

# Recursion (Part 1)



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

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## Beyond this lecture ...

- Fantastic resources for developing your recursive skills:

<http://codingbat.com/java/Recursion-1>

<http://codingbat.com/java/Recursion-2>

- The **best** long-term approaches for mastering recursion are:
  - learning a **functional programming** language  
[ e.g., Haskell: <https://www.haskell.org/tutorial/> ]
  - learning to develop a **compiler** (after learning **trees** in this course)  
[ e.g., ANTLR4 from [EECS4302](#) ]

# Background Study: Basic Recursion

- It is assumed that, in EECS2030, you learned about the basics of **recursion** in Java:
  - What makes a method recursive?
  - How to trace recursion using a **call stack**?
  - How to define and use **recursive helper methods** on arrays?
- If needed, review the above assumed basics from the relevant parts of EECS2030:
  - From **F'21**: Parts A – C, Lecture 8, Week 12
  - From **F'24**: Lecture 24, Sec. E (Tower of Hanoi)

## Tips.

- Skim the **slides**: watch lecture videos if needing explanations.
- Recursion lab from EECS2030-F22: **here** [Solution: **here**]
- Ask questions related to the assumed basics of **recursion**!
- Assuming that you know the basics of **recursion**, we will:
  - Look at an advanced example of **recursion on arrays** together.
  - Have you complete an assignment on the more advanced recursion problems.

# Learning Outcomes of this Lecture

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This module is designed to help you:

- Quickly review the *recursion basics*.
- Know about the resources on *recursion basics*.
- Get used to the more advanced use of recursion.

# Recursion: Principle

- **Recursion** is useful in expressing solutions to problems that can be **recursively** defined:
  - **Base Cases:** Small problem instances immediately solvable.
  - **Recursive Cases:**
    - Large problem instances *not immediately solvable*.
    - Solve by reusing *solution(s) to strictly smaller problem instances*.
- Similar idea learnt in high school: [ **mathematical induction** ]

# Tracing Method Calls via a Stack

- When a method is called, it is **activated** (and becomes **active**) and **pushed** onto the stack.
- When the body of a method makes a (helper) method call, that (helper) method is **activated** (and becomes **active**) and **pushed** onto the stack.
  - ⇒ The stack contains activation records of all **active** methods.
    - **Top** of stack denotes the current point of execution.
    - Remaining parts of stack are (temporarily) **suspended**.
- When entire body of a method is executed, stack is **popped**.
  - ⇒ The current point of execution is returned to the new **top** of stack (which was **suspended** and just became **active**).
- Execution terminates when the stack becomes **empty**.

# Tracing Method Calls via a Stack

- Can you identify the pattern of a Fibonacci sequence?

$$F = 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, \dots$$

- Here is the formal, **recursive** definition of calculating the  $n_{th}$  number in a Fibonacci sequence (denoted as  $F_n$ ):

$$F_n = \begin{cases} 1 & \text{if } n = 1 \\ 1 & \text{if } n = 2 \\ F_{n-1} + F_{n-2} & \text{if } n > 2 \end{cases}$$

- Your tasks are then to review how to
  - **implement** the above mathematical, recursive function in Java
  - **trace**, via a stack, the recursive execution at runtime

by studying **this video** ( $\approx 20$  minutes):

# Making Recursive Calls on an Array

- For **efficiency**, we exploit the feature of **call by value**, by:
  - passing the **reference** of the same array
  - specifying the **range of indices** to be considered

```
void m(int[] a, int from, int to) {  
    if(from > to) { /* base case */ }  
    else if(from == to) { /* base case */ }  
    else { m(a, from + 1, to) } }
```

- `m(a, 0, a.length - 1)` [ Initial call; entire array ]
- `m(a, 1, a.length - 1)` [ 1st r.c. on array of size `a.length - 1` ]
- `m(a, a.length-1, a.length-1)` [ Last r.c. on array of size 1 ]

- Required Task:**

Study the two examples `allPositive` and `isSorted` from the background study.

# A More Advanced Example on Recursion



**Assuming** that you will review the assumed basic, let's go over an advanced example from paper to Eclipse:

- **Problem Description:**

<https://www.eecs.yorku.ca/~wangcw/teaching/lectures/2025/W/EECS2101/exercises/EECS2101-W25-Problem-Recursion-splitArray-Spec.pdf>

- **Java starter project** (with hints and **JUnit tests**):

<https://www.eecs.yorku.ca/~wangcw/teaching/lectures/2025/W/EECS2101/exercises/ExtraRecursionProblemSplitArray Starter.zip>

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# Asymptotic Analysis of Algorithms



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# What You're Assumed to Know

- You will be required to **implement** Java classes and methods, and to **test** their correctness using JUnit.

Review them if necessary:

[https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2030\\_F21](https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2030_F21)

- Implementing classes and methods in Java [ Weeks 1 – 2 ]
  - Testing methods in Java [ Week 4 ]
- Also, make sure you know how to trace programs using a **debugger**:

<https://www.eecs.yorku.ca/~jackie/teaching/tutorials/index.html#java from scratch w21>

- Debugging actions (Step Over/Into/Return) [ Parts C – E, Week 2 ]

# Learning Outcomes

This module is designed to help you learn about:

- Notions of *Algorithms* and *Data Structures*
- Measurement of the “goodness” of an algorithm
- Measurement of the *efficiency* of an algorithm
- Experimental measurement vs. *Theoretical* measurement
- Understand the purpose of *asymptotic* analysis.
- Understand what it means to say two algorithms are:
  - equally efficient, **asymptotically**
  - one is more efficient than the other, **asymptotically**
- Given an algorithm, determine its *asymptotic upper bound* .

# Algorithm and Data Structure

- A **data structure** is:
  - A systematic way to store and organize data in order to facilitate **access** and **modifications**
  - Never suitable for all purposes: it is important to know its **strengths** and **limitations**
- A **well-specified computational problem** precisely describes the desired **input/output relationship**.
  - **Input:** A sequence of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$
  - **Output:** A permutation (reordering)  $\langle a'_1, a'_2, \dots, a'_n \rangle$  of the input sequence such that  $a'_1 \leq a'_2 \leq \dots \leq a'_n$
  - An **instance** of the problem:  $\langle 3, 1, 2, 5, 4 \rangle$
- An **algorithm** is:
  - A solution to a **well-specified** computational problem
  - A **sequence of computational steps** that takes value(s) as **input** and produces value(s) as **output**
- An **algorithm** manipulates some chosen **data structure(s)**.

# Measuring “Goodness” of an Algorithm

## 1. **Correctness**:

- Does the *algorithm* produce the expected output?
- Use *unit & regression testing* (e.g., JUnit) to ensure this.

## 2. Efficiency:

- **Time Complexity**: processor time required to complete
- **Space Complexity**: memory space required to store data

**Correctness** is always the priority.

How about efficiency? Is time or space more of a concern?

# Measuring Efficiency of an Algorithm

- **Time** is more of a concern than is **storage**.
- Solutions (run on computers) should be **as fast as possible**.
- Particularly, we are interested in how **running time** depends on two **input factors**:
  1. **size**  
e.g., sorting an array of 10 elements vs. 1m elements
  2. **structure**  
e.g., sorting an already-sorted array vs. a hardly-sorted array

Q. How does one determine the **running time** of an algorithm?

1. Measure time via **experiments**
2. Characterize time as a **mathematical function** of the input size

# Measure Running Time via Experiments

- Once the algorithm is implemented (e.g., in Java):
  - Execute program on **test inputs** of various **sizes** & **structures**.
  - For each test, record the **elapsed time** of the execution.

```
long startTime = System.currentTimeMillis();
/* run the algorithm */
long endTime = System.currentTimeMillis();
long elapsed = endTime - startTime;
```

- **Visualize** the result of each test.
- To make **sound statistical claims** about the algorithm's **running time**, the set of **test inputs** should be "**complete**".  
e.g., To experiment with the **RT** of a sorting algorithm:
  - **Unreasonable:** **only** consider small-sized and/or almost-sorted arrays
  - **Reasonable:** **also** consider large-sized, randomly-organized arrays

# Example Experiment

- **Computational Problem:**
  - **Input:** A character  $c$  and an integer  $n$
  - **Output:** A string consisting of  $n$  repetitions of character  $c$   
e.g., Given input '\*' and 15, output \*\*\*\*\*.
- **Algorithm 1** using String Concatenations:

```
public static String repeat1(char c, int n) {  
    String answer = "";  
    for (int i = 0; i < n; i++) { answer += c; }  
    return answer; }
```

- **Algorithm 2** using append from StringBuilder:

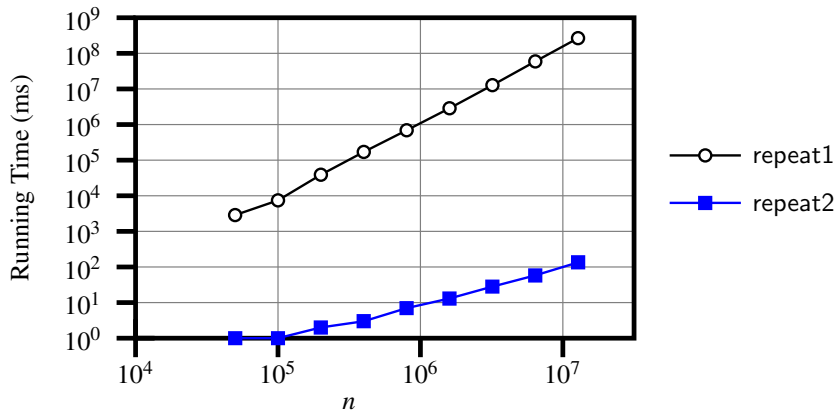
```
public static String repeat2(char c, int n) {  
    StringBuilder sb = new StringBuilder();  
    for (int i = 0; i < n; i++) { sb.append(c); }  
    return sb.toString(); }
```

## Example Experiment: Detailed Statistics

$n$	repeat1 (in ms)	repeat2 (in ms)
50,000	2,884	1
100,000	7,437	1
200,000	39,158	2
400,000	170,173	3
800,000	690,836	7
1,600,000	2,847,968	13
3,200,000	12,809,631	28
6,400,000	59,594,275	58
12,800,000	265,696,421 ( $\approx$ 3 days)	135

- As **input size** is doubled, **rates of increase** for both algorithms are **linear**:
  - Running time** of repeat1 increases by  $\approx$  5 times.
  - Running time** of repeat2 increases by  $\approx$  2 times.

# Example Experiment: Visualization



# Experimental Analysis: Challenges

1. An algorithm must be **fully implemented** (e.g., in Java) in order to study its runtime behaviour experimentally.
  - What if our purpose is to **choose among alternative** data structures or algorithms to implement?
  - Can there be a **higher-level analysis** to determine that one algorithm or data structure is more “**superior**” than others?
2. Comparison of multiple algorithms is only **meaningful** when experiments are conducted under the same working environment of:
  - **Hardware**: CPU, running processes
  - **Software**: OS, JVM version, Version of Compiler
3. Experiments can be done only on **a limited set of test inputs**.
  - What if **worst-case** inputs were not included in the experiments?
  - What if “**important**” inputs were not included in the experiments?

# Moving Beyond Experimental Analysis

- A better approach to analyzing the *efficiency* (e.g., *running time*) of algorithms should be one that:
  - Can be applied using a *high-level description* of the algorithm (without fully implementing it).  
[ e.g., Pseudo Code, Java Code (with “tolerances”) ]
  - Allows us to calculate the *relative efficiency* (rather than absolute elapsed time) of algorithms in a way that is *independent of* the hardware and software environment.
  - Considers *all* possible inputs (esp. the *worst-case scenario*).
- We will learn a better approach that contains 3 ingredients:
  1. Counting *primitive operations*
  2. Approximating running time as *a function of input size*
  3. Focusing on the *worst-case* input (requiring most running time)

# Counting Primitive Operations

- A **primitive operation** (**POs**) corresponds to a low-level instruction with a **constant execution time**.
  - (Variable) Assignment [e.g., `x = 5;`]
  - Indexing into an array [e.g., `a[i]`]
  - Arithmetic, relational, logical op. [e.g., `a + b`, `z > w`, `b1 && b2`]
  - Accessing an attribute of an object [e.g., `acc.balance`]
  - Returning from a method [e.g., `return result;`]

**Q:** Is a **method call** a primitive operation?

**A:** **Not** in general. It may be a call to:

- a “**cheap**” method (e.g., printing `Hello World`), or
- an “**expensive**” method (e.g., sorting an array of integers)
- **RT** of an **algorithm** is approximated as the number of **POs** involved (**despite** the execution environment).

## Example: Counting Primitive Operations (1)

```
1  int findMax (int[] a, int n) {
2      currentMax = a[0];
3      for (int i = 1; i < n; ) {
4          if (a[i] > currentMax) {
5              currentMax = a[i]; }
6          i ++ }
7      return currentMax; }
```

# of times  $i < n$  in **Line 3** is executed?  $[n]$

# of times the loop body (**Line 4** to **Line 6**) is executed?  $[n - 1]$

- **Line 2:**      $2$                                  [1 indexing + 1 assignment]
- **Line 3:**      $n + 1$                             [1 assignment +  $n$  comparisons]
- **Line 4:**      $(n - 1) \cdot 2$                   [1 indexing + 1 comparison]
- **Line 5:**      $(n - 1) \cdot 2$                   [1 indexing + 1 assignment]
- **Line 6:**      $(n - 1) \cdot 2$                   [1 addition + 1 assignment]
- **Line 7:**      $1$                                  [1 return]
- **Total # of Primitive Operations:**      $7n - 2$

## Example: Counting Primitive Operations (2)

Count the number of primitive operations for

```
1  boolean foundEmptyString = false;
2  int i = 0;
3  while (!foundEmptyString && i < names.length) {
4      if (names[i].length() == 0) {
5          /* set flag for early exit */
6          foundEmptyString = true;
7      }
8      i = i + 1;
9  }
```

- # times the stay condition of the `while` loop is checked?  
[ between 1 and `names.length + 1` ]  
[ **worst case**: `names.length + 1` times ]
- # times the body code of `while` loop is executed?  
[ between 0 and `names.length` ]  
[ **worst case**: `names.length` times ]

# From Absolute RT to Relative RT

- Each **primitive operation (PO)** takes approximately the same, constant amount of time to execute. [ say  $t$  ]

The absolute value of  $t$  depends on the **execution environment**.

**Q.** How do you relate the **number of POs** required by an algorithm and its **actual RT** on a specific working environment?

**A.** **Number of POs** should be proportional to the actual **RT**.

$$RT = t \cdot \text{number of POs}$$

- e.g., `findMax (int[] a, int n)` has  **$7n - 2$**  POs

$$RT = (7n - 2) \cdot t$$

- e.g., Say two algorithms with **RT**  $(7n - 2) \cdot t$  and **RT**  $(10n + 3) \cdot t$ :  
It suffices to compare their relative running time:

$$7n - 2 \text{ vs. } 10n + 3.$$

$\therefore$  To determine the **time efficiency** of an algorithm, we only focus on their **number of POs**.

## Example: Approx. # of Primitive Operations

- Given # of primitive operations counted precisely as  $7n - 2$ , we view it as

$$7 \cdot n^1 - 2 \cdot n^0$$

- We say
  - $n$  is the **highest power**
  - 7 and 2 are the **multiplicative constants**
  - 2 is the **lower term**
- When approximating a **function** [ e.g.,  $RT \approx f(n)$  ] (considering that **input size** may be very large):
  - Only** the **highest power** matters.
  - multiplicative constants** and **lower terms** can be dropped.

$\Rightarrow 7n - 2$  is approximately  $n$

**Exercise:** Consider  $7n + 2n \cdot \log n + 3n^2$ :

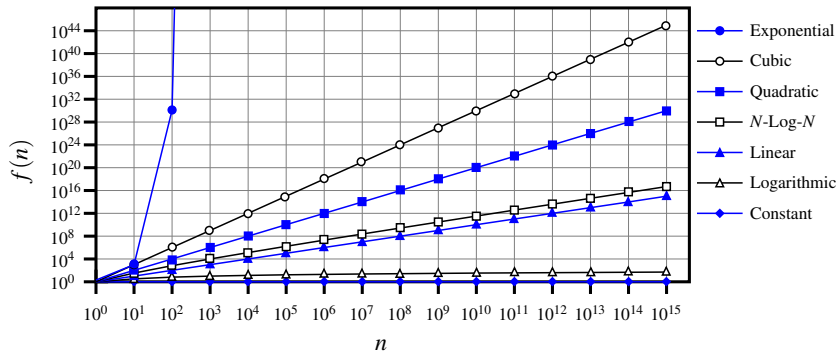
- highest power?** [  $n^2$  ]
- multiplicative constants?** [ 7, 2, 3 ]
- lower terms?** [  $7n, 2n \cdot \log n$  ]

# Approximating Running Time as a Function of Input Size

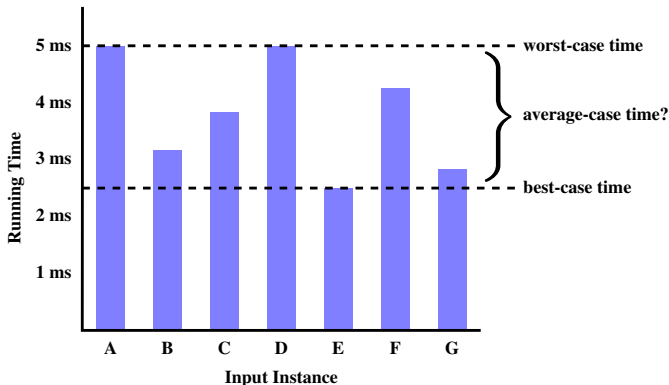
Given the **high-level description** of an algorithm, we associate it with a function  $f$ , such that  $f(n)$  returns the **number of primitive operations** that are performed on an **input of size  $n$** .

- $f(n) = 5$  [constant]
- $f(n) = \log_2 n$  [logarithmic]
- $f(n) = 4 \cdot n$  [linear]
- $f(n) = n^2$  [quadratic]
- $f(n) = n^3$  [cubic]
- $f(n) = 2^n$  [exponential]

# Rates of Growth: Comparison



# Focusing on the Worst-Case Input



- **Average-case** analysis calculates the expected running time based on the probability distribution of input values.
- **worst-case** analysis or **best-case** analysis?

# What is Asymptotic Analysis?

## Asymptotic analysis

- Is a method of describing behaviour towards the limit:
  - How the **running time** of the algorithm under analysis changes as the **input size** changes without bound
  - e.g., Contrast:  $RT_1(n) = n$  vs.  $RT_2(n) = n^2$
- Allows us to compare the relative performance of alternative algorithms:
  - For large enough inputs, the multiplicative constants and lower-order terms of an exact running time can be disregarded.
  - e.g.,  $RT_1(n) = 3n^2 + 7n + 18$  and  $RT_2(n) = 100n^2 + 3n - 100$  are considered **equally efficient**, **asymptotically**.
  - e.g.,  $RT_1(n) = n^3 + 7n + 18$  is considered **less efficient** than  $RT_2(n) = 100n^2 + 100n + 2000$ , **asymptotically**.

# Three Notions of Asymptotic Bounds

We may consider three kinds of *asymptotic bounds* for the *running time* of an algorithm:

- Asymptotic *upper* bound  $[O]$
- Asymptotic lower bound  $[\Omega]$
- Asymptotic tight bound  $[\Theta]$

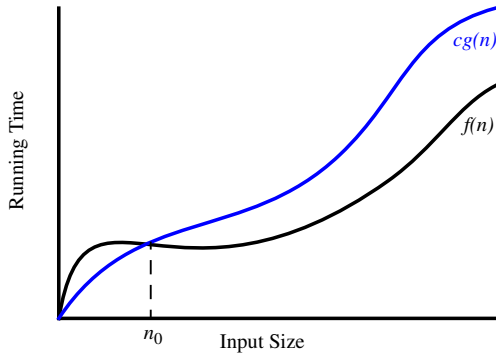
# Asymptotic Upper Bound: Definition

- Let  $f(n)$  and  $g(n)$  be functions mapping pos. integers (input size) to pos. real numbers (running time).
  - $f(n)$  characterizes the running time of some algorithm.
  - $O(g(n))$  :
    - denotes a collection of functions
    - consists of all functions that can be **upper bounded by  $g(n)$** , starting at some point, using some constant factor
- $f(n) \in O(g(n))$  if there are:
  - A real **constant**  $c > 0$
  - An integer **constant**  $n_0 \geq 1$
 such that:

$$f(n) \leq c \cdot g(n) \quad \text{for } n \geq n_0$$

- For each member function  $f(n)$  in  $O(g(n))$ , we say that:
  - $f(n) \in O(g(n))$  [f(n) is a member of "big-O of g(n)"]
  - $f(n)$  **is**  $O(g(n))$  [f(n) is "big-O of g(n)"]
  - $f(n)$  **is order of**  $g(n)$

# Asymptotic Upper Bound: Visualization



From  $n_0$ ,  $f(n)$  is *upper bounded by*  $c \cdot g(n)$ , so  $f(n)$  is  $O(g(n))$ .

# Asymptotic Upper Bound: Example (1)

**Prove:** The function  $8n + 5$  is  $O(n)$ .

**Strategy:** Choose a real constant  $c > 0$  and an integer constant  $n_0 \geq 1$ , such that for every integer  $n \geq n_0$ :

$$8n + 5 \leq c \cdot n$$

Can we choose  $c = 9$ ? What should the corresponding  $n_0$  be?

n	$8n + 5$	$9n$
1	13	9
2	21	18
3	29	27
4	37	36
5	45	45
6	53	54

...

Therefore, we prove it by choosing  $c = 9$  and  $n_0 = 5$ .

We may also prove it by choosing  $c = 13$  and  $n_0 = 1$ . Why?

# Asymptotic Upper Bound: Proposition

If  $f(n)$  is a polynomial of degree  $d$ , i.e.,

$$f(n) = a_0 \cdot n^0 + a_1 \cdot n^1 + \dots + a_d \cdot n^d$$

and  $a_0, a_1, \dots, a_d$  are integers, then  $f(n)$  is  $O(n^d)$ .

- We prove by choosing

$$\begin{aligned} c &= |a_0| + |a_1| + \dots + |a_d| \\ n_0 &= 1 \end{aligned}$$

- We know that for  $n \geq 1$ :  $n^0 \leq n^1 \leq n^2 \leq \dots \leq n^d$
- Upper-bound effect:  $n_0 = 1$ ?  $[f(1) \leq (|a_0| + |a_1| + \dots + |a_d|) \cdot 1^d]$

$$a_0 \cdot 1^0 + a_1 \cdot 1^1 + \dots + a_d \cdot 1^d \leq |a_0| \cdot 1^d + |a_1| \cdot 1^d + \dots + |a_d| \cdot 1^d$$

- Upper-bound effect holds?  $[f(n) \leq (|a_0| + |a_1| + \dots + |a_d|) \cdot n^d]$

$$a_0 \cdot n^0 + a_1 \cdot n^1 + \dots + a_d \cdot n^d \leq |a_0| \cdot n^d + |a_1| \cdot n^d + \dots + |a_d| \cdot n^d$$

## Asymptotic Upper Bound: Example (2)

**Prove:** The function  $f(n) = 5n^4 - 3n^3 + 2n^2 - 4n + 1$  is  $O(n^4)$ .

**Strategy:** Choose a real constant  $c > 0$  and an integer constant  $n_0 \geq 1$ , such that for every integer  $n \geq n_0$ :

$$5n^4 + 3n^3 + 2n^2 + 4n + 1 \leq c \cdot n^4$$

Using the proven **proposition**, choose:

- $c = |5| + |-3| + |2| + |-4| + |1| = 15$
- $n_0 = 1$

# Asymptotic Upper Bound: Families

- If a function  $f(n)$  is **upper bounded by** another function  $g(n)$  of degree  $d$ ,  $d \geq 0$ , then  $f(n)$  is also **upper bounded by** all other functions of a **strictly higher degree** (i.e.,  $d + 1$ ,  $d + 2$ , etc.).
  - e.g., Family of  $O(n)$  contains all  $f(n)$  that can be **upper bounded by**  $g(n) = n^1$ :
 

$n, 2n, 3n, \dots$	[ functions with degree 1 ]
$n^0, 2n^0, 3n^0, \dots$	[ functions with degree 0 ]
  - e.g., Family of  $O(n^2)$  contains all  $f(n)$  that can be **upper bounded by**  $g(n) = n^2$ :
 

$n^2, 2n^2, 3n^2, \dots$	[ functions with degree 2 ]
$n, 2n, 3n, \dots$	[ functions with degree 1 ]
$n^0, 2n^0, 3n^0, \dots$	[ functions with degree 0 ]
- Consequently:

$$O(n^0) \subset O(n^1) \subset O(n^2) \subset \dots$$

# Using Asymptotic Upper Bound Accurately

- Use the big-O notation to characterize a function (of an algorithm's running time) **as closely as possible**.

