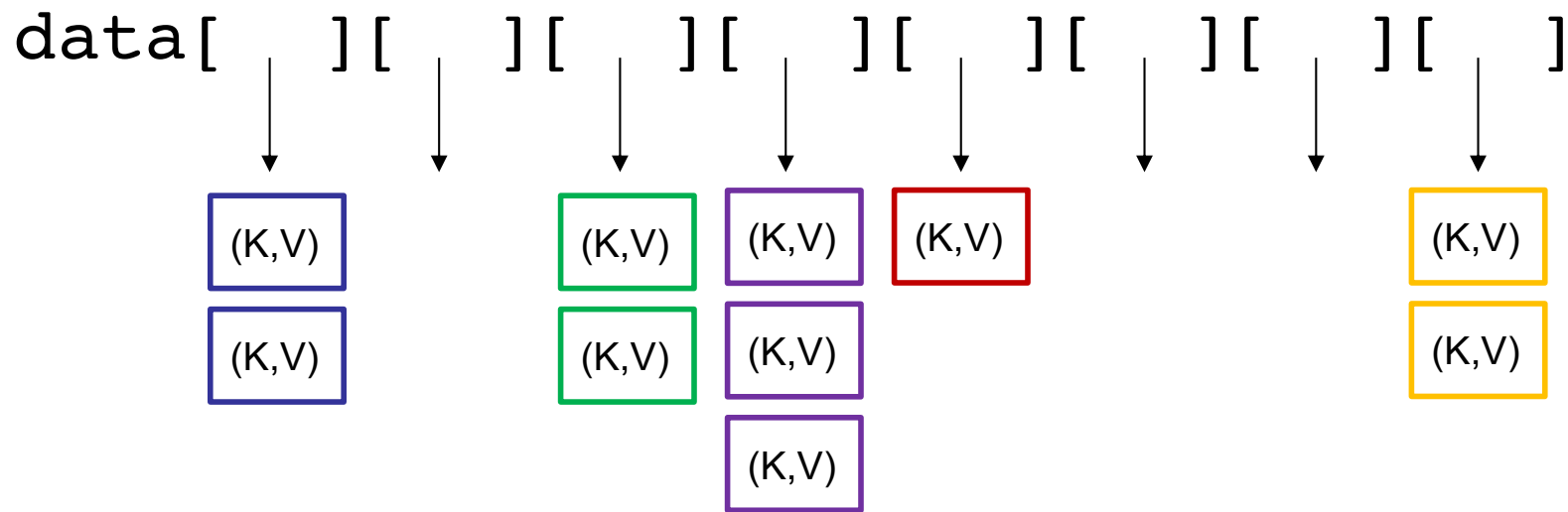
- External Chaining to resolve collisions
- Fun hashing applications (not on exam)
  - Cuckoo hashing
  - Bloom Filters
  - Verification/integrity
  - Deduplication

# Linear Probing Review

- A hash function maps a **key-value pair** to a **bin**
- If two keys hash to the same bin, we have a **collision**
- **Linear probing** scans and places the collided element in the first available bin, creating a **run**
  - When we remove, must add a placeholder so we don't artificially break up runs

# External Chaining

- Instead of runs, we store a list in each bin



- Everything that hashes to  $\text{bin}_i$  goes into  $\text{list}_i$ 
  - `get()`, `put()`, and `remove()` only need to check one slot's list
  - No placeholders!

# Probing vs. Chaining

What is the performance of:

- `put (K, V)`
  - LP:  $O(1 + \text{run length})$
  - EC:  $O(1 + \text{chain length})$
- `get (K)`
  - LP:  $O(1 + \text{run length})$
  - EC:  $O(1 + \text{chain length})$
- `remove (K)`
  - LP:  $O(1 + \text{run length})$
  - EC:  $O(1 + \text{chain length})$
- Run/Chain size is important. How do we control cluster/chain length?

# Load Factor

- Need to keep track of how full the table is
  - Why?
  - What happens when array fills completely?
- **Load factor** is a measure of how full the hash table is
  - $LF = (\# \text{ elements}) / (\text{table size})$
- When LF reaches some threshold, grow size of array (typically threshold = 0.6)
  - Challenges?



