VE281
Data Structures and Algorithms

Hashing: Basics and Hash Function

Learning Objectives:

- Know the purpose of hashing

- Understand the collision problem

- Know how to design a good hash function

Outline

- Review of Dictionary

- Hashing Basics

- Hash Function

Dictionary

- How do you use a dictionary?
  - Look up a "word" and find its meaning.
- We also have an abstract data type called dictionary.
  - It is a collection of pairs, each containing a **key** and an **value**

    (key, value)

  - **Important**: Different pairs have different keys.
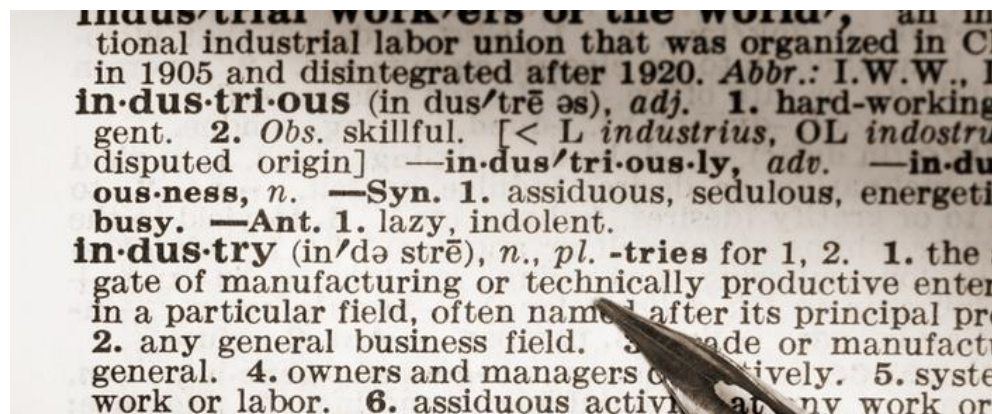
Dictionary

- Key space is usually more regular/structured than value space, so easier to search.

- Dictionary is optimized to quickly **add** `(key, value)` pair and **retrieve value** by key.

Methods

- **`Value find(Key k)`** : Return the value whose key is **k**. Return **`Null`** if none.

- **`void insert(Key k, Value v)`** : Insert a pair **`(k, v)`** into the dictionary. If the pair with key as **k** already exists, update its value.

- **`Value remove(Key k)`** : Remove the pair with key as **k** from the dictionary and return its value. Return **`Null`** if none.

Runtime for Array Implementation

`Pair Array[MAXSIZE]:` | a | b | c | d | | | |

- Unsorted array
  - find() $O(n)$
  - insert() $O(n)$: $O(n)$ to verify duplicate, $O(1)$ to put at the end
  - remove() $O(n)$: $O(n)$ to verify existence, $O(1)$ to exchange the "hole" with the last element

- Sorted array
  - find() $O(\log n)$: binary search
  - insert() $O(n)$: $O(\log n)$ to verify duplicate, $O(n)$ to insert
  - remove() $O(n)$: $O(\log n)$ to verify existence, $O(n)$ to remove

Can we do **`find`**, **`insert`**, and **`remove`** in $O(1)$ time?

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Hashing: High-Level Idea

- **Setup**: A universe U of objects
  - E.g., All names, all IP addresses, etc.
  - Generally, very BIG!
- **Goal**: Want to maintain an evolving set $S \subseteq U$
  - E.g., 200 students, 500 IP addresses
  - Generally, of reasonable size.
- Naïve solutions

1. Array-based solution (index by $u \in U$)
   - $\Theta(1)$ operation time, BUT $\Theta(|U|)$ space.
2. Linked list-based solution:
   - $\Theta(|S|)$ space, BUT $\Theta(|S|)$ operation time.

Can we get the best of both solutions?

