

Hashing (Hash Table): Implementation and Runtime Analysis

The Dictionary Data Structure

- Dictionary or map is a data structure.

Dictionary/map: aka. **associative array** is a data structure that consists of a collection of (key, value) pairs called an entry of a dictionary.

- Note: the key and the value can be composite.
- Usage of a dictionary:
 - The dictionary data structure is used to store (key, value) pairs where the user can loop up (=search) a value using the key.
 - To support its usage, a dictionary must provide the following operations:
 - `size()` : return the number of entries stored inside this dictionary.
 - `put(key, value)` : if `key` is found in the dictionary, it updates the value with `value` . Otherwise, it inserts (key, value) into the dictionary.
 - `get(key)` : return the corresponding `value` if `key` is found, and return `null` otherwise.
 - `remove(key)` : remove the dictionary entry containing `key` (and return the corresponding `value` if `key` is found).
 - The most frequently used operation is `get(key)` , so fast lookup is required!
- The Dictionary interface:

```
public interface Dictionary<K,V> {  
    public int size(); // return number of entries in the dictionary  
    public void put(K key, V value); // insert or update (key, value) pair  
    public V get(K key); // return value associated with key;  
                        // return null if not found  
    public V remove(K key); // remove entry with key and return corresponding value  
}
```

- We can implement the dictionary data structure using:
 - An array

- A linked list
- For simplicity, we will implement the dictionary using an array. To achieve it, we need to:
 - Definition of the `Entry` class to represent an entry in a dictionary.
 - Definition of a class to represent a dictionary using an array.
 - Implement the (support) methods of the `Dictionary` interface.
- The `Entry` class and the `ArrayMap` implementation:

```

public class ArrayMap<K,V> implements Dictionary<K,V> {
    /* ----- Nested Entry class ----- */
    private class Entry<K,V> {
        private K key; // The key (to look up)
        private V value; // the value (corresponding to the key)

        public Entry(K k, V, v) { // constructor
            key = k;
            value = v;
        }

        /** Accessor method fvor the key */
        public K getKey() {
            return key;
        }

        /** Accessor method for the value */
        public V getValue() {
            return value;
        }

        /** Mutator method for the value */
        public void setValue(V v) {
            this.value = v;
        }

        @Override
        public String toString() {
            return "(" + key + ", " + value + ")";
        }
    }

    /* ----- End of nested Entry class ----- */

    Entry<K,V>[] entry; // Dictionary
    int nEntries; // number of entries in the dictionary

    public ArrayMap(int N) { // Constructor
        entry = new Entry[N];
        nEntries = 0;
    }

    @Override
    public int size() {

```

```

        return nEntries;
    }

    @Override
    public void put(K k, V v) {
        for (int i = 0; i < nEntries; i++) {
            if (entry[i].getKey().equals(k)) {
                // Found
                entry[i].setValue(v); // update the value
                return;
            }
        }
        // Key not found
        entry[nEntries] = new Entry<K,V>(k, v); // insert (k,v)
        nEntries++;
    }

```

```

    @Override
    public V get(K k) {
        for (int i = 0; i < nEntries; i++) {
            if (entry[i].getKey().equals(k)) {
                // Found
                return entry[i].getValue();
            }
        }
        // Not found
        return null;
    }

```

```

    @Override
    public V remove(K k) {
        boolean found = false; // Indicate key not found
        int loc = -1; // contains index of key
        V ret = null; // contains the return value

        for (int i = 0; i < nEntires; i++) {
            if (entry[i].getKey().equals(k)) {
                found = true; // indicates key k was found
                loc = i; // remember the index of the entry
                break;
            }
        }
    }

```

```

    if (found) {
        // Key found
        ret = entry[loc].getValue(); // update return value
        for (int i = loc + 1; i < n Entries; i++) {
            // delete entry [loc]
            entry[i-1] = entry[i]; // shift array
        }
        nEntries--;
    }
    return ret;
}
}

```

- Problems with the `ArrayMap` implementation:
 - A dictionary is used to look up information (= value) for a given key .
 - The loop up operation `get()` is $\mathcal{O}(n)$.
 - Can we do better than $\mathcal{O}(n)$?
 - Yes, we can sort the array and use binary search to reduce the runtime to $\mathcal{O}(\log n)$.
 - Can we do better than $\mathcal{O}(\log n)$?
 - Yes, we can use hashing to reduce the runtime to $\mathcal{O}(1)$.

