

# Algorithms and Data Structures

## B4. Hash Tables

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## B4.1 Introduction

## B4.2 Chaining

## B4.3 Open Addressing

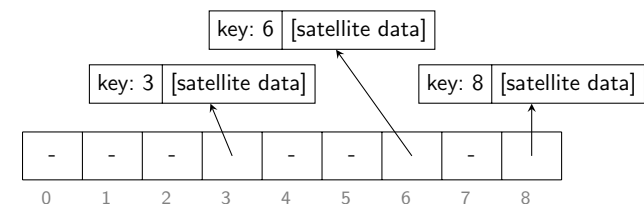
## B4.4 Hash Functions

## B4.5 Summary

## B4.1 Introduction

## Direct-address Table

- ▶ Assume you want to store elements that are associated with keys from a fixed universe  $U = \{0, 1, \dots, k\}$ .
- ▶ For every key, you need to store at most one element.
- ▶ **Idea:** Use array  $T$  (= direct access table), storing at position  $i$  a pointer to the element with key  $i$ .
- ▶ Inserting, removing and accessing the element for a key takes constant time.



## Disadvantages of Direct-address Table

- ▶ If the universe is large or infinite, storing a table of size  $|U|$  may be impractical or impossible.
- ▶ If the number of stored entries is small compared to the size of the universe, most space allocated for  $T$  would be wasted.

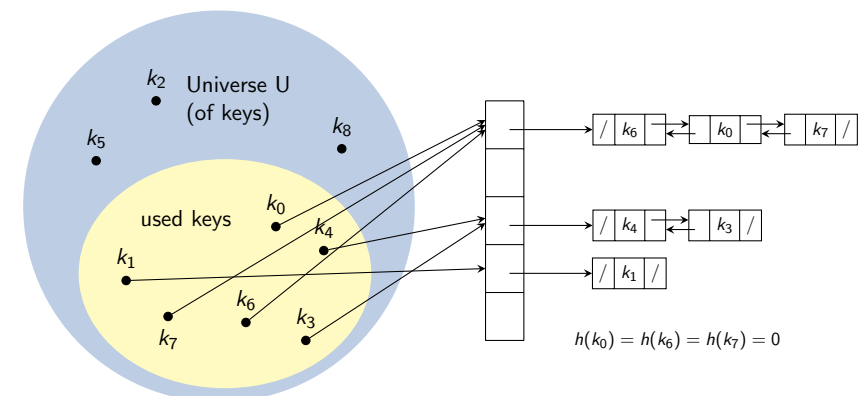
## Hash Table

- ▶ Use a smaller array  $T$  (= the **hash table**) of size  $m$ , and
- ▶ a **hash function**  $h : U \rightarrow \{0, \dots, m-1\}$ , mapping the universe of keys into the possible positions in  $T$ .  
For example  $h(k) = k \bmod m$
- ▶ We call  $h(k)$  the **hash value** of key  $k$ .
- ▶ **Problem:** possible **collisions**
  - ▶ Different keys mapped to same hash value.
  - ▶ Unavoidable if  $|U| > m$ .
- ▶ Need **collision resolution strategy**. We will cover two methods:
  - ▶ Chaining
  - ▶ Open Addressing

## B4.2 Chaining

## Hashing with Chaining

Every non-empty hash-table position  $i$  points to a doubly linked list (the **chain**) of all the keys whose hash value is  $i$ :



## Chaining: Implementation

- ▶ Search for an entry with key  $k$ 
  - ▶ Search for entry with key  $k$  in list  $T[h(k)]$ .
- ▶ Remove entry with key  $k$ 
  - ▶ Search for and remove element with key  $k$  from list  $T[h(k)]$ .
- ▶ Insert an entry  $e$  with key  $k$ 
  - ▶ Search for entry with key  $k$  in list  $T[h(k)]$ .
  - ▶ If found: update linked list node to hold  $e$ .
  - ▶ If not found: prepend entry to list at  $T[h(k)]$ .

