

## ALGORITHMS IN AI & DATA SCIENCE 1 (AKIDS 1)

# Hashing

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# Content

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- ADS: Associative Array
- Hash Table
- Hash Functions

# Dynamic sets

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- We go back to our **dynamic sets**
  - We need to store a set of data points (simple data types or complex ones)
- We've already seen several ADS that can store **dynamic sets**
  - **List**
  - **Stack**
  - **Queue**
  - **Priority Queue (Heap)**
- From these, only **list** has no constraints on **insertion** into and **removal** of elements from the **dynamic set**

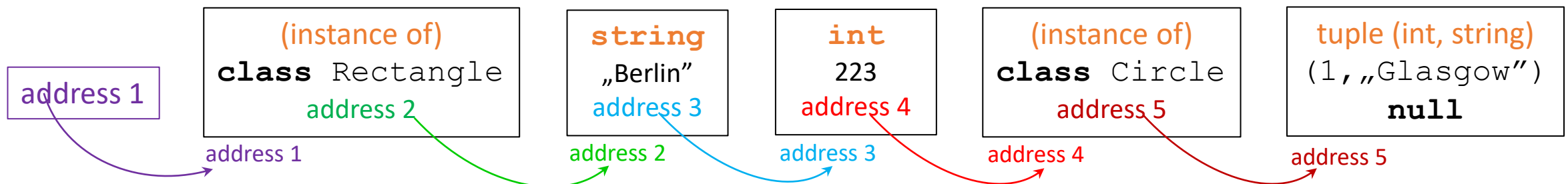
# Dynamic sets

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- In many applications/algorithms we need only **three basic operations** for manipulating the content of **dynamic sets**
  - 1. INSERT** – add new element to the dynamic set
    - In general, in no particular order
    - Constraints on positioning of elements: stacks, queues, heaps, search trees...
  - 2. SEARCH** – answer the question „is element X in the set“?
  - 3. DELETE** – remove an element from the set
    - In general, any element from the set can be removed
    - Constraints on (order of) element removal – stacks, queues, heaps, search trees...

# Recap – Lists: Arrays vs. Linked Lists

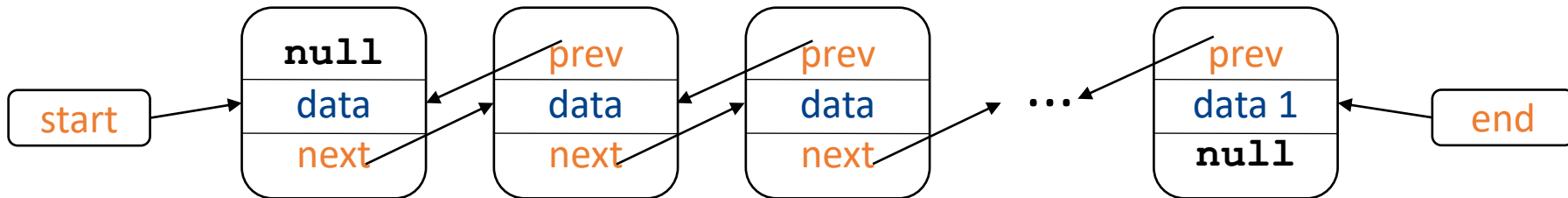
- ADT: **List** – a **linear** sequence of elements
  - When we design algorithms, we typically think in terms of ADTs
- **Linked List**
  - Consists of **nodes**: nodes contain both the data (values) and a **pointer** to the next node in the list
  - Nodes can contain values of different types
  - Dynamic data structure: „resizable” at run time
    - Non-contiguous memory allocation possible, space for **new nodes** can be allocated dynamically (on „per-need” basis)



# Dynamic Set with List

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- Dynamic set operations: **INSERT, DELETE, SEARCH**
- List implemented as a (bidirectional) **linked list**:



- Runtimes (in **Big-O** notation):
  - INSERT (assuming no constraints on where the element is to be inserted)?
  - SEARCH?
  - DELETE (assuming no constraints on where the element is to be inserted)?