The Hash Function

- Insight on how to improve the search performance of arrays.
 - Fact on arrays:
 - Array access is very fast if access uses an array index.
 - Fact on dictionaries:
 - Entries in a dictionary are looked up using its key.
 - The problem with the `ArrayMap` implementation of the dictionary is that entries of the dictionary are stored using an index that is not related to the key .
 - To improve the search operation for a dictionary stored in an array, we need to find a way to relate (=map) the key `k` to an index `h` of the array:

`h = hashFunction(k)`

- This way of storing data into an array is called hashing.
- Hashing functions `H()` :

- Hash function is a function that maps a key k to a number h in the range $[0, M-1]$, where M = length of the array. That is, $h = H(k)$, where $h \in [0, \dots, M-1]$.
 - $H()$ is consistent: always gives the same answer for a given key.
 - $H()$ is uniform: the function values are distributed evenly across $[0 \dots M-1]$.
- A hash function is usually specified as the composition of 2 functions: $H(k) = H_2(H_1(k))$, where
 - $H_1(k)$ is the **hash code** function that returns the integer value of the key k .
 - $H_2(x)$ is a compression function that maps a value x uniformly to the range $[0, M-1]$.
- The hash code of a key:
 - Fact: all data inside a computer is stored as abinary number.
 - The `Object` class in Java contains a `hashCode()` method that returns the data stored in the `Object` as an integer.
 - We can use the `hashCode()` method as our $H_1(k)$ function.
- The compression function $H_2(x)$
 - Notice from the previous discussion on the hash code $H_1(k)$: $H_1(k)$ uses the data stored in the key k to compute (deterministically) a hash code value.
 - The compression function $H_2(x)$ has two purposes:
 - Make sure that the return value is in the range $[0, M-1]$. (where M is the length of the array).
 - Scatter/randomize the input value $x = H_1(k)$, so that the value $H_2(x)$ is evenly/uniformly distributed over the range $[0, M-1]$.
 - Why do we use uniform randomization?
 - The array element used to stroe the diction entry is `array index = H(k) = H2(H1(k))`.
 - Uniform randomization will minimize the likelihoos/chance that 2 different keys being hashed to the same value (=array index) (a.k.a. **collision**).
 - A commonly used compression function is the **Multiply Add Divide (MAD)** function:

$$H_2(x) = (\underbrace{(ax + b) \% p}_{\text{randomizes}}) \% M, \quad \text{where } p = \text{some prime number}$$

- In the MAD function, a and b are random numbers, and p is a prime number.
- In the examples of this course, we will use $p = 109345121$ (a prime number), $a = 123$ and $b = 456$.
- Note: p must be greater than M (i.e., $p > M$), otherwise, we will not use the full capacity of the array.

Hash Table

- Terminologies:
 - **Hash function** $H()$: maps a key k to an integer in the range $[0, M-1]$.
 $H(k)$ = integer in $[0, M-1]$.
 - **Hash value** h : the value returned by the hash function $H()$: $h = H(k)$.
 - **Bucket**: the array element used to store an entry of the dictionary.
 - **Collision**: A collision occurs when 2 different keys k_1 and k_2 have the same hash value.
 $h_1 \neq h_2$ but $H(k_1) = H(k_2)$.
- If there are n entries in a hash table of size M , how likely is it that 2 entries hash into the same bucket?

$$\begin{aligned} P(\text{all } n \text{ entries use different buckets}) &= \frac{M(M-1) \cdots (M-n+1)}{M^n} \\ &= \frac{M!}{M^n(M-n)!} \\ P(2 \text{ entries use the same bucket}) &= 1 - \frac{M!}{M^n(M-n)!} \end{aligned}$$

- There are 2 techniques to handle collision in hashing:
 - Closed addressing (a.k.a. **Separable Chaining**):
 - Entries are always stored in their hash bucket.
 - Each bucket of the hash table is organized as a linked list.
 - Open addressing:
 - Entries are stored in a different bucket than their hash buckets.
 - A rehash algorithm is used to find an empty bucket.

