

## CS217 - Data Structures & Algorithm Analysis (DSAA)

Lecture #8

### ► Elementary Data Structures

Prof. Pietro S. Oliveto

Department of Computer Science and Engineering  
Southern University of Science and Technology (SUSTech)

olivetop@sustech.edu.cn  
<https://faculty.sustech.edu.cn/olivetop>

Reading: Part III Introduction & Chapter 10

### ► Data Structures

- **Dynamic sets** that can store and retrieve elements.
- Data structures are techniques for representing finite dynamic sets of elements
- Each element can contain:
  - a **key**, used to identify the element
  - **Satellite data**, carried around but unused by the data structure
  - **Attributes**, that are manipulated by the data structure eg., pointers to other objects
- Often keys stem from a **totally ordered set** (e. g. numbers)
  - Allows to define the minimum, successor and predecessor

### ► Aims of this lecture

- To introduce **data structures** and their typical operations.
- **Stacks, queues, priority queues** and **linked lists**.
- To work out the **running time** for operations on these data structures.
- To identify pros and cons for data structures in terms of efficiency.

### ► Data Structure Operations

- Operations on a dynamic sets  $S$  can be grouped into **queries** and **modifying operations**:
- Typical operations:
  - **Search( $S, k$ )**: returns a pointer to element  $x \in S$  with **key**  $k$ , ( $x.key = k$ ) or NIL
  - **Insert( $S, x$ )**: adds element pointed to by  $x$  to  $S$
  - **Delete( $S, x$ )**: given a pointer  $x$  to an element in  $S$  removes it from  $S$
  - **Minimum( $S$ ), Maximum( $S$ )**: return a pointer  $x$  to element resp. with smallest or largest key
  - **Successor( $S, x$ ), Predecessor( $S, x$ )**: next larger (smaller) than  $x.key$
- **Time** often measured using  $n$  as the number of elements in  $S$ .

## ► Data Structure Operations

- What's the runtime of each operation on an **array**?
  - **Search(S, k)**: returns a pointer to element  $x \in S$  with **key k**, ( $x.key = k$ ) or NIL  $\Theta(n)$
  - **Insert(S, x)**: adds element pointed to by  $x$  to  $S$   $\Theta(1)$
  - **Delete(S, x)**: given a pointer  $x$  to an element in  $S$  removes it from  $S$   $\Theta(1)$
  - **Minimum(S), Maximum(S)**: return a pointer  $x$  to element resp. with smallest or largest key  $\Theta(n)$
  - **Successor(S, x), Predecessor(S, x)**: next larger (smaller) than  $x.key$   $\Theta(n)$

## ► Data Structure Operations

- What's the runtime of each operation on a **sorted array**?
  - **Search(S, k)**: returns a pointer to element  $x \in S$  with **key k**, ( $x.key = k$ ) or NIL  $\Theta(\log n)$
  - **Insert(S, x)**: adds element pointed to by  $x$  to  $S$   $\Theta(n)$
  - **Delete(S, x)**: given a pointer  $x$  to an element in  $S$  removes it from  $S$   $\Theta(n)$
  - **Minimum(S), Maximum(S)**: return a pointer  $x$  to element resp. with smallest or largest key  $\Theta(1)$
  - **Successor(S, x), Predecessor(S, x)**: next larger (smaller) than  $x.key$   $\Theta(1)$

We'll now see some data structures that improve on the array implementation for many of the dynamic-set operations.

## ► Roadmap for the next lectures

- Simple data structures
  - Stacks
  - Queues
  - Linked lists
  - Binary search trees
  - Graphs
- Advanced data structures
  - Balanced trees
  - Priority queues

## ► Stacks

3
6
8



- Only the **top element** is accessible in a stack.
  - Last-in, first-out policy (LIFO)
- Insert is usually called **Push**, and Delete is called **Pop**.



## Stacks implemented using arrays

- Stacks can be implemented as an array  $S$  with attribute  $S.top$ .

PUSH( $S, x$ )

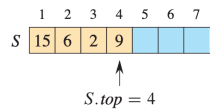
```
1 if  $S.top == S.size$ 
2   error "overflow"
3 else  $S.top = S.top + 1$ 
4    $S[S.top] = x$ 
```

STACK-EMPTY( $S$ )

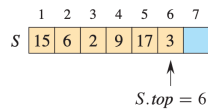
```
1 if  $S.top == 0$ 
2   return TRUE
3 else return FALSE
```

POP( $S$ )

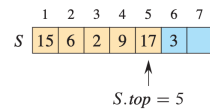
```
1 if STACK-EMPTY( $S$ )
2   error "underflow"
3 else  $S.top = S.top - 1$ 
4   return  $S[S.top + 1]$ 
```



(a)



(b)



(c)

- All stack operations take time  $O(1)$ .

