

# CSE 373: Data Structures and Algorithms

## Lecture 6: Finishing Amortized Analysis; Dictionaries ADT; Introduction to Hash Tables

Instructor: Lilian de Greef  
Quarter: Summer 2017

Today:

- Finish up Amortized Analysis
- Dictionary ADT
- Introduce Hash Tables

# Reminder: No class on Monday!

Unofficial holiday – have a good 4-day weekend! 😊



- Will ask you to do a ~30 minute activity to make up for last class time
- Remember that homework 2 is also due the day after we're back

# Amortized Analysis

How we calculate the average time!

# Amortized Cost

The **amortized cost** of  $n$  operations is the worst-case total cost of the operations divided by  $n$ .

Shorthand:

If  $T(n)$  = worst-case (upper bound) of total cost  
for  $n$  = number of operations

$$\Rightarrow \text{Amortized Cost} = T(n) / n$$

## Example: Array Stack

What's the amortized cost of calling `push()`  $n$  times if we double the array size when it's full?

$n$  operations:

- $n$  pushes at  $O(1)$  each  $\rightarrow$  total cost =  $n$
- cost of resizing =  $n + n/2 + n/4 + n/8 + \dots \leq 2n \rightarrow$  total cost  $\leq 2n$

$\rightarrow T(n) = n + 2n = 3n$

$\rightarrow$  Amortized cost =  $T(n)/n = 3n/n = 3$

$\rightarrow$  Amortized Running time =  $O(1)$

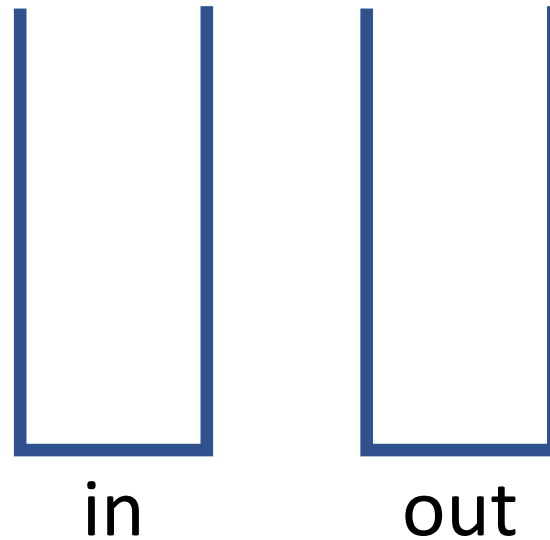
The **amortized cost** of  $n$  operations is the worst-case total cost of the operations divided by  $n$ .

## Example #2: Queue made of Stacks

A sneaky way to implement Queue using two Stacks

Example walk-through:

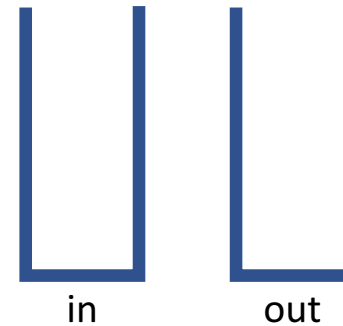
- enqueue A
- enqueue B
- enqueue C
- dequeue
- enqueue D
- enqueue E
- dequeue
- dequeue
- dequeue



## Example #2: Queue made of Stacks

A sneaky way to implement Queue using two Stacks

```
class Queue<E> {  
    Stack<E> in = new Stack<E>();  
    Stack<E> out = new Stack<E>();  
    void enqueue(E x) { in.push(x); }  
    E dequeue() {  
        if(out.isEmpty()) {  
            while(!in.isEmpty()) {  
                out.push(in.pop());  
            }  
        }  
        return out.pop();  
    }  
}
```



Wouldn't it be nice to have a queue of t-shirts to wear instead of a stack (like in your dresser)?