Hashing: High-Level Idea

- Solution:
  - Pick an array A of n buckets.
    - $n = c|S|$: a small multiple of $|S|$.
  - Choose a hash function $h: U \rightarrow \{0, 1, \ldots, n-1\}$
    - h is fast to compute.
    - The same key is always mapped to the **same** location.
  - Store item k in A[h(k)]
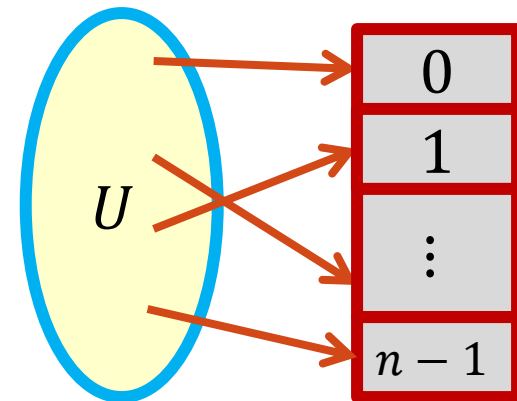
- The array is called **hash table**
  - An array of **buckets**, where each bucket contains items as assigned by a hash function.
  - h[k] is called the **home bucket** of key k.

Hashing Example

- Pairs are: (22,a), (33,b), (3,c), (73,d), (85,e)

- Hash table is `A[0:7]` and table size is `M = 8`

- Hash function is `h[key] = key/11`

- Every item with `key` is stored in the bucket `A[h(key)]`

(3,c)		(22,a)	(33,b)			(73,d)	(85,e)
[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]

Question: What is the time complexity for `find()`, `insert()`, and `remove()`?   O(1)

Achieves both fast runtime and efficient memory usage

What Can Go Wrong?

(3,c)		(22,a)	(33,b)			(73,d)	(85,e)
[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]

- Where does (35, g) go?
- Problem: The home bucket for (35, g) is already occupied!
  - This is a "**collision**".

Collision and Collision Resolution

- Collision occurs when the hash function maps two or more items—all having **different** search keys—into the **same** bucket.

- What to do when there is a collision?
  - **Collision-resolution scheme**: assigns distinct locations in the hash table to items involved in a collision.

- Two major schemes:
  - Separate chaining
  - Open addressing

Insight of Collision: Birthday Problem

- Consider n people with random birthdays (i.e., with each day of the year equally likely). What is the smallest n so that there is at least a 50% chance that two people have the same birthday?

  A. 23

  B. 57

  C. 184

  D. 367

Collision is inevitable!

Hash Table Issues

- Choice of the hash function.

- Collision resolution scheme.

- Size of the hash table and rehashing.

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Hash Function Design Criteria

- Must compute a bucket for every key in the universe.
- Must compute the same bucket for the same key.
- Should be easy and quick to compute.
- Minimizes collision ← The hardest criterion
  - Spread keys out evenly in hash table
  - **Gold standard**: **completely random hashing**
    - The probability that a randomly selected key has bucket i as its home bucket is $1/n$, $0 \leq i < n$.
    - Completely random hashing **minimizes** the likelihood of an collision when keys are selected at random.
    - However, completely random hashing is **<u>infeasible</u>** due to the need to remember the random bucket.

Bad Hash Functions

- Example: keys $=$ phone number in China (11 digits)
  - $|U| = 10^{11}$
  - **Terrible** hash function: $h(key) =$ first 3 digits of key, i.e., area code
    - The keys are not spread out evenly. Buckets 010, 021 may have a lot of keys mapped to them, while some buckets have no keys.
  - **Mediocre** hash function: $h(key) =$ last 3 digits of key.
    - Still vulnerable to patterns in last 3 digits.

Hash Functions

- Hash function ($h(key)$) maps key to buckets in two steps:

1. Convert key into an integer in case the key is not an integer.
   - A function $t(key)$ which returns an integer value, known as **hash code**.

2. **Compression map**: Map an integer (hash code) into a home bucket.
   - A function $c(hashcode)$ which gives an integer in the range $[0, n-1]$, where n is the number of buckets in the table.

