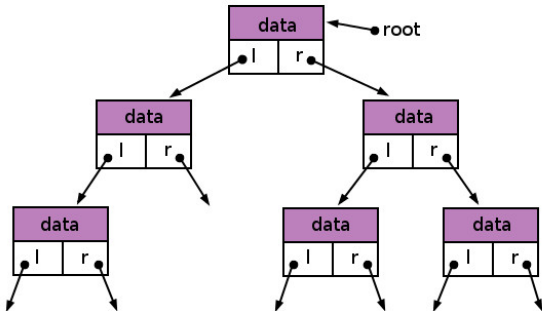


# Dynamic Data Structures

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# Dynamic Data Structures

Having efficient **data structures** is crucial for successful algorithms.

- The problems seen so far involved fixed length lists
- In most languages we have a simple way to implement this efficiently — arrays
- Our algorithms assumed some sort of array type was available

Other problems require **dynamic** data structures such as

- Lists, Stacks and Queues
- Sets and Dictionaries

These are designed to hold variable, essentially unlimited amounts of data.

# Ordered Data Structures

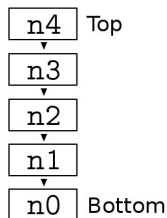
A *list* is an ordered collection of {nodes, items, elements}.

- The key property of a list is the ordering of the nodes
- A list might support operations such as
  - `push` adds an element to the end of the list
  - `pop` removes the last element of the list
  - `shift` removes the first element of the list
  - `unshift` adds an element to the front of the list
  - `insert` adds an element at a given position
  - `remove` removes the element at a given position
  - `iterate` returns the items in order
- Plus sorting, searching, copying, joining, splitting ...
- The most appropriate implementation depends on which operations are needed.

# Stacks

A *stack* is a *last-in first-out* (LIFO) list.

- Stacks support only
  - push* for adding elements
  - pop* for removing elements
- Stacks are usually pictured as a vertical (stacked!) structure



- Stacks support recursive algorithms including fundamental operations such as calling subprocedures and evaluating arithmetic expressions

# Stacks

## Question

How would you implement a stack?

- Must be able to add “unlimited” objects
- Push and Pop must implement LIFO behaviour

# Performance of Push

## Question

Given a stack containing  $N$  objects, what is the worst case time complexity of **push**?

- Assume: time to insert (copy, add) one object to array is  $c$
- Assume: initial capacity is 4

# Performance of Push

## Revised Question

Given an empty stack, what is the worst case time to push  $N$  objects?

- Assume: initial capacity is 4
- Assume: time to insert (copy, add) one object to array is  $c$

# Amortisation

The time for  $N$  pushes is  $O(N)$ , so:

- A single push is *effectively* a constant time operation
- More correctly: push is **amortised**  $\Theta(1)$
- NOT the same as  $\Theta(1)$

## Amortisation

- Related to accountancy method used to defer large costs
- **Amortised analysis** considers a sequence of operations
- Cost of individual ops is “amortised” across the sequence
- Unlike accountancy, must never be in debt



# Amortised Analysis

Rather than calculating cost of full sequence of  $N$  steps we can

- Pick a **representative** subsequence
- Subsequence is some “cycle” that repeats
- Pick an amortised cost for operations
- Show that paying amortised cost covers all costs (never in debt)

## Exercise

Find a representative cycle (subsequence) of pushes into the stack and show that the amortised cost of  $3c$  covers all costs.

# Amortised Analysis

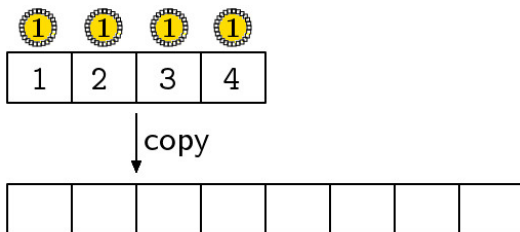
Begin cycle when ...

End cycle when ...

