

# 2

# Fundamental Data Structures

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Sebastian Wild

## Outline

# 2 Fundamental Data Structures

- 2.1 Stacks & Queues
- 2.2 Resizable Arrays
- 2.3 Priority Queues
- 2.4 Binary Search Trees
- 2.5 Summary

## 2.1 Stacks & Queues

# Abstract Data Types

## abstract data type (ADT)

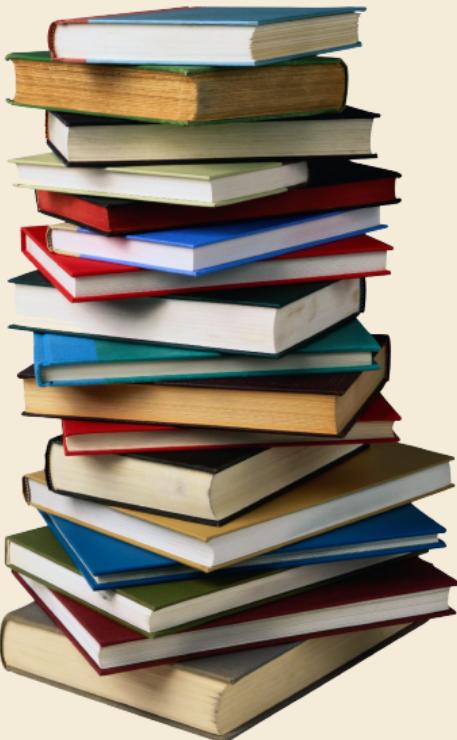
- ▶ list of supported operations
  - ▶ **what** should happen
  - ▶ **not:** how to do it
  - ▶ **not:** how to store data
- ≈ Java interface  
(with Javadoc comments)

*Why separate?*

- ▶ Can swap out implementations
- ~~ reusable code!
- ▶ (Often) better abstractions
- ▶ Prove generic lower bounds ( ~~ Unit 3)



# Stacks



## Stack ADT

- ▶ `top()`  
Return the topmost item on the stack  
Does not modify the stack.
- ▶ `push(x)`  
Add *x* onto the top of the stack.
- ▶ `pop()`  
Remove the topmost item from the stack  
(and return it).
- ▶ `isEmpty()`  
Returns true iff stack is empty.
- ▶ `create()`  
Create and return an new empty stack.

# Linked-list implementation for Stack

## Invariants:

- ▶ maintain top pointer to topmost element
- ▶ each element points to the element below it  
(or null if bottommost)

## Linked stacks:

- ▶ require  $\Theta(n)$  space when  $n$  elements on stack
- ▶ All operations take  $O(1)$  time

# Array-based implementation for Stack

Can we avoid extra space for pointers?

↝ array-based implementation

**Invariants:**

- ▶ maintain array  $S$  of elements, from bottommost to topmost
- ▶ maintain index  $\text{top}$  of position of topmost element in  $S$ .



What to do if stack is full upon pop?

**Array stacks:**

- ▶ require *fixed capacity*  $C$  (known at creation time)!
- ▶ require  $\Theta(C)$  space for a capacity of  $C$  elements
- ▶ all operations take  $O(1)$  time

## 2.2 Resizable Arrays

# Digression – Arrays as ADT

Arrays can also be seen as an ADT!

## Array operations:

- ▶ `create(n)`    *Java: A = new int[*n*];*  
Create a new array with *n* cells, with positions  $0, 1, \dots, n - 1$
  - ▶ `get(i)`    *Java: A[i]*  
Return the content of cell *i*
  - ▶ `set(i, x)`    *Java: A[i] = x;*  
Set the content of cell *i* to *x*.
- ~~ Arrays have fixed size (supplied at creation).

Usually directly implemented by compiler + operating system / virtual machine.



**Difference to others ADTs:** Implementation usually fixed  
to “a contiguous chunk of memory”.

# Doubling trick

*Can we have unbounded stacks based on arrays?* Yes!

## Invariants:

- ▶ maintain array  $S$  of elements, from bottommost to topmost
- ▶ maintain index  $\text{top}$  of position of topmost element in  $S$
- ▶ maintain capacity  $C = S.\text{length}$  so that  $\frac{1}{4}C \leq n \leq C$
- ~~ can always push more elements!

*How to maintain the last invariant?*

- ▶ before push
  - If  $n = C$ , allocate new array of size  $2n$ , copy all elements.
- ▶ after pop
  - If  $n < \frac{1}{4}C$ , allocate new array of size  $2n$ , copy all elements.
- ~~ “*Resizing Arrays*”
  - an implementation technique, not an ADT!

# Amortized Analysis

- ▶ Any individual operation push / pop can be expensive!  
 $\Theta(n)$  time to copy all elements to new array.
- ▶ **But:** An one expensive operation of cost  $T$  means  $\Omega(T)$  next operations are cheap!

Formally: consider “credits/potential”  $\Phi = \min\{n - \frac{1}{4}C, C - n\} \in [0, \frac{1}{2}n]$

- ▶ amortized cost of an operation = actual cost  $- 2 \cdot$  change in  $\Phi$ 
  - ▶ cheap push/pop: actual cost 1 array access, consumes  $\leq 1$  credits  $\rightsquigarrow$  amortized cost  $\leq 3$
  - ▶ copying push: actual cost  $2n + 1$  array accesses, creates  $n$  credits  $\rightsquigarrow$  amortized cost 1
  - ▶ copying pop: actual cost  $n + 1$  array accesses, creates  $\frac{1}{2}n$  credits  $\rightsquigarrow$  amortized cost 1
- ~ sequence of  $m$  operations: total actual cost  $\leq$  total amortized cost + initial credits
  - here:  $\leq 3n + \leq 2 \cdot \frac{1}{2}n = 4n$

# Queues

## Operations:

- ▶ enqueue( $x$ )

Add  $x$  at the end of the queue.