# Abstract Data Types

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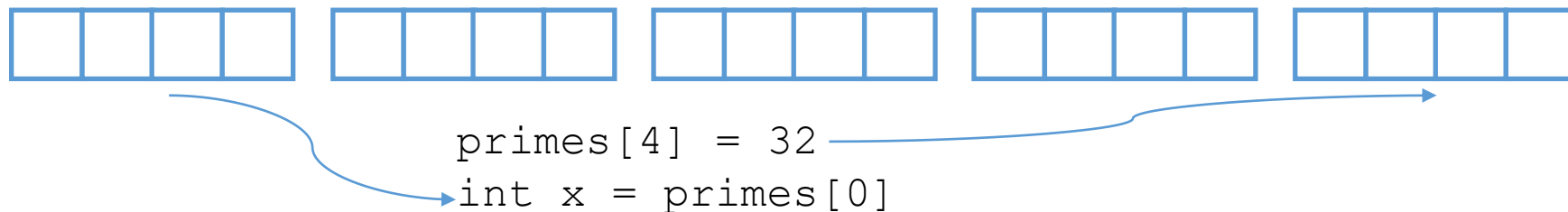
Abstract Data Type	Other Common Names	Commonly implemented as
List	Sequence	Array, Linked List
Queue		Array, Linked List
Double-ended Queue	Deque, Deque	Array, Doubly-linked List
Stack		Array, Linked List
Associative Array	Dictionary, Hash Map, Map	Hash Table
Set		Red-black Tree or Hash Table
Priority Queue	Heap	Heap

# Associative Array

## Associative array

An ADS for representing dynamic sets containing **(key, value)** pairs such that each key is **unique** in the array. Associative array, also known as **Dictionary** or **Map**, supports **direct addressing**: computation of memory location of **value** directly from the **key**.

- Isn't a regular **array** associative **by default**?
  - **key** = index of the array at which we find the element
  - Given the **key** (i.e., index), we can compute the memory address of the value





# Associative Array

## Associative array

An ADS for representing dynamic sets containing **(key, value)** pairs such that each key is **unique** in the array. Associative array, also known as **Dictionary** or **Map**, supports **direct addressing**: computation of memory location of **value** directly from the **key**.

- Isn't a regular **array** (aka **direct-access table**) associative by default?
  - **key** = index of the array at which I find the element
- What if we need to store a **very large number of elements**?
  - Memory reservation for **every possible key** → **large memory occupancy**
- What if the the **space/universe of keys** is virtually unlimited?
  - E.g., any string?

# Associative Array

## Associative array

An ADS for representing dynamic sets containing **(key, value)** pairs such that each key is **unique** in the array. Associative array, also known as **Dictionary** or **Map**, supports **direct addressing**: computation of memory location of **value** directly from the **key**.

- Let **S** be the set (possibly infinite) of all allowed keys
  - Defined commonly with some primitive data type: `int`, `string`, `float`
  - This basically means that key can be **any** value of the primitive type
- Let **K** be the set of keys we would have in any concrete dynamic set
  - In most practical problems,  $K \ll |S|$
  - If we would implement the dynamic set as the direct-access table (i.e., array), we'd have to allocate `|S|` slots in the memory and only **K** would be „used“
    - **Waste of memory**

# Content

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- ADS: Associative Array
- Hash Table
- Hash Functions

# Hash table

## Hash table

**Hash table** is a data structure (tightly coupled with a corresponding **hash function**) used for implementing **runtime and memory efficient associative arrays**.

- Direct-access tables store the element with **key**  $k$  at **index**  $k$
- **Hash tables** store the element with key  $k$  at index  $h(k)$
- $h(k)$  is the **hash(ing) function** that computes the array index at which to store/find the element's value  $v$  directly from its key  $k$
- Assuming a table (array)  $T$  of size  $m$  elements,  $h$  maps the universe of keys  $S$  to indices of the array

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

# Hash table

- Direct-access tables would need  $|S|$  elements
- Because of the hashing (i.e., mapping) function  $h$ , Hash table  $T$  can have  $m \ll |S|$  elements
- Hashing reduces the range of the indices, i.e., the size of the array
  - Memory (space saving)
- **Caveat:** what if two keys „hash” to the same value?  $h(k_2) = h(k_5)$ 
  - **Q:** How **likely** is that to happen? What does it depend on?

