# COMP20003 Workshop Week 4 Complexity + Stacks & Queues

- 1. Complexity Analysis
- 2. Another ADT: Stack
- 3. Yet Another ADT: Queue

#### LAB:

• Finish Assignment 1 OR do Week 4 Extras

# How fast is an algorithm?

- How to measure its the speed/efficiency of an algorithm?
- We have 2 algorithm, A and B, for solving the same problem. Which one is more efficient?

The running time is measured as time complexity

## **Strict Big-O definition**

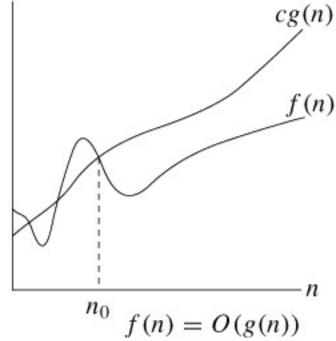
The complexity of an algorithm is the number of operations/steps and is expressed as a function f(n) of the *input size* n.

**Def:** We say f(n) belongs to the class O(g(n)), that is,  $f(n) \in O(g(n))$ , iif

• There are constants c and  $n_0$  such that  $f(n) \le c.g(n)$  for all  $n > n_0$ 

#### Underlying meaning:

- f(n) grows slower than, or as the same rate as, g(n)
- c.g(n) is an upper bound of f(n) for all large enough n



https://web.engr.oregonstate.edu/~huanlian/teaching/cs570

- Example:  $f(n) = 3n^2 + 6n + 20$ 
  - prove that  $f(n) \in O(n^2)$
  - any other g(n) such that  $f(n) \in O(f(n))$ ?

$$f(n) \in \begin{cases} O( ) \\ O( ) \\ O( ) \end{cases}$$

- We normally don't use the definition to find complexity.
   But if we want to prove, we need to rely on the definition.

## Big-O Notation: strict definition vs. CS meaning

**Def:** We say f(n) belongs to the class O(g(n)), that is,  $f(n) \in O(g(n))$ , iif

There are constants c and  $n_0$  such that  $f(n) \le c.g(n)$  for all  $n > n_0$ 

#### *Underlying meaning*:

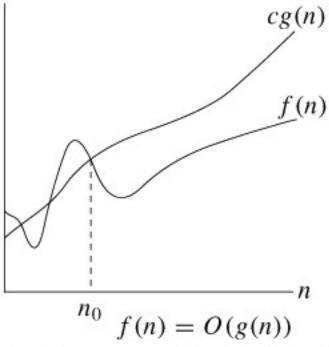
- f(n) grows slower than, or as the same rate as, g(n)
- c.g(n) is an upper bound of f(n) for all large enough n

#### CS meaning:

c.g(n) is the least upper bound of f(n) for all large enough n

Example:  $f(n) = 3n^2 + 6n + 20$ 

- strict Big-O:  $f(n) \in O(n^2)$ ,  $f(n) \in O(n^3)$ ,  $f(n) \in O(n.log n)$ , ...
- CS meaning:  $f(n) \in ?$



# Big- $\Omega$

The pitfall of Big-O is that it *only* describes an **upper bound** on an algorithm's running time, which often corresponds to the worst-case scenario.

algorithm A: max value	algorithm B: linear search
<pre>int max(int A[], int n) {   int max= A[0];   for (int i = 1; i<n; (a[i]="" i++)="" if=""> max )       max= A[i];   return max; }</n;></pre>	<pre>int search(int A[], int n, int key) {   for (int i = 0; i<n; (a[i]="=" )="" i++)="" i;="" if="" key="" notfound="" pre="" return="" }<=""></n;></pre>
O()?	does O() fully describe the performance?

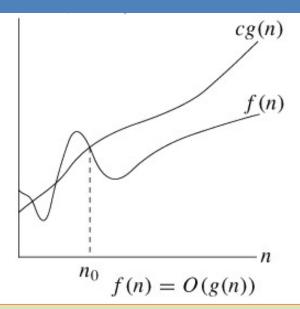
We use  $\Omega$  notation to specifically describe the lower bound of an algorithm's running time, which often corresponds to the best-case scenario.

When an algorithm's running time has the same upper bound and lower bound, we use **Theta** (O) **notation** to describe its tight bound.

# Big- $\Omega$

When an algorithm's running time has the *same upper bound and lower bound*, we use Theta (Θ) notation to describe its tight bound.