For example, say  $f(n) = 4n^3 + 3n^2 + 5$ :

- Recall:  $O(n^3) \subset O(n^4) \subset O(n^5) \subset \dots$
  - It is the **most accurate** to say that  $f(n)$  is  $O(n^3)$ .
  - It is **true**, but not very useful, to say that  $f(n)$  is  $O(n^4)$  and that  $f(n)$  is  $O(n^5)$ .
  - It is **false** to say that  $f(n)$  is  $O(n^2)$ ,  $O(n)$ , or  $O(1)$ .
- Do **not** include **constant factors** and **lower-order terms** in the big-O notation.

For example, say  $f(n) = 2n^2$  is  $O(n^2)$ , do not say  $f(n)$  is  $O(4n^2 + 6n + 9)$ .

# Asymptotic Upper Bound: More Examples

- $5n^2 + 3n \cdot \log n + 2n + 5$  is  $O(n^2)$  [ $c = 15, n_0 = 1$ ]
- $20n^3 + 10n \cdot \log n + 5$  is  $O(n^3)$  [ $c = 35, n_0 = 1$ ]
- $3 \cdot \log n + 2$  is  $O(\log n)$  [ $c = 5, n_0 = 2$ ]
  - Why can't  $n_0$  be 1?
  - Choosing  $n_0 = 1$  means  $\Rightarrow f(\boxed{1})$  **is** upper-bounded by  $c \cdot \log \boxed{1}$ :
    - We have  $f(\boxed{1}) = 3 \cdot \log 1 + 2$ , which is 2.
    - We have  $c \cdot \log \boxed{1}$ , which is 0.
  - $\Rightarrow f(\boxed{1})$  **is not** upper-bounded by  $c \cdot \log \boxed{1}$  [ Contradiction! ]
- $2^{n+2}$  is  $O(2^n)$  [ $c = 4, n_0 = 1$ ]
- $2n + 100 \cdot \log n$  is  $O(n)$  [ $c = 102, n_0 = 1$ ]

# Classes of Functions

upper bound	class	cost
$O(1)$	constant	<i>cheapest</i>
$O(\log(n))$	logarithmic	
$O(n)$	linear	
$O(n \cdot \log(n))$	"n-log-n"	
$O(n^2)$	quadratic	
$O(n^3)$	cubic	
$O(n^k), k \geq 1$	polynomial	
$O(a^n), a > 1$	exponential	<i>most expensive</i>

# Upper Bound of Algorithm: Example (1)

```
1  int maxOf (int x, int y) {  
2      int max = x;  
3      if (y > x) {  
4          max = y;  
5      }  
6      return max;  
7  }
```

- # of primitive operations: 4  
2 assignments + 1 comparison + 1 return = 4
- Therefore, the running time is  $O(1)$ .
- That is, this is a *constant-time* algorithm.

## Upper Bound of Algorithm: Example (2)

```
1  int findMax (int[] a, int n) {  
2      currentMax = a[0];  
3      for (int i = 1; i < n; ) {  
4          if (a[i] > currentMax) {  
5              currentMax = a[i]; }  
6          i ++ }  
7      return currentMax; }
```

- From last lecture, we calculated that the # of primitive operations is  $7n - 2$ .
- Therefore, the running time is  $O(n)$ .
- That is, this is a *linear-time* algorithm.

## Upper Bound of Algorithm: Example (3)

```
1  boolean containsDuplicate (int[] a, int n) {  
2      for (int i = 0; i < n; ) {  
3          for (int j = 0; j < n; ) {  
4              if (i != j && a[i] == a[j]) {  
5                  return true; }  
6              j ++; }  
7          i ++; }  
8      return false; }
```

- Worst case is when we reach Line 8.
- # of primitive operations  $\approx c_1 + n \cdot n \cdot c_2$ , where  $c_1$  and  $c_2$  are some constants.
- Therefore, the running time is  $O(n^2)$ .
- That is, this is a *quadratic* algorithm.

## Upper Bound of Algorithm: Example (4)

```
1  int sumMaxAndCrossProducts (int[] a, int n) {  
2      int max = a[0];  
3      for(int i = 1; i < n; i++) {  
4          if (a[i] > max) { max = a[i]; }  
5      }  
6      int sum = max;  
7      for (int j = 0; j < n; j++) {  
8          for (int k = 0; k < n; k++) {  
9              sum += a[j] * a[k]; } }  
10     return sum; }
```

- # of primitive operations  $\approx (c_1 \cdot n + c_2) + (c_3 \cdot n \cdot n + c_4)$ , where  $c_1$ ,  $c_2$ ,  $c_3$ , and  $c_4$  are some constants.
- Therefore, the running time is  $O(n + n^2) = O(n^2)$ .
- That is, this is a *quadratic* algorithm.

## Upper Bound of Algorithm: Example (5)

```
1  int triangularSum (int[] a, int n) {  
2      int sum = 0;  
3      for (int i = 0; i < n; i++) {  
4          for (int j = i; j < n; j++) {  
5              sum += a[j]; } }  
6      return sum; }
```

- # of primitive operations  $\approx n + (n - 1) + \dots + 2 + 1 = \frac{n \cdot (n+1)}{2}$
- Therefore, the running time is  $O(\frac{n^2+n}{2}) = O(n^2)$ .
- That is, this is a *quadratic* algorithm.

## Beyond this lecture ...

- You will be required to **implement** Java classes and methods, and to **test** their correctness using JUnit.

Review them if necessary:

[https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2030\\_F21](https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2030_F21)

- Implementing classes and methods in Java [ Weeks 1 – 2 ]
  - Testing methods in Java [ Week 4 ]
- Also, make sure you know how to trace programs using a **debugger**:

<https://www.eecs.yorku.ca/~jackie/teaching/tutorials/index.html#java from scratch w21>

- Debugging actions (Step Over/Into/Return) [ Parts C – E, Week 2 ]

# Index (1)

---

**What You're Assumed to Know**

**Learning Outcomes**

**Algorithm and Data Structure**

**Measuring “Goodness” of an Algorithm**

**Measuring Efficiency of an Algorithm**

**Measure Running Time via Experiments**

**Example Experiment**

**Example Experiment: Detailed Statistics**

**Example Experiment: Visualization**

**Experimental Analysis: Challenges**

**Moving Beyond Experimental Analysis**

# Index (2)

Counting Primitive Operations

Example: Counting Primitive Operations (1)

Example: Counting Primitive Operations (2)

From Absolute RT to Relative RT

Example: Approx. # of Primitive Operations

Approximating Running Time  
as a Function of Input Size

Rates of Growth: Comparison

Focusing on the Worst-Case Input

What is Asymptotic Analysis?

Three Notions of Asymptotic Bounds

# Index (3)

**Asymptotic Upper Bound: Definition**

**Asymptotic Upper Bound: Visualization**

**Asymptotic Upper Bound: Example (1)**

**Asymptotic Upper Bound: Proposition**

**Asymptotic Upper Bound: Example (2)**

**Asymptotic Upper Bound: Families**

**Using Asymptotic Upper Bound Accurately**

**Asymptotic Upper Bound: More Examples**

**Classes of Functions**

**Upper Bound of Algorithm: Example (1)**

**Upper Bound of Algorithm: Example (2)**

## Index (4)

Upper Bound of Algorithm: Example (3)

Upper Bound of Algorithm: Example (4)

Upper Bound of Algorithm: Example (5)

Beyond this lecture ...

# Basic Data Structures: Arrays vs. Linked-Lists



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

CHEN-WEI WANG

# Learning Outcomes of this Lecture

This module is designed to help you learn about:

- **basic data structures**: **Arrays** vs. **Linked Lists**
- Two **Sorting** Algorithms: Selection Sort vs. Insertion Sort
- **Linked Lists**: Singly-Linked vs. Doubly-Linked
- **Running Time**: Array vs. Linked-List Operations
- Java **Implementations**: **String** Lists vs. **Generic** Lists

# Basic Data Structure: Arrays

- An array is a sequence of indexed elements.
- **Size** of an array is fixed at the time of its construction.
  - e.g., `int[] numbers = new int[10];`
  - **Heads-Up**. Two **resizing** strategies: *increments* vs. *doubling*.
- Supported operations on an array:
  - **Accessing**: e.g., `int max = a[0];`  
Time Complexity:  **$O(1)$**  [ *constant-time* op. ]
  - **Updating**: e.g., `a[i] = a[i + 1];`  
Time Complexity:  **$O(1)$**  [ *constant-time* op. ]
  - **Inserting/Removing**:

```
String[] insertAt(String[] a, int n, String e, int i)
String[] result = new String[n + 1];
for(int j = 0; j <= i - 1; j++){ result[j] = a[j]; }
result[i] = e;
for(int j = i + 1; j <= n; j++){ result[j] = a[j-1]; }
return result;
```

Time Complexity:  **$O(n)$**  [ *linear-time* op. ]

# Array Case Study:

## Comparing Two Sorting Strategies

---

- The Sorting Problem:

**Input**: An array  $a$  of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$  (e.g.,  $\langle 3, 4, 1, 3, 2 \rangle$ )

**Output**: A permutation/reordering  $\langle a'_1, a'_2, \dots, a'_n \rangle$  of the input sequence s.t. elements are arranged in a non-descending order (e.g.,  $\langle 1, 2, 3, 3, 4 \rangle$ ):  $a'_1 \leq a'_2 \leq \dots \leq a'_n$

**Remark**. Variants of the *sorting problem* may require different *orderings*:

- non-descending
  - ascending/increasing
  - non-ascending
  - descending/decreasing
- Two alternative *implementation strategies* for solving this problem
  - At the end, choose one based on their *time complexities*.

# Sorting: Strategy 1 – Selection Sort

- Maintain a (initially empty) **sorted portion** of array *a*.
- From left to right in array *a*, select and insert the **minimum** element to the **end** of this sorted portion, so it remains sorted.

```

1 void selectionSort(int[] a, int n)
2   for (int i = 0; i <= (n - 2); i++)
3     int minIndex = i;
4     for (int j = i; j <= (n - 1); j++)
5       if (a[j] < a[minIndex]) { minIndex = j; }
6     int temp = a[i];
7     a[i] = a[minIndex];
8     a[minIndex] = temp;

```

- How many times does the body of **for-loop** (L4) run? [  $(n - 1)$  ]
- Running time? [  $O(n^2)$  ]

$$\underbrace{n}_{\text{find } \{a[0], \dots, a[n-1]\}} + \underbrace{(n-1)}_{\text{find } \{a[1], \dots, a[n-1]\}} + \dots + \underbrace{2}_{\text{find } \{a[n-2], a[n-1]\}}$$

- So **selection sort** is a **quadratic-time algorithm**.

## Sorting: Strategy 2 – Insertion Sort

- Maintain a (initially empty) **sorted portion** of array *a*.
- From left to right in array *a*, insert **one element** at a time into the **“correct” spot** in this sorted portion, so it remains sorted.

```

1 void insertionSort(int[] a, int n)
2   for (int i = 1; i < n; i++)
3     int current = a[i];
4     int j = i;
5     while (j > 0 && a[j - 1] > current)
6       a[j] = a[j - 1];
7       j--;
8     a[j] = current;

```

- **while-loop** (L5) exits when?  $[j \leq 0 \text{ or } a[j - 1] \leq \text{current}]$
- Running time?  $[O(n^2)]$

$$O(\underbrace{1}_{\text{insert into } \{a[0]\}} + \underbrace{2}_{\text{insert into } \{a[0], a[1]\}} + \dots + \underbrace{(n-1)}_{\text{insert into } \{a[0], \dots, a[n-2]\}})$$

- So **insertion sort** is a **quadratic-time algorithm**.

# Tracing Insertion & Selection Sorts in Java

- Given a fragment of Java code, you are expected to:
  - (1) **Derive** its *asymptotic upper bound*  
(by approximating the number of *POs*)
  - (2) **Trace** its *runtime execution*  
(by understanding how *variables* change)
- We did (1) in class.
- We discussed how, intuitively, the two sorting algorithms work.
- You are now **expected** to trace the Java code (both on paper and in Eclipse) on your own.
- If needed, you may consult with these videos:
  - Tracing Insertion Sort on paper [ [LINK](#) ]
  - Tracing Selection Sort on paper [ [LINK](#) ]
  - Tracing in Eclipse (using breakpoints/Debugger) [ [LINK](#) ]

# Comparing Insertion & Selection Sorts

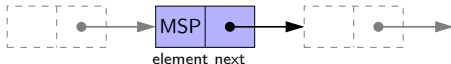
- *Asymptotically*, running times of *selection sort* and *insertion sort* are both  $O(n^2)$ .
- We will later see that there exist more efficient algorithms that can perform faster than quadratic:  $O(n \cdot \log n)$ .

## Exercise: Alternative Implementations?

- In the Java implementations of *selection sort* and *insertion sort*, we maintain the “*sorted portion*” from the *left* end.
  - For *selection sort*, we select the *minimum* element from the “*unsorted portion*” and insert it to the *end* of the “*sorted portion*”.
  - For *insertion sort*, we choose the *left-most* element from the “*unsorted portion*” and insert it at the “*correct spot*” in the “*sorted portion*”.
- **Exercise:** Modify the Java implementations, so that the “*sorted portion*” is:
  - arranged in a *non-ascending* order (e.g.,  $\langle 5, 3, 3, 1 \rangle$ ); and
  - maintained and grown from the *right* end instead.

# Basic Data Structure: Singly-Linked Lists

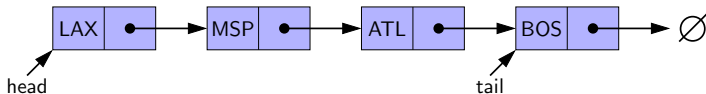
- We know that **arrays** perform:
  - well** in indexing
  - badly** in inserting and deleting
- We now introduce an alternative data structure to arrays.
- A **linked list** is a series of **connected nodes**, forming a **linear sequence**.  
**Remark.** At **runtime**, node **connections** are through **reference aliasing**.
- Each **node** in a **singly-linked list (SLL)** stores:
  - reference** to a **data object**; and
  - reference** to the **next node** in the list.**Contrast.** **relative** positioning of LL vs. **absolute** indexing of arrays



- The **last node** in a singly-linked list is different from others. How so?  
 Its reference to the **next node** is simply `null`.

# Singly-Linked List: How to Keep Track?

- Contrary to arrays, we do not keep track of all nodes in a SLL directly by indexing the *nodes*.
- Instead, we only store a *reference* to the *head* (i.e., *first node*) and the *tail* (i.e., *last node*), and access other parts of the list *indirectly*.



- Due to its “*chained*” *structure*, a SLL, when first being created, does not need to be specified with a fixed length.
- We can use a SLL to *dynamically* store and manipulate as many elements as we desire without the need to *resize*. We achieve this by:
  - e.g., *inserting* some node to the *beginning/middle/end* of a SLL
  - e.g., *deleting* some node from the *beginning/middle/end* of a SLL
- Exercise: Given the *head* reference of a SLL, describe how we may:
  - Count the number of nodes currently in the list. [ Running Time? ]
  - Find the reference to its *tail* (i.e., *last node*) [ Running Time? ]

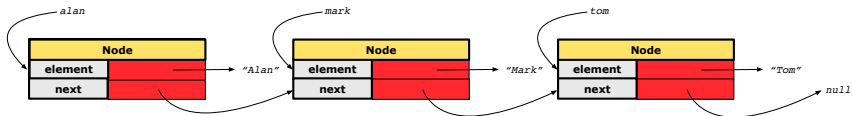
# Singly-Linked List: Java Implementation

We first implement a **SLL** storing strings only.

```
public class Node {  
    private String element;  
    private Node next;  
    public Node(String e, Node n) { element = e; next = n; }  
    public String getElement() { return element; }  
    public void setElement(String e) { element = e; }  
    public Node getNext() { return next; }  
    public void setNext(Node n) { next = n; }  
}
```

```
public class SinglyLinkedList {  
    private Node head;  
    public void setHead(Node n) { head = n; }  
    public int getSize() { ... }  
    public Node getTail() { ... }  
    public void addFirst(String e) { ... }  
    public Node getNodeAt(int i) { ... }  
    public void addAt(int i, String e) { ... }  
    public void removeLast() { ... }  
}
```

# Singly-Linked List: Constructing a Chain of Nodes



## Approach 1

```

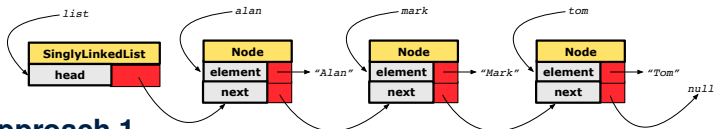
Node tom = new Node("Tom", null);
Node mark = new Node("Mark", tom);
Node alan = new Node("Alan", mark);
  
```

## Approach 2

```

Node alan = new Node("Alan", null);
Node mark = new Node("Mark", null);
Node tom = new Node("Tom", null);
alan.setNext(mark);
mark.setNext(tom);
  
```

# Singly-Linked List: Setting a List's Head



## Approach 1

```
Node tom = new Node("Tom", null);
Node mark = new Node("Mark", tom);
Node alan = new Node("Alan", mark);
SinglyLinkedList list = new SinglyLinkedList();
list.setHead(alan);
```

## Approach 2

```
Node alan = new Node("Alan", null);
Node mark = new Node("Mark", null);
Node tom = new Node("Tom", null);
alan.setNext(mark);
mark.setNext(tom);
SinglyLinkedList list = new SinglyLinkedList();
list.setHead(alan);
```

# Singly-Linked List: Counting # of Nodes (1)

**Problem:** Return the number of nodes currently stored in a SLL.

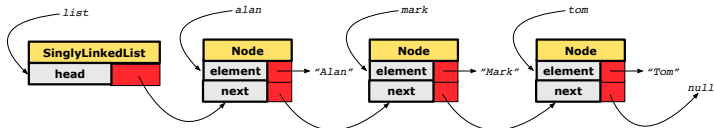
- **Hint.** Only the *last node* has a *null next* reference.
- Assume we are in the context of class `SinglyLinkedList`.

```

1  int getSize() {
2      int size = 0;
3      Node current = head;
4      while (current != null) {
5          current = current.getNext();
6          size ++;
7      }
8      return size;
9  }
```

- When does the while-loop (L4) exit? [ `current == null` ]
- RT of `getSize`:  $O(n)$  [ linear-time op. ]
- **Contrast:** RT of `a.length`:  $O(1)$  [ constant-time op. ]

# Singly-Linked List: Counting # of Nodes (2)



```

1 int getSize() {
2     int size = 0;
3     Node current = head;
4     while (current != null) { /* exit when current == null */
5         current = current.getNext();
6         size++;
7     }
8     return size;
9 }

```

Let's now consider `list.getSize()` :

current	current != null	End of Iteration	size
alan	true	1	1
mark	true	2	2
tom	true	3	3
null	false	—	—

# Singly-Linked List: Finding the Tail (1)

**Problem:** Retrieved the tail (i.e., last node) in a SLL.

- **Hint.** Only the *last node* has a *null next* reference.
- Assume we are in the context of class `SinglyLinkedList`.

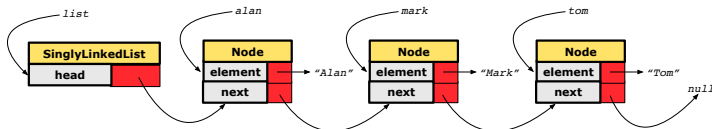
```

1  Node getTail() {
2      Node current = head;
3      Node tail = null;
4      while (current != null) {
5          tail = current;
6          current = current.getNext();
7      }
8      return tail;
9  }

```

- When does the while-loop (L4) exit? [ `current == null` ]
- RT of `getTail`:  $O(n)$  [ linear-time op. ]
- **Contrast:** RT of `a[a.length - 1]`:  $O(1)$  [ constant-time op. ]

# Singly-Linked List: Finding the Tail (2)



```

1 Node getTail() {
2     Node current = head;
3     Node tail = null;
4     while (current != null) { /* exit when current == null */
5         tail = current;
6         current = current.getNext();
7     }
8     return tail;
9 }

```

Let's now consider `list.getTail()` :

current	current != null	End of Iteration	tail
alan	true	1	alan
mark	true	2	mark
tom	true	3	tom
null	false	—	—

# Singly-Linked List: Can We Do Better?

- In practice, we may frequently need to:
  - Access the **tail** of a list. [e.g., customers joining a service queue]
  - Inquire the **size** of a list. [e.g., the service queue full?]

Both operations cost  $O(n)$  to run (with only **head** available).

- We may improve the **RT** of these two operations.

**Principle.** Trade **space** for **time**.

- Declare a new attribute **tail** pointing to the end of the list.
  - Declare a new attribute **size** denoting the number of stored nodes.
  - **RT** of these operations, accessing attribute values, are  **$O(1)$** !
- Why not declare attributes to store references of **all nodes** between **head** and **tail** (e.g., `secondNode`, `thirdNode`)?
  - No – at the **time of declarations**, we simply do **not** know how many nodes there will be at **runtime**.

# Singly-Linked List: Inserting to the Front (1)

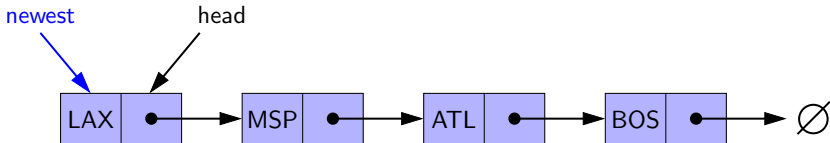
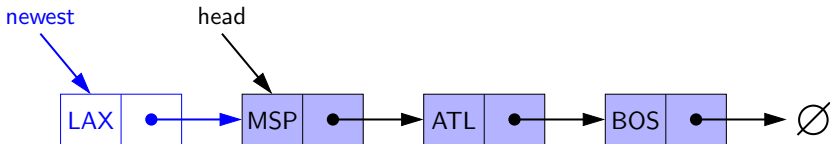
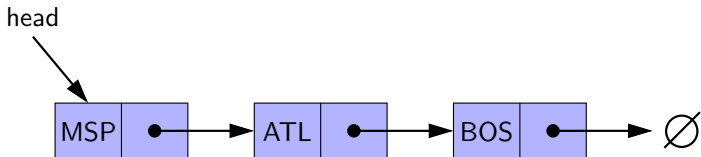
**Problem:** Insert a new string *e* to the front of the list.