## Chaining: Running Time I

- ▶ Assumption: Computing  $h(k)$  takes constant time.
- ▶ The running time of all operations is dominated by the running time of the linked-list operations.
- ▶ All operations linear in the size of the involved linked list.
- ▶ **Worst-case:** All entries have the same hash value.  
 $\rightsquigarrow$  worst-case running time linear in the number of entries

## Independent Uniform Hashing

- ▶ “Ideal” hash function: for each key  $k$ , hash value  $h(k)$  is randomly and independently chosen uniformly from the range  $\{0, \dots, m - 1\}$  (with  $m$  size of hash table).
- ▶ Subsequent calls of  $h(k)$  for the same key  $k$  give the same output.
- ▶ Such a  $h$  is called a **independent uniform hash function**.
- ▶ Cannot reasonably be implemented in practise but useful for theoretical analysis.

## Chaining: Running Time II

- ▶ **Load factor**  $\alpha$  is defined as  $n/m$ , where
  - ▶  $m$  is the number of positions (slots) in the hash table, and
  - ▶  $n$  is the number of stored elements.
- ▶  $\alpha$  is the average number of entries in a chain.

## Chaining: Running Time III

### Theorem

*In a hash table in which collisions are resolved by chaining, a search (successful or unsuccessful) takes  $\Theta(1 + \alpha)$  time on average, under the assumption of independent uniform hashing.*

### Consequence

If the number of elements  $n$  is at most proportional to the number of slots  $m$  ( $n \leq cm$  for constant  $c > 0$ ), then  $\alpha \leq cm/m \in O(1)$ .  
 → average running time of insertion, deletion and search is  $O(1)$ .

## Adapting the Size of the Hash Table

- ▶ To maintain an upper bound on the load factor (and thus constant average running times of operations), we may need to increase the size of the table.
- ▶ The change from the previous size  $m$  to size  $m'$  requires an adaptation of the hash function.
- ▶ In contrast to a size change of an array (where we just move every entry to the same index of the new memory range), we need to rehash all elements and insert them anew.

## B4.3 Open Addressing

## Open Addressing

- ▶ In contrast to chaining, with open addressing the entries are stored in the hash table itself.
- ▶ Hash table cannot hold more entries than size  $m$  (load factor cannot exceed 1).
- ▶ Size adaptation is analogous to chaining (need to rehash and reinsert all entries).
- ▶ To find a slot to insert an element, probe the hash table for the key until you find an empty slot:
  - ▶ If first choice for key occupied, try the second choice,
  - ▶ if second choice for key occupied, try the third choice,
  - ▶ ...
- ▶ To search for an element with key  $k$ , probe the table for the key until you find a slot that holds an element with key  $k$ .

## Hash Functions for Open Addressing

- ▶ The hash function contains the probe number as a second input:

$$h : U \times \{0, \dots, m-1\} \rightarrow \{0, \dots, m-1\}$$

- ▶ Probe sequence for key  $k$ :  $\langle h(k, 0), h(k, 1), h(k, 2), \dots, h(k, m-1) \rangle$ .
- ▶ For every key, the probe sequence must be a **permutation** of  $\{0, \dots, m-1\}$ : every position in the hash table included exactly once.

## Open Addressing: Insertion and Search

Assumption:  $key(e) = e$ . Fix hash function  $h$ , hash table size  $m$ .

```

1  def hash_insert(T, k):
2      for i in range(m): # i = 0, 1, ..., m-1
3          pos = h(k, i)
4          if T[pos] is None: # position empty
5              T[pos] = k
6              return pos
7          raise Exception("hash table overflow")

1  def hash_search(T, k):
2      for i in range(m):
3          pos = h(k, i)
4          if T[pos] == k:
5              return pos
6          if T[pos] is None:
7              break
8      return None # does not contain k

```

## Open Addressing: Deletion?

- ▶ When deleting the element, we may not simply set the slot to `None` (*Why?*).
- ▶ Can mark the slot as `deleted`.
  - ▶ Insertion treats it like an empty slot.
  - ▶ Search treats it as an occupied slot.
- ▶ Disadvantage: Search times no longer depend on load factor but can take longer.
- ▶ If keys need to be deleted: consider chaining instead.
- ▶ Linear probing (a special case of open addressing) avoids need for `deleted` (later today).