#### from Big-O to Big- $\Omega$ and Big- $\theta$

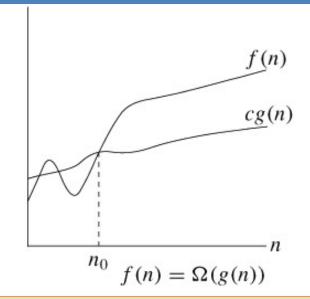


$$f(n) \in O(g(n))$$

**Def:** There are constants c and  $n_0$ such that

$$f(n) \le c.g(n)$$
 for all  $n > n_0$ 

- f(n) grows slower than, or as the same rate as, g(n)
- $f(n) \le c.g(n)$

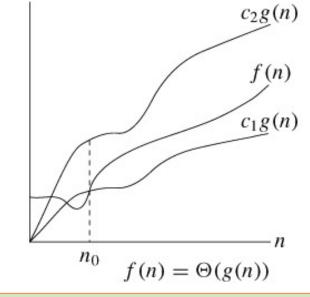


$$f(n) \in \Omega(g(n))$$
  
 $\Leftrightarrow g(n) \in O(f(n))$ 

**Def:** There are constants c and n<sub>0</sub> such that

$$f(n) \ge c.g(n)$$
 for all  $n > n_0$ 

- f(n) grows faster than, or as the same rate as, g(n)
- $f(n) \ge c.g(n)$



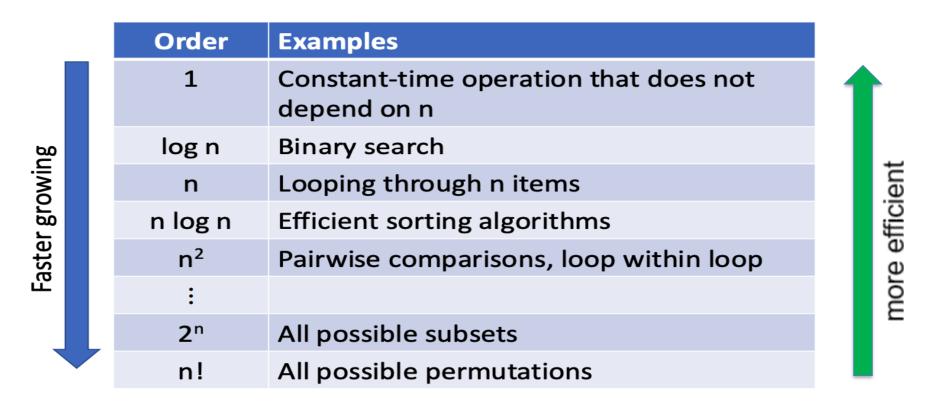
$$f(n) \in \theta(g(n))$$

$$\Leftrightarrow f(n) \in O(g(n)) \& f(n) \in \Omega(g(n))$$

- **Def:** There are constants  $c_1$ ,  $c_2$  and n<sub>0</sub> such that  $c_1.g(n) \le f(n) \le c_2.g(n)$  for all  $n > n_0$
- f(n) grows as the same rate as g(n)
- $c_1.g(n) \le f(n) \le c_2.g(n)$

Example: find some  $\Omega$ , O and  $\theta$  for:  $f(n)=3n^2+6n+20$ 

# **Complexity Classes**



When working with the complexity of an algorithm, we:

- **Derive a time complexity class**, which describes how the running time scales with the input size.
- Consider worst case and best case and use  $\Theta$  or (O and  $\Omega$ ) notation appropriately.

# Application 1: Comparing complexity functions f(n) and g(n)

- First reduce f(n) and g(n) to their simplest form (ie. complexity classes) using Big-O arithmetic
- •Then compare the classes using the increasing order: 1, log n, n, n log n, n<sup>2</sup>, n<sup>3</sup>, ..., 2<sup>n</sup>, n!

#### **Big-O Arithmetic**

- $\Theta(f(n)+g(n)) = \Theta(\max(f(n), g(n)))$
- $\Theta(c f(n)) = \Theta(f(n))$
- $\Theta(f) * \Theta(g) = \Theta(f*g)$

For a complexity function, we can:

- keep the most dominant term
- drop constants

Example: given

- $f(n) = 2n^2$
- g(n) = 9nlogn + 5n + 8

compare them in terms of complexity

Related Exercise: W4.3, W4.4

## Application 2: Finding Complexity of C codes (and algorithms in general)

#### Rules from lectures/Skiena:

- Each simple operation takes exactly one time step.
- Each memory access takes exactly one time step.