# Growing the Underlying Array

- Cannot just copy values
  - Why?
  - Key-value pairs' bins may change
  - Example: suppose `(key.hashCode() == 11)`
    - $11 \% 7 = 4$ ;
    - $11 \% 13 = 11$ ;
- **Result:** must recompute all hash codes, then reinsert key-value pairs into new array
- Also: try to keep array sizes relatively prime
  - Redistribute “clumps”

# Good Hashing Functions

- **Important point:** All of this hinges on using “good” hash functions that spread keys “evenly”
- **Good hash functions:**
  - Are fast to compute
  - Distribute keys uniformly
- We almost always have to test “goodness” empirically

# Example Hash Functions

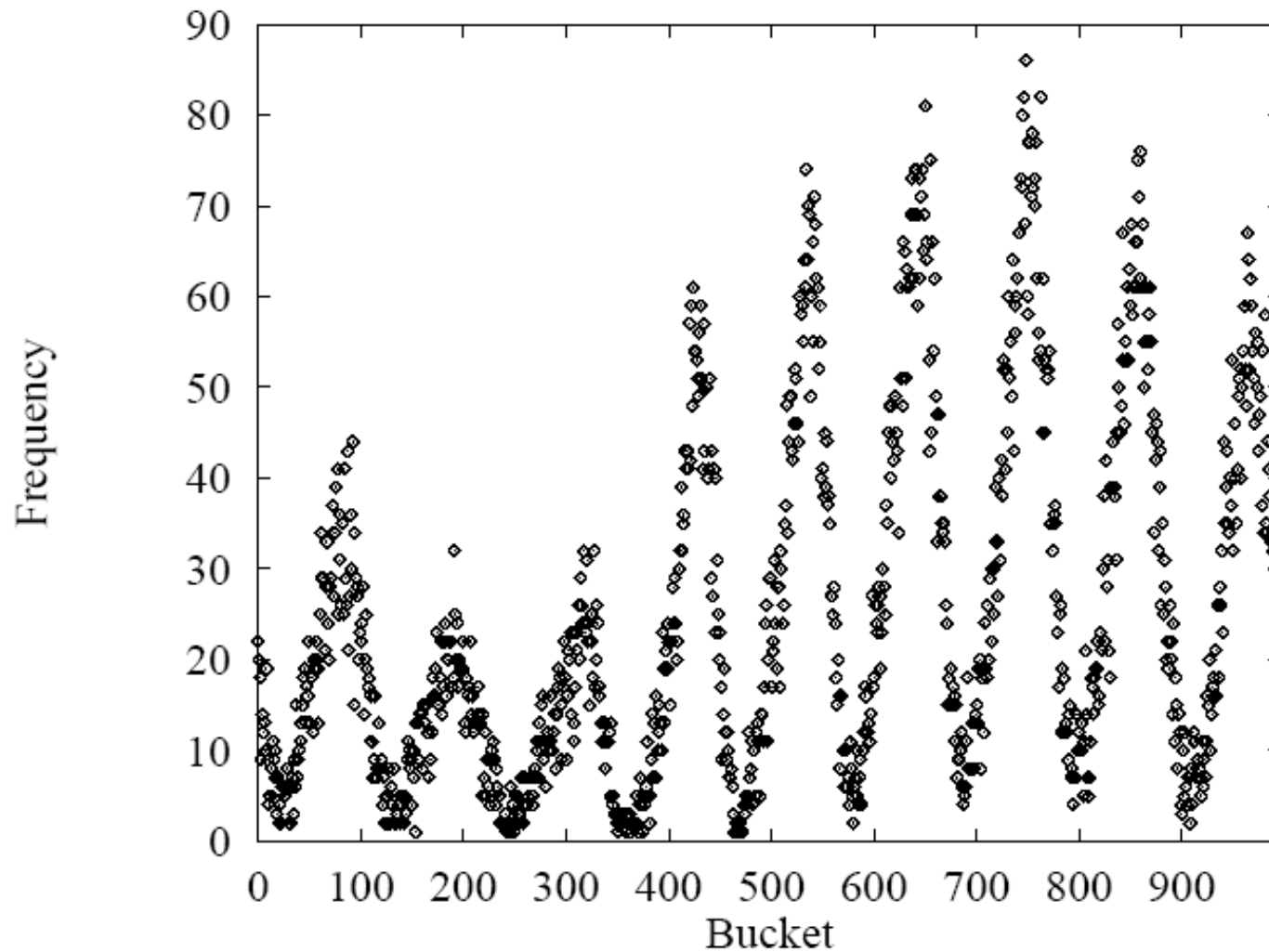
- What are some feasible hash functions for Strings?
  - Use the first char's ASCII value?
    - 0-255 only
    - Not uniform (some letters more popular than others)
  - Sum of all characters' ASCII values?
    - Not uniform - lots of small words
    - smile, limes, miles, slime are all the same