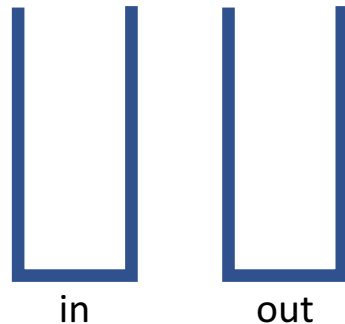
So have two stacks

- *in*: stack of t-shirts go after you wash them
- *out*: stack of t-shirts to wear
- if *out* is empty, reverse *in* into *out*



## Example #2: Queue made of Stacks (Analysis)

```
class Queue<E> {  
    Stack<E> in = new Stack<E>();  
    Stack<E> out = new Stack<E>();  
    void enqueue(E x) { in.push(x); }  
    E dequeue() {  
        if(out.isEmpty()) {  
            while(!in.isEmpty()) {  
                out.push(in.pop());  
            }  
        }  
        return out.pop();  
    }  
}
```



Assume stack operations are (amortized)  $O(1)$ .  
What's the worst-case for `dequeue()`?

What operations did we need to do to reach that condition (starting with an empty Queue)?

Hence, what is the amortized cost?

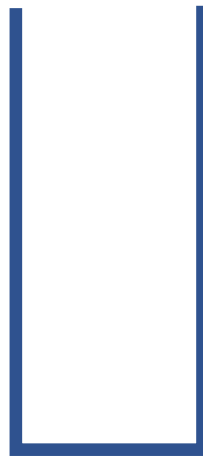
So the average time for `dequeue()` is:

## Example #2: Using “Currency” Analogy

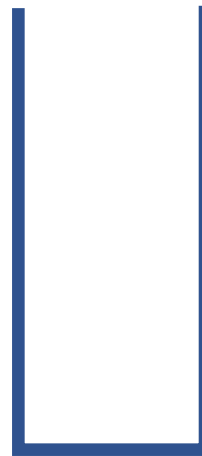
“Spend” \$2 for every `enqueue` – \$1 to the “computer”, \$1 to the “bank”.

Example walk-through:

- enqueue A
- enqueue B
- enqueue C
- enqueue D
- enqueue E
- dequeue



in



out



Potential Function

## Example #3: (Parody / Joke Example)

Lectures are 1 hour long, 3 times per week, so I'm supposed to lecture for 27 hours this quarter.

If I end the first 26 lectures 5 minutes early, then I'd have "saved up" 130 minutes worth of extra lecture time.

Then I could spend it all on the last lecture and can keep you here for 3 hours (bwahahahaha)!

(After all, each lecture would still be 1 hour amortized time)

# Wrapping up Amortized Analysis

- In what cases do we care more about the average / amortized run time?
- In what cases do we care more about the worst-case run time?

# What have we covered so far?

- Abstract Data Types (**ADTs**)
- Two data structures
  - **Stacks** (both using arrays and linked-lists)
  - **Queues** (including **circular queues**)
- Asymptotic Analysis
  - Intuition for **Big-O**
  - Formally proving Big-O using **Inductive Proofs**
  - Calculating Big-O for recursive methods using **Recurrence Relations**
  - Big-O's cousins: **Big-Ω**, **Big-θ**, **little-o**, **little-ω**
  - Average running time using **Asymptotic Analysis**

Whew!

That was a \*lot\* of algorithm analysis.

Now shifting gears completely...  
on to some new data structures!

# Dictionary ADT

# Dictionary ADT



# Uses of Dictionary ADT

Used to store information with some key and retrieve it efficiently – lots of programs do that!

Examples:

- Contacts in a phone (name: number, email)
- Orca/Husky cards (account number: balance)
- Genome maps (DNA sequence: location on genome)
- Lilian's database of your grades (student ID: assignments, grades)
- Networks (router tables), Operating Systems (page tables), Compilers (symbol tables), Databases
- ... and so much more!

Possibly the most widely used ADT!

# Motivating Hash Tables

Creative thinking time: how could you implement a dictionary using what you know so far (namely, linked-lists and arrays)?

e.g. map names (key) to phone numbers (value)

(extra space for notes)

# Motivating Hash Tables

Running times for Dictionary operations with  $n$  (key, value) pairs:

insert

find

delete

“Magic Array”

# “Magic Array”

Use key to compute array index for an item in  $O(1)$  time

Example: phone contacts (name, number)

`name`  $\rightarrow$  `index = computeIndex(name)`  $\rightarrow$  `array[index] = (name, number)`

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# “Magic Array”

Use key to compute array index for an item in  $O(1)$  time

Example: phone contacts (name, number)

`name`  $\rightarrow$  `index = computeIndex(name)`  $\rightarrow$  `array[index] = (name, number)`