- **Hint.** The list's new head should store *e* and point to the old head.
- Assume we are in the context of class `SinglyLinkedList`.

```
1 void addFirst (String e) {  
2     head = new Node(e, head);  
3     if (size == 0) {  
4         tail = head;  
5     }  
6     size ++;  
7 }
```

- Remember that RT of accessing *head* or *tail* is  $O(1)$
- RT of `addFirst` is  $O(1)$  [ constant-time op. ]
- **Contrast:** Inserting into an array costs  $O(n)$  [ linear-time op. ]

# Singly-Linked List: Inserting to the Front (2)



# Exercise

---



See `ExampleStringLinkedLists.zip`.

Compare and contrast two alternative ways to constructing a SLL: `testSLL_01` vs. `testSLL_02`.

# Exercise

- Complete the Java *implementations*, *tests*, and *running time analysis* for:
  - `void removeFirst()`
  - `void addLast(String e)`
- **Question:** The `removeLast()` method may not be completed in the same way as is `void addLast(String e)`. Why?

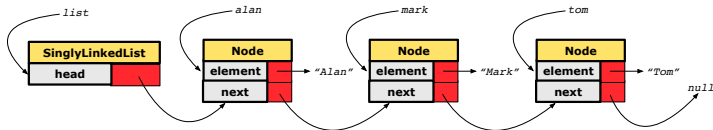
# Singly-Linked List: Accessing the Middle (1)

**Problem:** Return the node at index  $i$  in the list.

- **Hint.**  $0 \leq i < \text{list.getSize}()$
- Assume we are in the context of class `SinglyLinkedList`.

```
1  Node getNodeAt (int i) {
2      if (i < 0 || i >= size) {
3          throw new IllegalArgumentException("Invalid Index");
4      }
5      else {
6          int index = 0;
7          Node current = head;
8          while (index < i) { /* exit when index == i */
9              index++;
10             /* current is set to node at index i
11              * last iteration: index incremented from i - 1 to i
12              */
13             current = current.getNext();
14         }
15         return current;
16     }
17 }
```

# Singly-Linked List: Accessing the Middle (2)



```

1 Node getNodeAt (int i) {
2   if (i < 0 || i >= size) { /* error */ }
3   else {
4     int index = 0;
5     Node current = head;
6     while (index < i) { /* exit when index == i */
7       index ++;
8       current = current.getNext();
9     }
10    return current;
11  }
12 }

```

Let's now consider `list.getNodeAt(2)`:

current	index	index < 2	Beginning of Iteration
alan	0	true	1
mark	1	true	2
tom	2	false	—

# Singly-Linked List: Accessing the Middle (3)



- What is the **worst case** of the index  $i$  for `getNodeAt(i)`?
  - Worst case: `list.getNodeAt(list.size - 1)`
  - RT of `getNodeAt` is  $O(n)$  [ linear-time op. ]
- **Contrast:** Accessing an array element costs  $O(1)$  [ constant-time op. ]

# Singly-Linked List: Inserting to the Middle (1)

**Problem:** Insert a new element at index  $i$  in the list.

- **Hint 1.**  $0 \leq i \leq \text{list.getSize}()$
- **Hint 2.** Use `getNodeAt(?)` as a helper method.

```

1 void addAt (int i, String e) {
2     if (i < 0 || i > size) {
3         throw new IllegalArgumentException("Invalid Index.");
4     }
5     else {
6         if (i == 0) {
7             addFirst(e);
8         }
9         else {
10            Node nodeBefore = getNodeAt(i - 1);
11            Node newNode = new Node(e, nodeBefore.getNext());
12            nodeBefore.setNext(newNode);
13            size ++;
14        }
15    }
16 }

```

**Example.** See `testSLLAddAt` in `ExampleStringLinkedLists.zip`.

## Singly-Linked List: Inserting to the Middle (2)

- A call to `addAt(i, e)` may end up executing:
  - Line 3 (throw exception)  $[O(1)]$
  - Line 7 (`addFirst`)  $[O(1)]$
  - Lines 10 (`getNodeAt`)  $[O(n)]$
  - Lines 11 – 13 (setting references)  $[O(1)]$
- What is the **worst case** of the index  $i$  for `addAt(i, e)`?
  - A.** `list.addAt(list.getSize(), e)`  
 which requires `list.getNodeAt(list.getSize() - 1)`
- RT of `addAt` is  $O(n)$  [ linear-time op. ]
- **Contrast:** Inserting into an array costs  $O(n)$  [ linear-time op. ]  
 For arrays, when given the *index* to an element, the RT of inserting an element is always  $O(n)$  !

# Singly-Linked List: Removing from the End

**Problem:** Remove the last node (i.e., tail) of the list.

**Hint.** Using *tail* sufficient? Use `getNodeAt(?)` as a helper?

- Assume we are in the context of class `SinglyLinkedList`.

```
1 void removeLast () {  
2     if (size == 0) {  
3         throw new IllegalArgumentException("Empty List.");  
4     }  
5     else if (size == 1) {  
6         removeFirst();  
7     }  
8     else {  
9         Node secondLastNode = getNodeAt(size - 2);  
10        secondLastNode.setNext(null);  
11        tail = secondLastNode;  
12        size --;  
13    }  
14 }
```

Running time?  $O(n)$

# Singly-Linked List: Exercises

Consider the following two linked-list operations, where a *reference node* is given as an input parameter:

- `void insertAfter(Node n, String e)`

- Steps?

- Create a new node *nn*.
    - Set *nn*'s next to *n*'s next.
    - Set *n*'s next to *nn*.

- Running time?

[  $O(1)$  ]

- `void insertBefore(Node n, String e)`

- Steps?

- Iterate from the head, until *current.next == n*.
    - Create a new node *nn*.
    - Set *nn*'s next to *current*'s next (which is *n*).
    - Set *current*'s next to *nn*.

- Running time?

[  $O(n)$  ]

# Exercise

---



- Complete the Java *implementation*, *tests*, and *running time analysis* for `void removeAt(int i)`.

# Arrays vs. Singly-Linked Lists

DATA STRUCTURE		ARRAY	SINGLY-LINKED LIST
OPERATION			
get size			O(1)
get first/last element			
get element at index i		O(1)	O(n)
remove last element			
add/remove first element, add last element			O(1)
add/remove $i^{th}$ element	given reference to $(i - 1)^{th}$ element	O(n)	O(1)
	not given		O(n)

# Background Study: Generics in Java

- It is assumed that, in EECS2030, you learned about the basics of Java **generics**:
  - General collection (e.g., `Object[]`) vs. Generic collection (e.g., `E[]`)
  - How using generics minimizes **casts** and **instanceof checks**
  - How to implement and use generic classes
- If needed, review the above assumed basics from the relevant parts of EECS2030 ([https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2030\\_F21](https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2030_F21)):
  - Parts A1 – A3, Lecture 7, Week 10
  - Parts B – C, Lecture 7, Week 11

## Tips.

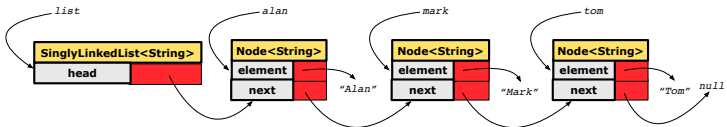
- Skim the **slides**; watch lecture videos if needing explanations.
- Ask questions related to the assumed basics of **generics**!
- Assuming that know the basics of Java **generics**, we will implement and use **generic SLL** and **DLL**.

# Generic Classes: Singly-Linked List (1)

```
public class Node<E> {  
    private E element;  
    private Node<E> next;  
    public Node(E e, Node<E> n) { element = e; next = n; }  
    public E getElement() { return element; }  
    public void setElement(E e) { element = e; }  
    public Node<E> getNext() { return next; }  
    public void setNext(Node<E> n) { next = n; }  
}
```

```
public class SinglyLinkedList<E> {  
    private Node<E> head;  
    private Node<E> tail;  
    private int size;  
    public void setHead(Node<E> n) { head = n; }  
    public void addFirst(E e) { ... }  
    Node<E> getNodeAt (int i) { ... }  
    void addAt (int i, E e) { ... }  
}
```

# Generic Classes: Singly-Linked List (2)



## Approach 1

```

Node<String> tom = new Node<String>("Tom", null);
Node<String> mark = new Node<>("Mark", tom);
Node<String> alan = new Node<>("Alan", mark);
SinglyLinkedList<String> list = new SinglyLinkedList<>();
list.setHead(alan);
  
```

## Approach 2

```

Node<String> alan = new Node<String>("Alan", null);
Node<String> mark = new Node<>("Mark", null);
Node<String> tom = new Node<>("Tom", null);
alan.setNext(mark);
mark.setNext(tom);
SinglyLinkedList<String> list = new SinglyLinkedList<>();
list.setHead(alan);
  
```

## Generic Classes: Singly-Linked List (3)

Assume we are in the context of class `SinglyLinkedList`.

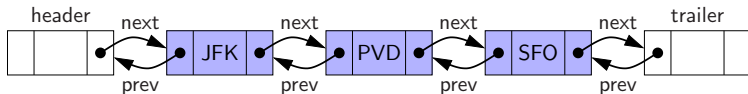
```
void addFirst (E e) {  
    head = new Node<E>(e, head);  
    if (size == 0) { tail = head; }  
    size ++;  
}
```

```
Node<E> getNodeAt (int i) {  
    if (i < 0 || i >= size) {  
        throw new IllegalArgumentException("Invalid Index"); }  
    else {  
        int index = 0;  
        Node<E> current = head;  
        while (index < i) {  
            index ++;  
            current = current.getNext();  
        }  
        return current;  
    }  
}
```

# Basic DS: Doubly-Linked Lists (1.1)

- We know that a *singly-linked list (SLL)* performs:
  - **WELL:**
    - inserting to the front/end [  $O(1)$  ]
    - removing from the front [ *head/tail* ]
    - inserting/deleting the middle [ *head* ]
  - **POORLY:**
    - accessing the middle [ given ref. to previous node ]
    - removing from the end [  $O(n)$  ]
    - inserting/deleting the middle [ `getNodeAt(i)` ]
- We may again improve the performance by trading *space* for *time* just like how attributes *size* and *tail* were introduced.

# Basic DS: Doubly-Linked Lists (1.2)



- Each **node** in a **doubly-linked list (DLL)** stores:

- A **reference** to an element of the sequence
- A **reference** to the next node in the list
- A **reference** to the **previous node** in the list

[ SYMMETRY ]

# Basic DS: Doubly-Linked Lists (2.1)

- Recall the need to handle **edge cases** for **singly**-linked lists:

```

1 void addFirst (E e) {
2     head = new Node<E>(e, head);
3     if (size == 0) { tail = head; }
4     size ++;
5 }

```

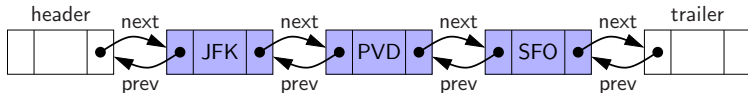
```

1 void removeFirst () {
2     if (size == 0) { /* error */ }
3     else if (size == 1) { head = null; tail = null; }
4     else {
5         Node<E> oldHead = head;
6         head = oldHead.getNext(); oldHead.setNext(null);
7     }
8     size --;
9 }

```

- Cases of empty **current** and **resulting** lists are **explicitly** coded.
- We can actually resolve this issue via a **small extension**!

## Basic DS: Doubly-Linked Lists (2.2)



- Each **DLL** stores:
  - A **reference** to a dedicated **header node** in the list
  - A **reference** to a dedicated **trailer node** in the list
- Remark.** Unlike SLL, **DLL** does not store refs. to **head** and **tail**.
- These two special nodes are called **sentinels** or **guards**:
  - They do **not** store data, but store node references:
    - The **header node** stores the **next** reference only
    - The **trailer node** stores **previous** reference only
  - They **always** exist, even in the case of empty lists.

# DLNs and DLLs: prev, header, trailer

- The node-level *prev reference* helps *improve the performance* of `removeLast()`.
  - ∴ The second last node can be accessed in *constant time*.  
`[ trailer.getPrev().getPrev() ]`
- The two list-level *sentinel/guard* nodes (*header* and *trailer*) do not help improve the performance.
  - Instead, they help *simplify the logic* of your code.
  - Each insertion/deletion can be treated
    - *Uniformly* : a node is always inserted/deleted in-between two nodes
    - Without worrying about dealing with edge cases by re-setting the *head* and *tail* of list

# Generic Doubly-Linked Lists in Java (1)

```
public class Node<E> {  
    private E element;  
    private Node<E> next;  
    public E getElement() { return element; }  
    public void setElement(E e) { element = e; }  
    public Node<E> getNext() { return next; }  
    public void setNext(Node<E> n) { next = n; }  
    private Node<E> prev;  
    public Node<E> getPrev() { return prev; }  
    public void setPrev(Node<E> p) { prev = p; }  
    public Node(E e, Node<E> p, Node<E> n) {  
        element = e;  
        prev = p;  
        next = n;  
    }  
}
```

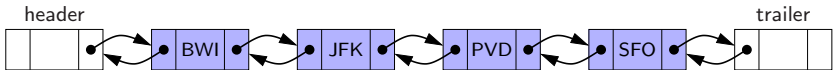
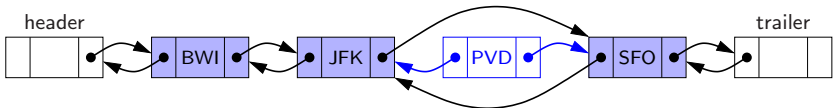
## Generic Doubly-Linked Lists in Java (2)

```
1 public class DoublyLinkedList<E> {  
2     private int size = 0;  
3     public void addFirst(E e) { ... }  
4     public void removeLast() { ... }  
5     public void addAt(int i, E e) { ... }  
6     private Node<E> header;  
7     private Node<E> trailer;  
8     public DoublyLinkedList() {  
9         header = new Node<>(null, null, null);  
10        trailer = new Node<>(null, header, null);  
11        header.setNext(trailer);  
12    }  
13 }
```

Lines 8 to 10 are equivalent to:

```
header = new Node<>(null, null, null);  
trailer = new Node<>(null, null, null);  
header.setNext(trailer);  
trailer.setPrev(header);
```

# Doubly-Linked List: Insertions



# Doubly-Linked List: Inserting to Front/End

```
1 void addBetween(E e, Node<E> pred, Node<E> succ) {  
2     Node<E> newNode = new Node<>(e, pred, succ);  
3     pred.setNext(newNode);  
4     succ.setPrev(newNode);  
5     size++;  
6 }
```

Running Time?  $O(1)$

```
void addFirst(E e) {  
    addBetween(e, header, header.getNext())  
}
```

Running Time?  $O(1)$

```
void addLast(E e) {  
    addBetween(e, trailer.getPrev(), trailer)  
}
```

Running Time?  $O(1)$

# Doubly-Linked List: Inserting to Middle

```

1 void addBetween(E e, Node<E> pred, Node<E> succ) {
2     Node<E> newNode = new Node<>(e, pred, succ);
3     pred.setNext(newNode);
4     succ.setPrev(newNode);
5     size++;
6 }

```

Running Time?  $O(1)$

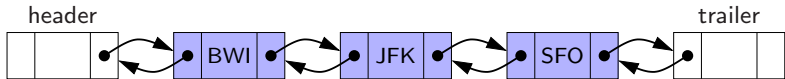
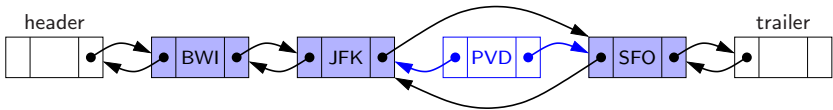
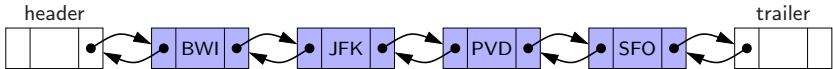
```

addAt(int i, E e) {
    if (i < 0 || i > size) {
        throw new IllegalArgumentException("Invalid Index.");
    } else {
        Node<E> pred = getNodeAt(i - 1);
        Node<E> succ = pred.getNext();
        addBetween(e, pred, succ);
    }
}

```

Running Time? Still  $O(n)$  !!!

# Doubly-Linked List: Removals



# Doubly-Linked List: Removing from Front/End

```

1  void remove (Node<E> node) {
2      Node<E> pred = node.getPrev();
3      Node<E> succ = node.getNext();
4      pred.setNext(succ); succ.setPrev(pred);
5      node.setNext(null); node.setPrev(null);
6      size--;
7  }

```

Running Time?  $O(1)$

```

void removeFirst() {
    if (size == 0) { throw new IllegalArgumentException("Empty"); }
    else { remove(header.getNext()); }
}

```

Running Time?  $O(1)$

```

void removeLast() {
    if (size == 0) { throw new IllegalArgumentException("Empty"); }
    else { remove(trailer.getPrev()); }
}

```

Running Time? Now  $O(1)$  !!!

# Doubly-Linked List: Removing from Middle

```
1 void remove (Node<E> node) {  
2     Node<E> pred = node.getPrev();  
3     Node<E> succ = node.getNext();  
4     pred.setNext(succ); succ.setPrev(pred);  
5     node.setNext(null); node.setPrev(null);  
6     size --;  
7 }
```

Running Time?  $O(1)$

```
removeAt (int i) {  
    if (i < 0 || i >= size) {  
        throw new IllegalArgumentException("Invalid Index.");  
    }  
    else {  
        Node<E> node = getNodeAt(i);  
        remove (node);  
    }  
}
```

Running Time? Still  $O(n)$  !!!

# Reference Node: To be Given or Not to be Given

**Exercise 1:** Compare the steps and running times of:

- *Not given* a reference node:
  - `addNodeAt(int i, E e)` [  $O(n)$  ]
- *Given* a reference node:
  - `addNodeBefore(Node<E> n, E e)` [ SLL:  $O(n)$ ; DLL:  $O(1)$  ]
  - `addNodeAfter(Node<E> n, E e)` [  $O(1)$  ]

**Exercise 2:** Compare the steps and running times of:

- *Not given* a reference node:
  - `removeNodeAt(int i)` [  $O(n)$  ]
- *Given* a reference node:
  - `removeNodeBefore(Node<E> n)` [ SLL:  $O(n)$ ; DLL:  $O(1)$  ]
  - `removeNodeAfter(Node<E> n)` [  $O(1)$  ]
  - `removNode(Node<E> n)` [ SLL:  $O(n)$ ; DLL:  $O(1)$  ]

# Arrays vs. (Singly- and Doubly-Linked) Lists

DATA STRUCTURE		ARRAY	SINGLY-LINKED LIST	DOUBLY-LINKED LIST
OPERATION				
size			O(1)	
first/last element				
element at index i		O(1)	O(n)	O(n)
remove last element				O(1)
add/remove first element, add last element			O(1)	
add/remove $i^{th}$ element	given reference to $(i - 1)^{th}$ element	O(n)		
	not given			O(n)

# Beyond this lecture ...



- In Eclipse, *implement* and *test* the assigned methods in `SinglyLinkedList` class and `DoublyLinkedList` class.
- Modify the *insertion sort* and *selection sort* implementations using a SLL or DLL.

# Index (1)

**Learning Outcomes of this Lecture**

**Basic Data Structure: Arrays**

**Array Case Study:**

**Comparing Two Sorting Strategies**

**Sorting: Strategy 1 – Selection Sort**

**Sorting: Strategy 2 – Insertion Sort**

**Tracing Insertion & Selection Sorts in Java**

**Comparing Insertion & Selection Sorts**

**Exercise: Alternative Implementations?**

**Basic Data Structure: Singly-Linked Lists**

**Singly-Linked List: How to Keep Track?**

## Index (2)

**Singly-Linked List: Java Implementation**

**Singly-Linked List:**

**Constructing a Chain of Nodes**

**Singly-Linked List: Setting a List's Head**

**Singly-Linked List: Counting # of Nodes (1)**

**Singly-Linked List: Counting # of Nodes (2)**

**Singly-Linked List: Finding the Tail (1)**

**Singly-Linked List: Finding the Tail (2)**

**Singly-Linked List: Can We Do Better?**

**Singly-Linked List: Inserting to the Front (1)**

**Singly-Linked List: Inserting to the Front (2)**

# Index (3)

**Exercise**

**Exercise**

**Singly-Linked List: Accessing the Middle (1)**

**Singly-Linked List: Accessing the Middle (2)**

**Singly-Linked List: Accessing the Middle (3)**

**Singly-Linked List: Inserting to the Middle (1)**

**Singly-Linked List: Inserting to the Middle (2)**

**Singly-Linked List: Removing from the End**

**Singly-Linked List: Exercises**

**Exercise**

**Arrays vs. Singly-Linked Lists**

# Index (4)

**Background Study: Generics in Java**

**Generic Classes: Singly-Linked List (1)**

**Generic Classes: Singly-Linked List (2)**

**Generic Classes: Singly-Linked List (3)**

**Basic DS: Doubly-Linked Lists (1.1)**

**Basic DS: Doubly-Linked Lists (1.2)**

**Basic DS: Doubly-Linked Lists (2.1)**

**Basic DS: Doubly-Linked Lists (2.2)**

**DLNs and DLLs: prev, header, trailer**

**Generic Doubly-Linked Lists in Java (1)**

**Generic Doubly-Linked Lists in Java (2)**

## Index (5)

**Doubly-Linked List: Insertions**

**Doubly-Linked List: Inserting to Front/End**

**Doubly-Linked List: Inserting to Middle**

**Doubly-Linked List: Removals**

**Doubly-Linked List: Removing from Front/End**

**Doubly-Linked List: Removing from Middle**

**Reference Node:**

**To be Given or Not to be Given**

**Arrays vs. (Singly- and Doubly-Linked) Lists**

**Beyond this lecture ...**

# Abstract Data Types (ADTs), Stacks, Queues



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

CHEN-WEI WANG

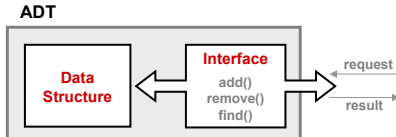
# Learning Outcomes of this Lecture

This module is designed to help you learn about:

- The notion of *Abstract Data Types (ADTs)*
- **ADTs**: Stack vs. Queue
- Implementing Stack and Queue in Java [ interface, classes ]
- Applications of Stacks vs. Queues
- *Circular* Arrays
- Optional (but highly **encouraged**):
  - Criterion of *Modularity*, Modular Design
  - *Dynamic* Arrays, *Amortized* Analysis

# Abstract Data Types (ADTs)

- Given a problem, decompose its solution into **modules**.
- Each **module** implements an **abstract data type (ADT)**:
  - filters out **irrelevant** details
  - contains a list of declared **data** and well-specified **operations**



- Supplier's **Obligations**:
  - Implement all operations
  - Choose the **"right"** data structure [ e.g., arrays vs. SLL vs. DLL ]
  - The internal details of an implemented **ADT** should be **hidden**.
- Client's **Benefits**:
  - Correct** output
  - Efficient** performance

# Java API Approximates ADTs (1)

## Interface List<E>

### Type Parameters:

E - the type of elements in this list

### All Superinterfaces:

Collection<E>, Iterable<E>

### All Known Implementing Classes:

AbstractList, AbstractSequentialList, ArrayList, AttributeList, CopyOnWriteArrayList, LinkedList, RoleList, RoleUnresolvedList, Stack, Vector

```
public interface List<E>  
    extends Collection<E>
```

An ordered collection (also known as a *sequence*). The user of this interface has precise control over where in the list each element is inserted. The user can access elements by their integer index (position in the list), and search for elements in the list.