# Example Hash Functions

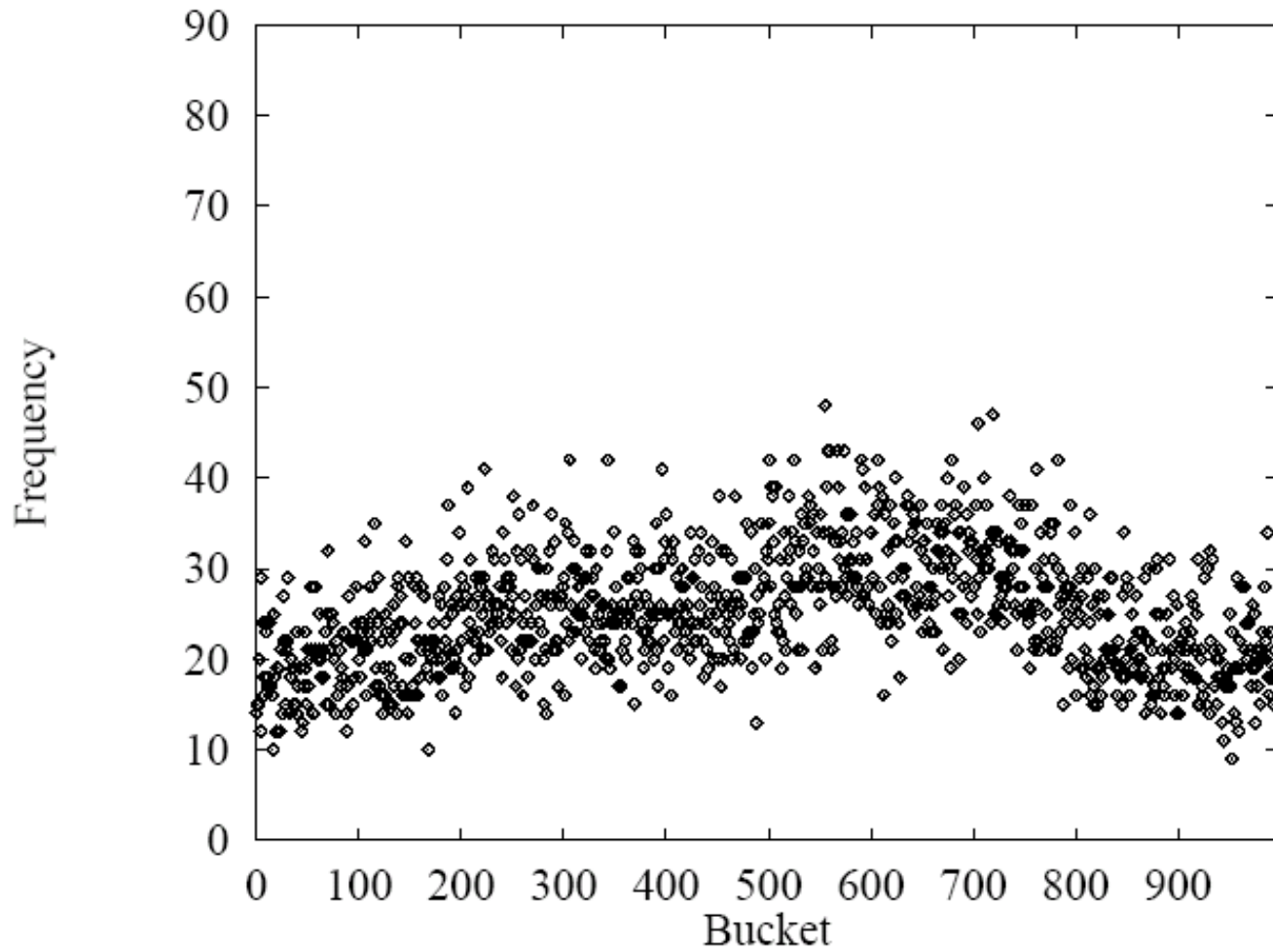
- String hash functions commonly use weighted sums
  - Character values weighted by position in string
    - Long words get bigger codes
    - Distributes keys better than non-weighted sum
  - Let's look at different weights...

$$\sum_{i=0}^{n=s.length()} s.charAt(i)$$

Hash of all words in UNIX  
spelling dictionary (997  
buckets)