#### Practically:

Any combination of memory accesses, assignments, expressions is just  $\theta(1)$ if the total number of operations does not depend on the input size n

#### Examples: are they $\theta(1)$ ?

$$a= (b+c)*d - x;$$
if  $(a+b > c << 10)$ 
 $a= x + a*b;$ 

- Single operation or memory access is 1 step
- Loops are not considered as simple operations. Instead, they are the composition of many single-step operations.

```
1 for (i=0; i<n; i++)
2 sum = sum + A[i];
```

What's is the complexity of the above algorithm?

#### Model of computation:

- Single operation or memory access is 1 step
- Loops are not considered as simple operations. Instead, they are the composition of many single-step operations.

```
for (i=0; i<n; i++)
   sum = sum + A[i];
```

```
// find complexity of the following linear search
     for (i=0; i<n; i++) {
        if (x==A[i]) return i;
3
4
     return -1;
```

```
Solutions:
```

Related Exercises: W4.5

```
// find complexity of the following linear search
     for (i=0; i<n; i++) {
        if (x==A[i]) return i;
3
4
     return -1;
```

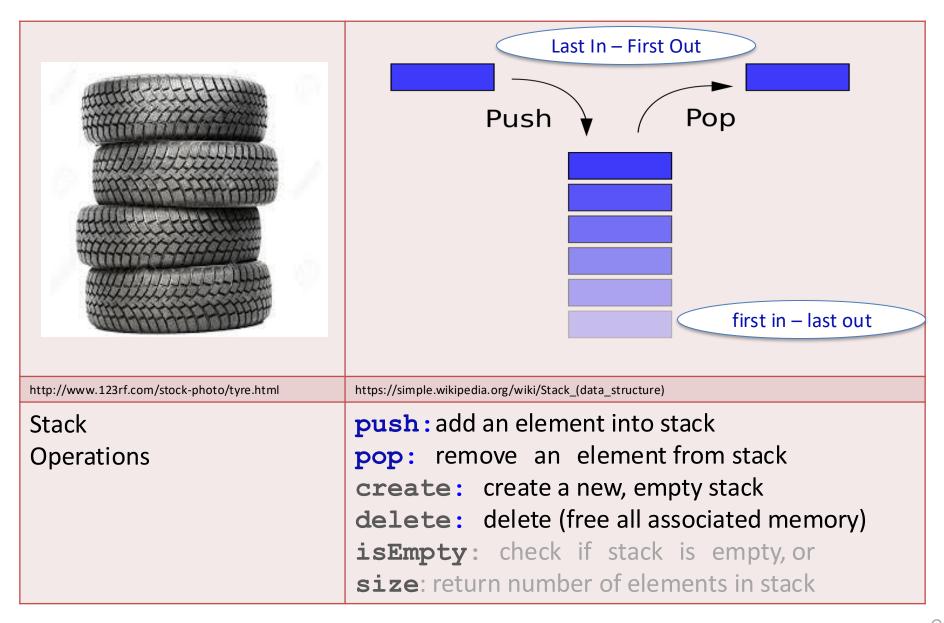
## Peer Activity W4.10

What is the strongest statement on the time complexity of this code snippet?

- A. O(n)
- B.  $O(n \log n)$
- C.  $O(n^2)$
- D.  $O(n^2 \log n)$

```
int ops = 0;
for (int i = 0; i < n - 1; i++) {
for (j = 1 << i; j < n; j++)
// 1<<i has value 2<sup>i</sup>
ops++;
```