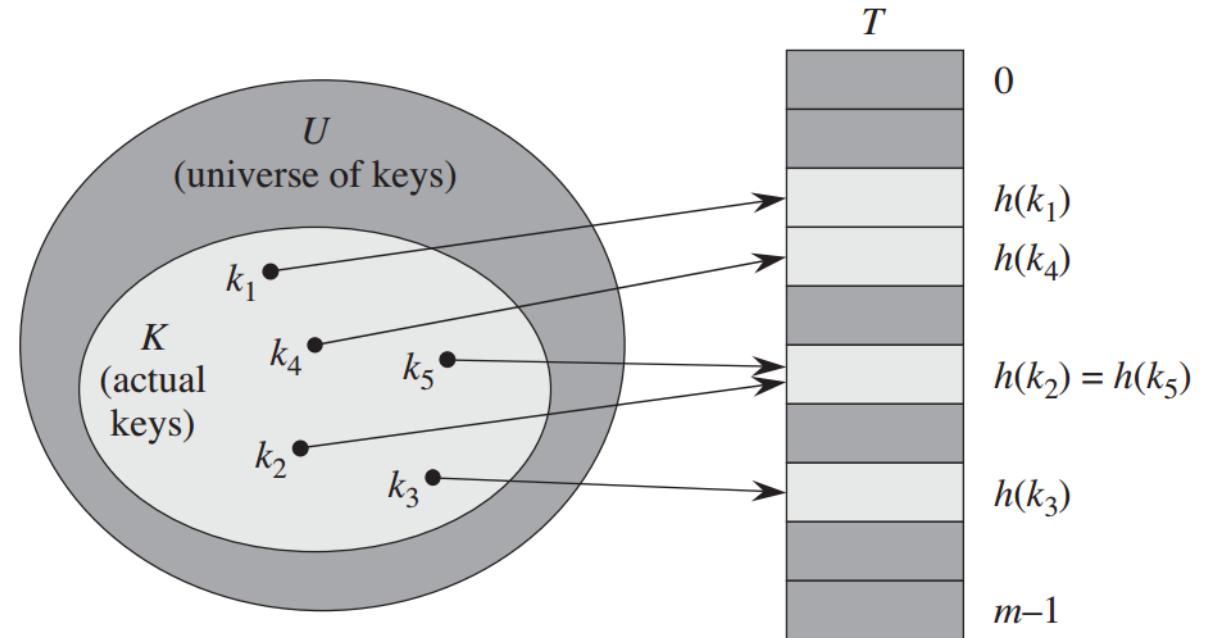


Image source: Cormen et al.

# Collisions and hash functions

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- **Collision**: when two (or more) keys get the same **hash**,  $h(k_1) = h(k_2)$
- The frequency of collisions depends on
  - (1) The size  $m$  of the table (array)  $T$
  - (2) The properties of the hashing function  $h$
  - (3) The concrete set of keys to be hashed,  $K = \{k_1, k_2, \dots, k_K\}$
- Obviously, we want to **avoid** or at least **minimize** collisions
- We typically have no control over (3), some control over (1), and only full control over (2): the **selection/design** of the hash function

# Collisions and hash functions

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- **Crucial:** hash functions **must be deterministic**. What does that mean?
- We typically have no control over the set of keys **K** for which the values need to be stored in the hash table
  - Because of this, we **cannot guarantee** the absence of collisions
  - We can reduce the probability of collisions. **How?**
- If  $m$  (size of **T**)  $\geq |K|$  it is theoretically possible to store all allowed keys without collision
  - **Problem:** we don't actually know **K** in advance, we know only **S** (the universe of all possible/allowed keys)

# Collisions and hash functions

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- We typically have no control over the set of keys  $K$  for which the values need to be stored in the hash table
  - Because of this, we **cannot guarantee** the absence of collisions
  - **Q1**: If  $m$  (size of  $T$ )  $< |K|$  what is the minimal possible number of collisions?
  - **Q2**: If keys in  $K$  are integers  $\{0, 1, 2, \dots, |K|-1\}$  and the size of the hash table  $T$  is  $m < |K|$ , find a hash  $h$  that guarantees this minimal number of collisions?
- Since we cannot guarantee no collisions, how do we handle them?