It is useful to have:

- A **generic collection class** where the **homogeneous type** of elements are parameterized as E.
- A reasonably **intuitive overview** of the ADT.

# Java API Approximates ADTs (2)

**E**                      `set(int index, E element)`  
 Replaces the element at the specified position in this list with the specified element (optional operation).

## set

`E set(int index,  
       E element)`

Replaces the element at the specified position in this list with the specified element (optional operation).

### Parameters:

`index` - index of the element to replace

`element` - element to be stored at the specified position

### Returns:

the element previously at the specified position

### Throws:

`UnsupportedOperationException` - if the set operation is not supported by this list

`ClassCastException` - if the class of the specified element prevents it from being added to this list

`NullPointerException` - if the specified element is null and this list does not permit null elements

`IllegalArgumentException` - if some property of the specified element prevents it from being added to this list

`IndexOutOfBoundsException` - if the index is out of range (`index < 0 || index >= size()`)

Methods described in a *natural language* can be *ambiguous*.

# Building ADTs for Reusability

- **ADTs** are *reusable software components* that are common for solving many real-world problems.  
e.g., Stacks, Queues, Lists, Tables, Trees, Graphs
- An **ADT**, once thoroughly tested, can be reused by:
  - *Clients* of Applications
  - *Suppliers* of other ADTs
- As a supplier, you are obliged to:
  - *Implement* standard ADTs [ ≈ lego building bricks ]  
**Note.** Recall the basic data structures: arrays vs. SLLs vs. DLLs
  - *Design* algorithms using standard ADTs [ ≈ lego houses, ships ]
- For each standard **ADT**, you should know its **interface**:
  - Stored *data*
  - For each *operation* manipulating the stored data
    - How are *clients* supposed to use the method? [ **preconditions** ]
    - What are the services provided by *suppliers*? [ **postconditions** ]
    - Time (and sometimes space) *complexity*

# What is a Stack?

- A **stack** is a collection of objects.
- Objects in a **stack** are inserted and removed according to the **last-in, first-out (LIFO)** principle.
  - **Cannot** access arbitrary elements of a stack
  - **Can** only access or remove the **most-recently added** element



# The Stack ADT

- *top*

[ *precondition*: stack is not empty ]

[ *postcondition*: return item last pushed to the stack ]

- *size*

[ *precondition*: none ]

[ *postcondition*: return number of items pushed to the stack ]

- *isEmpty*

[ *precondition*: none ]

[ *postcondition*: return whether there is no item in the stack ]

- *push(item)*

[ *precondition*: stack is not full ]

[ *postcondition*: push the input item onto the top of the stack ]

- *pop*

[ *precondition*: stack is not empty ]

[ *postcondition*: remove and return the top of stack ]

# Stack: Illustration

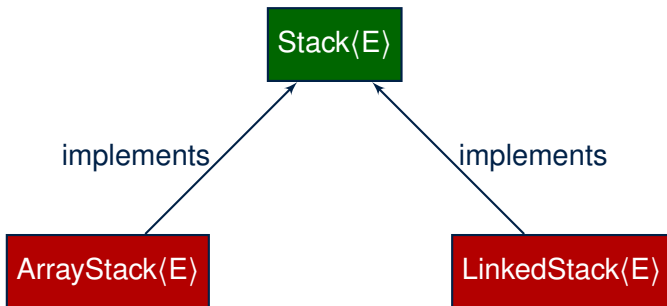
OPERATION	RETURN VALUE	STACK CONTENTS
—	—	∅
isEmpty	<i>true</i>	∅
push(5)	—	5
push(3)	—	3 5
push(1)	—	1 3 5
size	3	1 3 5
top	1	1 3 5
pop	1	3 5
pop	3	5
pop	5	∅

# Generic Stack: Interface

```
public interface Stack<E> {  
    public int size();  
    public boolean isEmpty();  
    public E top();  
    public void push(E e);  
    public E pop();  
}
```

The **Stack** ADT, declared as an **interface**, allows **alternative implementations** to conform to its method headers.

# Generic Stack: Architecture



# Implementing Stack: Array (1)

```
public class ArrayStack<E> implements Stack<E> {
    private final int MAX_CAPACITY = 1000;
    private E[] data;
    private int t; /* index of top */
    public ArrayStack() {
        data = (E[]) new Object[MAX_CAPACITY];
        t = -1;
    }

    public int size() { return (t + 1); }
    public boolean isEmpty() { return (t == -1); }

    public E top() {
        if (isEmpty()) { /* Precondition Violated */ }
        else { return data[t]; }
    }
    public void push(E e) {
        if (size() == MAX_CAPACITY) { /* Precondition Violated */ }
        else { t++; data[t] = e; }
    }
    public E pop() {
        E result;
        if (isEmpty()) { /* Precondition Violated */ }
        else { result = data[t]; data[t] = null; t--; }
        return result;
    }
}
```

# Implementing Stack: Array (2)

- Running Times of *Array*-Based *Stack* Operations?

<i>ArrayStack</i> Method	Running Time
size	$O(1)$
isEmpty	$O(1)$
top	$O(1)$
push	$O(1)$
pop	$O(1)$

- Exercise** This version of implementation treats the *end* of array as the *top* of stack. Would the RTs of operations change if we treated the *beginning* of array as the *top* of stack?
- Q.** What if the preset capacity turns out to be insufficient?  
**A.** `IllegalArgumentException` occurs and it takes  $O(1)$  time to respond.
- At the end, we will explore the alternative of a *dynamic array*.

# Implementing Stack: Singly-Linked List (1)

```
public class LinkedStack<E> implements Stack<E> {
    private SinglyLinkedList<E> list;
    ...
}
```

## Question:

Stack Method	Singly-Linked List Method	
	Strategy 1	Strategy 2
size	list.size	
isEmpty	list.isEmpty	
top	list.first	list.last
push	list.addFirst	list.addLast
pop	list.removeFirst	list.removeLast

Which **implementation strategy** should be chosen?

## Implementing Stack: Singly-Linked List (2)

- If the *front of list* is treated as the *top of stack*, then:
  - All stack operations remain  $O(1)$  [  $\therefore$  removeFirst takes  $O(1)$  ]
- If the *end of list* is treated as the *top of stack*, then:
  - The *pop* operation takes  $O(n)$  [  $\therefore$  removeLast takes  $O(n)$  ]
- But in both cases, given that a linked, *dynamic* structure is used, **no resizing** is necessary!

# Generic Stack: Testing Implementations

```
@Test
public void testPolymorphicStacks() {
    Stack<String> s = new ArrayStack<>();
    s.push("Alan"); /* dynamic binding */
    s.push("Mark"); /* dynamic binding */
    s.push("Tom"); /* dynamic binding */
    assertTrue(s.size() == 3 && !s.isEmpty());
    assertEquals("Tom", s.top());

    s = new LinkedStack<>();
    s.push("Alan"); /* dynamic binding */
    s.push("Mark"); /* dynamic binding */
    s.push("Tom"); /* dynamic binding */
    assertTrue(s.size() == 3 && !s.isEmpty());
    assertEquals("Tom", s.top());
}
```

# Polymorphism & Dynamic Binding

```
1 Stack<String> myStack;  
2 myStack = new ArrayStack<String>();  
3 myStack.push("Alan");  
4 myStack = new LinkedStack<String>();  
5 myStack.push("Alan");
```

- **Polymorphism**

An object may change its “**shape**” (i.e., **dynamic type**) at runtime.

Which lines? 2, 4

- **Dynamic Binding**

Effect of a method call depends on the “**current shape**” of the target object.

Which lines? 3, 5

# Stack Application: Reversing an Array

- **Implementing** a **generic** algorithm:

```
public static <E> void reverse(E[] a) {  
    Stack<E> buffer = new ArrayStack<E>();  
    for (int i = 0; i < a.length; i++) {  
        buffer.push(a[i]);  
    }  
    for (int i = 0; i < a.length; i++) {  
        a[i] = buffer.pop();  
    }  
}
```

- **Testing** the **generic** algorithm:

```
@Test  
public void testReverseViaStack() {  
    String[] names = {"Alan", "Mark", "Tom"};  
    String[] expectedReverseOfNames = {"Tom", "Mark", "Alan"};  
    StackUtilities.reverse(names);  
    assertEquals(expectedReverseOfNames, names);  
  
    Integer[] numbers = {46, 23, 68};  
    Integer[] expectedReverseOfNumbers = {68, 23, 46};  
    StackUtilities.reverse(numbers);  
    assertEquals(expectedReverseOfNumbers, numbers);  
}
```

# Stack Application: Matching Delimiters (1)

- **Problem**

Opening delimiters: (, [, {

Closing delimiters: ), ], }

e.g., **Correct**: () (()) { ([() ] ) }

e.g., **Incorrect**: ( { [ ] ) }

- **Sketch of Solution**

- When a new **opening** delimiter is found, **push** it to the stack.
- **Most-recently** found delimiter should be matched first.
- When a new **closing** delimiter is found:
  - If it matches the **top** of the stack, then **pop** off the stack.
  - Otherwise, an error is found!
- Finishing reading the input, an empty stack means a success!

# Stack Application: Matching Delimiters (2)

- **Implementing** the algorithm:

```
public static boolean isMatched(String expression) {
    final String opening = "[{(";
    final String closing = ")]}";
    Stack<Character> openings = new LinkedStack<Character>();
    int i = 0;
    boolean foundError = false;
    while (!foundError && i < expression.length()) {
        char c = expression.charAt(i);
        if (opening.indexOf(c) != -1) { openings.push(c); }
        else if (closing.indexOf(c) != -1) {
            if (openings.isEmpty()) { foundError = true; }
            else {
                if (opening.indexOf(openings.top()) == closing.indexOf(c)) { openings.pop(); }
                else { foundError = true; } }
        }
        i++;
    }
    return !foundError && openings.isEmpty();
}
```

- **Testing** the algorithm:

```
@Test
public void testMatchingDelimiters() {
    assertTrue(StackUtilities.isMatched(""));
    assertTrue(StackUtilities.isMatched("{}{}({})");
    assertFalse(StackUtilities.isMatched("{}[]{}");
    assertFalse(StackUtilities.isMatched("{}[]{}");
    assertFalse(StackUtilities.isMatched("{}[]{}");
}
```

# Stack Application: Postfix Notations (1)

**Problem:** Given a postfix expression, calculate its value.

Infix Notation	Postfix Notation
Operator <i>in-between</i> Operands	Operator <i>follows</i> Operands
Parentheses force precedence	Order of evaluation embedded
3	3
3 + 4	3 4 +
3 + 4 + 5	3 4 + 5 +
3 + (4 + 5)	3 4 5 + +
3 - 4 * 5	3 4 5 * -
(3 - 4) * 5	3 4 - 5 *

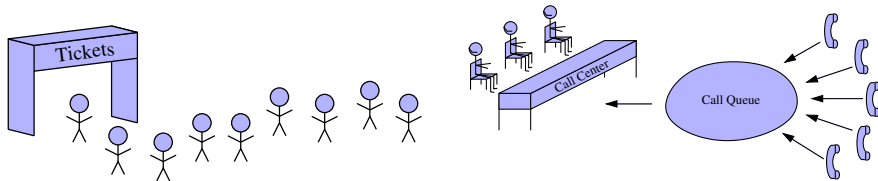
# Stack Application: Postfix Notations (2)

## Sketch of Solution

- When input is an **operand** (i.e., a number), **push** it to the stack.
- When input is an **operator**, obtain its two **operands** by **popping** off the stack **twice**, evaluate, then **push** the result back to stack.
- When finishing reading the input, there should be **only one** number left in the stack.
- **Error** if:
  - Not enough items left in the stack for the operator [ e.g.,  $523++$  ]
  - When finished, two or more numbers left in stack [ e.g.,  $53+6$  ]

# What is a Queue?

- A **queue** is a collection of objects.
- Objects in a **queue** are inserted and removed according to the **first-in, first-out (FIFO)** principle.
  - Each new element joins at the **back/end** of the queue.
  - **Cannot** access arbitrary elements of a queue
  - **Can** only access or remove the **least-recently inserted (or longest-waiting)** element



# The Queue ADT

- *first* ≈ *top* of stack  
[ *precondition*: queue is not empty ]  
[ *postcondition*: return item first enqueued ]
- *size*  
[ *precondition*: none ]  
[ *postcondition*: return number of items enqueued ]
- *isEmpty*  
[ *precondition*: none ]  
[ *postcondition*: return whether there is no item in the queue ]
- *enqueue(item)* ≈ *push* of stack  
[ *precondition*: queue is not full ]  
[ *postcondition*: enqueue item as the “last” of the queue ]
- *dequeue* ≈ *pop* of stack  
[ *precondition*: queue is not empty ]  
[ *postcondition*: remove and return the first of the queue ]

# Queue: Illustration

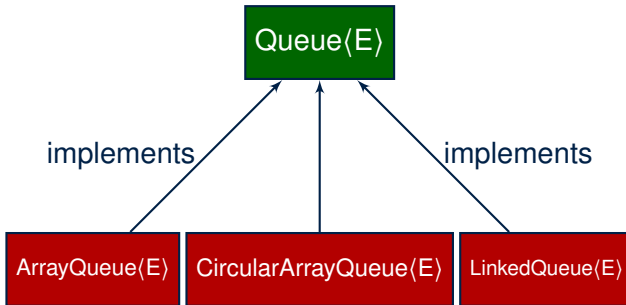
Operation	Return Value	Queue Contents
—	—	$\emptyset$
isEmpty	<i>true</i>	$\emptyset$
enqueue(5)	—	(5)
enqueue(3)	—	(5, 3)
enqueue(1)	—	(5, 3, 1)
size	3	(5, 3, 1)
dequeue	5	(3, 1)
dequeue	3	1
dequeue	1	$\emptyset$

# Generic Queue: Interface

```
public interface Queue<E> {  
    public int size();  
    public boolean isEmpty();  
    public E first();  
    public void enqueue(E e);  
    public E dequeue();  
}
```

The *Queue* ADT, declared as an *interface*, allows *alternative implementations* to conform to its method headers.

# Generic Queue: Architecture



# Implementing Queue ADT: Array (1)

```
public class ArrayQueue<E> implements Queue<E> {
    private final int MAX_CAPACITY = 1000;
    private E[] data;
    private int r; /* rear index */
    public ArrayQueue() {
        data = (E[]) new Object[MAX_CAPACITY];
        r = -1;
    }
    public int size() { return (r + 1); }
    public boolean isEmpty() { return (r == -1); }
    public E first() {
        if (isEmpty()) { /* Precondition Violated */ }
        else { return data[0]; }
    }
    public void enqueue(E e) {
        if (size() == MAX_CAPACITY) { /* Precondition Violated */ }
        else { r++; data[r] = e; }
    }
    public E dequeue() {
        if (isEmpty()) { /* Precondition Violated */ }
        else {
            E result = data[0];
            for (int i = 0; i < r; i++) { data[i] = data[i + 1]; }
            data[r] = null; r--;
            return result;
        }
    }
}
```

# Implementing Queue ADT: Array (2)

- Running Times of *Array*-Based *Queue* Operations?

<i>ArrayQueue</i> Method	Running Time
size	$O(1)$
isEmpty	$O(1)$
first	$O(1)$
enqueue	$O(1)$
dequeue	$O(n)$

- Exercise** This version of implementation treats the *beginning* of array as the *first* of queue. Would the RTs of operations change if we treated the *end* of array as the *first* of queue?
- Q.** What if the preset capacity turns out to be insufficient?  
**A.** `IllegalArgumentException` occurs and it takes  $O(1)$  time to respond.
- At the end, we will explore the alternative of a *dynamic array*.

# Implementing Queue: Singly-Linked List (1)

```
public class LinkedListQueue<E> implements Queue<E> {
    private SinglyLinkedList<E> list;
    ...
}
```

## Question:

Queue Method	Singly-Linked List Method	
	Strategy 1	Strategy 2
size	list.size	
isEmpty	list.isEmpty	
first	list.first	list.last
enqueue	list.addLast	list.addFirst
dequeue	list.removeFirst	list.removeLast

Which **implementation strategy** should be chosen?

## Implementing Queue: Singly-Linked List (2)

- If the *front of list* is treated as the *first of queue*, then:
  - All queue operations remain  $O(1)$  [  $\because$  removeFirst takes  $O(1)$  ]
- If the *end of list* is treated as the *first of queue*, then:
  - The *dequeue* operation takes  $O(n)$  [  $\because$  removeLast takes  $O(n)$  ]
- But in both cases, given that a linked, *dynamic* structure is used, **no resizing** is necessary!

# Generic Queue: Testing Implementations

```
@Test
public void testPolymorphicQueues() {
    Queue<String> q = new ArrayQueue<>();
    q.enqueue("Alan"); /* dynamic binding */
    q.enqueue("Mark"); /* dynamic binding */
    q.enqueue("Tom"); /* dynamic binding */
    assertTrue(q.size() == 3 && !q.isEmpty());
    assertEquals("Alan", q.first());

    q = new LinkedQueue<>();
    q.enqueue("Alan"); /* dynamic binding */
    q.enqueue("Mark"); /* dynamic binding */
    q.enqueue("Tom"); /* dynamic binding */
    assertTrue(q.size() == 3 && !q.isEmpty());
    assertEquals("Alan", q.first());
}
```

# Polymorphism & Dynamic Binding

```
1 Queue<String> myQueue;  
2 myQueue = new CircularArrayQueue<String>();  
3 myQueue.enqueue("Alan");  
4 myQueue = new LinkedList<String>();  
5 myQueue.enqueue("Alan");
```

- **Polymorphism**

An object may change its “**shape**” (i.e., *dynamic type*) at runtime.

Which lines? 2, 4

- **Dynamic Binding**

Effect of a method call depends on the “*current shape*” of the target object.

Which lines? 3, 5

## Exercise: Implementing a Queue using Two Stacks

```
public class StackQueue<E> implements Queue<E> {  
    private Stack<E> inStack;  
    private Stack<E> outStack;  
    ...  
}
```

- For **size**, add up sizes of inStack and outStack.
- For **isEmpty**, are inStack and outStack both empty?
- For **enqueue**, **push** to inStack.
- For **dequeue**:
  - **pop** from outStack  
If outStack is empty, we need to first **pop** all items from inStack and **push** them to outStack.

Exercise: Why does this work? [ **implement** and **test** ]

Exercise: Running Time? [ see analysis on **dynamic arrays** ]

# Implementing Queue ADT: Circular Array (1)

- Maintain two indices:  $f$  for *front*;  $r$  for *next available slot*.
- Maximum size:**  $N - 1$  [  $N = \text{data.length}$  ]
- Empty Queue:** when  $r = f$



- Full Queue:** when  $(r + 1) \% N = f$

- When  $r > f$ :



- When  $r < f$ :



- Size of Queue:**

- If  $r = f$ : 0

- If  $r > f$ :  $r - f$



- If  $r < f$ :  $r + (N - f)$



# Implementing Queue ADT: Circular Array (2)



Running Times of *CircularArray*-Based **Queue** Operations?

<i>CircularArrayQueue</i> Method	Running Time
size	$O(1)$
isEmpty	$O(1)$
first	$O(1)$
enqueue	$O(1)$
dequeue	<b><math>O(1)</math></b>

**Exercise:** Create a Java class `CircularArrayQueue` that implements the `Queue` interface using a *circular array*.

# Limitations of Queue

- Say we use a *queue* to implement a *waiting list*.
  - What if we *dequeue* the front customer, but find that we need to *put them back to the front* (e.g., seat is still not available, the table assigned is not satisfactory, *etc.*)?
  - What if the customer at the end of the queue decides not to wait and leave, how do we *remove them from the end of the queue*?
- **Solution:** A new ADT extending the *Queue* by supporting:
  - *insertion* to the *front*
  - *deletion* from the *end*

# The Double-Ended Queue ADT

- **Double-Ended Queue** (or **Deque**) is a queue-like data structure that supports **insertion** and **deletion** at both the **front** and the **end** of the queue.

```
public interface Deque<E> {  
    /* Queue operations */  
    public int size();  
    public boolean isEmpty();  
    public E first();  
    public void addLast(E e); /* enqueue */  
    public E removeFirst(); /* dequeue */  
    /* Extended operations */  
    public void addFirst(E e);  
    public E removeLast();  
}
```

- **Exercise**: Implement **Deque** using a **circular array**.
- **Exercise**: Implement **Deque** using a **SLL** and/or **DLL**.

# Optional Materials

*These topics are useful for your knowledge about  
**ADTs, stacks, and Queues.***

You are **encouraged** to follow through these online lectures:

[https://www.eecs.yorku.ca/~jackie/teaching/  
lectures/index.html#EECS2011\\_W22](https://www.eecs.yorku.ca/~jackie/teaching/lectures/index.html#EECS2011_W22)

- **Design by Contract** and **Modularity**
  - Week 5: Lecture 3, Parts A2 - A3
- **Dynamic Arrays** and **Amortized Analysis**
  - Week 6: Lecture 3, Parts E1 - E5

# Terminology: Contract, Client, Supplier

- A **supplier** implements/provides a service (e.g., microwave).
- A **client** uses a service provided by some supplier.
  - The client is required to follow certain instructions to obtain the service (e.g., supplier **assumes** that client powers on, closes door, and heats something that is not explosive).
  - If instructions are followed, the client would **expect** that the service does what is guaranteed (e.g., a lunch box is heated).
  - The client does not care how the supplier implements it.
- What are the **benefits** and **obligations** of the two parties?

	<b>benefits</b>	<b>obligations</b>
CLIENT	obtain a service	follow instructions
SUPPLIER	assume instructions followed	provide a service

- There is a **contract** between two parties, violated if:
  - The instructions are not followed. [ Client's fault ]
  - Instructions followed, but service not satisfactory. [ Supplier's fault ]

# Client, Supplier, Contract in OOP (1)

```
class Microwave {
    private boolean on;
    private boolean locked;
    void power() {on = true;}
    void lock() {locked = true;}
    void heat(Object stuff) {
        /* Assume: on && locked */
        /* stuff not explosive. */
    }
}
```

```
class MicrowaveUser {
    public static void main(...) {
        Microwave m = new Microwave();
        Object obj = ???;
        m.power(); m.lock();
        m.heat(obj);
    }
}
```

Method call **m.heat(obj)** indicates a client-supplier relation.

- **Client:** resident class of the method call [MicrowaveUser]
- **Supplier:** type of context object (or call target) **m** [Microwave]

## Client, Supplier, Contract in OOP (2)

```
class Microwave {
    private boolean on;
    private boolean locked;
    void power() {on = true;}
    void lock() {locked = true;}
    void heat(Object stuff) {
        /* Assume: on && locked */
        /* stuff not explosive. */}}

```

```
class MicrowaveUser {
    public static void main(...) {
        Microwave m = new Microwave();
        Object obj = [???];
        m.power(); m.lock();
        m.heat(obj);
    }
}

```

- The **contract** is *honoured* if:

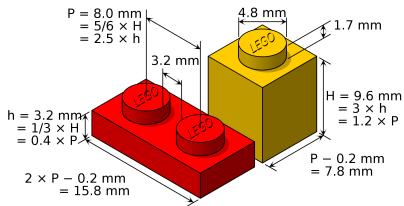
Right **before** the method call :

- State of `m` is as assumed: `m.on==true` and `m.locked==ture`
- The input argument `obj` is valid (i.e., not explosive).

