

# CSE 332: Data Structures and Parallelism

Fall 2022

Richard Anderson

Lecture 2: Stacks, Queues and  
Algorithm Analysis

# Announcements

- My office hours (CSE2 344)
  - Tuesday, 10 AM – 11 AM
  - Friday, 3 PM - 4 PM
- Exercise #1: Due Monday
  - Computer set up and some programming
- Reading, Weiss
  - Algorithm Analysis: 2.1-2.4
  - Stacks and Queues: 3.1-3.7

# First Example: Queue ADT

- FIFO: First In First Out
- Queue operations

create

enqueue

dequeue

is\_empty

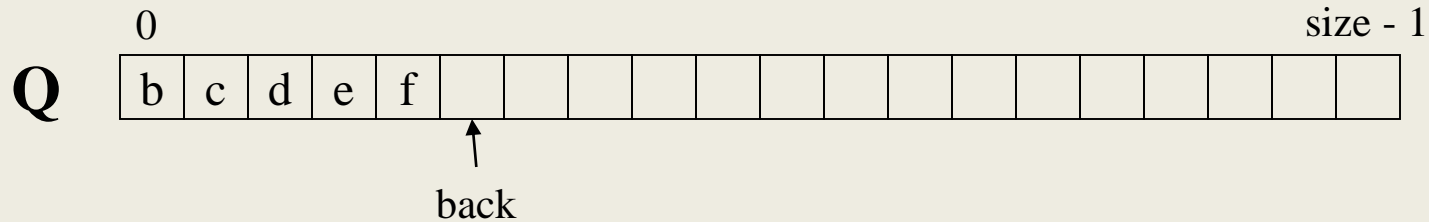


# Queues in practice

- Print jobs
- File serving
- Phone calls and operators

(Later, we will consider “priority queues.”)

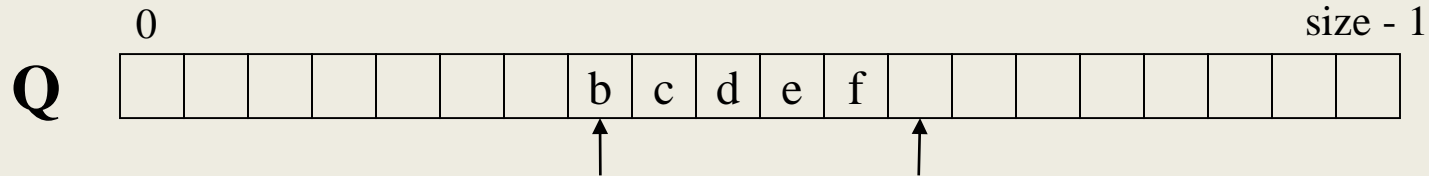
# Array Queue Data Structure



```
enqueue (Object x) {  
    Q[back] = x  
    back = back + 1  
}
```

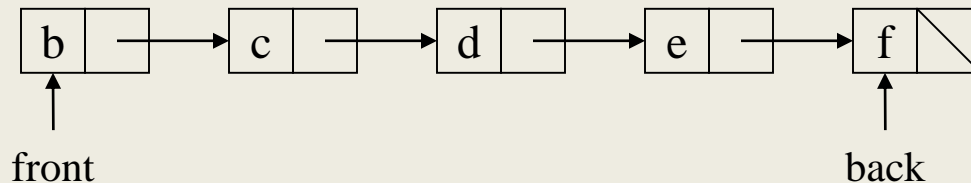
```
dequeue () {  
    x = Q[0]  
    shiftLeftOne()  
    back = back - 1  
    return x  
}
```

# Circular Array Queue Data Structure



```
enqueue(Object x) {  
    assert(!is_full())  
    Q[back] = x  
    back = (back + 1) % Q.size  
}  
  
dequeue() {  
    assert(!is_empty())  
    x = Q[front]  
    front = (front + 1) % Q.size  
    return x  
}
```

# Linked List Queue Data Structure



```
void enqueue(Object x) {  
    if (is_empty())  
        front = back = new Node(x)  
    else {  
        back.next = new Node(x)  
        back = back.next  
    }  
}  
  
bool is_empty() {  
    return front == null  
}
```

```
Object dequeue() {  
    assert(!is_empty())  
    return_data = front.data  
    temp = front  
    front = front.next  
    delete temp  
    return return_data  
}
```