Closed Addressing (Separate Chaining)

- Previously, we used the `Entry<K,V>` class in the `ArrayMapo<K,V>` implementation to store the dictionary entries.
 - In order to support separate chaining, the `Entry<K,V>` class must be modified to support a linked list.

```

public class HashTableSC<K,V> implements Dictionary<K,V> {
    /* ----- Nested Entry class ----- */
    private class Entry<K,V> {
        private K key; // key
        private V value; // value
        private Entry<K,V> next; // link to create a linked list

        public Entry(K k, V V) { // constructor
            key = k;
            value = v;
        }
        /** Accessor method fvor the key */
        public K getKey() {
            return key;
        }
        /** Accessor method for the value */
        public V getValue() {
            return value;
        }
        /** Mutator method for the value */
        public void setValue(V v) {
            this.value = v;
        }
        @Override
        public String toString() {
            return "(" + key + ", " + value + ")";
        }
    }
    /* ----- End of nested Entry class ----- */

    public Entry<K,V>[] bucket; // The hash table
    public int capacity; // capacity = bucket.length
    int NItems; // number of entries in the hash table

    // MAD formula: (Math.abs(a * hashCode + b) % p) % M
    public int MAD_p; // prime number in the MAD alg
    public int MAD_a; // multiplier in the MAD alg
    public int MAD_b; // offset in the MAD alg

    public HashTableSC(int M) { // create a hash table of size M
        bucket = (Entry[]) new Entry[M]; // create hash table of size M
        capacity = bucket.length; // capacity of has table
        NItems = 0; // number of entries in the hash table
    }
}

```



```

    // Initialize MAD parameters
    MAD_p = 109345121; // prime number
    MAD_a = 123; // multiplier
    MAD_b = 456; // offset
}

/** Hash function H(k) */
public int hashCode(K key) {
    int x = key.hashCode(); // hash code of the key
    return (Math.abs(x * MAD_a + MAD_b) % MAD_p) % capacity;
}

/* -----
The help method findEntry(k): find the Entry containing key in the hash table
return: Entry object containing key if found
return: null if not found
----- */
public Entry findEntry(K k) {
    int hashIdx = hashCode(k); // get hash index using key k
    Entry<K,V> curr = bucket[hashIdx]; // curr = first of linked list

    while (curr != null) {
        if (curr.getKey().equals(k)) {
            return curr;
        }
        curr = curr.next;
    }
    return null; // not found
}

@Override
public int size() {
    return NItems;
}

@Override
public void put(K k, V v) {
    int hashIdx = hashCode(k);
    Entry<K,V> h = findEntry(k);
    if (h != null) {
        h.setValue(v); // update value with v
    } else {

```

```

        // Add newEntry as first element in the list at bucket[hashIdx]
        Entry<K,V> newEntry = new Entry<>(k, v); // make new entry
        newEntry.next = bucket[hashIdx]; // point to the first bucket
        bucket[hashIdx] = newEntry; // make newEntry the first bucket
        NItems++; // increment number of entries
    }
}

@Override
public V get(K k) {
    Entry<K, V> h = findEntry(k);
    if (h != null) {
        return h.getValue();
    } else {
        return null;
    }
}

@Override
public V remove(K k) {
    int hashIdx = hashCode(k);
    // General case delete from linked list
    Entry<K,V> previous = bucket[hashIdx];
    Entry<K,V> current = bucket[hashIdx];

    while (current != null && !current.getValue().equals(k)) {
        previous = current;
        current = current.next;
    }

    if (current != null) { // found
        previous.next = current.next; // unlink current
        NItems--; // decrement number of entries
        return current.getValue();
    }
    return null; // not found
}
}

```

- Runtime analysis:

- Consider a hash table using separate chaining. Due to randomization of the hash value,
 - Some entries in the hash table has no keys