Right **after** the method call : `obj` is properly heated.

- If any of these fails, there is a **contract violation**.
  - `m.on` or `m.locked` is false  $\Rightarrow$  MicrowaveUser's fault.
  - `obj` is an explosive  $\Rightarrow$  MicrowaveUser's fault.
  - A fault from the client is identified  $\Rightarrow$  Method call will not start.
  - Method executed but `obj` not properly heated  $\Rightarrow$  Microwave's fault

# Modularity (1): Childhood Activity



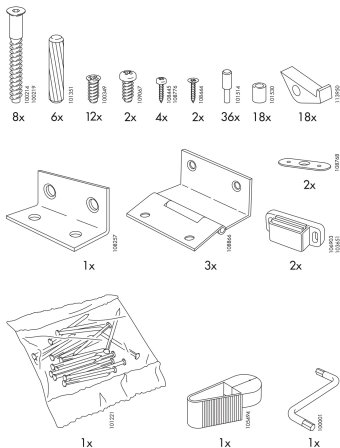
(INTERFACE) SPECIFICATION



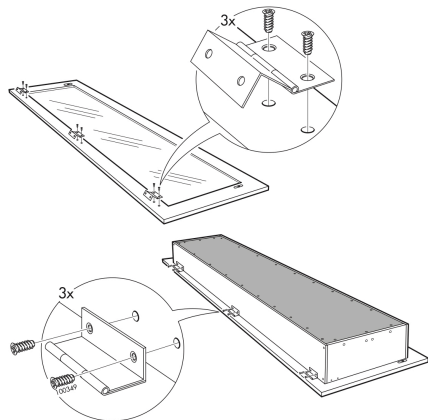
(ASSEMBLY) ARCHITECTURE

Sources: <https://commons.wikimedia.org> and <https://www.wish.com>

# Modularity (2): Daily Construction



(INTERFACE) SPECIFICATION

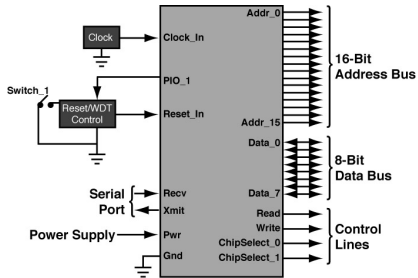


(ASSEMBLY) ARCHITECTURE

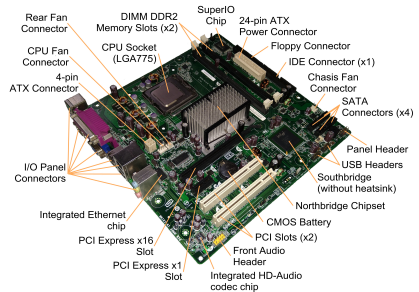
Source: <https://usermanual.wiki/>

# Modularity (3): Computer Architecture

*Motherboards* are built from functioning units (e.g., *CPUs*).



(INTERFACE) SPECIFICATION

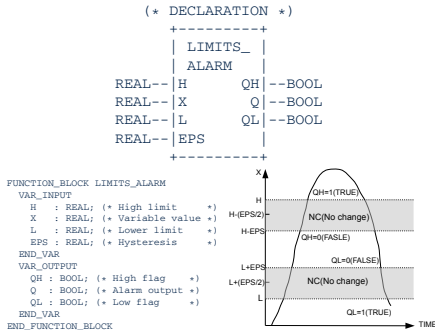


(ASSEMBLY) ARCHITECTURE

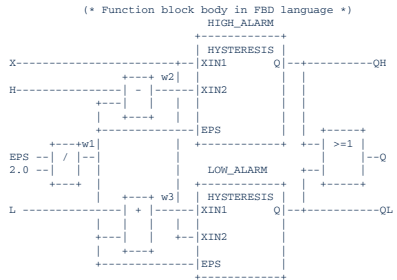
Sources: [www.embeddedlinux.org.cn](http://www.embeddedlinux.org.cn) and <https://en.wikipedia.org>

# Modularity (4): System Development

Safety-critical systems (e.g., *nuclear shutdown systems*) are built from *function blocks*.



(INTERFACE) SPECIFICATION

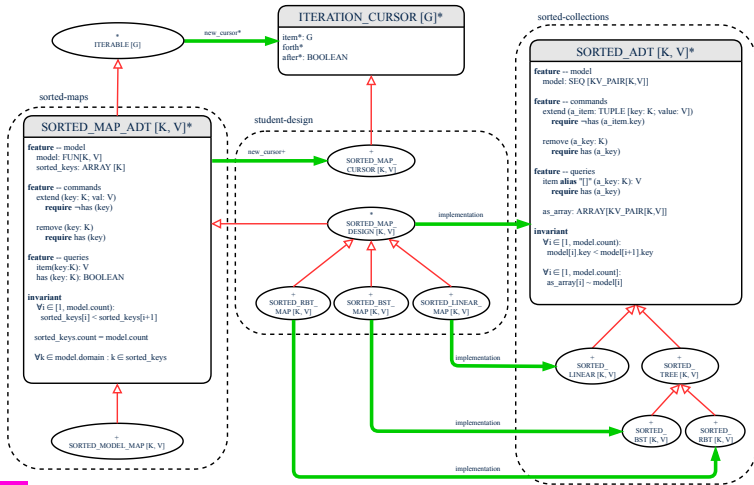


(ASSEMBLY) ARCHITECTURE

Sources: <https://plcopen.org/iec-61131-3>

# Modularity (5): Software Design

Software systems are composed of *well-specified classes*.



# Design Principle: Modularity

- **Modularity** refers to a sound quality of your design:
  1. **Divide** a given complex **problem** into inter-related **sub-problems** via a logical/justifiable functional decomposition.  
e.g., In designing a game, solve sub-problems of: 1) rules of the game; 2) actor characterizations; and 3) presentation.
  2. **Specify** each **sub-solution** as a **module** with a clear **interface**: inputs, outputs, and input-output relations.
    - The UNIX principle: Each command does one thing and does it well.
    - In object-oriented design (OOD), each class serves as a module.
  3. **Conquer** original **problem** by assembling **sub-solutions**.
    - In OOD, classes are assembled via client-supplier relations (aggregations or compositions) or inheritance relations.
- A **modular design** satisfies the criterion of modularity and is:
  - **Maintainable**: fix issues by changing the relevant modules only.
  - **Extensible**: introduce new functionalities by adding new modules.
  - **Reusable**: a module may be used in different compositions
- Opposite of modularity: A **superman module** doing everything.

# Array Implementations: Stack and Queue

- When implementing *stack* and *queue* via *arrays*, we imposed a maximum capacity:

```
public class ArrayStack<E> implements Stack<E> {  
    private final int MAX_CAPACITY = 1000;  
    private E[] data;  
    ...  
    public void push(E e) {  
        if (size() == MAX_CAPACITY) { /* Precondition Violated */ }  
        else { ... }  
    }  
    ...  
}
```

```
public class ArrayQueue<E> implements Queue<E> {  
    private final int MAX_CAPACITY = 1000;  
    private E[] data;  
    ...  
    public void enqueue(E e) {  
        if (size() == MAX_CAPACITY) { /* Precondition Violated */ }  
        else { ... }  
    }  
    ...  
}
```

- This made the *push* and *enqueue* operations both cost  $O(1)$ .

# Dynamic Array: Constant Increments

Implement **stack** using a **dynamic array** resizing itself by a constant increment:

```

1 public class ArrayStack<E> implements Stack<E> {
2     private int I;
3     private int C;
4     private int capacity;
5     private E[] data;
6     public ArrayStack() {
7         I = 1000; /* arbitrary initial size */
8         C = 500; /* arbitrary fixed increment */
9         capacity = I;
10        data = (E[]) new Object[capacity];
11        t = -1;
12    }
13    public void push(E e) {
14        if (size() == capacity) {
15            /* resizing by a fixed constant */
16            E[] temp = (E[]) new Object[capacity + C];
17            for(int i = 0; i < capacity; i++) {
18                temp[i] = data[i];
19            }
20            data = temp;
21            capacity = capacity + C
22        }
23        t++;
24        data[t] = e;
25    }
26 }

```

- This alternative strategy **resizes** the array, whenever needed, by a **constant** amount.
- L17 – L19 make **push** cost  **$O(n)$** , in the **worst case**.
- However, given that **resizing** only happens rarely, how about the average running time?
- We will refer L14 – L22 as the **resizing** part and L23 – L24 as the **update** part.

# Dynamic Array: Doubling

Implement **stack** using a **dynamic array** resizing itself by doubling:

```

1 public class ArrayStack<E> implements Stack<E> {
2     private int I;
3     private int capacity;
4     private E[] data;
5     public ArrayStack() {
6         I = 1000; /* arbitrary initial size */
7         capacity = I;
8         data = (E[]) new Object[capacity];
9         t = -1;
10    }
11    public void push(E e) {
12        if (size() == capacity) {
13            /* resizing by doubling */
14            E[] temp = (E[]) new Object[capacity * 2];
15            for(int i = 0; i < capacity; i++) {
16                temp[i] = data[i];
17            }
18            data = temp;
19            capacity = capacity * 2
20        }
21        t++;
22        data[t] = e;
23    }
24 }

```

- This alternative strategy **resizes** the array, whenever needed, by **doubling** its current size.
- L15 – L17 make **push** cost  **$O(n)$** , in the worst case.
- However, given that **resizing** only happens rarely, how about the average running time?
- We will refer L12 – L20 as the resizing part and L21 – L22 as the update part.

# Avg. RT: Const. Increment vs. Doubling

- Without loss of generality, assume: There are  $n$  **push** operations, and the **last push** triggers the **last resizing** routine.

	Constant Increments	Doubling
RT of exec. <u>update</u> part for $n$ pushes	$O(n)$	
RT of executing 1st <u>resizing</u>	$I$	
RT of executing 2nd <u>resizing</u>	$I + C$	$2 \cdot I$
RT of executing 3rd <u>resizing</u>	$I + 2 \cdot C$	$4 \cdot I$
RT of executing 4th <u>resizing</u>	$I + 3 \cdot C$	$8 \cdot I$
RT of executing $k^{\text{th}}$ <u>resizing</u>	$I + (k - 1) \cdot C$	$2^{k-1} \cdot I$
RT of executing last <u>resizing</u>	$n$	
# of <u>resizing</u> needed (solve $k$ for $RT = n$ )	$O(n)$	$O(\log_2 n)$
Total RT for $n$ pushes	$O(n^2)$	$O(n)$
Amortized/Average RT over $n$ pushes	<b><math>O(n)</math></b>	<b><math>O(1)</math></b>

- Over  $n$  push operations, the **amortized** / **average** running time of the **doubling** strategy is more efficient.

## Beyond this lecture ...

---



- Attempt the exercises throughout the lecture.
- Implement the *Postfix Calculator* using a stack.

# Index (1)

**Learning Outcomes of this Lecture**

**Abstract Data Types (ADTs)**

**Java API Approximates ADTs (1)**

**Java API Approximates ADTs (2)**

**Building ADTs for Reusability**

**What is a Stack?**

**The Stack ADT**

**Stack: Illustration**

**Generic Stack: Interface**

**Generic Stack: Architecture**

**Implementing Stack: Array (1)**

## Index (2)

**Implementing Stack: Array (2)**

**Implementing Stack: Singly-Linked List (1)**

**Implementing Stack: Singly-Linked List (2)**

**Generic Stack: Testing Implementations**

**Polymorphism & Dynamic Binding**

**Stack Application: Reversing an Array**

**Stack Application: Matching Delimiters (1)**

**Stack Application: Matching Delimiters (2)**

**Stack Application: Postfix Notations (1)**

**Stack Application: Postfix Notations (2)**

**What is a Queue?**

# Index (3)

**The Queue ADT**

**Queue: Illustration**

**Generic Queue: Interface**

**Generic Queue: Architecture**

**Implementing Queue ADT: Array (1)**

**Implementing Queue ADT: Array (2)**

**Implementing Queue: Singly-Linked List (1)**

**Implementing Queue: Singly-Linked List (2)**

**Generic Queue: Testing Implementations**

**Polymorphism & Dynamic Binding**

## Index (4)

**Exercise:**

**Implementing a Queue using Two Stacks**

**Implementing Queue ADT: Circular Array (1)**

**Implementing Queue ADT: Circular Array (2)**

**Limitations of Queue**

**The Double-Ended Queue ADT**

**Optional Materials**

**Terminology: Contract, Client, Supplier**

**Client, Supplier, Contract in OOP (1)**

**Client, Supplier, Contract in OOP (2)**

**Modularity (1): Childhood Activity**

## Index (5)

**Modularity (2): Daily Construction**

**Modularity (3): Computer Architecture**

**Modularity (4): System Development**

**Modularity (5): Software Design**

**Design Principle: Modularity**

**Array Implementations: Stack and Queue**

**Dynamic Array: Constant Increments**

**Dynamic Array: Doubling**

**Avg. RT: Const. Increment vs. Doubling**

**Beyond this lecture ...**

# Recursion (Part 2)



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

CHEN-WEI WANG

# Learning Outcomes of this Lecture

This module is designed to help you:

- Learn about the more intermediate *recursive algorithms*:
  - Binary Search
  - Merge Sort
  - Quick Sort

# Recursion: Binary Search (1)

- **Searching Problem**

Given a numerical key  $k$  and an array  $a$  of  $n$  numbers:

**Precondition:** Input array  $a$  **sorted** in a non-descending order  
i.e.,  $a[0] \leq a[1] \leq \dots \leq a[n-1]$

**Postcondition:** Return whether or not  $k$  exists in the input array  $a$ .

- **Q.** RT of a search on an **unsorted** array?

**A.**  $O(n)$  (despite being iterative or recursive)

- **A Recursive Solution**

**Base Case:** Empty array  $\rightarrow$  **false**.

**Recursive Case:** Array of size  $\geq 1 \rightarrow$

- Compare the **middle** element of array  $a$  against key  $k$ .
  - All elements to the left of **middle** are  $\leq k$
  - All elements to the right of **middle** are  $\geq k$
- If the **middle** element **is** equal to key  $k \rightarrow$  **true**
- If the **middle** element **is not** equal to key  $k$ :
  - If  $k < \text{middle}$ , recursively **search** key  $k$  on the left half.
  - If  $k > \text{middle}$ , recursively **search** key  $k$  on the right half.

## Recursion: Binary Search (2)

```
boolean binarySearch(int[] sorted, int key) {  
    return binarySearchH(sorted, 0, sorted.length - 1, key);  
}  
boolean binarySearchH(int[] sorted, int from, int to, int key) {  
    if (from > to) { /* base case 1: empty range */  
        return false; }  
    else if (from == to) { /* base case 2: range of one element */  
        return sorted[from] == key; }  
    else {  
        int middle = (from + to) / 2;  
        int middleValue = sorted[middle];  
        if (key < middleValue) {  
            return binarySearchH(sorted, from, middle - 1, key);  
        }  
        else if (key > middleValue) {  
            return binarySearchH(sorted, middle + 1, to, key);  
        }  
        else { return true; }  
    }  
}
```

# Running Time: Binary Search (1)

We define  $T(n)$  as the *running time function* of a *binary search*, where  $n$  is the size of the input array.

$$\begin{cases} T(0) &= 1 \\ T(1) &= 1 \\ T(n) &= T(\frac{n}{2}) + 1 \quad \text{where } n \geq 2 \end{cases}$$

To solve this recurrence relation, we study the pattern of  $T(n)$  and observe how it reaches the *base case(s)*.

## Running Time: Binary Search (2)

*Without loss of generality*, assume  $n = 2^i$  for some  $i \geq 0$ .

$$\begin{aligned}
 T(n) &= T\left(\frac{n}{2}\right) + 1 \\
 &= \underbrace{\left(T\left(\frac{n}{4}\right) + 1\right)}_{T\left(\frac{n}{2}\right)} + \underbrace{1}_{1 \text{ time}} \\
 &= \underbrace{\left(\left(T\left(\frac{n}{8}\right) + 1\right) + 1\right)}_{T\left(\frac{n}{4}\right)} + \underbrace{1}_{2 \text{ times}} \\
 &= \dots \\
 &= \left( \left( \left( \underbrace{1}_{T\left(\frac{n}{2^{\log n}}\right) = T(1)} \right) + 1 \right) \dots \right) + 1 \\
 &\qquad\qquad\qquad \log n \text{ times}
 \end{aligned}$$

$\therefore T(n)$  is  $O(\log n)$

# Recursion: Merge Sort

- **Sorting Problem**

Given a list of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$ :

**Precondition:** NONE

**Postcondition:** A permutation of the input list  $\langle a'_1, a'_2, \dots, a'_n \rangle$   
**sorted** in a non-descending order (i.e.,  $a'_1 \leq a'_2 \leq \dots \leq a'_n$ )

- **A Recursive Algorithm**

**Base Case 1:** Empty list  $\rightarrow$  Automatically sorted.

**Base Case 2:** List of size 1  $\rightarrow$  Automatically sorted.

**Recursive Case:** List of size  $\geq 2 \rightarrow$

1. **Split** the list into two (**unsorted**) halves: **L** and **R**.
2. **Recursively sort** **L** and **R**, resulting in: **sortedL** and **sortedR**.
3. Return the **merge** of **sortedL** and **sortedR**.

# Recursion: Merge Sort in Java (1)

```

/* Assumption: L and R are both already sorted. */
private List<Integer> merge(List<Integer> L, List<Integer> R) {
    List<Integer> merge = new ArrayList<>();
    if(L.isEmpty() || R.isEmpty()) { merge.addAll(L); merge.addAll(R); }
    else {
        int i = 0;
        int j = 0;
        while(i < L.size() && j < R.size()) {
            if(L.get(i) <= R.get(j)) { merge.add(L.get(i)); i++; }
            else { merge.add(R.get(j)); j++; }
        }
        /* If i >= L.size(), then this for loop is skipped. */
        for(int k = i; k < L.size(); k++) { merge.add(L.get(k)); }
        /* If j >= R.size(), then this for loop is skipped. */
        for(int k = j; k < R.size(); k++) { merge.add(R.get(k)); }
    }
    return merge;
}

```

RT(merge)?

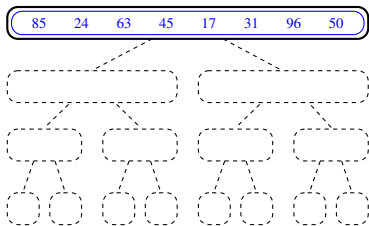
[  $O(L.size() + R.size())$  ]

## Recursion: Merge Sort in Java (2)

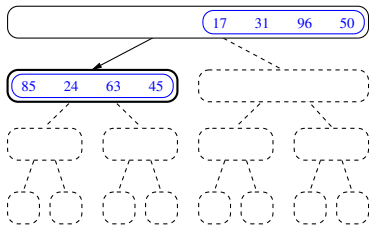
```
public List<Integer> sort(List<Integer> list) {  
    List<Integer> sortedList;  
    if(list.size() == 0) { sortedList = new ArrayList<>(); }  
    else if(list.size() == 1) {  
        sortedList = new ArrayList<>();  
        sortedList.add(list.get(0));  
    }  
    else {  
        int middle = list.size() / 2;  
        List<Integer> left = list.subList(0, middle);  
        List<Integer> right = list.subList(middle, list.size());  
        List<Integer> sortedLeft = sort(left);  
        List<Integer> sortedRight = sort(right);  
        sortedList = merge(sortedLeft, sortedRight);  
    }  
    return sortedList;  
}
```

# Recursion: Merge Sort Example (1)

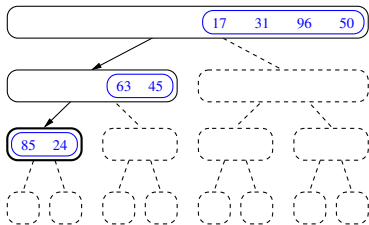
(1) Start with input list of size 8



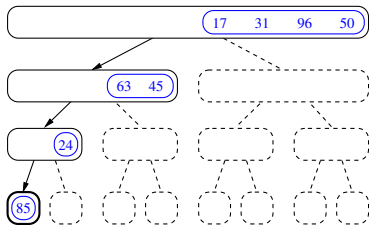
(2) Split and recur on L of size 4



(3) Split and recur on L of size 2

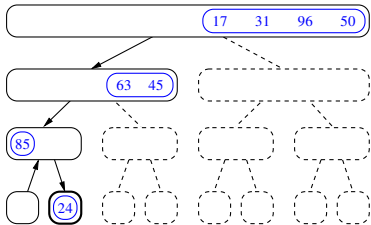


(4) Split and recur on L of size 1, *return*

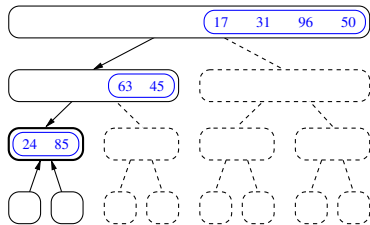


# Recursion: Merge Sort Example (2)

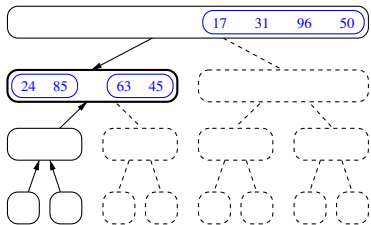
(5) Recur on R of size 1 and *return*



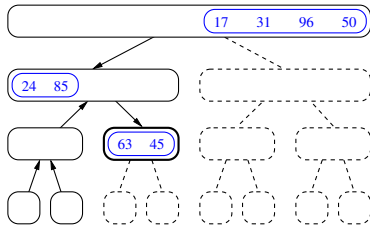
(6) Merge sorted L and R of sizes 1



(7) Return merged list of size 2

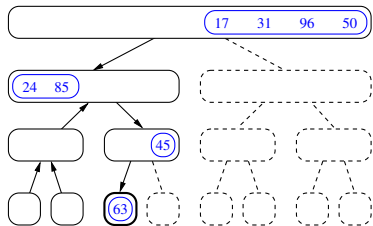


(8) Recur on R of size 2

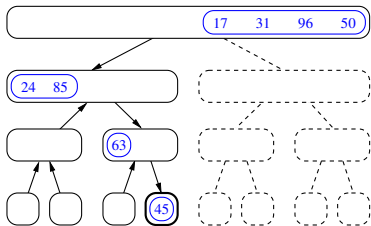


# Recursion: Merge Sort Example (3)

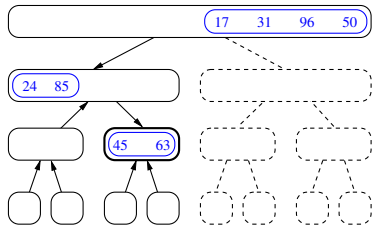
(9) Split and recur on L of size 1, *return*