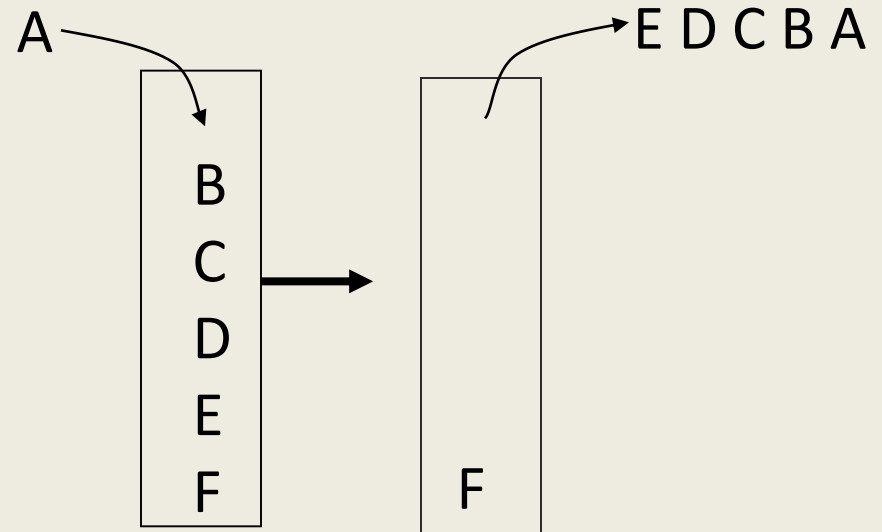
# Circular Array vs. Linked List

- Advantages of a circular array
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
  
- Advantages of a linked list



# Second Example: Stack ADT

- LIFO: Last In First Out
- Stack operations
  - create
  - push
  - pop
  - top
  - is\_empty



# Stacks in Practice

- Function call stack
- Removing recursion
- Balancing symbols (parentheses)
- Evaluating postfix or “reverse Polish” notation

# Algorithm Analysis

# Algorithm Analysis

- Correctness:
  - Does the algorithm do what is intended.
- Performance:
  - Speed: **time complexity**
  - Memory: **space complexity**
- Why analyze?
  - To make good design decisions
  - Enable you to look at an algorithm (or code) and identify the bottlenecks, etc.

# How to measure performance?

# Analyzing Performance

- Focus on **Worst Case Time Complexity**

<b>Basic operations</b>	Constant time
<b>Consecutive statements</b>	Sum of times
<b>Conditionals</b>	Test, plus larger branch cost
<b>Loops</b>	Sum of iterations
<b>Function calls</b>	Cost of function body
<b>Recursive functions</b>	Solve recurrence relation...

# Worst case complexity

Problem size **N**

- **Worst-case complexity:** **max** # steps algorithm takes on “most challenging” input of size **N**

# Exercise - Searching

2	3	5	16	37	50	73	75
---	---	---	----	----	----	----	----

```
bool ArrayContains(int array[], int n, int key) {  
  
  
  
  
  
  
  
  
  
}
```



# Linear Search Analysis

```
bool LinearArrayContains(int array[], int n, int key ) {  
    for( int i = 0; i < n; i++ ) {  
        if( array[i] == key )  
            // Found it!  
            return true;  
    }  
    return false;  
}
```

Best Case:

Worst Case:

# Binary Search Analysis

2	3	5	16	37	50	73	75
---	---	---	----	----	----	----	----

```
bool BinArrayContains( int array[], int low, int high, int key ) {  
    // The subarray is empty  
    if( low > high ) return false;  
  
    // Search this subarray recursively  
    int mid = (high + low) / 2;  
    if( key == array[mid] ) {  
        return true;  
    } else if( key < array[mid] ) {  
        return BinArrayFind( array, low, mid-1, key );  
    } else {  
        return BinArrayFind( array, mid+1, high, key );  
    }  
}
```

Best case:

Worst case:

# Solving Recurrences

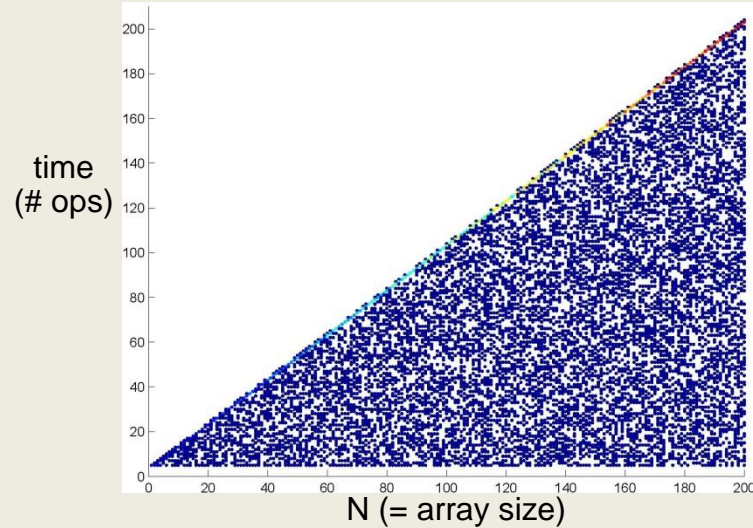
$$T(n) = T(n/2) + 7; \quad T(1) = 9$$

1. Determine the recurrence relations and base cases
2. Expand relation in terms of number of expansions  $k$
3. Find a closed form by setting  $k$  to value that reduces problem to the base case

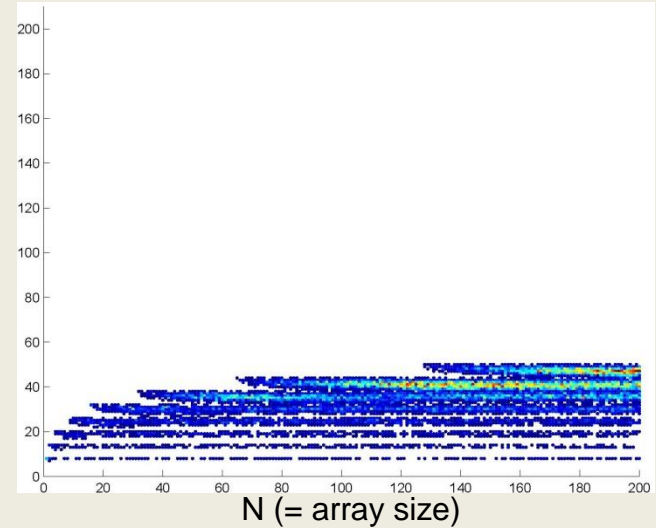
# Linear Search vs Binary Search

	Linear Search	Binary Search
Best Case	4	5 at [middle]
Worst Case	$3n+3$	$7 \lfloor \log n \rfloor + 9$

# Empirical comparison



Linear search



Binary search

# Asymptotic Analysis

- Consider only the *order* of the running time
  - A valuable tool when the input gets “large”
  - **Ignores** the effects of *different machines* or *different implementations* of same algorithm

# Asymptotic Analysis

- To find the asymptotic runtime, throw away the constants and low-order terms
  - Linear search is  $T_{worst}^{LS}(n) = 3n + 3 \in O(n)$
  - Binary search is  $T_{worst}^{BS}(n) = 7\lfloor \log_2 n \rfloor + 9 \in O(\log n)$

*Remember: the “fastest” algorithm has the slowest growing function for its runtime*

# Asymptotic Analysis

Eliminate low order terms

$$- 4n + 5 \Rightarrow$$

$$- 0.5 n \log n + 2n + 7 \Rightarrow$$

$$- n^3 + 3 \cdot 2^n + 8n \Rightarrow$$

Eliminate coefficients

$$- 4n \Rightarrow$$

$$- 0.5 n \log n \Rightarrow$$

$$- 3 \cdot 2^n \Rightarrow$$



# Properties of Logs

Basic:

- $A^{\log_A B} = B$
- $\log_A A =$

Independent of base:

- $\log(AB) =$
- $\log(A/B) =$
- $\log(A^B) =$
- $\log((A^B)^C) =$

# Properties of Logs

Changing base  $\rightarrow$  multiply by constant

- For example:  $\log_2 x = 3.22 \log_{10} x$
- More generally

$$\log_A n = \left( \frac{1}{\log_B A} \right) \log_B n$$

- Means we can ignore the base for asymptotic analysis (since we're ignoring constant multipliers)