## Open Addressing: Running Time I

- ▶ Assumptions for running time analysis:
  - ▶  $\alpha < m$  (at least one slot empty)
  - ▶ no deletions
  - ▶ independent uniform permutation hashing: the probe sequence for a key is equally likely to be any permutation of  $\{0, \dots, m-1\}$ .
- ▶ Unsuccessful search: every probe but the last accesses an occupied slot (not containing the search key), last slot is empty.
- ▶ Successful search: some probe in the probe sequence accesses a slot with the searched key.

## Open Addressing: Running Time II

### Theorem

For a open-address hash table with load factor  $\alpha = n/m < 1$ , the **expected number of probes in an unsuccessful search is at most  $1/(1 - \alpha)$** , assuming independent uniform permutation hashing and no deletions.

### Intuition:

$$1/(1 - \alpha) = 1 + \alpha + \alpha^2 + \alpha^3 + \dots$$

First probe always occurs, with probability  $\alpha$  the probed slot is occupied, so a second probe occurs, ...

### Corollary

Under the same assumption as in the theorem, **inserting an element requires at most  $1/(1 - \alpha)$  probes on average.**

## Open Addressing: Running Time III

### Theorem

For a open-address hash table with load factor  $\alpha < 1$ , the **expected number of probes in a successful search is at most  $\frac{1}{\alpha} \log_e \frac{1}{1-\alpha}$** , assuming independent uniform permutation hashing with no deletions and assuming that each key in the table is equally likely to be searched for.

## Double Hashing

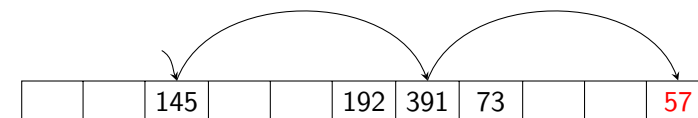
- ▶ Double hashing uses **two auxiliary hash functions**  $h_1 : U \rightarrow \{0, \dots, m - 1\}$  and  $h_2 : U \rightarrow \{0, \dots, m - 1\}$ .
- ▶ **Hash function**  $h(k, i) = (h_1(k) + i \cdot h_2(k)) \bmod m$
- ▶ Initial probe position  $h_1(k)$  and step size  $h_2(k)$  depend on  $k$ .
- ▶  $h_2(k)$  must be **relatively prime** to  $m$  (the only common divisor of  $h_2(k)$  and  $m$  is 1).

For example:

- ▶  $m$  power of 2 and  $h(k)$  odd number, or
- ▶  $m$  prime and  $h(h)$  positive integer less than  $m$ .

## Double Hashing: Example

- ▶  $m = 11$ ,  $h_1(k) = k \bmod 11$ ,  $h_2(k) = 1 + k \bmod 9$
- ▶ Insert  $k = 57$ .
  - ▶  $57 \bmod 11 = 2$
  - ▶  $57 \bmod 9 = 3$



## Special Case: Linear Probing

Use hash function  $h_1 : U \rightarrow \{0, \dots, m - 1\}$

- ▶ Probe sequence:  
 $\langle h_1(k), h_1(k) + 1, \dots, h_1(m - 1), h_1(0), h_1(1), \dots, h_1(k) - 1 \rangle$
- ▶  $h(k, i) = (h_1(k) + i) \bmod m$

Why is this a special case of double hashing?

## Linear Probing: Deletion I

- ▶ Use function  $g(k, q) = (q - h_1(k)) \bmod m$ .
- ▶ If  $h(k, i) = q$  then  $g(k, q) = i$

## Linear Probing: Deletion II

```

1 def linear_probing_hash_delete(T, q): # delete entry at position q
2     T[q] = None
3     pos = q
4
5     # search for a key that would have been inserted at position q
6     # instead of its current position if q had been free.
7     while True:
8         pos = (pos + 1) % m # next slot in linear probing
9         if T[pos] is None:
10            # there is no key that would have been inserted at q.
11            return
12        key = T[pos] # this could be such a key
13        if g(key, q) < g(key, pos):
14            # indeed, this key should be moved to q.
15            break
16        # otherwise continue with next position
17
18    T[q] = key # move key into slot p
19    linear_probing_hash_delete(T, pos) # now pos needs to be emptied

```

## Linear Probing: (Dis-)Advantage

Disadvantage: Primary clustering

- ▶ An empty slot occurring after  $i$  full slots gets filled next with probability  $(i + 1)/m$ .
- ▶ Linear probing has a tendency to build up long runs of occupied slots (so-called clusters).
- ▶ Running time of search depends on size of clusters.