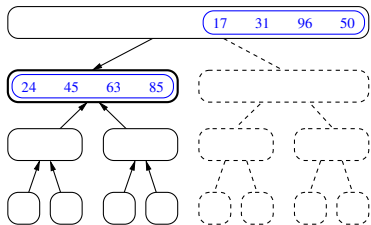
(10) Recur on R of size 1, *return*



(11) Merge sorted L and R of sizes 1, *return*

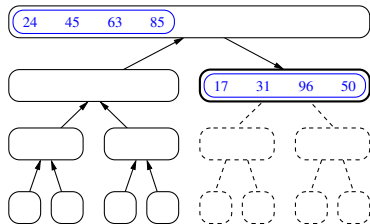


(12) Merge sorted L and R of sizes 2

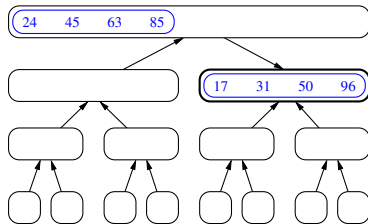


# Recursion: Merge Sort Example (4)

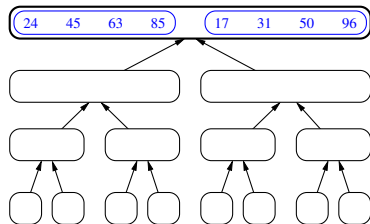
(13) Recur on R of size 4



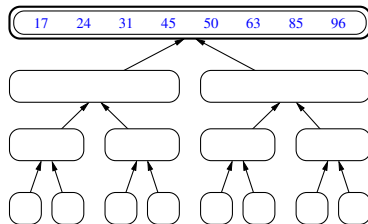
(14) *Return* a sorted list of size 4



(15) Merge sorted L and R of sizes 4



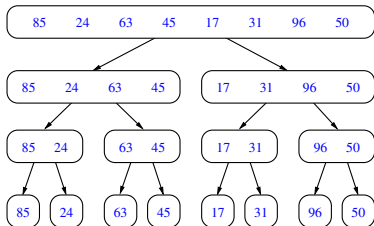
(16) *Return* a sorted list of size 8



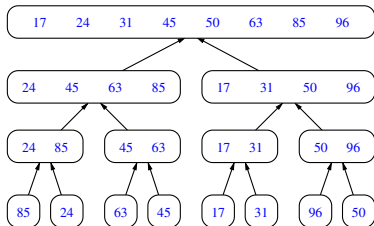
# Recursion: Merge Sort Example (5)

Let's visualize the two *critical phases* of **merge sort** :

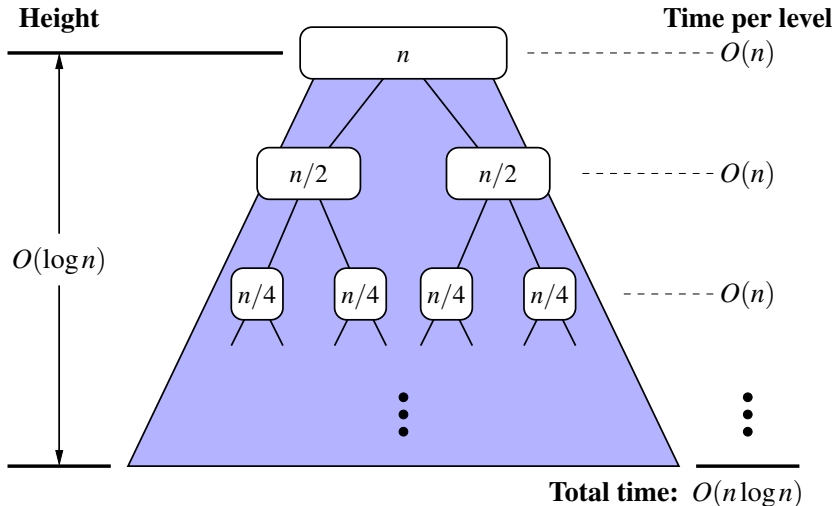
(1) After *Splitting Unsorted* Lists



(2) After *Merging Sorted* Lists



# Recursion: Merge Sort Running Time (1)



# Recursion: Merge Sort Running Time (2)

- **Base Case 1:** Empty list  $\rightarrow$  Automatically sorted. [  $O(1)$  ]
- **Base Case 2:** List of size 1  $\rightarrow$  Automatically sorted. [  $O(1)$  ]
- **Recursive Case:** List of size  $\geq 2 \rightarrow$ 
  1. **Split** the list into two (**unsorted**) halves: **L** and **R**; [  $O(1)$  ]
  2. **Recursively sort** **L** and **R**, resulting in: **sortedL** and **sortedR**  
Q. # times to **split** until **L** and **R** have size 0 or 1?  
A. [  $O(\log n)$  ]
  3. Return the **merge** of **sortedL** and **sortedR**. [  $O(n)$  ]

## Running Time of Merge Sort

$$\begin{aligned}
 &= (\text{RT each RC}) \times (\# \text{ RCs}) \\
 &= (\text{RT merging } \text{sortedL} \text{ and } \text{sortedR}) \times (\# \text{ splits until bases}) \\
 &= O(n \cdot \log n)
 \end{aligned}$$

## Recursion: Merge Sort Running Time (3)

We define  $T(n)$  as the *running time function* of a **merge sort**, where  $n$  is the size of the input array.

$$\begin{cases} T(0) = 1 \\ T(1) = 1 \\ T(n) = 2 \cdot T\left(\frac{n}{2}\right) + n \quad \text{where } n \geq 2 \end{cases}$$

To solve this recurrence relation, we study the pattern of  $T(n)$  and observe how it reaches the *base case(s)*.

# Recursion: Merge Sort Running Time (4)

*Without loss of generality*, assume  $n = 2^i$  for some  $i \geq 0$ .

$$\begin{aligned}
 T(n) &= 2 \times T\left(\frac{n}{2}\right) + n \\
 &= \underbrace{2 \times \left(2 \times T\left(\frac{n}{4}\right) + \frac{n}{2}\right)}_{\substack{2 \text{ terms} \\ 2 \text{ terms}}} + n \\
 &= \underbrace{2 \times \left(2 \times \left(2 \times T\left(\frac{n}{8}\right) + \frac{n}{4}\right) + \frac{n}{2}\right)}_{\substack{3 \text{ terms} \\ 3 \text{ terms}}} + n \\
 &= \dots \\
 &= \underbrace{2 \times \left(2 \times \left(2 \times \dots \times \left(2 \times T\left(\frac{n}{2^{\log n}}\right) + \frac{n}{2^{(\log n)-1}}\right) + \dots + \frac{n}{4} + \frac{n}{2}\right) + \frac{n}{2}\right)}_{\substack{\log n \text{ terms} \\ \log n \text{ terms}}} + n \\
 &= \underbrace{2 \cdot \frac{n}{2} + 2^2 \cdot \frac{n}{4} + \dots + 2^{(\log n)-1} \cdot \frac{n}{2^{(\log n)-1}} + \underbrace{n}_{2^{\log n} \cdot \frac{n}{2^{\log n}}}}_{\log n \text{ terms}} \\
 &= \underbrace{n + n + \dots + n + n}_{\log n \text{ terms}}
 \end{aligned}$$

$\therefore T(n)$  is  $O(n \cdot \log n)$

# Recursion: Quick Sort

- **Sorting Problem**

Given a list of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$ :

**Precondition:** NONE

**Postcondition:** A permutation of the input list  $\langle a'_1, a'_2, \dots, a'_n \rangle$   
**sorted** in a non-descending order (i.e.,  $a'_1 \leq a'_2 \leq \dots \leq a'_n$ )

- **A Recursive Algorithm**

**Base Case 1:** Empty list  $\rightarrow$  Automatically sorted.

**Base Case 2:** List of size 1  $\rightarrow$  Automatically sorted.

**Recursive Case:** List of size  $\geq 2 \rightarrow$

1. Choose a **pivot** element. [ ideally the **median** ]
2. **Split** the list into two (**unsorted**) halves: **L** and **R**, s.t.:  
 All elements in **L** are less than or equal to ( $\leq$ ) the **pivot**.  
 All elements in **R** are greater than ( $>$ ) the **pivot**.
3. **Recursively sort** **L** and **R**: **sortedL** and **sortedR**;
4. Return the **concatenation** of: **sortedL** + **pivot** + **sortedR**.

# Recursion: Quick Sort in Java (1)

```
List<Integer> allLessThanOrEqualTo(int pivotIndex, List<Integer> list)
{
    List<Integer> sublist = new ArrayList<>();
    int pivotValue = list.get(pivotIndex);
    for(int i = 0; i < list.size(); i++) {
        int v = list.get(i);
        if(i != pivotIndex && v <= pivotValue) { sublist.add(v); }
    }
    return sublist;
}

List<Integer> allLargerThan(int pivotIndex, List<Integer> list) {
    List<Integer> sublist = new ArrayList<>();
    int pivotValue = list.get(pivotIndex);
    for(int i = 0; i < list.size(); i++) {
        int v = list.get(i);
        if(i != pivotIndex && v > pivotValue) { sublist.add(v); }
    }
    return sublist;
}
```

RT(allLessThanOrEqualTo)?

[  $O(n)$  ]

RT(allLargerThan)?

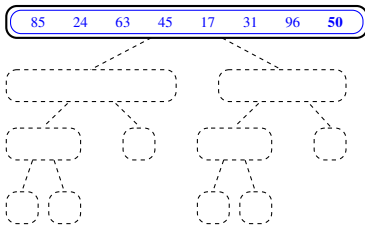
[  $O(n)$  ]

## Recursion: Quick Sort in Java (2)

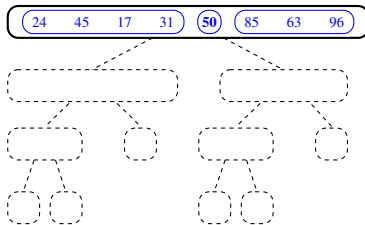
```
public List<Integer> sort(List<Integer> list) {  
    List<Integer> sortedList;  
    if(list.size() == 0) { sortedList = new ArrayList<>(); }  
    else if(list.size() == 1) {  
        sortedList = new ArrayList<>(); sortedList.add(list.get(0)); }  
    else {  
        int pivotIndex = list.size() - 1;  
        int pivotValue = list.get(pivotIndex);  
        List<Integer> left = allLessThanOrEqualTo(pivotIndex, list);  
        List<Integer> right = allLargerThan(pivotIndex, list);  
        List<Integer> sortedLeft = sort(left);  
        List<Integer> sortedRight = sort(right);  
        sortedList = new ArrayList<>();  
        sortedList.addAll(sortedLeft);  
        sortedList.add(pivotValue);  
        sortedList.addAll(sortedRight);  
    }  
    return sortedList;  
}
```

# Recursion: Quick Sort Example (1)

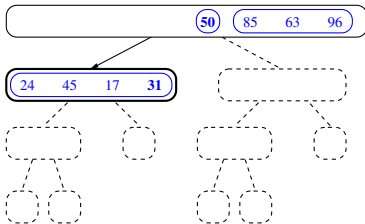
(1) Choose pivot 50 from list of size 8



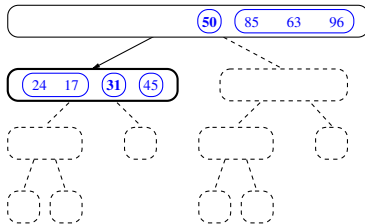
(2) Split w.r.t. the chosen pivot 50



(3) Recur on L of size 4, choose pivot 31

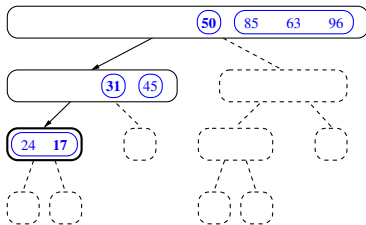


(4) Split w.r.t. the chosen pivot 31

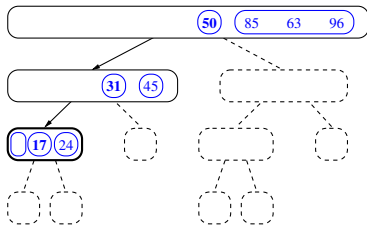


## Recursion: Quick Sort Example (2)

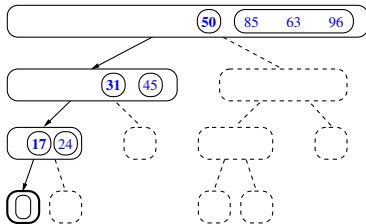
(5) Recur on L of size 2, choose pivot 17



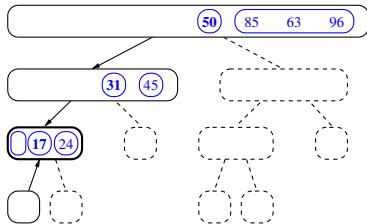
(6) Split w.r.t. the chosen pivot 17



(7) Recur on L of size 0

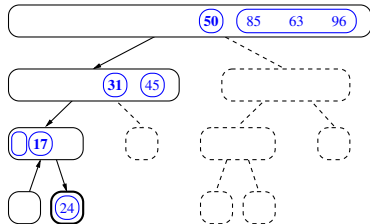


(8) *Return* empty list

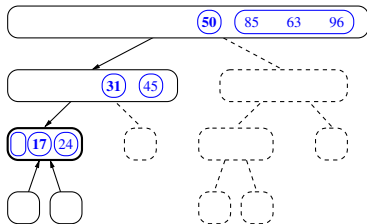


# Recursion: Quick Sort Example (3)

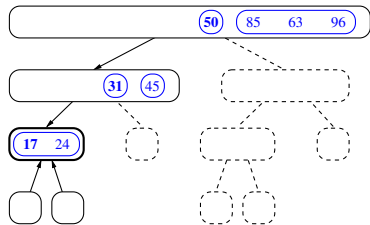
(9) Recur on R of size 1



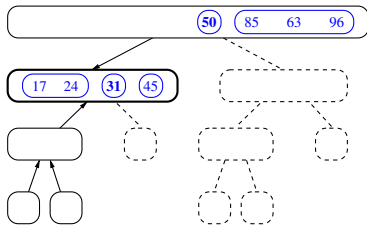
(10) *Return* singleton list <24>



(11) Concatenate {}, <17>, and <24>

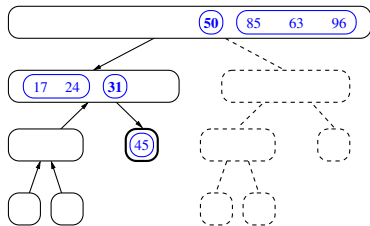


(12) *Return* concatenated list of size 2

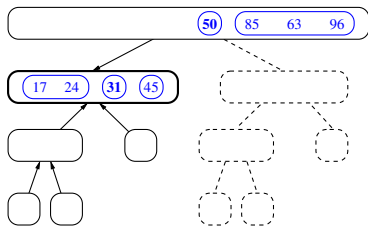


# Recursion: Quick Sort Example (4)

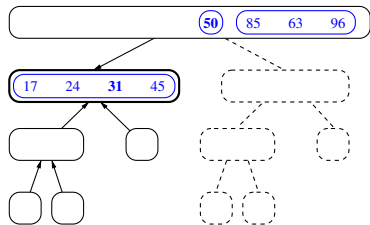
(13) Recur on R of size 1



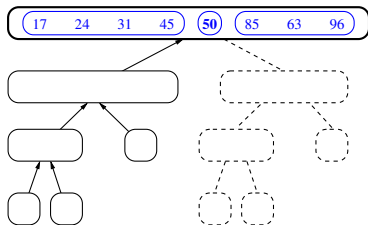
(14) *Return* singleton list <45>



(15) Concatenate <17, 24>, <31>, and <45>

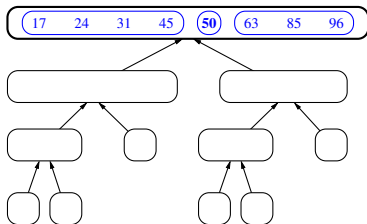


(16) *Return* concatenated list of size 4

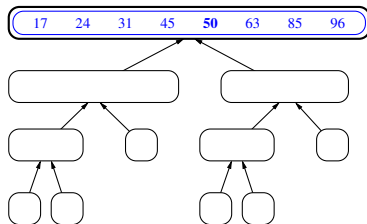


## Recursion: Quick Sort Example (5)

(15) Recur on R of size 3



(16) *Return* sorted list of size 3

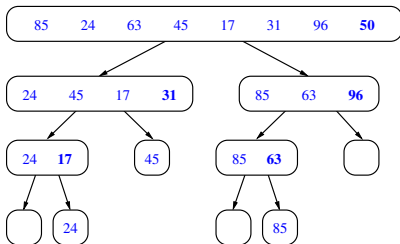


(17) Concatenate  $\langle 17, 24, 31, 45 \rangle$ ,  $\langle 50 \rangle$ , and  $\langle 63, 85, 96 \rangle$ , then *return*

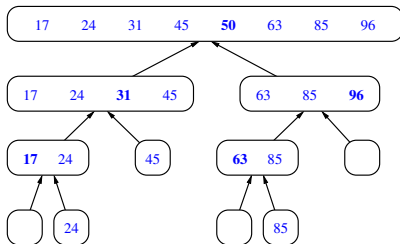
# Recursion: Quick Sort Example (6)

Let's visualize the two *critical phases* of *quick sort* :

(1) After *Splitting Unsorted* Lists

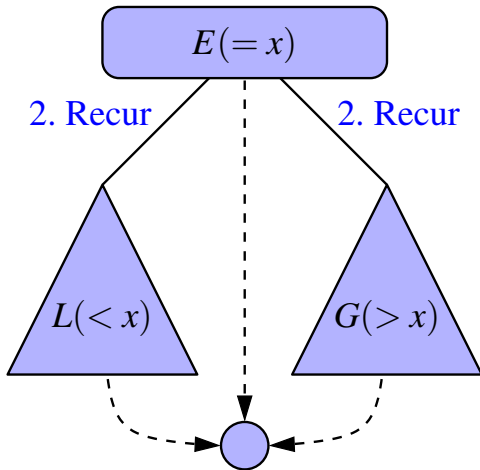


(2) After *Concatenating Sorted* Lists



# Recursion: Quick Sort Running Time (1)

1. Split using pivot  $x$



# Recursion: Quick Sort Running Time (2)

- **Base Case 1:** Empty list  $\rightarrow$  Automatically sorted. [  $O(1)$  ]
- **Base Case 2:** List of size 1  $\rightarrow$  Automatically sorted. [  $O(1)$  ]
- **Recursive Case:** List of size  $\geq 2 \rightarrow$ 
  1. Choose a **pivot** element (e.g., rightmost element) [  $O(1)$  ]
  2. **Split** the list into two (**unsorted**) halves: **L** and **R**, s.t.:
    - All elements in **L** are less than or equal to ( $\leq$ ) the **pivot**. [  $O(n)$  ]
    - All elements in **R** are greater than ( $>$ ) the **pivot**. [  $O(n)$  ]
  3. **Recursively sort** **L** and **R**: **sortedL** and **sortedR**;  
     **Q.** # times to **split** until **L** and **R** have size 0 or 1?  
     **A.**  $O(\log n)$  [ **if** pivots chosen are close to **median values** ]
  4. Return the **concatenation** of: **sortedL** + **pivot** + **sortedR**. [  $O(1)$  ]

## Running Time of Quick Sort

$$\begin{aligned}
 &= (\text{RT each RC}) \times (\# \text{ RCs}) \\
 &= (\text{RT splitting into } L \text{ and } R) \times (\# \text{ splits until bases}) \\
 &= O(n \cdot \log n)
 \end{aligned}$$

# Recursion: Quick Sort Running Time (3)

- We define  $T(n)$  as the *running time function* of a **quick sort**, where  $n$  is the size of the input array.

- Worst Case**

- If the pivot is s.t. the two sub-arrays are “**unbalanced**” in sizes:  
e.g., rightmost element in a reverse-sorted array  
 (“**unbalanced**” splits/partitions: 0 vs.  $n - 1$  elements)

$$\begin{cases} T(0) &= 1 \\ T(1) &= 1 \\ T(n) &= T(n-1) + n \quad \text{where } n \geq 2 \end{cases}$$

- As efficient as Selection/Insertion Sorts:  $O(n^2)$

[ EXERCISE ]

- Best Case**

If the pivot is s.t. it is close to the **median** value:

$$\begin{cases} T(0) &= 1 \\ T(1) &= 1 \\ T(n) &= 2 \cdot T\left(\frac{n}{2}\right) + n \quad \text{where } n \geq 2 \end{cases}$$

- As efficient as Merge Sort:  $O(n \cdot \log n)$
  - Even with partitions such as  $\frac{n}{10}$  vs.  $\frac{9 \cdot n}{10}$  elements, RT remains  $O(n \cdot \log n)$ .

# Index (1)

**Learning Outcomes of this Lecture**

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**Recursion: Binary Search (2)**

**Running Time: Binary Search (1)**

**Running Time: Binary Search (2)**

**Recursion: Merge Sort**

**Recursion: Merge Sort in Java (1)**

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**Recursion: Quick Sort Example (6)**

**Recursion: Quick Sort Running Time (1)**

**Recursion: Quick Sort Running Time (2)**

**Recursion: Quick Sort Running Time (3)**

# General Trees and Binary Trees



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

CHEN-WEI WANG

# Learning Outcomes of this Lecture

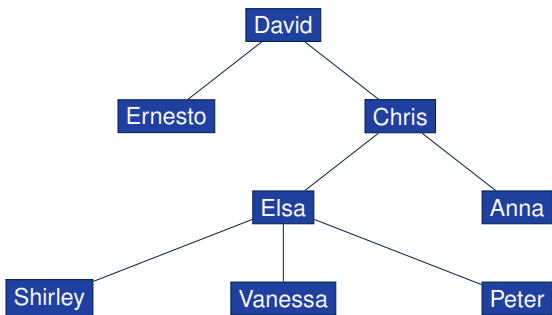
This module is designed to help you understand:

- **Linear** DS (e.g., arrays, LLs) vs. **Non-Linear** DS (e.g., trees)
- Terminologies: **General** Trees vs. **Binary** Trees
- Implementation of a **Generic** Tree
- Mathematical Properties of Binary Trees
- Tree **Traversals**

# General Trees

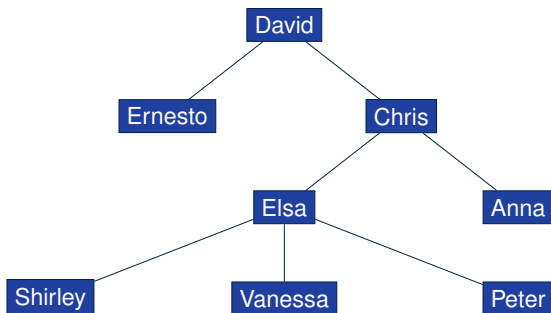
- A **linear** data structure is a sequence, where stored objects can be related via notions of “**predecessor**” and “**successor**”.
  - e.g., arrays
  - e.g., Singly-Linked Lists (SLLs)
  - e.g., Doubly-Linked Lists (DLLs)
- The **Tree ADT** is a **non-linear** collection of nodes/positions.
  - Each node stores some data object.
  - **Nodes** in a **tree** are organized into **levels**: some nodes are “**above**” others, and some are “**below**” others.
  - Think of a **tree** forming a **hierarchy** among the stored **nodes**.
- Terminology of the **Tree ADT** borrows that of **family trees**:
  - e.g., root
  - e.g., parents, siblings, children
  - e.g., ancestors, descendants

# General Trees: Terminology (1)



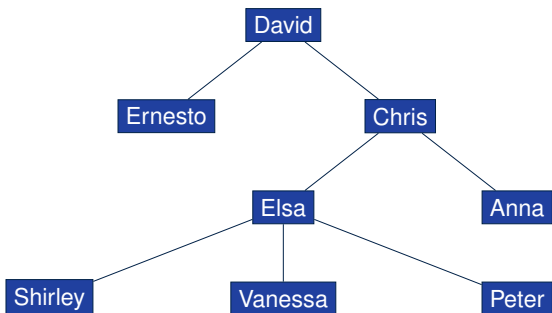
- **top** element of the tree [ *root of tree* ]  
e.g., root of the above family tree: David
- **the** node *immediately above* node *n* [ *parent of n* ]  
e.g., parent of Vanessa: Elsa
- **all** nodes *immediately below* node *n* [ *children of n* ]  
e.g., children of Elsa: Shirley, Vanessa, and Peter  
e.g., children of Ernesto:  $\emptyset$

## General Trees: Terminology (2)



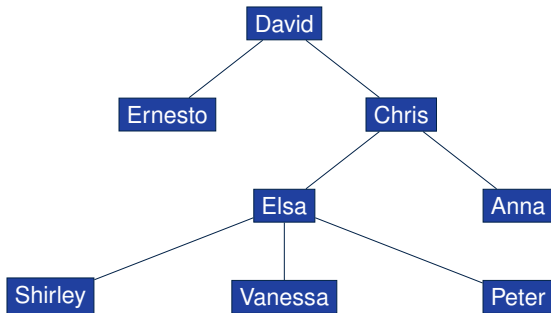
- Union of  $n$ ,  $n$ 's **parent**,  $n$ 's **grand parent**, ..., **root** [  $n$ 's **ancestors** ]  
e.g., ancestors of Vanessa: Vanessa, Elsa, Chris, and David  
e.g., ancestors of David: David
- Union of  $n$ ,  $n$ 's **children**,  $n$ 's **grand children**, ... [  $n$ 's **descendants** ]  
e.g., descendants of Vanessa: Vanessa  
e.g., descendants of David: the entire family tree
- By the above definitions, a **node** is both its **ancestor** and **descendant**.