- Some entries in the hash table has exactly 1 key.
- Some entries in the hash table has more than 1 key.
- Operations on a hash table always uses the hash value. The hash value will select one specific hash bucket.
 - The search key will be:
 - Found in this hash bucket, or
 - Not found in this hash bucket.
- Therefore, operations on a hash table will always examine all keys in one search bucket.
- Therefore, the running time of operations on a hash table is equal to the number of entries stored inside one bucket in the hash table.
 - Problem: how many entries will be stored inside 1 bucket?
 - Fact: A search key that has hash value k is stored in the bucket k .
 - Therefore, number of entries in bucket k is the number of keys where $H(\text{key}) = k$.
 - Now, let's estimate the number of entries stored in a bucket.
 - By the uniformity assumption, the random hash value $H(\text{key})$ is uniformly distributed over the range $[0, M-1]$. Then, each outcome is equally likely with probability of $1/M$.
 - Suppose there are a total of n items/entries hashed and stored in the hash table. According to the theory of probability, the number of items/entries in any bucket has a binomial probability distribution of $\text{BIN}\left(n, p = \frac{1}{M}\right)$.
 - Then, the average number of entries in 1 bucket is $\frac{n}{M}$. So, the average running time for hash operations is $\frac{n}{M} \sim \mathcal{O}(n)$.

Open Addressing

- Closed addressing vs Open addressing:
 - Closed addressing:
 - In closed addressing, each key is always stored in the hash bucket where the key is hashed to.
 - Close addressing must use some data structure (e.g. linked list) to store multiple entries in the same bucket.
 - Open addressing:
 - In open addressing, each hash bucket will store at most one hash table entry.
 - In open addressing, a key may be stored in different bucket than where they key was hashed to.
 - Entries used in open addressing:
 - Since in open addressing, each hash bucket will store at most one hash table entry, the entries stored in open address do not have a link variable.

- Therefore, the `Entry<K,V>` class used in open addressing is different from the `Entry<K,V>` class used in closed addressing. In fact, we can use the `Entry<K,V>` defined the `ArrayMap<K,V>` implementation.
- Collision resolution in Open Addressing:
 - If a key is hashed to a bucket that is already occupied, we need to find another bucket to store the key. This process will be completed with an insert algorithm.
 - The insert algorithm will start at the hash index and find the next variable hash bucket that can be used to store the key.
 - The procedure to find the next available hash bucket is called **rehashing**.
 - Note: rehashing is not random but deterministic (=computable).
- Commonly used Rehashing Algorithms to Resolve Collision in Open Addressing:
 - Linear Probing: in linear probing, the hash table is searched sequentially starting from the hash index value.
 - In other words, the rehash function is $\text{Rehash}(\text{key}) = (h + i) \% M$, where $h = H(\text{key})$ and $i = 1, 2, \dots$
 - Quadratic Probing: uses the following rehash function: $\text{Rehash}(\text{key}) = (h + i^2) \% M$, where $h = H(\text{key})$ and $i = 1, 2, \dots$
 - Double hashing: uses the following rehash function: $\text{Rehash}(\text{key}) = (h + i * H2(\text{key})) \% M$, where $h = H(\text{key})$, $h' = H'(\text{key})$ is a second hash function, and $i = 1, 2, \dots$
- The code for linear probing without `remove()` :

```

public class HashTableLP<K,V> {
    /* ----- Nested Entry class ----- */
    private class Entry<K,V> {
        private K key;    // The key (to loop up)
        private V value; // The value (corresponding to the key)
        public Entry(K k, V v) { // Constructor
            key = k;
            value = v;
        }
        public K getKey() { // Accessor method for the key
            return key;
        }
        public V getValue() { // Accessor method for the value
            return value;
        }
        public void setValue(V value) { // Mutator method for the value
            this.value = value;
        }
        public String toString() {
            return "(" + key + "," + value + ")";
        }
    }
    /* ----- End of nested Entry class ----- */

    public Entry<K,V>[] bucket; // The Hash table
    public int capacity;        // capacity == bucket.length
    int NItems;                 // # items in hash table
    // MAD formula: ( Math.abs(a * hashCode + b) % p ) % M
    public int MAD_p;           // Prime number in the Multiply Add Divide alg
    public int MAD_a;           // Multiplier in the Multiply Add Divide alg
    public int MAD_b;           // Offset in the Multiply Add Divide alg

    // Constructor
    public HashTableLP(int M) { // Create a hash table of size M
        bucket = (Entry[]) new Entry[M]; // Create a hash table of size M
        capacity = bucket.length;        // Capacity of this hash table
        NItems = 0;                       // # items in hash table

        MAD_p = 109345121;                // We pick this prime number...
        MAD_a = 123;                      // a = non-zero random number
        MAD_b = 456;                      // b = random number
    }
}

