# **ALGORITHMS AND DATA STRUCTURES**

# HASH TABLES

#### IMPLEMENTATION OF ASSOCIATIVE CONTAINERS

- What data structure is more suitable to implement a version of these containers?
  - > std::set<data\_type>
  - > std::map<key\_type, value\_type>

#### IMPLEMENTATION OF ASSOCIATIVE CONTAINERS

- What data structure is more suitable to implement a version of these containers?
  - > std::set<data\_type>
  - > std::map<key\_type, value\_type>
- What are the best complexities we can aspire to?
  - Quadratic, linear, linearithmic, logarithmic, constant...

#### ARRAY-BASED IMPLEMENTATION OF MAP

- Let us say you want to implement a version of std::map<key\_type, value\_type> using an array
  - How to go about implementing a map<string, string>, for instance?
  - Basic operations
    - find, insert, and remove

```
template <typename KeyType, typename ValueType>
class Map {
public:
    Map();
    ~Map();
    int size();
    bool isEmpty();
    void clear();
    void put(KeyType key, ValueType value);
    ValueType get(KeyType key);
    bool containsKey(KeyType key);
private:
    // private stuff
};
```

```
template <typename KeyType, typename ValueType>
class Map {
public:
    // public stuff
private:
    struct KeyValuePair {
        KeyType key;
        ValueType value;
    };
    KeyValuePair *array;
    int capacity;
    int count;
    void expandCapacity();
    int findKey(KeyType key);
};
```

```
template <typename KeyType, typename ValueType>
void Map<KeyType,ValueType>::put(KeyType key,
                                 ValueType value) {
    int index = findKey(key);
    if (index == -1) {
        if (count == capacity) expandCapacity();
        index = count++;
        array[index] key = key;
    }
    array[index].value = value;
}
template <typename KeyType, typename ValueType>
ValueType Map<KeyType,ValueType>::get(KeyType key) {
    int index = findKey(key);
    if (index == -1) error("get: No value for key");
    return array[index].value;
```

```
template <typename KeyType, typename ValueType>
bool Map<KeyType, ValueType>::containsKey(KeyType key) {
    return findKey(key) !=-1;
// all relevant methods depend on this method
template <typename KeyType, typename ValueType>
int Map<KeyType,ValueType>::findKey(KeyType key) {
    for (int i = 0; i < count; i++) {
        if (array[i].key == key) return i;
    }
    return -1;
```

#### ARRAY-BASED IMPLEMENTATION OF MAP

- All of the most relevant methods rely on findkey()
  - findkey(): linear search method, which makes
    - $\triangleright$  get, put, and containsKey all O(N)
  - binary search could improve get and contains Key to  $O(\log N)$ , but not put that is still O(N), why?
- We need to propose a more efficient strategy
  - one way to do that is with lookup tables...

- are programming structures to obtain a desired value in a quick lookup, <u>quick?</u>
  - by computing the appropriate index in a table
  - are typically an array or a grid
- replaces runtime computation with a simpler array indexing operation
- Example: thumb-tabs in a dictionary or simple lookup array

- **Example**: Translate two-letter codes into city names
  - What data structure to use?
    - ▶ Take advantage of the two-letter structure of the key

AK Alaska	HI Hawaii	ME Maine	NJ New Jersey	SD South Dakota
AL Alabama	IA Iowa	MI Michigan	NM New Mexico	TN Tennessee
AR Arkansas	ID Idaho	MN Minnesota	NV Nevada	TX Texas
AZ Arizona	IL Illinois	MO Missouri	NY New York	UT Utah
CA California	IN Indiana	MS Mississippi	OH Ohio	VA Virginia
CO Colorado	KS Kansas	MT Montana	OK Oklahoma	VT Vermont
CT Connecticut	KY Kentucky	NC North Carolina	OR Oregon	WA Washington
DE Delaware	LA Louisiana	ND North Dakota	PA Pennsylvania	WI Wisconsin
FL Florida	MA Massachusetts	NE Nebraska	RI Rhode Island	WV West Virginia
GA Georgia	MD Maryland	NH New Hampshire	SC South Carolina	WY Wyoming

Roberts, Eric. (2013). Programming Abstractions in C++. Pearson.

	A	В	С	D	E	F	G	Н	I	
A										0
В										1
С	California									2
D					Delaware					3
E										4
F										5
G	Georgia									6
Н									Hawaii	7
I	Iowa			Idaho						8
J										9
K										10
L	Louisiana									11
M	Massachusetts			Maryland	Maine				Michigan	12
N			North Carolina	North Dakota	Nebraska			New Hampshire		13
0								Ohio		14
P	Pennsylvania									15
Q										16
R									Rhode Island	17
s			South Carolina	South Dakota						18
T										19
U										20
V	Virginia									21
W	Washington								Wisconsin	22
X										23
Y						Roberts, Eric	c. (2013). Progra	mming Abstraction	ns in C++. Pearso	1 0n. 4
Z							(2020):210914	2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2 2	and a carso	25
	0	1	2	3	4	5	6	7	8	

- We could use map<string, string>
  - Array-based implementation?
  - Try a lookup table!
    - consider the two-letter code 'XY' as coordinates (X,Y)
       in a two-dimensional array that contains city names
    - use X and Y as indices in a two-dimensional grid
    - remember to convert letters (chars) to integers

- Works the same as looking up in an array
  - Simple arithmetics and then looking up
  - It then runs in constant time 0(1)
- How to generalize beyond two-letter codes?