# Collision resolution by chaining

- At each *hash slot* (index in  $\{0, \dots, m-1\}$ , aka **bucket**) we put a pointer to the beginning of a **linked list**
- Slot/bucket  $i \in \{0, \dots, m-1\}$  contains a **pointer to the head** of the linked list of all stored elements whose keys hash to  $i$ 
  - If no keys have been hashed to  $i$ , the pointer is **null**

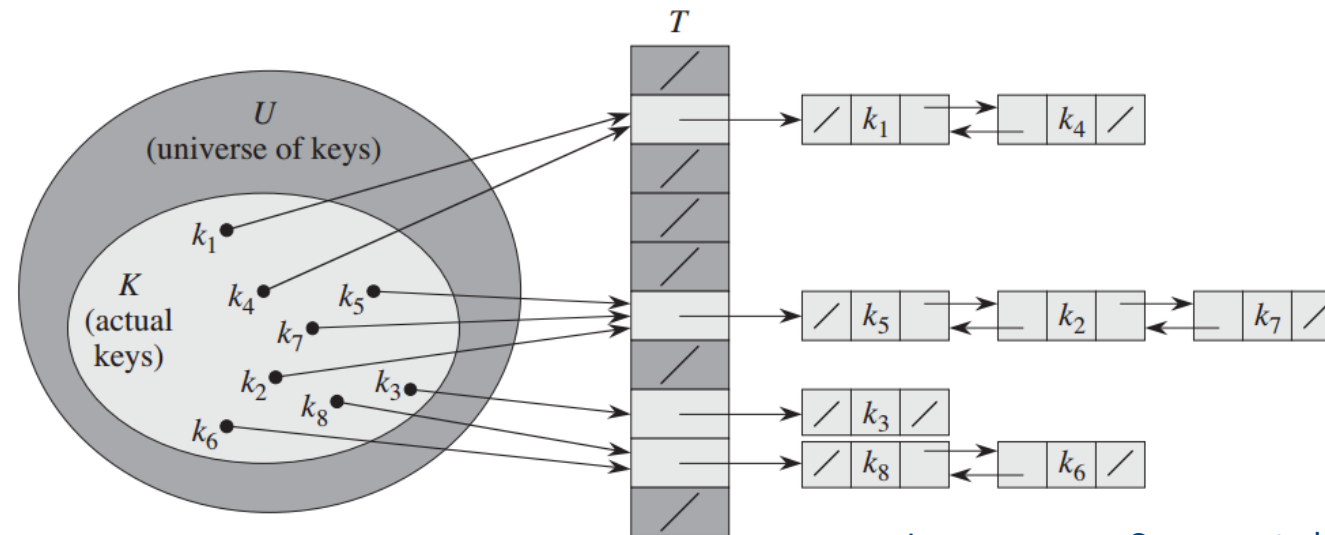
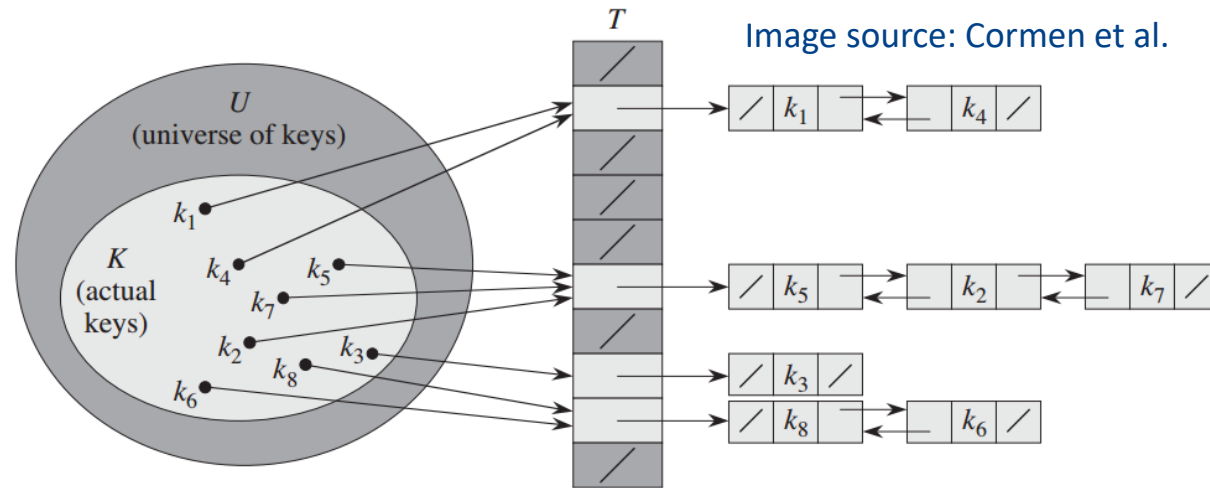