Advantage: Data locality

- ▶ Memory accessed by modern CPUs has a number of levels (registers, cache, main memory, ...).
- ▶ For example, the cache always fetches entire cache blocks from the main memory.
- ▶ Linear probing mostly “reuses” the same fetched block, avoiding frequent (slow) access to the main memory.

## B4.4 Hash Functions

## Static Hashing: Division and Multiplication Method

For the moment, we consider keys that are non-negative integers that fit in a machine word (32 or 64 bits).

**Static hashing** uses a single, fixed hash function.

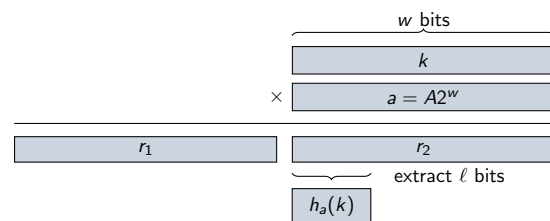
Examples ( $m =$  hash table size):

- ▶ **Division method:**  $h(k) = k \bmod m$ 
  - ▶ Works well when  $m$  is a prime not too close to a power of 2.
- ▶ **Multiplication method:** pick some  $A$  with  $0 < A < 1$ . Then

$$h(k) = \lfloor m(kA - \lfloor kA \rfloor) \rfloor.$$

- ▶  $kA - \lfloor kA \rfloor$ : fractional part of  $kA$ .
- ▶ Works best if  $m = 2^\ell$ , where  $\ell \leq w$ , where  $w$  is the number of bits in a machine word.

## Static Hashing: Multiply-shift Method



- ▶  $m = 2^\ell$  for integer  $\ell < w$ , where  $w$  is the number of bits in a machine word.
- ▶ For  $0 < A < 1$ , the result of  $k \cdot A2^2$  is an integer with  $\leq 2w$  bits (= 2 words).
- ▶ Use  $\ell$  most significant bits of the low-order word of the product as hash value.
- ▶ Fast but no formal guarantees.

## Random Hashing

- ▶ For every static hash function, an adversary can choose a sequence of keys that are all hashed to the same slot.
- ▶ **Random hashing** chooses the hash function randomly and independently of the keys that are going to be stored
- ▶ The special case of **universal hashing** guarantees good average performance, independent of the sequence of keys.



## Random Hashing: Universal Hashing

- ▶ A family  $\mathcal{H}$  of hash functions mapping universe  $U$  into slots  $\{0, \dots, m-1\}$  is **universal** if for each pair of distinct keys  $k, k' \in U$  there are at most  $|\mathcal{H}|/m$  hash functions  $h \in \mathcal{H}$  such that  $h(k) = h(k')$ .
- ▶ Universal hashing can be achieved in practise (e.g. using multiply-shift).
- ▶ With universal hashing and chaining, **any sequence of  $s$  insert, delete and search operations** takes  $\Theta(s)$  **expected time**, if it starts from an empty hash table with  $m$  slots and includes at most  $O(m)$  insert operations

## Cryptographic Hashing

- ▶ **Cryptographic hash functions** are complex pseudorandom functions, designed for applications requiring properties beyond those needed here.
- ▶ Some CPUs contain specific instructions to support a fast computation of some cryptographic functions.
- ▶ A cryptographic hash function takes as input an arbitrary byte string and returns a fixed-length output.
  - ▶ E.g. SHA-256 produces a 256-bit (32-byte) output for any input.
  - ▶ We can use  $h(k) = \text{SHA-256}(k) \bmod m$ , or
  - ▶ create a family of such hash functions by prepending different “salt” strings  $a$  to  $k$ .

## B4.5 Summary

## Summary

- ▶ **Hash functions** map the keys of the universe to the  $m$  possible slots of the hash table.
- ▶ Since there typically are more possible keys than slots, **collisions** are unavoidable.
- ▶ We deal with them by **chaining** and **open addressing** (e.g. using **linear probing**).
- ▶ Designing good hash functions is non-trivial and often uses a random selection from a family of functions.
- ▶ With a good hash function and load factor management, insertion and (successful) search is possible in constant amortized time on average (logarithmic in the worst case).