## Stacks Application (1): Bracket Balance Checking

- $1 + \{2 * [x + (4y - z)] * [5x - (5y + z)] - 5t\}$
- $\{[()] [()]\}$
- Are the brackets correctly balanced or not?
- Read the expression: **Push** each opening bracket and **pop** for each closing bracket
- If the type of popped bracket always matches return **true**, else return **false**
- What's the runtime of the algorithm?

## Stacks Application (2): Postfix expression

- $5 * ((9 + 3) * (4 * 2) + 7)$  (infix expression)
- $5\ 9\ 3 + 4\ 2 * * 7 + *$  (postfix expression)
- Parsing postfix expressions is somewhat easier than infix expressions. Why?
- Read the tokens one at a time:
  - If it is an operand, **push** it on the stack
  - If it is a binary operator **pop** twice, apply the operator, and **push** the result back on the stack
- What is the runtime of the algorithm?

## Stacks Application (2): Postfix expression

- $5 * ((9 + 3) * (4 * 2) + 7)$  (infix expression)
- $5\ 9\ 3 + 4\ 2 * * 7 + *$  (postfix expression)

Stack operations	Stack elements
push(5)	5
push(9)	5 9
push(3)	5 9 3
push(pop() + pop())	5 12
push(4)	5 12 4
push(2)	5 12 4 2
push(pop() * pop())	5 12 8
push(pop() * pop())	5 96
push(7)	5 96 7
push(pop() + pop())	5 103
push(pop() * pop())	515

## ► Queues



head 

3	6	8
---	---	---

 tail

- The British love them 😊
- The first element in a queue is accessible.
  - First-in, first-out policy (FIFO)
- Insert is called **Enqueue**, Delete is called **Dequeue**.
- Queues have a **head** and a **tail**, like in a supermarket
  - Elements are added to the tail
  - Elements are extracted from the head

## ► Queues implemented using arrays

- Queues can be stored in an array “**wrapped around**”.

ENQUEUE( $Q, x$ )

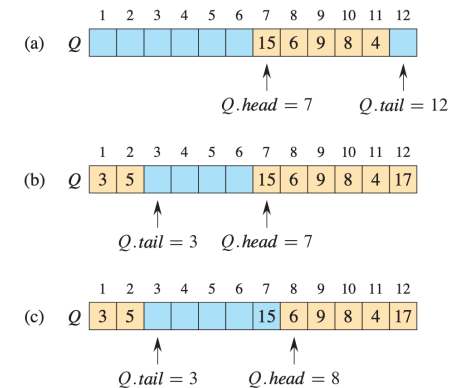
```

1   $Q[Q.tail] = x$ 
2  if  $Q.tail == Q.size$ 
3       $Q.tail = 1$ 
4  else  $Q.tail = Q.tail + 1$ 
```

DEQUEUE( $Q$ )

```

1   $x = Q[Q.head]$ 
2  if  $Q.head == Q.size$ 
3       $Q.head = 1$ 
4  else  $Q.head = Q.head + 1$ 
5  return  $x$ 
```



- All queue operations take time  $O(1)$ .

## ► Queues: Applications

- Playlists (eg., iTunes)
- Dispensing requests on a shared resource (eg., a printer, a server, a processor etc.,)
- Data buffers (eg., streaming services)
- What if I have priorities on the use of the resource?

## ► Priority Queues: Motivation

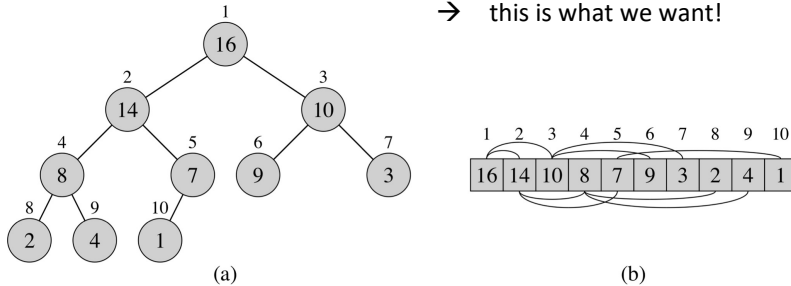
- Schedule jobs on a computer shared among multiple users
- A max-priority queue keeps track of the jobs to be performed and their relative priorities
- When a job is finished the scheduler selects the job with highest priority from those pending
- Jobs can be added to the scheduler at any time

Job	Owner	Priority (key)
Job 1	Tang Ke	35
Job 12	Oliveto Pietro	2
Job 24	Hao Qi	22
Job 25	Yu Shiqi	18
Job 72	Tang Ke	30

- **Use a heap!**

## ➤ Heap Properties

- **Max-heap property:** for every node other than the root, the parent is no smaller than the node,  $A[\text{Parent}(i)] \geq A[i]$ .
- In a max-heap, the **root** always stores a **largest** element.  
→ this is what we want!