Image source: Cormen et al.

# Collision resolution by chaining



- Let  $x$  be a set element with key  $x.key$  and value  $x.value$ 
  - $x.next$  the pointer to the next element in the collision chain of the slot
- Let  $T$  be the hash table with collision resolution by chaining
- **INSERT**: add to the beginning of the chain

Insert( $T, x$ ):

pointer =  $T[h(x.key)]$

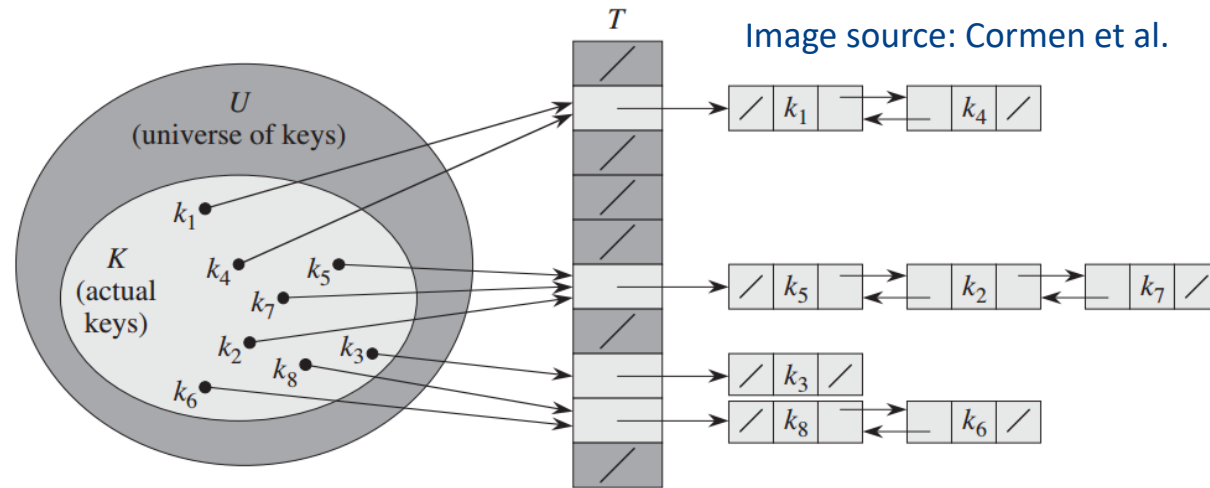
$x.next$  = pointer

pointer = address( $x$ )

**Q1**: Worst case running time?

**Q2**: What if  $x$  already in the list?

# Collision resolution by chaining



- Let  $x$  be a set element with key  $x.key$  and value  $x.value$ 
  - $x.next$  the pointer to the next element in the collision chain of the slot
- Let  $T$  be the hash table with chaining resolution of collisions
- **SEARCH**: find in the linked list of the chain

```
Search(T, key):  
    pointer = T[h(key)]  
    while pointer != null  
        x = read_memory(pointer)  
        if x.key == key  
            return x # or x.value  
        else  
            pointer = x.next  
    return null
```

Q: Worst case running time?