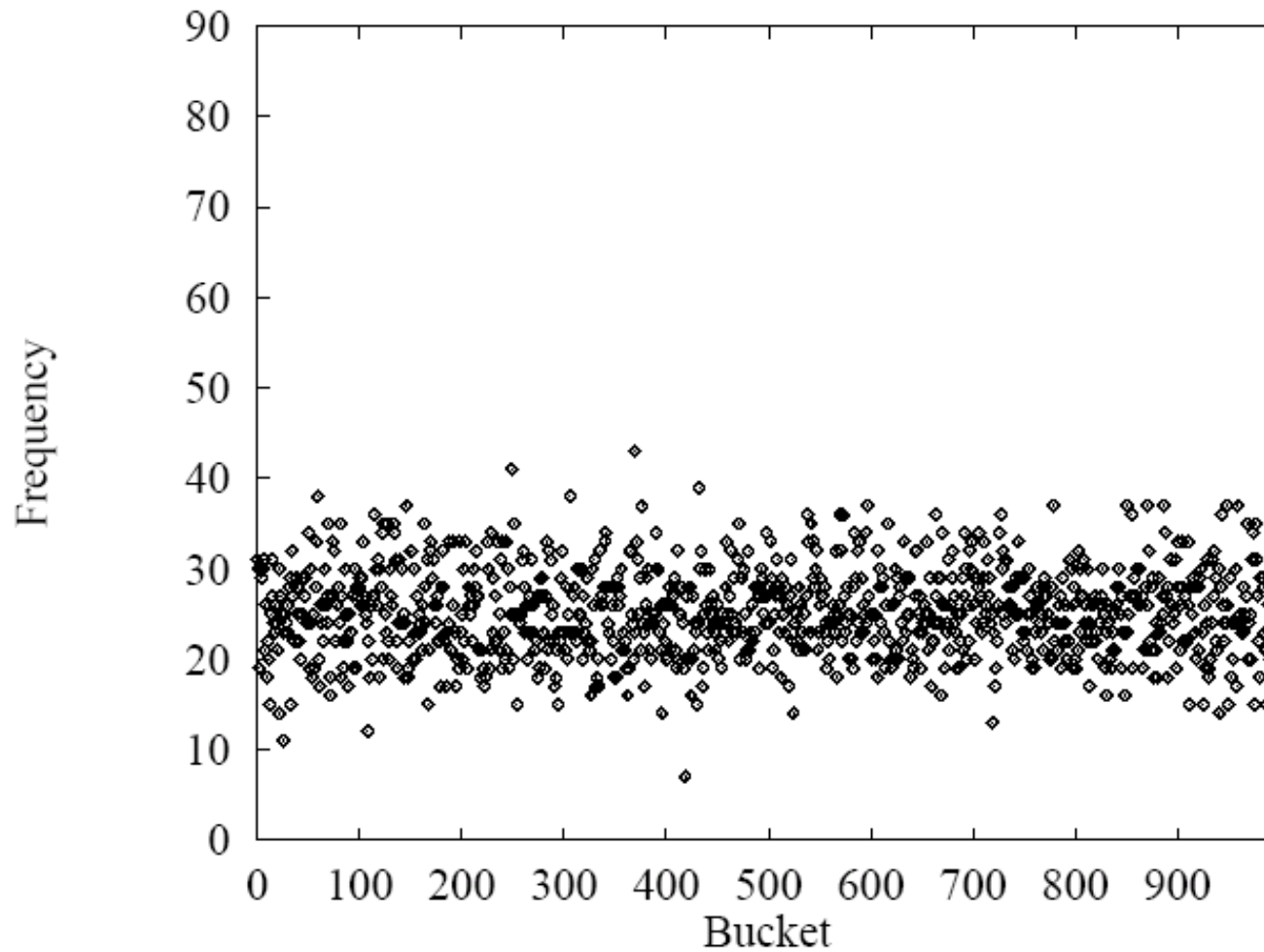


$$\sum_{i=0}^n \text{s.charAt}(i) * 2^i$$



$$\sum_{i=0}^n \text{s.charAt}(i) * 256^i$$

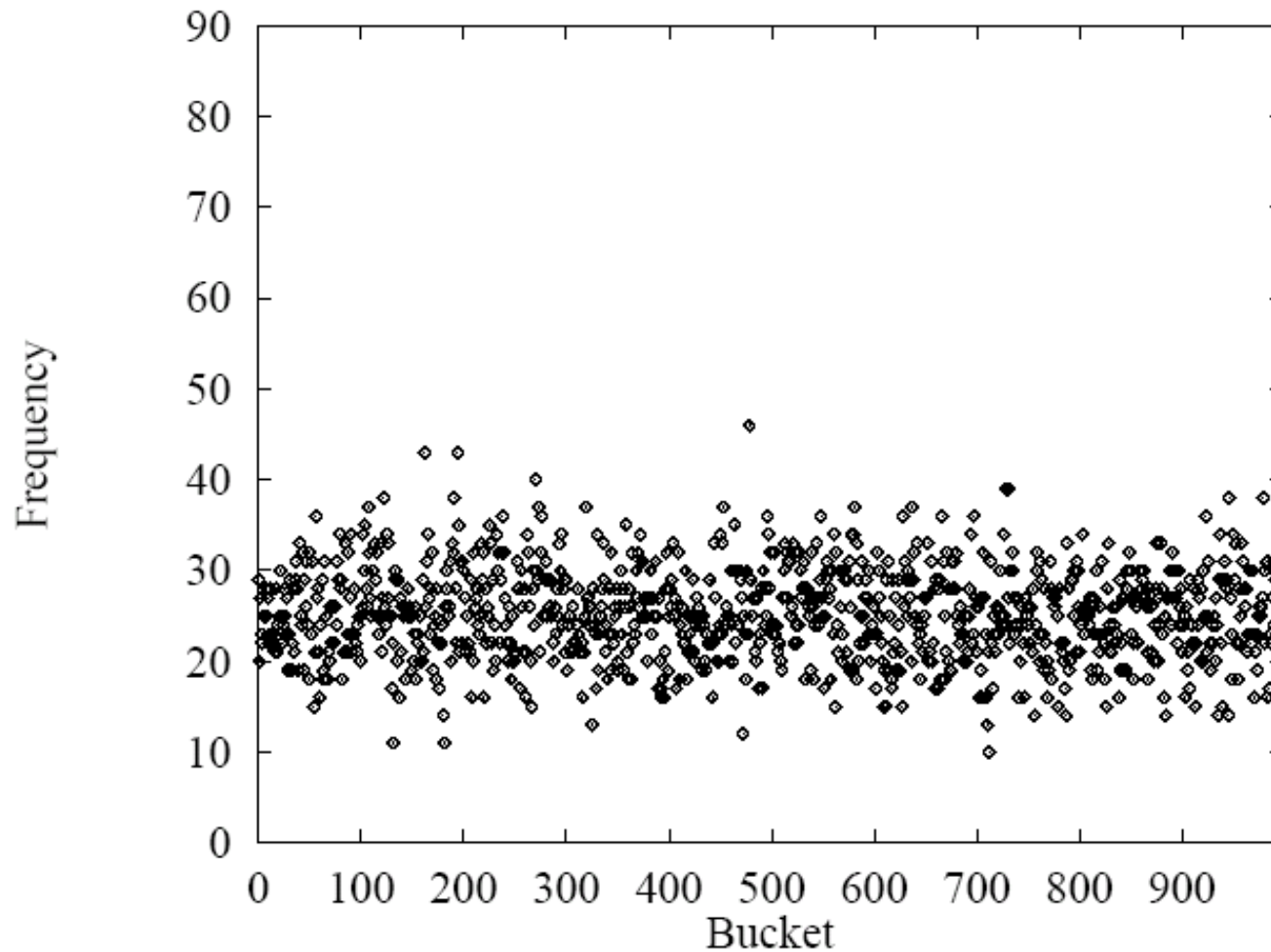
This looks pretty good, but  $256^i$  is big...



$$\sum_{i=0}^n s.\text{charAt}(i) * 31^i$$

Java uses:

$$\sum_{i=0}^n s.\text{charAt}(i) * 31^{(n-i-1)}$$





# Hashtables: $O(1)$ operations?

- How long does it take to compute a String's hashCode?
  - $O(s.length())$
- Given an object's hash code, how long does it take to find that object?
  - $O(\text{run length})$  or  $O(\text{chain length})$  PLUS cost of `.equals()` method
- Conclusion: for a good hash function (fast, uniformly distributed) and a low load factor (short runs/chains), we *say* hashtables are  $O(1)$

# Summary

	put	get	space
unsorted vector	$O(n)$	$O(n)$	$O(n)$
unsorted list	$O(n)$	$O(n)$	$O(n)$
sorted vector	$O(n)$	$O(\log n)$	$O(n)$
balanced BST	$O(\log n)$	$O(\log n)$	$O(n)$
array indexed by key	$O(1)$	$O(1)$	$O(\text{key range})$