- In summary, $h(key) = c(t(key))$, which gives an index in the table.

Map Non-integers into Hash Code

- String: use the ASCII (or UTF-8) encoding of each char and then perform arithmetic on them.

- Floating-point number: treat it as a string of bits.

- Images, (viral) code snippets, (malicious) Web site URLs: in general, treat the representation as a bit-string, using all of it or **extracting** parts of it (i.e., www.**sjtu**.**edu**.**cn**).

Strings to Integers

- Simple scheme: adds up all the ASCII codes for all the chars in the string.
  - Example: t("He") = 72 + 101 = 173.

- Not good. Why?
  - Consider English words "post", "pots", "spot", "stop", "tops".

Strings to Integers

- A better strategy: Polynomial hash code taking **positional** info into account.

$$t(s[]) = s[0]a^{k-1} + s[1]a^{k-2} + \cdots + s[k-2]a + s[k-1]$$

where a is a constant.

- If $a = 33$, the hash codes for "post" and "stop" are
$$t(\text{post}) = 112 \cdot 33^3 + 111 \cdot 33^2 + 115 \cdot 33 + 116 = 4149734$$
$$t(\text{stop}) = 115 \cdot 33^3 + 116 \cdot 33^2 + 111 \cdot 33 + 112 = 4262854$$

Strings to Integers

$$t(s[]) = s[0]a^{k-1} + s[1]a^{k-2} + \cdots + s[k-2]a + s[k-1]$$

- Good choice of a for English words: 31, 33, 37, 39, 41
  - What does it mean for a to be a **good** choice? Why are these particular values **good**?
  - Answer: according to statistics on 50,000 English words, each of these constants will produce less than 7 collisions.

- In Java, its **string** class has a built-in **hashCode()** function. It takes $a = 31$. Why?
  - Multiplication by 31 can be replaced by a shift and a subtraction for **better performance**: `31*i == (i << 5) - i`

Hash function criteria: Should be easy and quick to compute.

Compression Map

- Map an integer (hash code) into a home bucket.
- The most common method is by **modulo arithmetic**.

  `homeBucket = c(hashcode) = hashcode % n`

  where n is the **number of buckets** in the hash table.

- Example: Pairs are (22,a), (33,b), (3,c), (55,d), (79,e). Hash table size is 7.

	(22,a)	(79,e)	(3,c)		(33,b)	(55,d)
[0]	[1]	[2]	[3]	[4]	[5]	[6]

Hashing by Modulo

- In practice, keys of an application tend to have a specific pattern
  - For example, memory address in computer is multiple of 4.

- The choice of the hash table size n will affect the distribution of home buckets.

Hashing by Modulo

- Suppose the keys of an application are more likely to be mapped into even integers.

  - E.g., memory address is always a multiple of 4.

- When the hash table size n is an **even** number, **even** integers are hashed into **even** home buckets.

  - E.g., n = 14: 20%14 = 6, 30%14 = 2, 8%14 = 8

- The bias in the keys results in a bias toward the **even** home buckets.

  - All **odd** buckets are **<u>guaranteed</u>** to be empty.
  - The distribution of home buckets is not uniform!

Hashing by Modulo

- However, when the hash table size n is **odd**, even (or odd) integers may be hashed into both odd and even home buckets.
  - E.g., n = 15: $20\%15 = 5$, $30\%15 = 0$, $8\%15 = 8$
    $$15\%15 = 0, 3\%15 = 3, 23\%15 = 8$$

- The bias in the keys does not result in a bias toward either the odd or even home buckets.
  - Better chance of uniform distribution of home buckets.
- So **<u>do not</u>** use an even hash table size n.

Hashing by Modulo

- Similar **biased** distribution of home buckets happens in practice when the hash table size n is a multiple of small prime numbers.

- The effect of each prime divisor p of n **decreases** as p gets **larger**.

- Ideally, choose the hash table size n as a **large prime number**.