```

```

// The hash function for the hash table
public int hashCode(K key) {
    int x = key.hashCode(); // Uses Object.hashCode()
    return ((Math.abs(x*MAD_a + MAD_b) % MAD_p) % capacity);
}

public int size() {
    return NItems;
}

public void put(K k, V v) {
    int hashIdx = hashCode(k); // find the hash index for key k
    int i = hashIdx;
    do {
        if (bucket[i] == null) { // is entry empty?
            bucket[i] = new Entry<K,V>(k, v);
            return;
        } else if (bucket[i].getKey().equals(k)) { // is entry k?
            bucket[i].setValue(v); // update value
            return;
        }
        i = (i + 1) % capacity; // rehash
    } while (i != hashIdx); // all entries searched!
    System.out.println("Full");
}

public V get(K k) {
    int hashIdx = hashCode(k); // find the hash index for key k
    int i = hashIdx;
    do {
        if (bucket[i] == null) { // is entry empty?
            return null;
        } else if (bucket[i].getKey().equals(k)) { // is entry k?
            return bucket[i].getValue(); // return value
        }
        i = (i + 1) % capacity; // rehash
    } while (i != hashIdx); // all entries searched!
    return null; // not found
}
}

```

- Now, let's consider the `remove()` method. If we remove the entry stored in `bucket[i]`, then we will not be able to find the entry stored in `bucket[i+1]`.
 - Therefore, we need to move the entry stored in `bucket[i+1]` to `bucket[i]`.

- However, if we move the entry stored in `bucket[i+1]` to `bucket[i]` , then we will not be able to find the entry stored in `bucket[i+2]` .
- That means, instead of simply moving the entry stored in `bucket[i+1]` to `bucket[i]` , we need alternative method to solve this problem.
- To solve the deletion problem, a hash table using open addressing uses a special entry called AVAILABLE :

```
public Entry<K,V> AVAILABLE = new Entry<>(null, null);
```

- When an existing entry in the hash table is removed, the entry is replaced by the AVAILABLE entry.
- When we are searching for key `k` , then
 - AVAILABLE must be treated as an empty bucket (i.e., it does not contain any key).
 - The rehash algorithm must continue with the next search location.

```

public class HashTableLP<K,V> implements Dictionary<K,V> {
    /* ----- Nested Entry class ----- */
    private class Entry<K,V> {
        private K key;    // The key (to loop up)
        private V value; // The value (corresponding to the key)
        public Entry(K k, V v) { // Constructor
            key = k;
            value = v;
        }
        public K getKey() { // Accessor method for the key
            return key;
        }
        public V getValue() { // Accessor method for the value
            return value;
        }
        public void setValue(V value) { // Mutator method for the value
            this.value = value;
        }
        public String toString() {
            return "(" + key + "," + value + ")";
        }
    }
    /* ----- End of nested Entry class ----- */

    public Entry<K,V>[] bucket; // The Hash table
    public int capacity;        // capacity == bucket.length
    int NItems;                 // # items in hash table
    // MAD formula: ( Math.abs(a * hashCode + b) % p ) % M
    public int MAD_p;           // Prime number in the Multiply Add Divide alg
    public int MAD_a;           // Multiplier in the Multiply Add Divide alg
    public int MAD_b;           // Offset in the Multiply Add Divide alg

    public Entry<K,V> AVAILABLE = new Entry<>(null, null); // special entry for remove(

    // Constructor
    public HashTableLP(int M) { // Create a hash table of size M
        bucket = (Entry[]) new Entry[M]; // Create a hash table of size M
        capacity = bucket.length;        // Capacity of this hash table
        NItems = 0;                       // # items in hash table

        MAD_p = 109345121;                // We pick this prime number...
        MAD_a = 123;                       // a = non-zero random number
        MAD_b = 456;                       // b = random number
    }
}