# Amortised Analysis

Argument only works because array is initially empty and size is **doubled**

- Say we have  $N$  objects on stack after a copy
- Before next copy we always push  $N$  more
- This is how cost is covered



- Multiplying by **any factor** will do - will affect amortisation constant

# Queues

A *queue* is also a list, but the next object removed is either:

- The *earliest* one added (FIFO Queue)
- The one with highest *priority* (Priority Queue)

## Questions

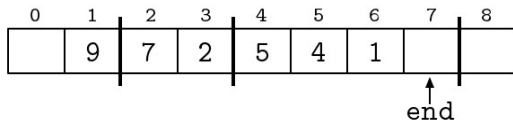
- How could you implement a priority queue (PQ)?
- Given a PQ following your design that contains  $N$  objects, what would be the worst case time to add a new object? (Each object has a key attribute that determines its priority.)

# Priority Queue Design

# Heap: a Tree in an Array

We want to know where the “end” of the tree is:

- Build a tree within an array

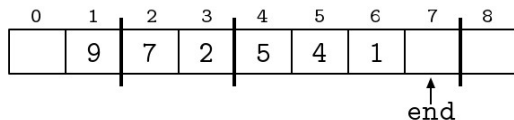


- Track end using “stack pointer”
- Navigate by indices
- Leaving  $a[0]$  blank means:
  - parent of  $a[n]$  is  $a[n/2]$
  - children of  $a[n]$  are  $a[2*n]$  and  $a[2*n+1]$

## Exercise

How should a new object be added to a **max** binary heap? (i.e. the greatest key should be at the root).

# Heap: a Tree in an Array

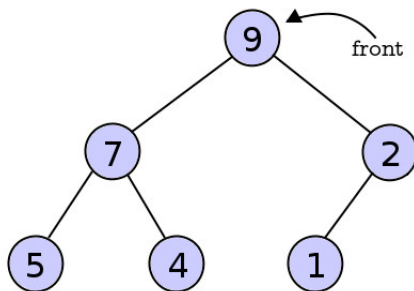


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  - parent of  $a[n]$  is  $a[n/2]$
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## Exercise

How should the object with the greatest key be removed from a **max** binary heap?

# Binary Heap Performance



## Question

Given a heap containing  $N$  objects, what is the time complexity for adding or removing one object?



# Heapsort

Heaps also provide us with the **Heapsort** algorithm (JWJ Williams, 1964)

## Heapsort (given a list $L$ )

- Create an empty heap  $H$
  - Remove each element of  $L$  and add it to  $H$
  - Remove each element of  $H$  and add it to  $L$
  - HALT
- 
- What could be simpler?!
  - Performance is again  $\Theta(N \log_2 N)$
  - Can also be implemented **in place** by setting up list and heap partitions within a single array

# Sets

A *set* is an unordered collection of objects each having a *unique* key.

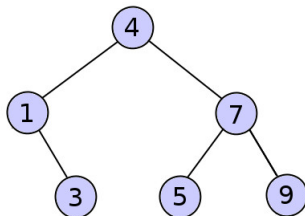
- Should have “unlimited” capacity
- Want to *put* and *get* by key
- A key could be any type that defines  $<$  and  $=$

## Questions

- How could you implement a set?
- Given a set following your design that contains  $N$  objects, what would be the worst case time to get the object with key  $k$ ?

# A Search Tree?

A tree will divide the data but need a different ordering



- Start at the root (it's a tree)
- Go right: find/add larger keys
- Go left: find/add smaller keys

# Binary Search Tree

In a **Binary Search Tree**

- Go right: find/add larger keys
- Go left: find/add smaller keys

## Exercise

- Draw the (integer) binary search tree implied by the following code:

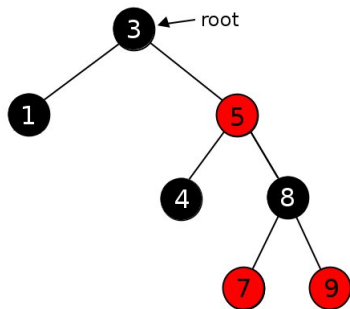
```
bst = new BST
keys = [5,3,10,1,6,9,8,0,4]
for i = 0 to 8
    bst.put(keys[i])
```

- What is the worst case time complexity of the put procedure?