How to improve this piece of code?

How to generalize beyond two-letter codes? <a href="Hashing!">Hashing!</a>

# HASH TABLES

- are abstract data structures that allows to perform
  - find
  - insert
  - remove

operations in 0(1) average time

- are associative arrays that maps keys to values
- can be used to efficiently implement a map

#### **HASHING**

- is a technique used for performing insertions, deletions, and searches in *constant average time*
- is a computational strategy that operates as follows
  - 1. Take a function (hash function) to transform a key into an integer (hash code [of that key])
  - 2. Use the hash code as a starting point as you search for a matching key in the table.
- A hash table is an implementation of map that uses hashing

# HASH TABLES

- > can be considered as a generalization of arrays
- is a data structure of fixed size that contains a collection of items
- can be implemented using vectors, pointers combined with trees, lists, etc.
- come in different flavors
  - chaining hash tables (today)
  - probing hash tables (exercise)

# HASH TABLES: CHAINING

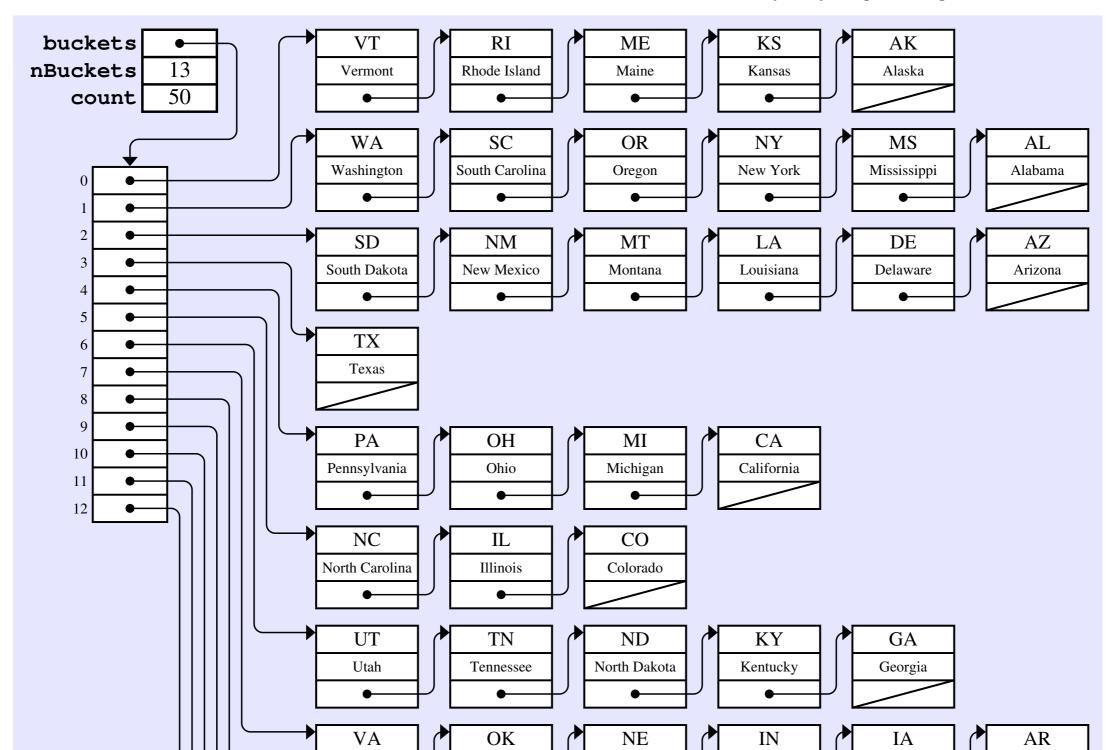
- Each hash code serves as an index into an array of linked lists
  - each list is called a bucket
- The right bucket is found by invoking the hash function to the key

#### int bucket = hashCode(key) % nBuckets;

- bucket is an index to the array of lists (the key hashes to a bucket)
- When two or more different keys hash to the same bucket is called <u>collision</u>