```



```

}

// The hash function for the hash table
public int hashCode(K key) {
    int x = key.hashCode(); // Uses Object.hashCode()
    return ((Math.abs(x*MAD_a + MAD_b) % MAD_p) % capacity);
}

@Override
public int size() {
    return NItems;
}

@Override
public void put(K k, V v) {
    int hashIdx = hashCode(k); // find the hash index for key k
    int i = hashIdx;
    int firstAvail = -1; // -1 means: no AVAILABLE entry found

    do { // search for key k
        if (bucket[i] == null) { // is entry empty?
            if (firstAvail == -1 ) { // No AVAILABLE entry found
                bucket[i] = new Entry<K,V>(k, v);
                // insert (k,v) in this empty bucket
            } else { // AVAILABLE entry found
                bucket[firstAvail] = new Entry<K,V>(k, v);
                // insert (k,v) in the first AVAILABLE bucket
            }
            return;
        } else if (bucket[i] == AVAILABLE) {
            if (firstAvail == -1) {
                firstAvail = i; // remember the first AVAILABLE entry
            }
        } else if (entry[i].getKey().equals(k)) { // is entry k?
            entry[i].setValue(v); // update value
            return;
        }
        i = (i + 1) % capacity; // rehash
    } while (i != hashIdx); // all entires searched!

    if (firstAvail == -1) {
        System.out.println("Full");
    } else {

```

```

        bucket[fiatAvail] = new Entry<>(k,v);
    }
}

```

```

@Override
public V get(K k) {
    int hashIdx = hashValue(k); // find the hash index for key k
    int i = hashIdx;
    do {
        if (bucket[i] == null) { // is entry empty?
            return null;
        } else if (bucket[i] == AVAILABLE) {
            // Do NOT Test bucket[i]
            // continue
        } else if (entry[i].getKey().equals(k)) { // is entry k?
            return entry[i].getValue(); // return value
        }
        i = (i + 1) % capacity; // rehash
    } while (i != hashIdx); // all entires searched!
    return null; // not found
}

```

```

@Override
public V remove(K k) {
    int hashIdx = hashValue(k);
    int i = hashIdx;

    do {
        if (bucket[i] == null) { // Is bucket empty?
            return null; // Not found
        } else if (bucket[i] == AVAILABLE) {
            // Do NOT Test bucket[i]
            // continue
        } else if (bucket[i].getKey().equals(k)) { // does bucket contain k?
            V retVal = bucekt[i].getValue();
            bucket[i] = AVAILABLE; // mark as deleted
            return retVal;
        }
        i = (i + 1) % capacity; // rehash
    } while (i != hashIdx); // all entires searched!
    return null; // not found
}
}

```

- Clustering in Learning Hashing:

- Suppose the hash table currently stores the entries as follows:

	0	1	2	3	4	5	6	7	8	9
entry[] =		A	B	C	D	E	F	G	H	

- Then, if we want to insert a key k with a hash value in the range $[1 \dots 9]$ we will have to store it in the bucket 9.
- This is called **clustering**.
- To alleviate clustering, other rehashing methods can be used:
 - Quadratic Probing
 - Double hashing.

Running Time Analysis

- Strength and Weakness of a Hash Table
 - A hash table is fast when entries are not clustered.
 - In this case, the running time of operations such as `get()`, `put()` and `remove()` is $\mathcal{O}(1)$.
 - The search will find the key immediately in the hash bucket.
 - Or else, the search will terminate in the next step because it finds an empty (`null`) bucket.
 - A hash table is slower when entries are clustered. In those cases, we need more comparison operations.
- Worse case running time of hashing with linear probing: when the hash table is full.
 - Then, `get()`, `put()`, and `remove()` may need to scan the entire hash table to find the entry.
 - Therefore, worse case running time of linear probing is $n/2$: The scan will examine approximately half of all the entries.
- Average case running time analysis of linear probing:
 - Consider the `get()` algorithm using linear probing. The `get()` method will return when it find
 - an empty bucket, or
 - the key k
 - Consider the `put()` algorithm using linear probing. The `put()` method will return when it find
 - an empty bucket, or
 - the key k
 - Consider the `remove()` algorithm using linear probing. The `remove()` method will return when it find