# Another ADT: Stack (LIFO)

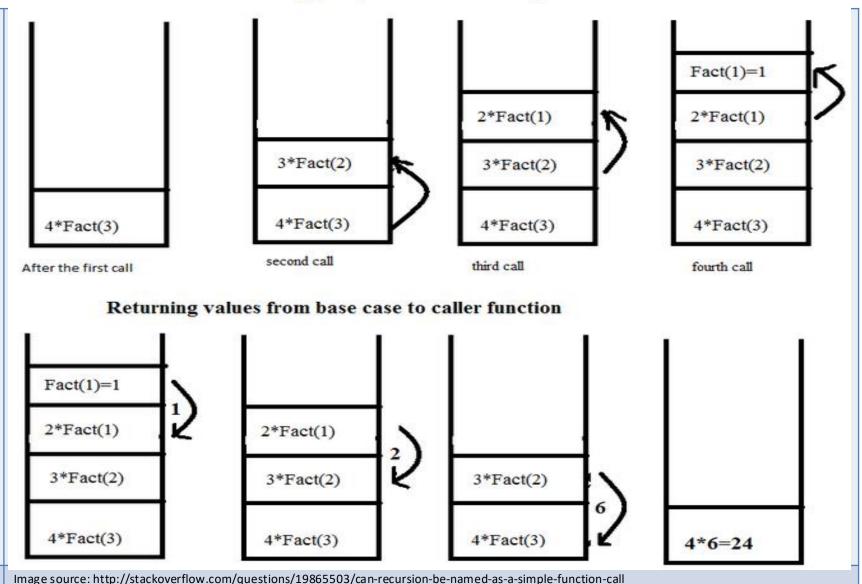


# Example: stack in function (and recursive function) calls

Stack is widely used in implementation of programming systems. For example, compilers employ stacks for keeping track of function calls and execution.

# Stack for: fact(4) int fact( int n ) { if ( n<=1 ) return 1; return n\*fact(n-1);

#### When function call happens previous variables gets stored in stack



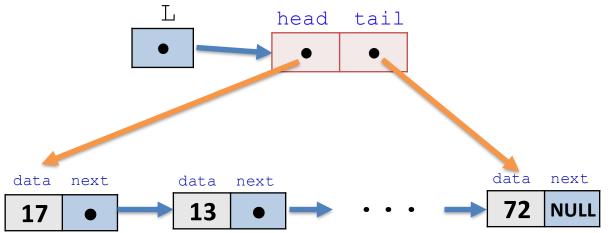
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## Stacks: Implementation using linked lists (exercise W4.6 + more)

```
// interface list.h given in WS3-page-8
// declare struct node
// declare struct list
struct list *create();
void prepend(struct list *, int);
void append(struct list *, int);
int deleteHead(struct list *);
int deleteTail(struct list *);
void freeList(struct list *);
```

#### Questions:

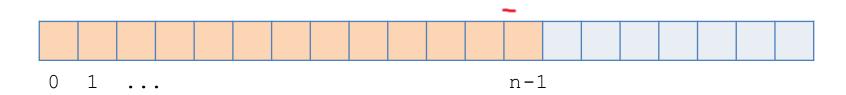
- How to efficiently implement Stacks using Linked Lists?
- What's the complexity of push? of pop?
- Using the above list interface, fill in the missing spaces in the RHS implementation.



```
#include "list.h"
typedef stru
stack_ADT createStack() {
void push(stack_ADT s, int data){
```

# Stacks: Implementation using arrays

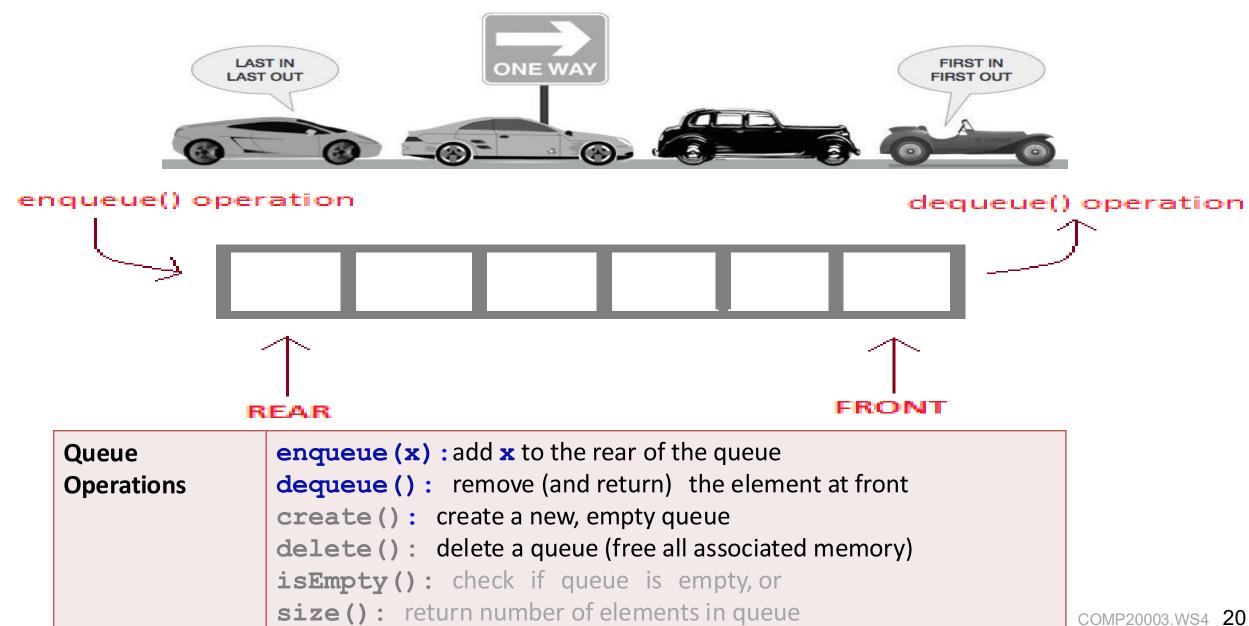
#### How to implement stacks using arrays?