## General Trees: Terminology (3)



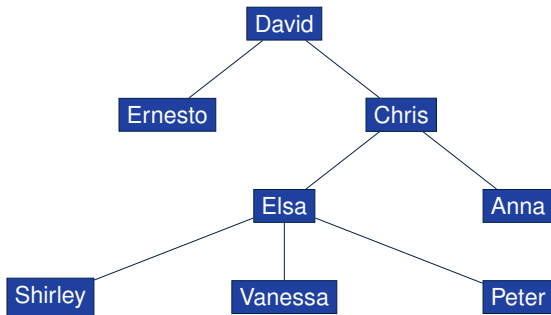
- all nodes with the **same parent** as  $n$ 's [ **siblings of node  $n$**  ]  
e.g., siblings of Vanessa: Shirley and Peter
- the tree formed by **descendants** of  $n$  [ **subtree rooted at  $n$**  ]
- nodes with **no children** [ **external nodes (leaves)** ]  
e.g., leaves of the above tree: Ernesto, Anna, Shirley, Vanessa, Peter
- nodes with **at least one child** [ **internal nodes** ]  
e.g., non-leaves of the above tree: David, Chris, Elsa

# General Trees: Terminology (4)



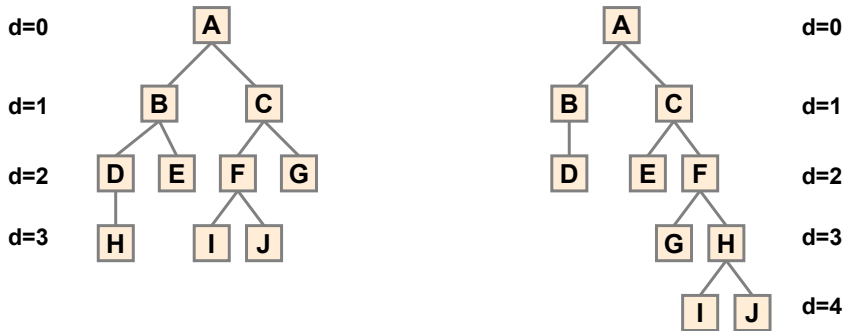
- a pair of **parent** and **child** nodes [ *an edge of tree* ]  
e.g., (David, Chris), (Chris, Elsa), (Elsa, Peter) are three edges
- a sequence of nodes where any two consecutive nodes form an **edge** [ *a path of tree* ]  
e.g., { David, Chris, Elsa, Peter } is a path  
e.g., Elsa's **ancestor path**: { Elsa, Chris, David }

## General Trees: Terminology (5)



- number of **edges** from the **root** to node  $n$  [ **depth of  $n$**  ]  
alternatively: number of  $n$ 's **ancestors** of  $n$  minus one  
 e.g., depth of David (root): 0  
 e.g., depth of Shirley, Vanessa, and Peter: 3
- maximum depth** among all nodes [ **height of tree** ]  
 e.g., Shirley, Vanessa, and Peter have the maximum depth

# General Trees: Example Node Depths



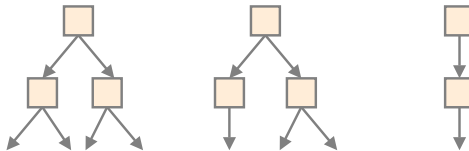
# General Tree: Definition

A **tree**  $T$  is a set of **nodes** satisfying **parent-child** properties:

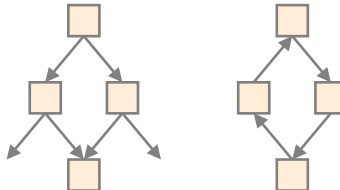
1. If  $T$  is **empty**, then it does not contain any nodes.
2. If  $T$  is **nonempty**, then:
  - $T$  contains at least its **root** (a special node with no parent).
  - Each node  $\underline{n}$  of  $T$  that is not the root has **a unique parent node**  $\underline{w}$ .
  - Given two nodes  $\underline{n}$  and  $\underline{w}$ ,  
if  $\underline{w}$  is the **parent** of  $\underline{n}$ , then **symmetrically**,  $\underline{n}$  is one of  $\underline{w}$ 's **children**.

# General Tree: Important Characteristics

There is a *single, unique path* from the *root* to any particular node in the same tree.



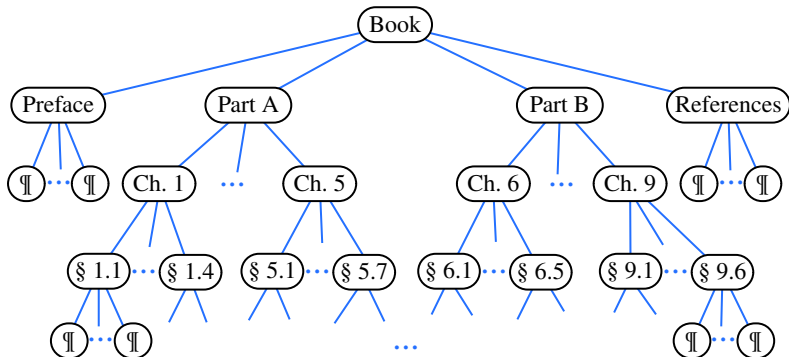
legal tree organization



illegal tree organization (nontrees)

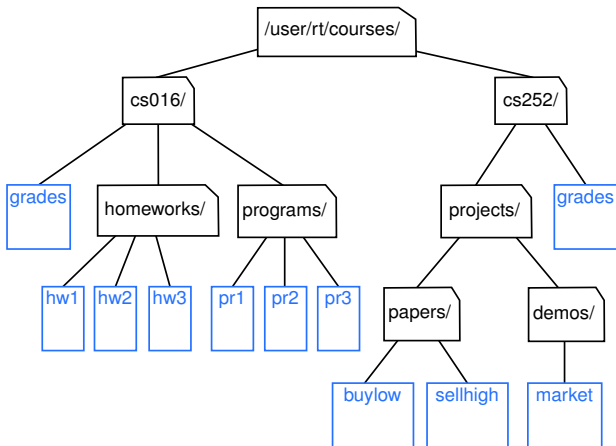
# General Trees: Ordered Trees

A tree is **ordered** if there is a meaningful **linear order** among the **children** of each **internal node**.



# General Trees: Unordered Trees

A tree is **unordered** if the order among the **children** of each **internal node** does **not** matter.



# Implementation: Generic Tree Nodes (1)

```

1 public class TreeNode<E> {
2     private E element; /* data object */
3     private TreeNode<E> parent; /* unique parent node */
4     private TreeNode<E>[] children; /* list of child nodes */
5
6     private final int MAX_NUM_CHILDREN = 10; /* fixed max */
7     private int noc; /* number of child nodes */
8
9     public TreeNode(E element) {
10         this.element = element;
11         this.parent = null;
12         this.children = (TreeNode<E>[])
13             Array.newInstance(this.getClass(), MAX_NUM_CHILDREN);
14         this.noc = 0;
15     }
16     ...
17 }

```

Replacing **L13** with the following results in a **ClassCastException**:

```
this.children = (TreeNode<E>[]) new Object[MAX_NUM_CHILDREN];
```

## Implementation: Generic Tree Nodes (2)

```
public class TreeNode<E> {  
    private E element; /* data object */  
    private TreeNode<E> parent; /* unique parent node */  
    private TreeNode<E>[] children; /* list of child nodes */  
  
    private final int MAX_NUM_CHILDREN = 10; /* fixed max */  
    private int noc; /* number of child nodes */  
  
    public E getElement() { ... }  
    public TreeNode<E> getParent() { ... }  
    public TreeNode<E>[] getChildren() { ... }  
  
    public void setElement(E element) { ... }  
    public void setParent(TreeNode<E> parent) { ... }  
    public void addChild(TreeNode<E> child) { ... }  
    public void removeChildAt(int i) { ... }  
}
```

**Exercise:** Implement void removeChildAt(int i).

# Testing: Connected Tree Nodes

Constructing a **tree** is similar to constructing a **SLL**:

```
@Test
public void test_general_trees_construction() {
    TreeNode<String> agnarr = new TreeNode<>("Agnarr");
    TreeNode<String> elsa = new TreeNode<>("Elsa");
    TreeNode<String> anna = new TreeNode<>("Anna");

    agnarr.addChild(elsa);
    agnarr.addChild(anna);
    elsa.setParent(agnarr);
    anna.setParent(agnarr);

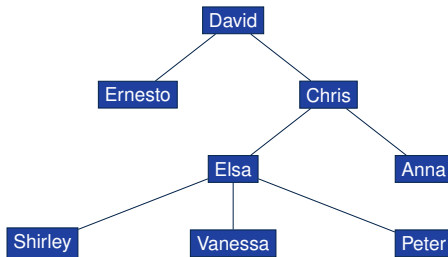
    assertNull(agnarr.getParent());
    assertTrue(agnarr == elsa.getParent());
    assertTrue(agnarr == anna.getParent());
    assertTrue(agnarr.getChildren().length == 2);
    assertTrue(agnarr.getChildren()[0] == elsa);
    assertTrue(agnarr.getChildren()[1] == anna);
}
```

# Problem: Computing a Node's Depth

- Given a node  $n$ , its **depth** is defined as:
  - If  $n$  is the **root**, then  $n$ 's depth is 0.
  - Otherwise,  $n$ 's **depth** is the **depth** of  $n$ 's parent plus one.
- Assuming under a **generic** class `TreeUtilities<E>`:

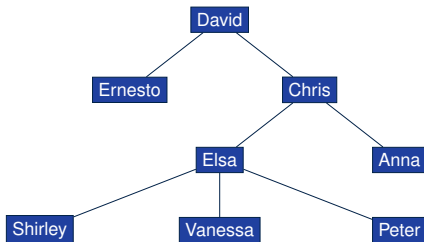
```
1 public int depth(TreeNode<E> n) {  
2     if(n.getParent() == null) {  
3         return 0;  
4     }  
5     else {  
6         return 1 + depth(n.getParent());  
7     }  
8 }
```

# Testing: Computing a Node's Depth



```
@Test
public void test_general_trees_depths() {
    ... /* constructing a tree as shown above */
    TreeUtilities<String> u = new TreeUtilities<>();
    assertEquals(0, u.depth(david));
    assertEquals(1, u.depth(ernesto));
    assertEquals(1, u.depth(chris));
    assertEquals(2, u.depth(elsa));
    assertEquals(2, u.depth(anna));
    assertEquals(3, u.depth(shirley));
    assertEquals(3, u.depth(vanessa));
    assertEquals(3, u.depth(peter));
}
```

# Unfolding: Computing a Node's Depth



```

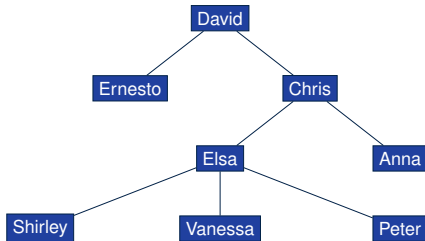
depth(vanessa)
= { vanessa.getParent() == elsa }
  1 + depth(elsa)
= { elsa.getParent() == chris }
  1 + 1 + depth(chris)
= { chris.getParent() == david }
  1 + 1 + 1 + depth(David)
= { David is the root }
  1 + 1 + 1 + 0
= 3
  
```

# Problem: Computing a Tree's Height

- Given node  $n$ , the **height** of subtree rooted at  $n$  is defined as:
  - If  $n$  is a **leaf**, then the **height** of subtree rooted at  $n$  is 0.
  - Otherwise, the height of subtree rooted at  $n$  is one plus the **maximum height** of all subtrees rooted at  $n$ 's children.
- Assuming under a **generic** class `TreeUtilities<E>`:

```
1 public int height(TreeNode<E> n) {  
2     TreeNode<E>[] children = n.getChildren();  
3     if(children.length == 0) { return 0; }  
4     else {  
5         int max = 0;  
6         for(int i = 0; i < children.length; i++) {  
7             int h = 1 + height(children[i]);  
8             max = h > max ? h : max;  
9         }  
10        return max;  
11    }  
12 }
```

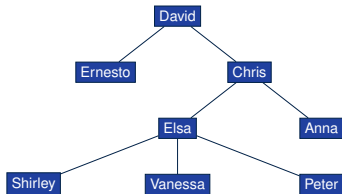
# Testing: Computing a Tree's Height



```

@Test
public void test_general_trees_heights() {
    ... /* constructing a tree as shown above */
    TreeUtilities<String> u = new TreeUtilities<>();
    /* internal nodes */
    assertEquals(3, u.height(david));
    assertEquals(2, u.height(chris));
    assertEquals(1, u.height(elsa));
    /* external nodes */
    assertEquals(0, u.height(ernesto));
    assertEquals(0, u.height(anna));
    assertEquals(0, u.height(shirley));
    assertEquals(0, u.height(vanessa));
    assertEquals(0, u.height(peter));
}
  
```

# Unfolding: Computing a Tree's Height



$$\begin{aligned}
 & \text{height}(\text{subtree rooted at chris}) \\
 &= \{ \text{chris is not a leaf} \} \\
 & \quad \text{MAX} \left( \begin{array}{l} 1 + \text{height}(\text{subtree rooted at elsa}), \\ 1 + \text{height}(\text{subtree rooted at anna}) \end{array} \right) \\
 &= \{ \text{elsa is not a leaf, anna is a leaf} \} \\
 & \quad \text{MAX} \left( \begin{array}{l} 1 + \text{MAX} \left( \begin{array}{l} 1 + \text{height}(\text{subtree rooted at shirley}), \\ 1 + \text{height}(\text{subtree rooted at vanessa}), \\ 1 + \text{height}(\text{subtree rooted at peter}) \end{array} \right), \\ 1 + 0 \end{array} \right) \\
 &= \{ \text{shirley, vanessa, and peter are all leaves} \} \\
 & \quad \text{MAX} \left( \begin{array}{l} 1 + \text{MAX} \left( \begin{array}{l} 1 + 0, \\ 1 + 0, \\ 1 + 0 \end{array} \right), \\ 1 + 0 \end{array} \right) \\
 &= 2
 \end{aligned}$$

# Exercises on General Trees

- Implement and test the following *recursive* algorithm:

```
public TreeNode<E> [ ] ancestors(TreeNode<E> n)
```

which returns the list of *ancestors* of a given node  $n$ .

- Implement and test the following *recursive* algorithm:

```
public TreeNode<E> [ ] descendants(TreeNode<E> n)
```

which returns the list of *descendants* of a given node  $n$ .

# Binary Trees (BTs): Definitions

A **binary tree (BT)** is an **ordered tree** satisfying the following:

1. Each node has **at most two** ( $\leq 2$ ) children.
2. Each **child node** is labeled as either a **left child** or a **right child**.
3. A **left child** precedes a **right child**.

A **binary tree (BT)** is either:

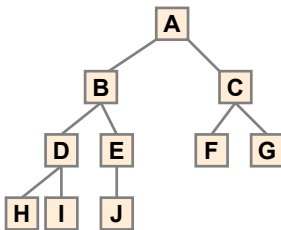
- An **empty** tree; or
- A **nonempty** tree with a **root** node  $r$  which has:
  - a **left subtree** rooted at its **left child**, if any
  - a **right subtree** rooted at its **right child**, if any

# BT Terminology: LST vs. RST

For an *internal* node (with at least one child):

- Subtree rooted at its *left child*, if any, is called *left subtree*.
- Subtree rooted at its *right child*, if any, is called *right subtree*.

e.g.,

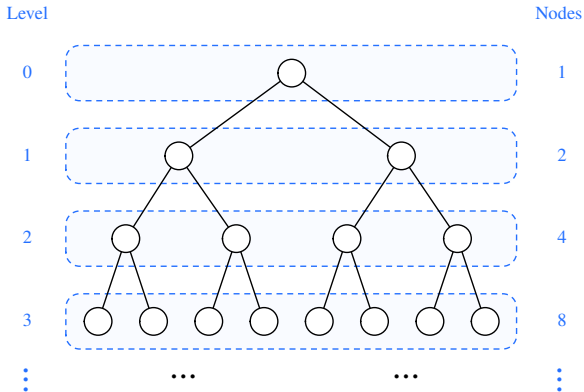


Node A has:

- a *left subtree* rooted at node B
- a *right subtree* rooted at node C

# BT Terminology: Depths, Levels

The set of nodes with the same *depth*  $d$  are said to be at the same *level*  $d$ .



# Background: Sum of Geometric Sequence

- Given a **geometric sequence** of  $n$  terms, where the initial term is  $a$  and the common factor is  $r$ , the **sum** of all its terms is:

$$\sum_{k=0}^{n-1} (a \cdot r^k) = a \cdot r^0 + a \cdot r^1 + a \cdot r^2 + \dots + a \cdot r^{n-1} = a \cdot \left( \frac{r^n - 1}{r - 1} \right)$$

[ See [here](#) to see how the formula is derived. ]

- For the purpose of **binary trees**, **maximum** numbers of nodes at all **levels** form a **geometric sequence**:
  - $a = 1$  [ the **root** at **Level 0** ]
  - $r = 2$  [  $\leq 2$  children for each **internal** node ]
  - e.g., **Max** total # of nodes at **levels** 0 to 4 =  $1 + 2 + 4 + 8 + 16 = 1 \cdot \left( \frac{2^5 - 1}{2 - 1} \right) = 31$

# BT Properties: Max # Nodes at Levels

Given a **binary tree** with **height**  $h$ :

- At each level:
  - **Maximum** number of nodes at **Level 0**:  $2^0 = 1$
  - **Maximum** number of nodes at **Level 1**:  $2^1 = 2$
  - **Maximum** number of nodes at **Level 2**:  $2^2 = 4$
  - ...
  - **Maximum** number of nodes at **Level  $h$** :  $2^h$
- Summing all levels:

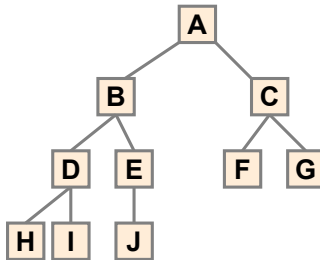
**Maximum** total number of nodes:

$$\underbrace{2^0 + 2^1 + 2^2 + \dots + 2^h}_{h+1 \text{ terms}} = 1 \cdot \left( \frac{2^{h+1} - 1}{2 - 1} \right) = 2^{h+1} - 1$$

# BT Terminology: Complete BTs

A **binary tree** with **height**  $h$  is considered as **complete** if:

- Nodes with **depth**  $\leq h - 2$  has two children.
- Nodes with **depth**  $h - 1$  may have zero, one, or two child nodes.
- **Children** of nodes with **depth**  $h - 1$  are filled from left to right.

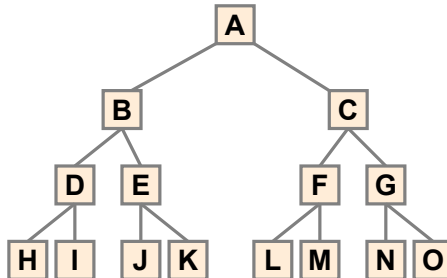


Q1: **Minimum** # of nodes of a **complete** BT?  $(2^h - 1) + 1 = 2^h$

Q2: **Maximum** # of nodes of a **complete** BT?  $2^{h+1} - 1$

# BT Terminology: Full BTs

A **binary tree** with **height**  $h$  is considered as **full** if:  
Each node with **depth**  $\leq h - 1$  has two child nodes.  
 That is, all **leaves** are with the same **depth**  $h$ .



Q1: **Minimum** # of nodes of a complete BT?  $2^{h+1} - 1$

Q2: **Maximum** # of nodes of a complete BT?  $2^{h+1} - 1$

# BT Properties: Bounding # of Nodes

Given a **binary tree** with **height**  $h$ , the **number of nodes**  $n$  is bounded as:

$$h + 1 \leq n \leq 2^{h+1} - 1$$

- Shape of BT with **minimum** # of nodes?  
A “one-path” tree (each **internal node** has exactly one child)
- Shape of BT with **maximum** # of nodes?  
A tree completely filled at each level

# BT Properties: Bounding Height of Tree

Given a **binary tree** with  $n$  nodes, the **height**  $h$  is bounded as:

$$\log(n + 1) - 1 \leq h \leq n - 1$$

- Shape of BT with **minimum** height?

A tree completely filled at each level

$$\begin{aligned} n &= 2^{h+1} - 1 \\ \iff n + 1 &= 2^{h+1} \\ \iff \log(n + 1) &= h + 1 \\ \iff \log(n + 1) - 1 &= h \end{aligned}$$

- Shape of BT with **maximum** height?

A “one-path” tree (each **internal node** has exactly one child)

# BT Properties: Bounding # of Ext. Nodes

Given a **binary tree** with **height**  $h$ , the **number of external nodes**  $n_E$  is bounded as:

$$1 \leq n_E \leq 2^h$$

- Shape of BT with **minimum** # of external nodes?  
A tree with only one node (i.e., the **root**)
- Shape of BT with **maximum** # of external nodes?  
A tree whose bottom level (with **depth**  $h$ ) is completely filled

# BT Properties: Bounding # of Int. Nodes

Given a **binary tree** with **height**  $h$ , the **number of internal nodes**  $n_I$  is bounded as:

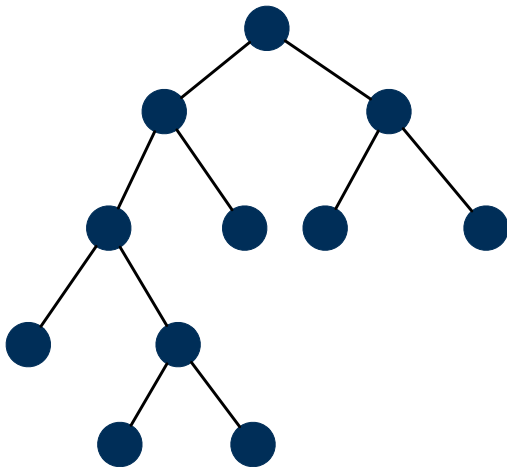
$$h \leq n_I \leq 2^h - 1$$

- Shape of BT with **minimum** # of internal nodes?
  - Number of nodes in a “one-path” tree  $(h + 1)$  minus one
  - That is, the “deepest” leaf node excluded
- Shape of BT with **maximum** # of internal nodes?
  - A tree whose  $\leq h - 1$  **levels** are all completely filled
  - That is:  $2^0 + 2^1 + \dots + 2^{h-1} = 2^h - 1$

$\underbrace{\hspace{10em}}_{n \text{ terms}}$

## BT Terminology: Proper BT

A **binary tree** is **proper** if each **internal node** has two children.