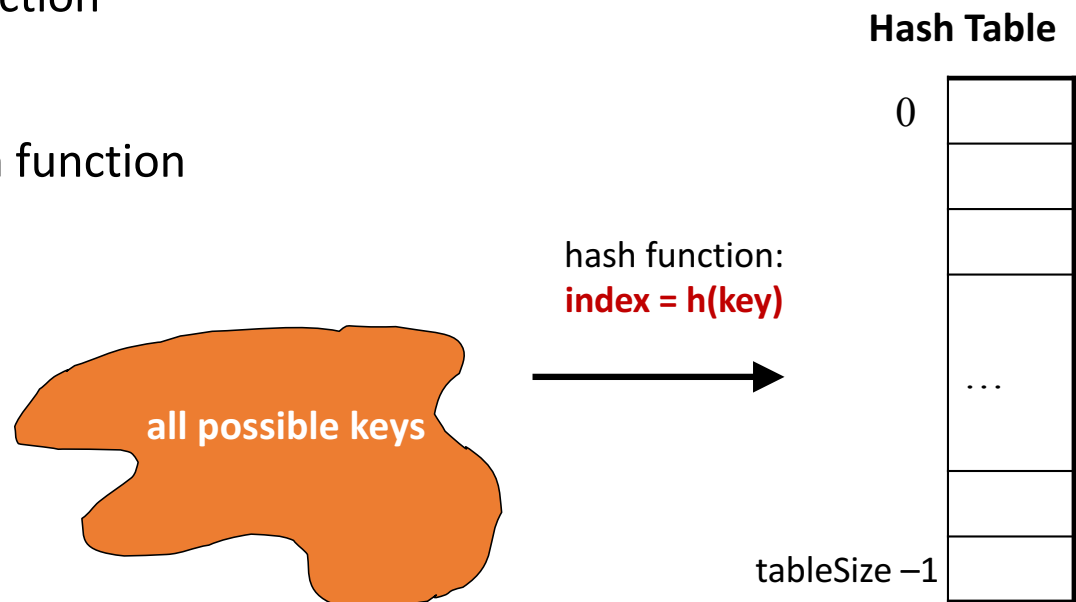
What would be important about the indices from `computeIndex`?

# Introducing... Hash Tables!

Closest thing to our “magic array”

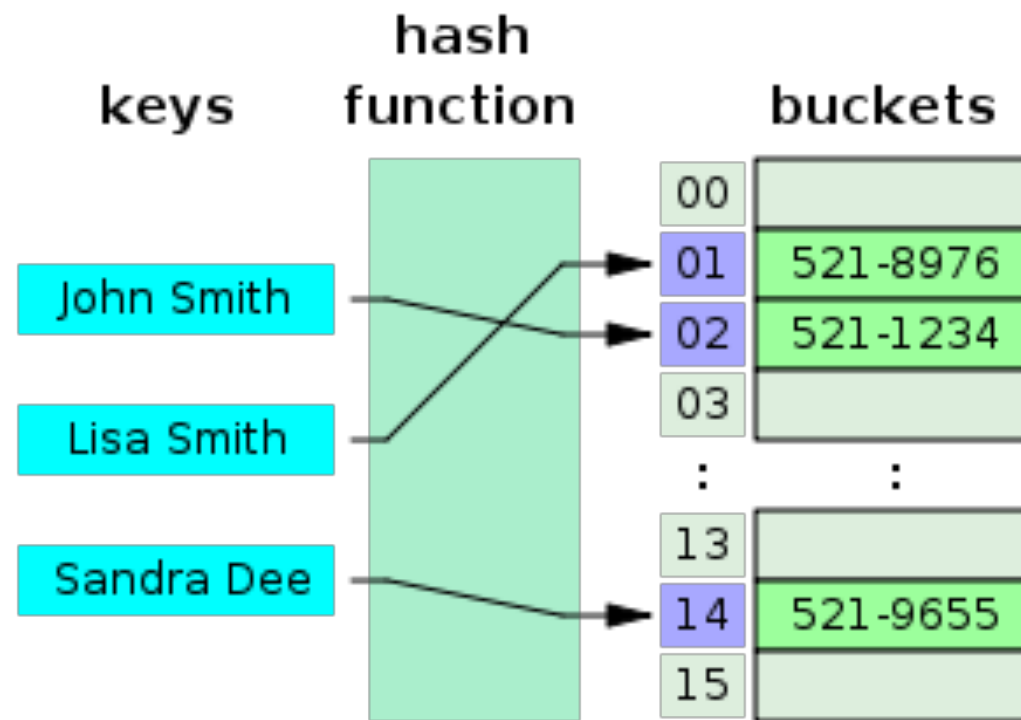
# Hash Tables: closest thing to our “Magic Array”