# HASH TABLES: CHAINING

Roberts, Eric. (2013). *Programming Abstractions in C++*. Pearson.



```
template <typename KeyType, typename ValueType>
class HashMap {
   struct Cell {
        KeyType key;
        ValueType value;
        Cell *link;
    };
    Cell **buckets;
    int nBuckets;
    int count;
    Cell *findCell(int bucket, ValueType key) {
        Cell *cp = buckets[bucket];
        while (cp != NULL && key != cp->key)
            cp = cp->link;
        return cp;
```

```
const int INITIAL_BUCKET_COUNT = 101;
template <typename KeyType, typename ValueType>
HashMap<KeyType,ValueType>::HashMap() {
    nBuckets = INITIAL_BUCKET_COUNT;
    buckets = new Cell*[nBuckets];
    for (int i = 0; i < nBuckets; i++)
        buckets[i] = NULL;
    count = 0;
template <typename KeyType, typename ValueType>
HashMap<KeyType,ValueType>::~HashMap() {
    clear();
```

```
template <typename KeyType, typename ValueType>
void HashMap<KeyType,ValueType>::clear() {
    for (int i = 0; i < nBuckets; i++) {
        Cell *cp = buckets[i];
        while (cp != NULL) {
            Cell *oldCell = cp;
            cp = cp->link;
            delete oldCell;
    count = 0;
```

```
template <typename KeyType, typename ValueType>
void HashMap<KeyType,ValueType>::put(KeyType key,
                                      ValueType val) {
    int bucket = hashCode(key) % nBuckets;
    Cell *cp = findCell(bucket, key);
    if (cp == NULL) {
        cp = new Cell;
        cp->key = key;
        cp->link = buckets[bucket];
        buckets[bucket] = cp;
        count++;
    cp->value = val;
```

```
template <typename KeyType, typename ValueType>
ValueType HashMap<KeyType,ValueType>::get(KeyType key)
    int bucket = hashCode(key) % nBuckets;
    Cell *cp = findCell(bucket, key);
    if (cp == NULL) error("get: no value for key");
    return cp->value;
template <typename KeyType, typename ValueType>
bool HashMap<KeyType,ValueType>::containsKey(KeyType key)
    int bucket = hashCode(key) % nBuckets;
    return findCell(bucket, key) != NULL;
```

#### **NUMBER OF BUCKETS**

- As the hash function, also change the probability of collisions
  - if nBuckets is small → more collisions
  - if nBuckets is large → waste resources
- It is key to compare to the number of entries
- Load factor of the hash table:  $\lambda = \frac{N_{\text{keys}}}{N_{\text{buckets}}} = \frac{n}{m}$

time-space tradeoff

# **NUMBER OF BUCKETS**

- A nice sweet spot for  $\lambda \approx 0.7$  leading to O(1) operations
- Most of the times it is not possible to determine the number of entries
  - Empirically obtain an estimate for nBuckets
- Rehashing: Dynamically change nBuckets as λ grows too large (exercise)
  - might impact performance but not too much if done sparsely

#### HASH TABLES: CHAINING

- Theorem 1: In a hash table in which collisions are resolved by chaining, an *unsuccessful* search takes  $\Theta(1 + \lambda)$  averagecase time, under the assumption of simple uniform hashing and constant-time hash code
- Theorem 2: In a hash table in which collisions are resolved by chaining, a *successful* search takes  $\Theta(1 + \lambda)$  averagecase time, under the assumption of simple uniform hashing and constant-time hash code

# HASH TABLES: CHAINING

- Meaning: If  $n = O(m) \rightarrow \alpha = O(1)$ 
  - Searching takes constant time <u>on average</u>
  - Since insertion takes O(1) worst-case time and deletion takes O(1) worst-case time (for doubly LL)
  - ▶ **ALL** dictionary operations in O(1) time <u>on average</u>
- What is the difference with lookup tables, in terms of space and time complexities?

- lies at the heart of the implementation
  - 1. must be computable in constant time (i.e., independent of the number of items in the hash table)
  - 2. must distribute its items uniformly among the array slots or buckets (sample uniform hashing)
- does not have to necessarily take advantage of all the information provided by the key

- Universal hash functions guarantee a uniformly distributed occupation of buckets
- is usually defined as a free function, not part of the HashMap interface
- has to be as unpredictable as possible for a given set of keys

- Every hash function will profoundly affect the efficiency of the implementation
- Although the correctness of the implementation is not affected by the unpredictability of the hash function
- It is often a good idea to choose the size of the table to be a primer number

Unpredictable, less collisions

```
const int HASH SEED = 5381;
const int HASH_MULTIPLIER = 33;
const int HASH MASK = unsigned(-1) >> 1;
int hashCode(string str) {
    unsigned hash = HASH SEED;
    int n = str.length();
    for (int i = 0; i < n; i++)
        hash = HASH_MULTIPLIER * hash + str[i];
    return int(hash & HASH_MASK);
```

This one would produce more collisions

```
int hashCode(string str) {
   int hash = 0;
   int n = str.length();

for (int i = 0; i < n; i++)
   hash += str[i];

return hash;
}</pre>
```

act, cat, tac