- **Min-heap property:** for every node other than the root, the parent is no larger than the node,  $A[\text{Parent}(i)] \leq A[i]$ .

## ➤ Priority Queues: handles

Operation	Time
Insert( $S, x, k$ ) – inserts $x$ with key $k$ into $S$	$O(\log n)$
Maximum( $S$ ) – returns the element in $S$ with the largest key	$O(1)$
Extract-Max( $S$ ) – removes and returns element in $S$ with the largest key	$O(\log n)$
Increase-Key( $S, x, k$ ) – increases the key of $x$ to a larger value $k$ (element may float up in the heap)	$O(\log n)$

- Elements in the priority queue correspond to objects (eg., jobs)
- For an operation such as Increase-Key we need a way to **map** objects to and from their position in the heap (and update it as it moves in the heap)
- One way is to use **handles**: extra information stored in the object that allows to do the mapping
- A Job  $x$ :  **$x.\text{key}$**  (priority)  **$x.\text{heap\_index}$**  (handle)  $x.\text{satellite\_data}$ ;
- The heap needs to contain for each element **a pointer to the object**

## ➤ Priority Queue based on max-heap

- A data structure for maintaining a set  $S$  of elements with an associated element called key (the priority).

Operation	Time
Insert( $S, x, k$ ) – inserts $x$ with key $k$ into $S$	$O(\log n)$
Maximum( $S$ ) – returns the element in $S$ with the largest key	$O(1)$
Extract-Max( $S$ ) – removes and returns element in $S$ with the largest key	$O(\log n)$
Increase-Key( $S, x, k$ ) – increases the key of $x$ to a larger value $k$ (element may float up in the heap)	$O(\log n)$

**Min-priority queue** based on min-heap also exist: we will use them in graph algorithms (eg., Dijkstra, Prim)

## ➤ Find and extract next job

MAX-HEAP-MAXIMUM( $A$ )

```

1  if  $A.\text{heap-size} < 1$ 
2      error "heap underflow"
3  return  $A[1]$ 
```

MAX-HEAP-EXTRACT-MAX( $A$ )

```

1   $max = \text{MAX-HEAP-MAXIMUM}(A)$ 
2   $A[1] = A[A.\text{heap-size}]$ 
3   $A.\text{heap-size} = A.\text{heap-size} - 1$ 
4  MAX-HEAPIFY( $A, 1$ )
5  return  $max$ 
```

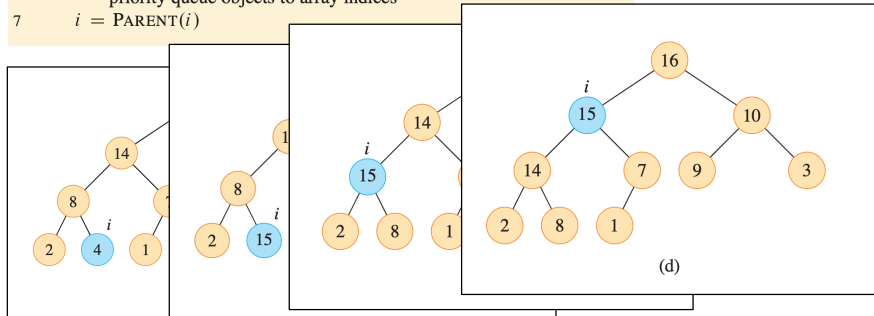
## ➤ Increase job priority

MAX-HEAP-INCREASE-KEY( $A, x, k$ )

```

1  if  $k < x.key$ 
2      error "new key is smaller than current key"
3   $x.key = k$ 
4  find the index  $i$  in array  $A$  where object  $x$  occurs
5  while  $i > 1$  and  $A[PARENT(i)].key < A[i].key$ 
6      exchange  $A[i]$  with  $A[PARENT(i)]$ , updating the information that maps
        priority queue objects to array indices
7   $i = PARENT(i)$ 

```



## ➤ Insert new job

MAX-HEAP-INSERT( $A, x, n$ )

```

1  if  $A.heap-size == n$ 
2      error "heap overflow"
3   $A.heap-size = A.heap-size + 1$ 
4   $k = x.key$ 
5   $x.key = -\infty$ 
6   $A[A.heap-size] = x$ 
7  map  $x$  to index  $heap-size$  in the array
8  MAX-HEAP-INCREASE-KEY( $A, x, k$ )