- an empty bucket, or
 - the key k
- Simplifying assumption: to keep the running time analysis simple, we will assume that there are no AVAILABLE entries in the hash table.
- From the observation of `get()`, `put()`, and `remove()` algorithms:
 - The running time of them depends on the number of entries we need to check in order to find the key k or an empty bucket.
 - So, the worst case running time is when the search ends by finding an empty bucket (takes longer time).
 - Therefore, average running time of `get()`, `put()`, and `remove()` = average number of compare operations to find an empty bucket.
- Load factor and the probability of finding an empty bucket.

Load factor: a.k.a. **occupancy level** is defined as

$$\alpha = \frac{\text{number of entries in hash table}}{\text{size of the hash table}} = \frac{n}{M}.$$

- The load factor α is a measure of how full the hash table is.
- Then, the probability (=likelihood) that a hash bucket is occupied is

$$\mathbf{P}(\text{bucket } i \text{ is occupied}) = \frac{\text{number of entries in the hash table}}{\text{total number of buckets in the hash table}} = \alpha.$$
- So, the probability (=likelihood) that a hash bucket is empty is $\mathbf{P}(\text{bucket } i \text{ is empty}) = 1 - \alpha.$
- The average running time of `get()`, `put()`, and `remove()` is found by computing:
 - How often (frequent) do we need to check 1 entry to find an empty slot ($=f_1$)? How many operations did we perform in this case? ($=c_1$)
 - How often (frequent) do we need to check 2 entry to find an empty slot ($=f_2$)? How many operations did we perform in this case? ($=c_2$)
 - ...
 - The average running time of `get()`, `put()`, and `remove()` is equal to

$$\text{Average running time} = f_1c_1 + f_2c_2 + f_3c_3 + \dots$$

- How often do we need to check 1 entry to find an empty slot?
 - The probability of finding a bucket to be empty $= 1 - \alpha.$
 - We check 1 entry (=the hash bucket) and find an empty bucket.

$$\begin{aligned} \mathbf{P}(\text{check 1 bucket to find an empty bucket}) &= 1 - \alpha = f_1 \\ \text{number of check operations performed} &= 1 = c_1 \end{aligned}$$

- Similarly, in the case of checking 2 entries to find an empty bucket, we have:

$$\begin{aligned} \mathbf{P}(\text{check 2 bucket to find an empty bucket}) &= \alpha(1 - \alpha) = f_2 \\ \text{number of check operations performed} &= 2 = c_2 \end{aligned}$$

- So, we know the average running time of `get()`, `put()`, and `remove()` is equal to

$$\begin{aligned} \text{Average running time} &= f_1 c_1 + f_2 c_2 + \dots + f_n c_n \\ &= (1 - \alpha) \cdot 1 + \alpha(1 - \alpha) \cdot 2 + \alpha^2(1 - \alpha) \cdot 3 + \dots \\ &= (1 - \alpha)[1 + 2\alpha^1 + 3\alpha^2 + 4\alpha^3 + \dots] \end{aligned}$$

- Suppose

$$S = 1 + 2\alpha^1 + 3\alpha^2 + 4\alpha^3 + \dots$$

To compute the sum, we used MATLAB

```
syms a k
assume(a > 0 & a < 1)
symsum((k+1)*(a^k), k, 0, inf)

>>> ans =
1/(a - 1)^2
```

- So, the average running time of `get()`, `put()`, and `remove()` is equal to

$$(1 - \alpha) \cdot \frac{1}{(1 - \alpha)^2} = \frac{1}{1 - \alpha}.$$

- Summary:

- α = the load factor or occupancy level.
- The probability (=likelihood) of finding a bucket to be empty = $1 - \alpha$.
- The average runtime of `get()`, `put()`, and `remove()` is the average number of compare operations performed to find an empty bucket. This quantity is equal to $\frac{1}{1 - \alpha}$.
- Example: If $\alpha = 10\%$, then (because 90% of the time we find an empty bucket), average number buckets searched is $1/(1-0.1) = 1/0.9 \approx 1.1$.