# Red-Black Trees

Red-Black Trees are binary search trees that maintain **balance**

- A BST can become (very) unbalanced, resulting in long branches
- Searching a BST takes  $O(N)$  time in the worst case
- The branches of a balanced tree remain as short as possible



# Red-Black Tree Properties

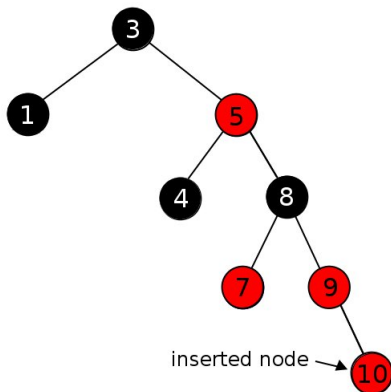
## Definition (Red-Black Tree)

A binary search tree  $T$  is a **red-black tree** iff  $T$  satisfies the following five properties:

- 1 All nodes (including nulls) are either red or black
- 2 The root node is black
- 3 Every leaf (all null) is black
- 4 Both children of a red node are black
- 5 All paths from a node to a descendant leaf contain the same number of black nodes

# Insertion

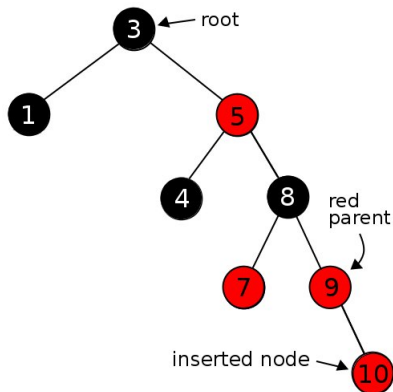
A node is inserted using the ordinary BST procedure



- A new node is always colored red

# Insertion

The insertion may result in a violation of the red-black tree properties

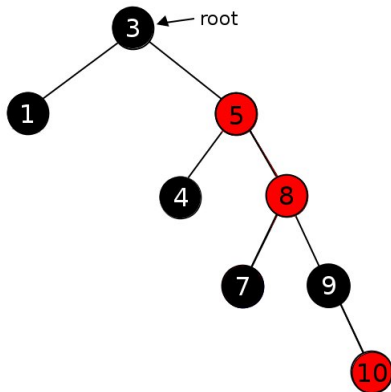


- The root might be coloured red
- A red node might have a red child



# Insertion

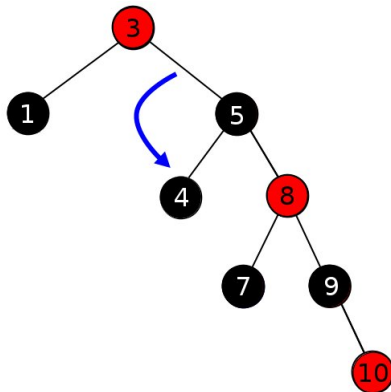
Either recolour  $\Theta(1)$  nodes



- There is still a red node with a red parent
- The problem has moved closer to the root (continue)

# Insertion

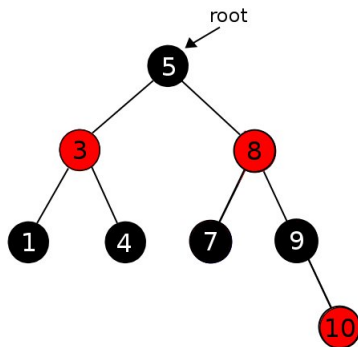
Or perform a **rotation** of  $\Theta(1)$  nodes and **Stop**



- Reduces height of the tree
- Preserves key ordering

# Insertion

The properties are restored



# Performance

By maintaining the red-black tree properties, we have  $h \leq 2\log_2(N + 1)$

- Get procedure is the same as for BST
- Height constraint means it is now  $O(\log_2 N)$

For Put, only the last part is different

- The extra work is still localised to one branch
- So, Put also runs in  $O(\log_2 N)$  time