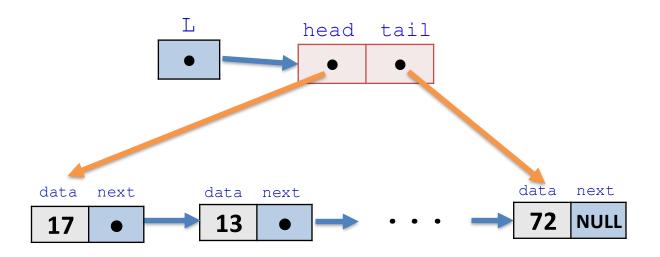
```
push (x) \Leftrightarrow \dots, complexity=?
pop ⇔ ..., complexity=?
```

Any potential problem?

#### Yet another ADT: Queue (FIFO) (exercise W4.7)



# Queue: implementation using linked list

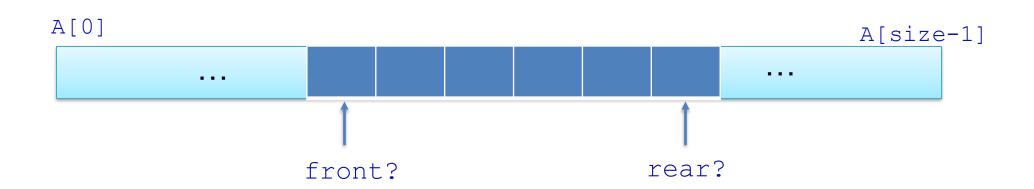


## How to implement queues using Linked Lists:

enqueue ⇔ dequeue ⇔

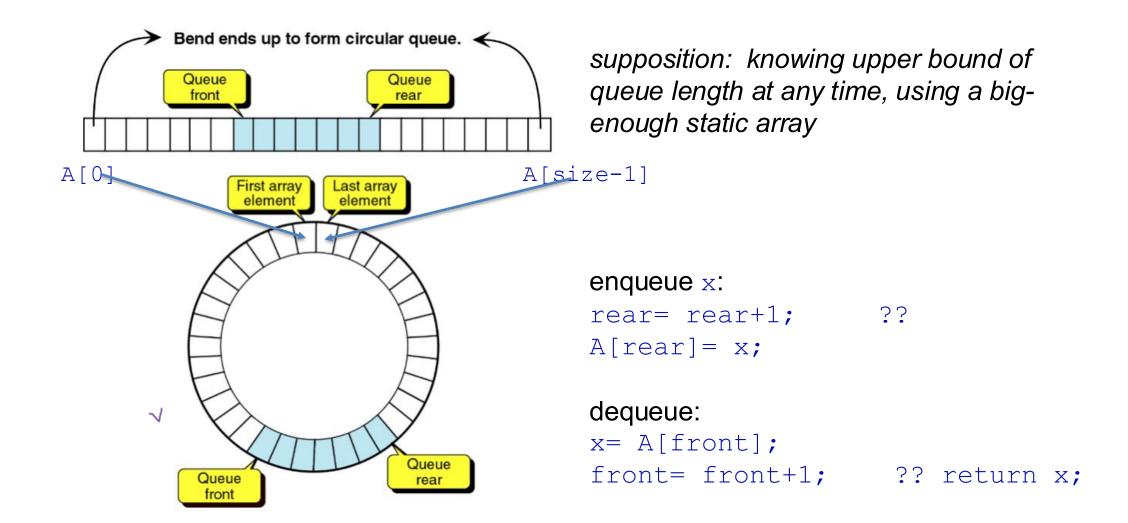
# Queue: implementation using array

Describe how to implement enqueue and dequeue using an unsorted array, ensuring  $\Theta(1)$  for enqueue & dequeue.



```
enqueue x: rear= rear+1; A[rear]= x;
dequeue: x = A[front]; front = front + 1; return x;
any problem?
```

# Queue: using circular arrays (if known the maximal size)



## Lab Time

Finish and/or refine Assignment 1

#### If Assignment 1 done:

- get all green ticks for Week 4 Workshop
- Do exercises in Week 4 Extras