# Collision resolution by chaining

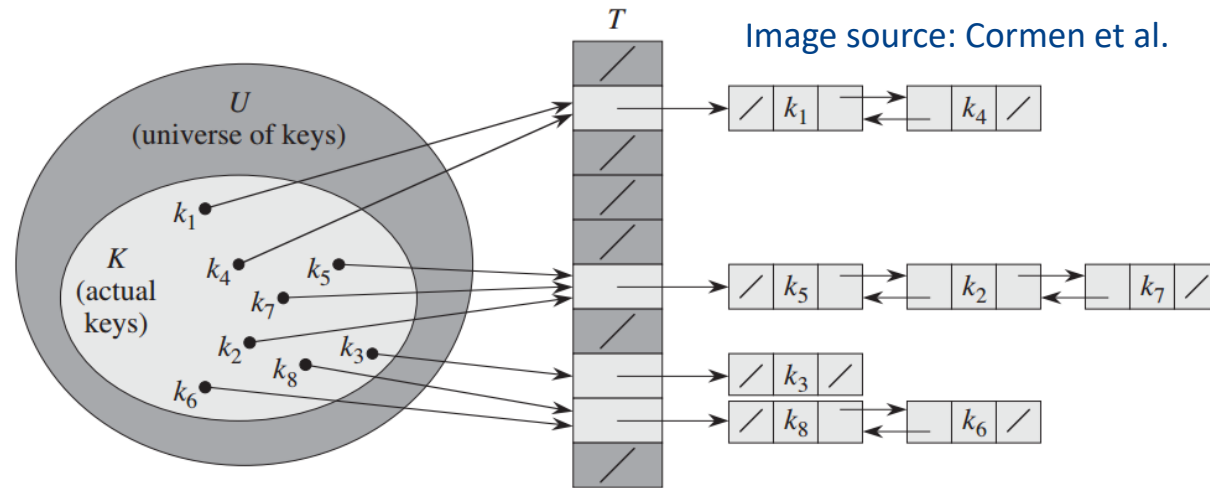


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- Let  $x$  be a set element with key  $x.key$  and value  $x.value$ 
  - $x.next$  the pointer to the next element in the collision chain of the slot
- Let  $T$  be the hash table with chaining resolution of collisions
- **DELETE**: find in the linked list of the chain

```
Delete(T, key):  
    pointer = T[h(key)]  
    while pointer != null  
        x = read_memory(pointer)  
        if x.key == key  
            # & - at the address where the pointer points  
            # we set the value x.next  
            &pointer = x.next  
        return  
    else  
        pointer = x.next
```

**Q:** Worst case running time?

# Collision resolution by chaining

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- **Runtimes** for hash tables with chaining
- **Load factor:**  $\alpha = |K| / m$ 
  - The average number of elements chained in linked lists (chains)
  - Runtime analysis in terms of  $\alpha$
- What is the **worst case** for hashing with chaining?
  - A hash function  $h$  that would map all  $n = |K|$  keys into the same hash slot
  - Search and Delete:  **$O(n)$**
  - Insert is always  **$O(1)$**

# Collision resolution by chaining

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- **Runtimes** for hash tables with chaining
- **Load factor:**  $\alpha = |K| / m$ 
  - The average number of elements chained in linked lists (chains)
  - Runtime analysis in terms of  $\alpha$
- What is the **average case** for hashing with chaining?
  - Depends on how well the hashing function  $h$  distributes keys across the  $m$  hash slots/buckets
  - **Assumption: simple uniform hashing**
    - $h$  equally likely to map a **key** to any of the  $m$  buckets
    - Buckets will have roughly the same number of elements:  $\alpha$
    - Runtime of Search (and Delete) is  **$O(\alpha)$**

