

Hash tables

Dictionary ADT

- Reminder: A dictionary stores pairs (key, information)
- Operations:
 - find(key k)
 - insert(key k, info i)
 - remove(key k)
- Binary Search Trees implement all these operations in time $O(h)$, where h is the height of the tree, which is $O(\log n)$ if we maintain the tree balanced
- We can sometimes do better...

Hash tables

- Suppose keys are integers between 0 and $K-1$
- Then, use an array $A[0..K-1]$ containing elements of type "info" to store the dictionary:
 - insert(key k, info i): $A[k] = i;$
 - remove(key k): $A[k] = \text{null};$
 - find(key k): return $A[k]$
- Running time: All operations are $O(1)$
- It's a miracle! Except that...

Problems with direct array implementation

- If K is large, the array will be very big
 - For McGill student ID, $K = 1\,000\,000\,000$
- The amount of memory needed (K) is essentially independent of the number of items in the dictionary.
- Idea: compress the array...

Hash functions

Idea: Map the K possible keys to N integers, with N being much smaller than K

Hash function $f: [0..K-1] \rightarrow [0..N-1]$

Space of keys: 0 1 2 $K-1$

Hash function

Hashed key 0 1 2 $N-1$

insert(key k, info i): $A[f(k)] = i;$

remove(key k): $A[f(k)] = \text{null};$

find(key k): return $A[f(k)];$

Hash tables

- Collisions! Many keys map to the same index
- Solution: Each element of the array is itself a dictionary (called a bucket), implemented with linked-list, binary search tree, or a hash table...

Hash table:

insert(key k, info i): $A[f(k)].\text{insert}(k,i);$

remove(key k, info i): $A[f(k)].\text{remove}(k);$

find(key k): return $A[f(k)].\text{find}(k);$

Hash tables - example

- Hashing student IDs:
 - $K = 1,000,000,000$
- $N = 10$
- $\text{Hash}(\text{ID}) = \text{lastDigit}(\text{ID})$
- Bucket implemented as linked-lists

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Insert(260053665, "Mathieu")

Hash tables - example

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Insert(260625329, "John")

Hash tables - example

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Insert(260313595, "Laura")

Hash tables - example

- Hashing student IDs:
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- $\text{Hash}(\text{ID}) = \text{lastDigit}(\text{ID})$
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Insert(260435215, "Julie")

Hash tables - example

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Find(260435215)

Hash tables - example

- Hashing student IDs:
 - $K = 1,000,000,000$
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Find(260435215)

Hash tables - example

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Find(260435215)

Analysis of Hashing with Chaining

- **Search time** = compute hash function + search the list.
- Time to compute hash function: $O(1)$.
- Worst time for searching happens when all keys go in the same bucket. We need to scan the full list $\Rightarrow O(n)$.
- **Search time** = $O(1) + O(n) = O(n)$
- **Insertion:** $O(1)$ time.
- **Deletion:** $O(1) + \text{Search time}$.

Importance of good hash functions

- Worst case complexity for hash table containing n elements
 - if all keys end up in the same bucket and we use a linked-list to store buckets??
 - if keys are evenly spread among the N buckets??
- We want a hash function that spreads the keys evenly among the buckets.
- Example: $N = 100$, key = student ID #
 - $f(\text{key } k) = \lfloor k/10\,000\,000 \rfloor$ = first 2 digits
 - $f(\text{key } k) = k \bmod 100$ = last 2 digits
 - $f(\text{key } k) = (\text{sum of digits of } k) \bmod 100$

Good hash functions

- Choice of hash function depends on application
- In general, $f(k) = k \bmod N$ is good choice when N is a prime number
- Example: For student Ids, choose $N = 101$
 - $f(k) = k \bmod 101$
- What if the key is not an integer (e.g. a String)?
 - map key to integer first with some function $g(\text{key})$
 - use $f()$ to map the integer to $[0..N-1]$

Hash functions on Strings

- We need a function $g: \text{String} \rightarrow \text{Integers}$ that minimizes collisions
 - Linear code:
 - $g(\text{key } k) = \text{sum of ASCII values of each char.}$
 - Problem:
 - Polynomial code: Choose a small prime number a
 - If key $k = k_0k_1k_2\dots k_e$, choose
 - $g(k) = k_0 + k_1 a + k_2 a^2 + \dots + k_e a^e$