```

## ► Linked Lists: Array Disadvantages

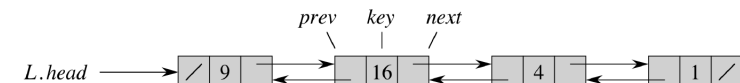
- You need to specify an initial size
- Changing the size of an array is troublesome
- Inserting and deleting elements in specific positions is difficult
- Let's say we want to delete 10 and keep the order of the rest:

$A$	5	8	10	13	16	19	27	46	51	86
$A$	5	8		13	16	19	27	46	51	86
$A$	5	8	13	16	19	27	46	51	86	

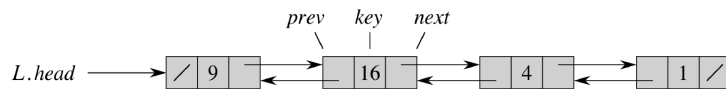
- What's the time complexity?

## ► Linked Lists

- Objects are linked using **pointers to the next element**.
- Linked lists can be **singly linked** or **doubly linked**: pointers to next and previous elements.
- Each element  $x$  has attributes
  - $x.key$  – the key used to identify the element
  - $x.next$  – a pointer to the next element
  - $x.prev$  – a pointer to the previous element
  - Optional: further satellite data



## ► Linked Lists: Searching



- Search inspects all elements in sequence and stops when the key has been found or the end of the list is reached.

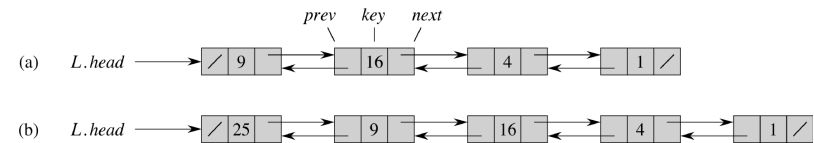
```

LIST-SEARCH( $L, k$ )
1:  $x = L.head$ 
2: while  $x \neq \text{NIL}$  and  $x.key \neq k$  do
3:    $x = x.next$ 
4: return  $x$ 

```

- The worst-case time is  $\Theta(n)$ , since it may have to search the entire list.

## ► Linked Lists: Inserting at the front



- New elements are added to the front of the list.

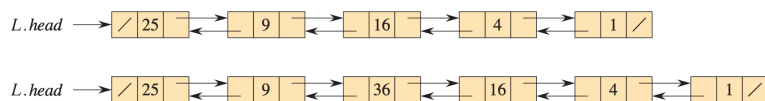
```

LIST-PREPEND( $L, x$ )
1  $x.next = L.head$ 
2  $x.prev = \text{NIL}$ 
3 if  $L.head \neq \text{NIL}$ 
4    $L.head.prev = x$ 
5  $L.head = x$ 

```

- The time for an insertion is  $O(1)$ .

## ► Linked Lists: Inserting after element x



- New element added after element y.

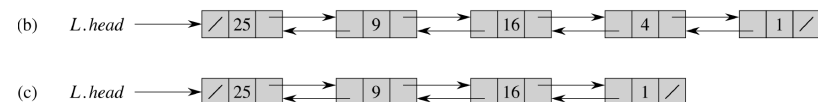
```

LIST-INSERT( $x, y$ )
1  $x.next = y.next$ 
2  $x.prev = y$ 
3 if  $y.next \neq \text{NIL}$ 
4    $y.next.prev = x$ 
5  $y.next = x$ 

```

- The time for an insertion is  $O(1)$  if you know the pointer to y

## ► Linked Lists: Deleting



- If element x is known, update pointers to take it out.

```

LIST-DELETE( $L, x$ )
1: if  $x.prev \neq \text{NIL}$  then
2:    $x.prev.next = x.next$ 
3: else
4:    $L.head = x.next$ 
5: if  $x.next \neq \text{NIL}$  then
6:    $x.next.prev = x.prev$ 

```

- The time for a deletion is  $O(1)$ .  
But if we only have the key and need to search the element x, it's time  $\Theta(n)$  in the worst case.

## ► Summary

- **Stacks** and **Queues** are simple data structures that can
  - be implemented efficiently in arrays (modulo space issues)
  - Have a restricted set of operations, but these run in time  $O(1)$ .
- **Priority Queues**: all operations in at most  $O(\log n)$  time
- Linked lists form an **unordered list** of elements
  - **Insertion** is fast if not important where it occurs: time  $O(1)$ .
  - **Searching** takes worst-case time  $\Theta(n)$ .
  - **Deletion** runs in time  $O(1)$  if the element is known, otherwise we need to run a search beforehand and incur time  $\Theta(n)$ .