- ▶ dequeue()

Remove item at the front of the queue and return it.



Implementations similar to stacks.

# Bags

*What do Stack and Queue have in common?*

They are special cases of a **Bag**!

**Operations:**

- ▶ `insert(x)`  
Add *x* to the items in the bag.
- ▶ `delAny()`  
Remove any one item from the bag and return it.  
(Not specified which; any choice is fine.)
- ▶ roughly similar to Java's Collection



Sometimes it is useful to state that order is irrelevant  $\rightsquigarrow$  Bag  
Implementation of Bag usually just a Stack or a Queue

## 2.3 Priority Queues

# Priority Queue ADT

Now: elements in the bag have different *priorities*.

(Max-oriented) Priority Queue (MaxPQ):

- ▶ `insert( $x, p$ )`

Insert item  $x$  with priority  $p$  into PQ.

- ▶ `max()`

Return item with largest priority.

(Does not modify the PQ.)

- ▶ `delMax()`

Remove the item with largest priority and return it.

- ▶ `changeKey( $x, p'$ )`

Update  $x$ 's priority to  $p'$ .

Sometimes restricted to *increasing* priority.

- ▶ `isEmpty()`

Fundamental building block in many applications.



# PQ implementations

## Elementary implementations

- unordered list  $\rightsquigarrow \Theta(1)$  insert, but  $\Theta(n)$  delMax
- sorted list  $\rightsquigarrow \Theta(1)$  delMax, but  $\Theta(n)$  insert

Can we get something between these extremes? Like a “slightly sorted” list?

Yes! *Binary heaps*.

### Array view

Heap = array  $A$  with  
 $\forall i \in [n] : A[\lfloor i/2 \rfloor] \geq A[i]$

  
store nodes  
in level order  
in  $A[1..n]$

### Tree view

Heap = tree that is  
(i) a complete binary tree  
(ii) heap ordered

all but last level full  
last level flush left

father  $\geq$  children

## Binary heap example

# Why heap-shaped trees?

## Why complete binary tree shape?

- ▶ only one possible tree shape  $\rightsquigarrow$  keep it simple!
- ▶ complete binary trees have minimal height among all binary trees
- ▶ simple formulas for moving from a node to parent or children:

For a node at index  $k$  in  $A$

- ▶ parent at  $\lfloor k/2 \rfloor$
- ▶ left child at  $2k$
- ▶ right child at  $2k + 1$

## Why heap ordered?

- ▶ Maximum must be at root!  $\rightsquigarrow \max()$  is trivial!
- ▶ But: Sorted only along paths of the tree; leaves lots of leeway for fast inserts

how? ... stay tuned

# Insert

## Delete Min

# Heap construction

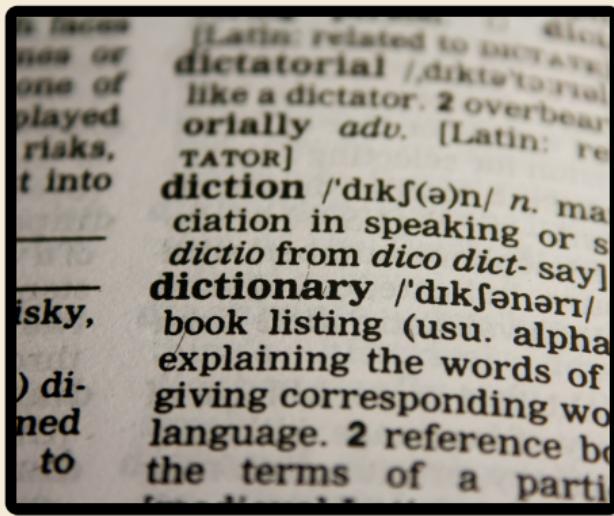
# Analysis

## 2.4 Binary Search Trees

# Symbol table ADT

Java: `java.util.Map<K,V>`

Symbol table / Dictionary / Map / Associative array / key-value store:



- ▶ `put(k, v)`      Python dict: `d[k] = v`  
Put key-value pair (*k*, *v*) into table
- ▶ `get(k)`      Python dict: `d[k]`  
Return value associated with key *k*
- ▶ `delete(k)`  
Remove key *k* (any associated value) from table
- ▶ `contains(k)`  
Returns whether the table has a value for key *k*
- ▶ `isEmpty(), size()`
- ▶ `create()`



*Most fundamental building block in computer science.*

(Every programming library has a symbol table implementation.)

# Symbol tables vs mathematical functions

- ▶ similar interface
- ▶ but: mathematical functions are static (never change their mapping)  
(Different mapping is a *different* function)
- ▶ symbol table = *dynamic* mapping  
Function may change over time

# Elementary implementations

## Unordered (linked) list:

👍 Fast put

👎  $\Theta(n)$  time for get

~~ Too slow to be useful

## Sorted *linked* list:

👎  $\Theta(n)$  time for put

👎  $\Theta(n)$  time for get

~~ Too slow to be useful

~~ *Sorted order does not help us at all?!*

# Binary search

# Binary search trees

Binary search trees (BSTs)  $\approx$  dynamic sorted array

- ▶ binary tree
  - ▶ Each node has left and right child
  - ▶ Either can be empty (`null`)
- ▶ Keys satisfy *search-tree property*

all keys in left subtree  $\leq$  root key  $\leq$  all keys in right subtree

## BST example & find

## BST insert

## BST delete

# Ordered symbol tables

# Augmented BSTs

# Analysis

## Balanced BSTs

## 2.5 Summary

# BSTs vs. Heaps