- *Average case*  $O(1)$  find, insert, and delete  
(when under some often-reasonable *assumptions*)
- Our `computeIndex` function is called a
- The `index` from the hash function is called a





# Hash Tables: Example Illustration



# Hash Functions

Hash functions need to...

- 
- 
- 

For a person's name, would it be a good hash function to...

- Use the ASCII values of first and second letter?
- Use the number of letters in the name?

# Example Hash Function

Hash function “djb2”:

```
unsigned long
hash(unsigned char *str)
{
    unsigned long hash = 5381;
    int c;

    while (c = *str++)
        hash = ((hash << 5) + hash) + c; /* hash * 33 + c */

    return hash;
}
```

# Hash Functions

- Many datatypes and Objects are hashable
- When writing a class, can make it hashable!

Do so by implementing `hashCode` method

- We'll focus on `ints` and `Strings` in this class.

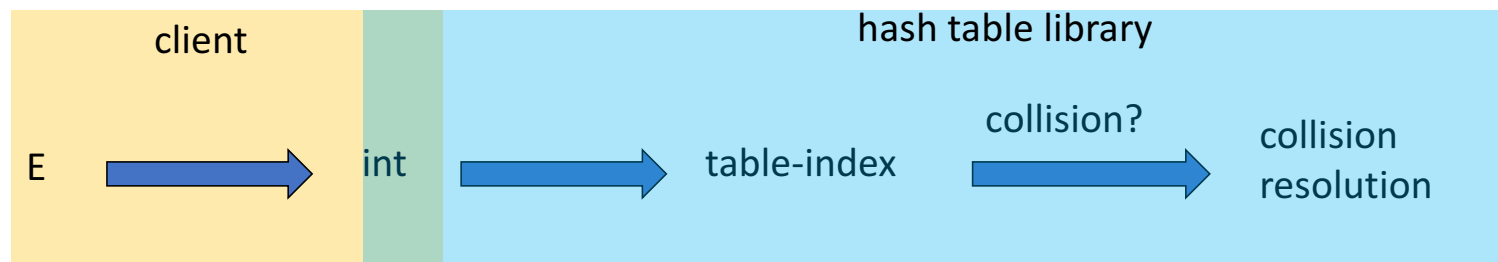
# Collisions

Happens when two elements get the same index (unavoidable in practice)

*Homework: come up with a strategy, write it down on paper, and bring it to class on Weds*

# Hash Table roles

When hash tables are a reusable library, the division of responsibility generally breaks down into two roles:



We will learn both roles, but most programmers “in the real world” spend more time as clients while understanding the library

# Hash Tables

- There are  $m$  possible keys ( $m$  typically large, even infinite)
- We expect our table to have only  $n$  items
- $n$  is much less than  $m$  (often written  $n \ll m$ )

## Many dictionaries have this property

- Compiler: All possible identifiers allowed by the language vs. those used in some file of one program
- Database: All possible student names vs. students enrolled
- AI: All possible chess-board configurations vs. those considered by the current player
- ...

# Hash Table Size

- How can we keep hash values (i.e. the indices) within the table size?

–

- Table size usually prime
  - Real-life data tends to have a pattern
  - "Multiples of 61" probably less likely than "multiples of 60"
  - Helpful for a collision-handling strategy we'll see next week



## Review: Hash Tables thus far...

- The hash table is one of the most important data structures – Supports only `find`, `insert`, and `delete` efficiently
  - Have to search entire table for other operations
- Important to use a good hash function
- Important to keep hash table at a good size
- Side-comment: hash functions have uses beyond hash tables – Examples: Cryptography, check-sums
- Big remaining topic: Handling collision

## Homework

Come up with a collision-resolution strategy, write it down on paper, and bring it to class on Wednesday

Goal: prime our brains for learning the most common collision-resolution strategies.