# Content

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# Hash functions

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- What would be a **good** hash function  $h$ ?
  - One that creates the minimal number of collisions
- If it satisfies the assumption of **simple uniform hashing (SUH)**, it will create a small number of collisions
  - Keys are equally likely to be hashed into any of the  $m$  buckets, independently of where the other keys have been hashed
  - **As if** you were randomly drawing a bucket for every key
    - Though we cannot do that, as this is **not deterministic**
  - If **SUH**, then we can always reduce the runtime (of `Search`) by increasing  $m$ : for large enough  $m$  (**trading space for time**),  **$O(1)$**
- Unfortunately, **no easy way to prove/check** if some  $h$  results with **SUH**
  - It also depends on the actual keys being hashed



# Designing hash functions

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- **Heuristic design**

- Kind of „**trial and error**” – we create  $h$  that „**makes intuitive sense**”
- Then we test it on various key sets  $\mathbf{K}$  to see whether it roughly exhibits the SUH property – that all bucket chains are similarly long, roughly  $\alpha = |\mathbf{K}|/m$

- Most hashing functions assume **integer keys**

- Other types (e.g., **float**, **string**) are converted to (natural) integers first
- For a **string**, let's assume a charset of  $\mathbf{N}$  characters (e.g.,  $\mathbf{N} = 128$ )
  - Each character corresponds to one integer between 0 and 127
  - „ $c_1c_2c_3$ ”  $\rightarrow 128^2 * c_1 + 128 * c_2 + c_3$

# Hash function: division method

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- Hash is the **remainder of the division** of the key with the size of the array  $m$ :  $h(\text{key}) = \text{key} \% m$
- It is common to avoid certain values of  $m$ , as this can violate the assumption of **simple uniform hashing**
  - Or make the violation more likely for arbitrary key sets
  - For example, we avoid  $m = 2^p$  because the value  $h(\text{key}) = \text{key} \% 2^p$  depends only on the lowest  $p$  bits of the **key**
    - this puts all keys with the same lowest  $p$  bits into the same bucket
- **Good choice** for  $m$ : a prime number not too close to any power of **2**
  - Example: **3000** keys, and we'd be ok with average chains of length  $\alpha = 4$ 
    - Good  $m$  would be **751** or **757**

# Hash function: multiplication method

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- Simple heuristic hashing like division, but choice of  $m$  less critical
  1. We choose a multiplier real number  $M$ ,  $0 < M < 1$
  2. We multiply the key  $k$  with  $M$  and take the **decimal remainder**  $r$ 
    - We will denote the decimal remainder of a float  $f$  as  $f\%1$
  3. We multiply  $r$  with the hash table size  $m$  and take the first integer smaller than that number
    - We denote the first integer smaller than a float  $f$  with  $\lfloor f \rfloor$  („floor“ of  $f$ )

$$h(k) = \lfloor (k * M \% 1) * m \rfloor$$

- For **multiplication hashing**, it's actually **desirable** that  $m = 2^p$ . **Q**: Why?

# Hashing functions: universal hashing

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- Multiplication and division hashing lead to uniform hashing as long as the input keys to be hashed are **not „rigged”** in a particular way
- Assume there's an **adversary** who wants to slow down your program that works with a hash table
  - By making **all/many** keys hashed to the same value
  - If they **guess your hash function**, they can easily do that
  - E.g., if they know you use **division method** with  $m = 751$ , they can send keys that leave the same remainder when divided with 751: {2, 753, 1504, ...}
- **Universal hashing**: no fixed hashing function
  - In every execution, choosing  $h$  randomly from a set/family of **carefully designed** hash functions  $H$  (but in a way **independent** of the keys)
  - Algorithm behaves differently in every execution, so no single input will always cause worst case running time

# Re-Hashing

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- Assume we set a hash table of size  $m$  in advance and then start hashing incoming keys and store corresponding values
- Assume we receive  $|K| = n \gg m$  to hash, so  $\alpha = n/m$  becomes larger
  - Search becomes **slower**
- **Re-hashing**
  - If we could increase  $m$ , that would reduce  $\alpha$
  - Creating a new hash table with larger  $m$ , re-hashing all existing keys
  - Trading space for time