# BT Properties: #s of Ext. and Int. Nodes

Given a **binary tree** that is:

- **nonempty** and **proper**
- with  $n_I$  **internal nodes** and  $n_E$  **external nodes**

We can then expect that:  $n_E = n_I + 1$

Proof by **mathematical induction**:

## • Base Case:

A **proper** BT with only the **root** (an **external node**):  $n_E = 1$  and  $n_I = 0$ .

## • Inductive Case:

- Assume a **proper** BT with  $n$  nodes ( $n > 1$ ) with  $n_I$  **internal nodes** and  $n_E$  **external nodes** such that  $n_E = n_I + 1$ .
- Only one way to create a **larger** BT (with  $n + 2$  nodes) that is still **proper** (with  $n'_E$  **external nodes** and  $n'_I$  **internal nodes**):

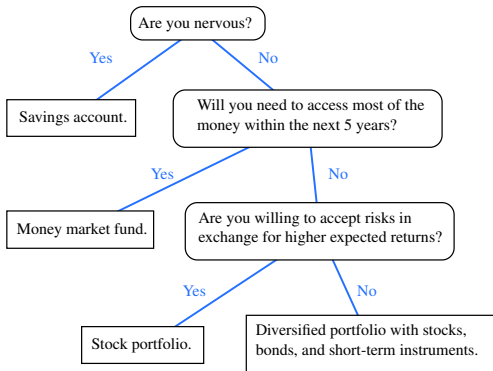
Convert an external node into an **internal** node.

$$n'_E = (n_E - 1) + 2 = n_E + 1 \wedge n'_I = n_I + 1 \Rightarrow n'_E = n'_I + 1$$

# Binary Trees: Application (1)

A **decision tree** is a proper binary tree used to express the decision-making process:

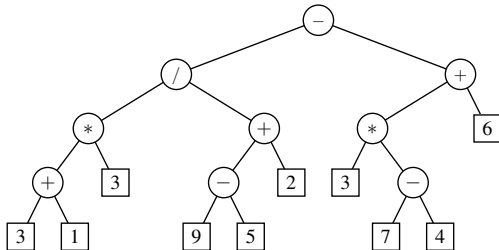
- Each **internal node** denotes a decision point: yes or no.
- Each **external node** denotes the consequence of a decision.



## Binary Trees: Application (2)

An *infix arithmetic expression* can be represented using a binary tree:

- Each **internal node** denotes an operator (unary or binary).
- Each **external node** denotes an operand (i.e., a number).



- To evaluate the expression that is represented by a binary tree, certain **traversal** over the entire tree is required.

# Tree Traversal Algorithms: Definition

- A **traversal** of a **tree**  $T$  systematically **visits all**  $T$ 's nodes.
- Visiting each **node** may be associated with an **action**: e.g.,
  - **Print** the node element.
  - **Determine** if the node element satisfies certain property (e.g., positive, matching a key).
  - **Accumulate** the node element values for some global result.

# Tree Traversal Algorithms: Common Types

Three common traversal orders:

- **Preorder**: Visit parent, then visit child subtrees.

```
preorder (n)
```

```
visit and act on position n
```

```
for child c: children(n) { preorder (c) }
```

- **Postorder**: Visit child subtrees, then visit parent.

```
postorder (n)
```

```
for child c: children(n) { postorder (c) }
```

```
visit and act on position n
```

- **Inorder** (for **BT**): Visit left subtree, then parent, then right subtree.

```
inorder (n)
```

```
if (n has a left child lc) { inorder (lc) }
```

```
visit and act on position n
```

```
if (n has a right child rc) { inorder (rc) }
```

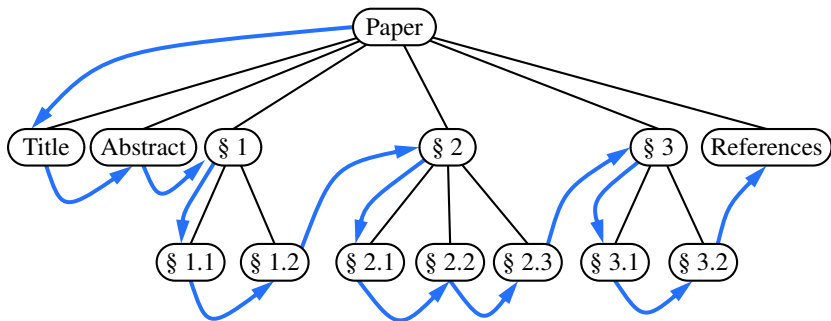
# Tree Traversal Algorithms: Preorder

**Preorder:** Visit parent, then visit child subtrees.

```
preorder (n)
```

```
  visit and act on position n
```

```
  for child c: children(n) { preorder (c) }
```



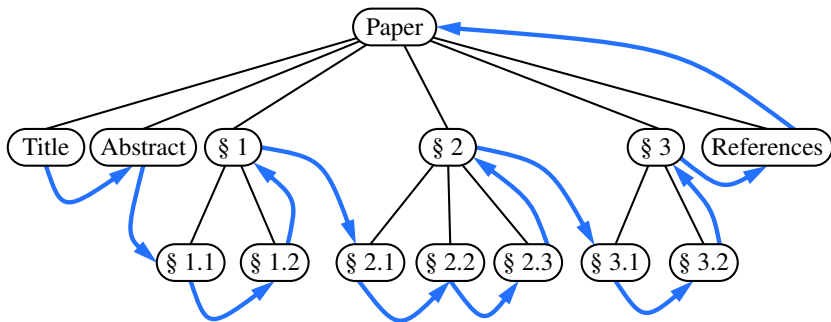
# Tree Traversal Algorithms: Postorder

**Postorder:** Visit child subtrees, then visit parent.

```
postorder (n)
```

```
  for child c: children(n) { postorder (c) }
```

```
  visit and act on position n
```



# Tree Traversal Algorithms: Inorder

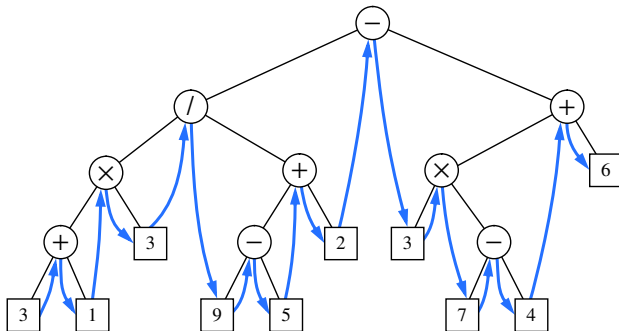
**Inorder** (for BT): Visit left subtree, then parent, then right subtree.

```
inorder(n)
```

```
if (n has a left child lc) { inorder(lc) }
```

```
visit and act on position n
```

```
if (n has a right child rc) { inorder(rc) }
```



# Index (1)

## Learning Outcomes of this Lecture

### General Trees

#### General Trees: Terminology (1)

#### General Trees: Terminology (2)

#### General Trees: Terminology (3)

#### General Trees: Terminology (4)

#### General Trees: Terminology (5)

#### General Trees: Example Node Depths

#### General Tree: Definition

#### General Tree: Important Characteristics

#### General Trees: Ordered Trees

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**Implementation: Generic Tree Nodes (2)**

**Testing: Connected Tree Nodes**

**Problem: Computing a Node's Depth**

**Testing: Computing a Node's Depth**

**Unfolding: Computing a Node's Depth**

**Problem: Computing a Tree's Height**

**Testing: Computing a Tree's Height**

**Unfolding: Computing a Tree's Height**

**Exercises on General Trees**

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**Binary Trees (BTs): Definitions**

**BT Terminology: LST vs. RST**

**BT Terminology: Depths, Levels**

**Background: Sum of Geometric Sequence**

**BT Properties: Max # Nodes at Levels**

**BT Terminology: Complete BTs**

**BT Terminology: Full BTs**

**BT Properties: Bounding # of Nodes**

**BT Properties: Bounding Height of Tree**

**BT Properties: Bounding # of Ext. Nodes**

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**BT Terminology: Proper BT**

**BT Properties: #s of Ext. and Int. Nodes**

**Binary Trees: Application (1)**

**Binary Trees: Application (2)**

**Tree Traversal Algorithms: Definition**

**Tree Traversal Algorithms: Common Types**

**Tree Traversal Algorithms: Preorder**

**Tree Traversal Algorithms: Postorder**

**Tree Traversal Algorithms: Inorder**

# Binary Search Trees



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

CHEN-WEI WANG

# Learning Outcomes of this Lecture

This module is designed to help you understand:

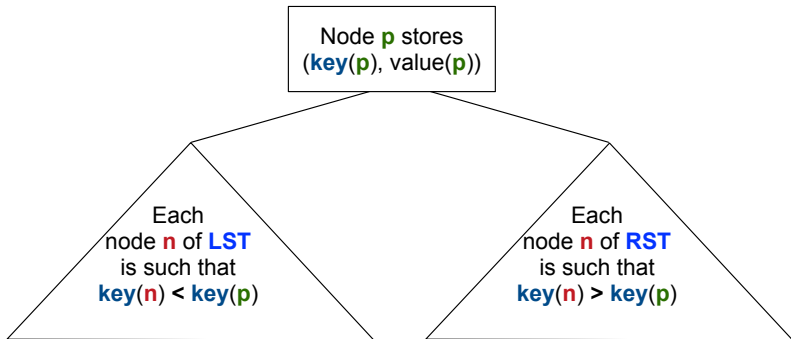
- **Binary Search Trees (BSTs)** = **BTs** + **Search Property**
- Implementing a **Generic** BST in Java
- BST Operations:
  - **Searching**: Implementation, Visualization, RT
  - **Insertion**: (Sketch of) Implementation, Visualization, RT
  - **Deletion**: (Sketch of) Implementation, Visualization, RT

# Binary Search Tree: Recursive Definition

A **Binary Search Tree** (**BST**) is a **BT** satisfying the **search property**:

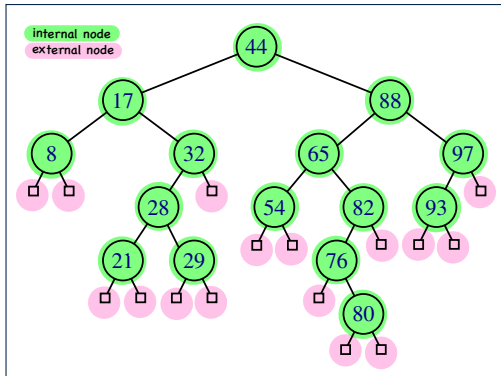
Each **internal node**  $p$  stores an **entry**, a key-value pair  $(k, v)$ , such that:

- For each node  $n$  in the **LST** of  $p$ :  $\text{key}(n) < \text{key}(p)$
- For each node  $n$  in the **RST** of  $p$ :  $\text{key}(n) > \text{key}(p)$



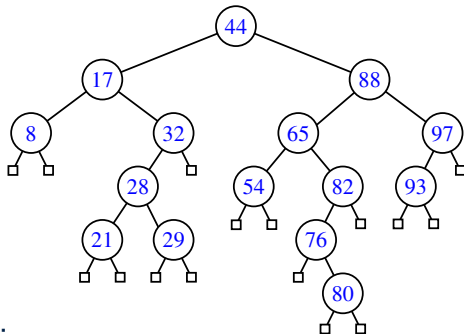
# BST: Internal Nodes vs. External Nodes

- We store key-value pairs only in *internal nodes*.
- Recall how we treat *header* and *trailer* in a DLL.
- We treat *external nodes* as *sentinels*, in order to simplify the *coding logic* of BST algorithms.



# BST: Sorting Property

- An *in-order traversal* of a **BST** will result in a sequence of nodes whose *keys* are arranged in an *ascending order*.
- Unless necessary, we may only show *keys* in BST nodes:



## Justification:

- In-Order Traversal: Visit **LST**, then **root**, then **RST**.
- Search Property of BST: keys in **LST/RST** **</>** **root's** key

# Implementation: Generic BST Nodes

```
public class BSTNode<E> {
    private int key; /* key */
    private E value; /* value */
    private BSTNode<E> parent; /* unique parent node */
    private BSTNode<E> left; /* left child node */
    private BSTNode<E> right; /* right child node */

    public BSTNode() { ... }
    public BSTNode(int key, E value) { ... }

    public boolean isExternal() {
        return this.getLeft() == null && this.getRight() == null;
    }
    public boolean isInternal() {
        return !this.isExternal();
    }
    public int getKey() { ... }
    public void setKey(int key) { ... }
    public E getValue() { ... }
    public void setValue(E value) { ... }
    public BSTNode<E> getParent() { ... }
    public void setParent(BSTNode<E> parent) { ... }
    public BSTNode<E> getLeft() { ... }
    public void setLeft(BSTNode<E> left) { ... }
    public BSTNode<E> getRight() { ... }
    public void setRight(BSTNode<E> right) { ... }
}
```

# Implementation: BST Utilities – Traversal

```
import java.util.ArrayList;
public class BSTUtilities<E> {
    public ArrayList<BSTNode<E>> inOrderTraversal(BSTNode<E> root) {
        ArrayList<BSTNode<E>> result = null;
        if(root.isInternal()) {
            result = new ArrayList<>();

            if(root.getLeft().isInternal()) {
                result.addAll(inOrderTraversal(root.getLeft()));
            }

            result.add(root);

            if(root.getRight().isInternal()) {
                result.addAll(inOrderTraversal(root.getRight()));
            }
        }
        return result;
    }
}
```

# Testing: Connected BST Nodes

Constructing a **BST** is similar to constructing a **General Tree**:

```
@Test
public void test_binary_search_trees_construction() {
    BSTNode<String> n28 = new BSTNode<>(28, "alan");
    BSTNode<String> n21 = new BSTNode<>(21, "mark");
    BSTNode<String> n35 = new BSTNode<>(35, "tom");
    BSTNode<String> extN1 = new BSTNode<>();
    BSTNode<String> extN2 = new BSTNode<>();
    BSTNode<String> extN3 = new BSTNode<>();
    BSTNode<String> extN4 = new BSTNode<>();
    n28.setLeft(n21); n21.setParent(n28);
    n28.setRight(n35); n35.setParent(n28);
    n21.setLeft(extN1); extN1.setParent(n21);
    n21.setRight(extN2); extN2.setParent(n21);
    n35.setLeft(extN3); extN3.setParent(n35);
    n35.setRight(extN4); extN4.setParent(n35);
    BSTUtilities<String> u = new BSTUtilities<>();
    ArrayList<BSTNode<String>> inOrderList = u.inOrderTraversal(n28);
    assertTrue(inOrderList.size() == 3);
    assertEquals(21, inOrderList.get(0).getKey());
    assertEquals("mark", inOrderList.get(0).getValue());
    assertEquals(28, inOrderList.get(1).getKey());
    assertEquals("alan", inOrderList.get(1).getValue());
    assertEquals(35, inOrderList.get(2).getKey());
    assertEquals("tom", inOrderList.get(2).getValue());
}
```

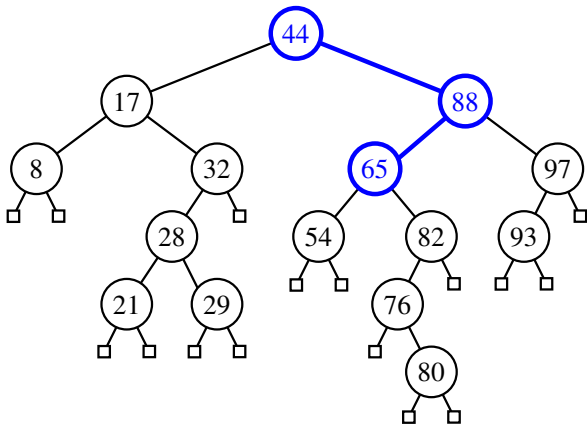
# Implementing BST Operation: Searching

Given a **BST** rooted at node **p**, to locate a particular **node** whose **key** matches **k**, we may view it as a **decision tree**.

```
public BSTNode<E> search(BSTNode<E> p, int k) {  
    BSTNode<E> result = null;  
    if(p.isExternal()) {  
        result = p; /* unsuccessful search */  
    }  
    else if(p.getKey() == k) {  
        result = p; /* successful search */  
    }  
    else if(k < p.getKey()) {  
        result = search(p.getLeft(), k); /* recur on LST */  
    }  
    else if(k > p.getKey()) {  
        result = search(p.getRight(), k); /* recur on RST */  
    }  
    return result;  
}
```

# Visualizing BST Operation: Searching (1)

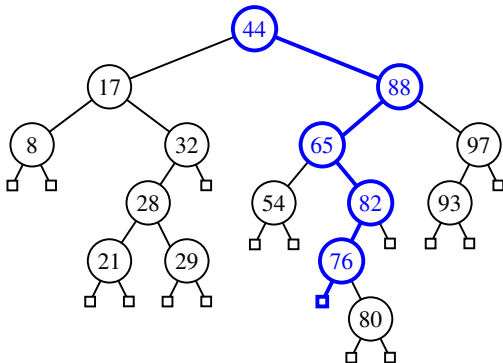
A **successful** search for **key 65**:



The **internal node** storing key 65 is returned.

## Visualizing BST Operation: Searching (2)

- An **unsuccessful** search for **key 68**:



The **external, left child node** of the **internal node** storing **key 76** is **returned**.

- Exercise**: Provide **keys** for different **external nodes** to be **returned**.

# Testing BST Operation: Searching

```
@Test
public void test_binary_search_trees_search() {
    BSTNode<String> n28 = new BSTNode<>(28, "alan");
    BSTNode<String> n21 = new BSTNode<>(21, "mark");
    BSTNode<String> n35 = new BSTNode<>(35, "tom");
    BSTNode<String> extN1 = new BSTNode<>();
    BSTNode<String> extN2 = new BSTNode<>();
    BSTNode<String> extN3 = new BSTNode<>();
    BSTNode<String> extN4 = new BSTNode<>();
    n28.setLeft(n21); n21.setParent(n28);
    n28.setRight(n35); n35.setParent(n28);
    n21.setLeft(extN1); extN1.setParent(n21);
    n21.setRight(extN2); extN2.setParent(n21);
    n35.setLeft(extN3); extN3.setParent(n35);
    n35.setRight(extN4); extN4.setParent(n35);

    BSTUtilities<String> u = new BSTUtilities<>();
    /* search existing keys */
    assertTrue(n28 == u.search(n28, 28));
    assertTrue(n21 == u.search(n28, 21));
    assertTrue(n35 == u.search(n28, 35));
    /* search non-existing keys */
    assertTrue(extN1 == u.search(n28, 17)); /* *17* < 21 */
    assertTrue(extN2 == u.search(n28, 23)); /* 21 < *23* < 28 */
    assertTrue(extN3 == u.search(n28, 33)); /* 28 < *33* < 35 */
    assertTrue(extN4 == u.search(n28, 38)); /* 35 < *38* */
}
```



## RT of BST Operation: Searching (2)

- Recursive calls of `search` are made on a **path** which
  - Starts from the **root**
  - Goes down one **level** at a time
    - RT of deciding from each node to go to LST or RST? [  $O(1)$  ]
  - Stops when the key is found or when a **leaf** is reached
    - Maximum** number of nodes visited by the search? [  $h + 1$  ]
- $\therefore$  RT of **search on a BST** is  $O(h)$
- Recall: Given a BT with  $n$  nodes, the **height  $h$**  is bounded as:
 
$$\log(n + 1) - 1 \leq h \leq n - 1$$
  - Best** RT of a **binary search** is  $O(\log(n))$  [ **balanced** BST ]
  - Worst** RT of a **binary search** is  $O(n)$  [ **ill-balanced** BST ]
- Binary search** on non-linear vs. linear structures:

	Search on a <b>BST</b>	Binary Search on a <b>Sorted Array</b>
START	Root of BST	Middle of Array
PROGRESS	LST or RST	Left Half or Right Half of Array
BEST RT	$O(\log(n))$	$O(\log(n))$
WORST RT	$O(n)$	

# Sketch of BST Operation: Insertion

To **insert** an **entry** (with **key**  $k$  & **value**  $v$ ) into a BST rooted at **node**  $n$ :

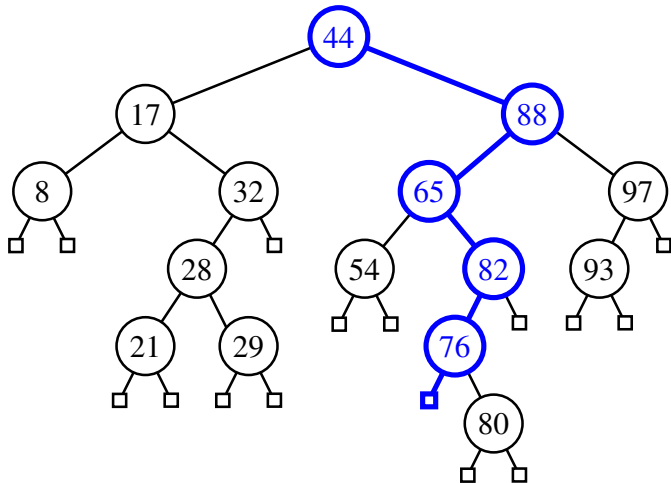
- Let node  $p$  be the return value from `search( $n$ ,  $k$ )`.
- If  $p$  is an **internal node**
  - ⇒ Key  $k$  exists in the BST.
  - ⇒ Set  $p$ 's value to  $v$ .
- If  $p$  is an **external node**
  - ⇒ Key  $k$  does **not** exist in the BST.
  - ⇒ Set  $p$ 's key and value to  $k$  and  $v$ .

Running time?

[  $O(h)$  ]

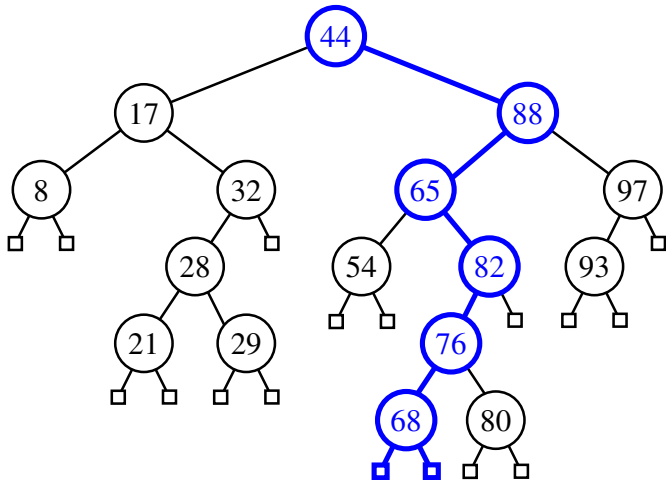
# Visualizing BST Operation: Insertion (1)

Before *inserting* an entry with **key 68** into the following BST:



## Visualizing BST Operation: Insertion (2)

After *inserting* an entry with **key 68** into the following BST:



# Exercise on BST Operation: Insertion

Exercise: In `BSTUtilities` class, *implement* and *test* the

```
void insert(BSTNode<E> p, int k, E v)
```

 method.

# Sketch of BST Operation: Deletion

To **delete** an **entry** (with **key**  $k$ ) from a BST rooted at **node**  $n$ :

Let node  $p$  be the return value from `search(n, k)`.

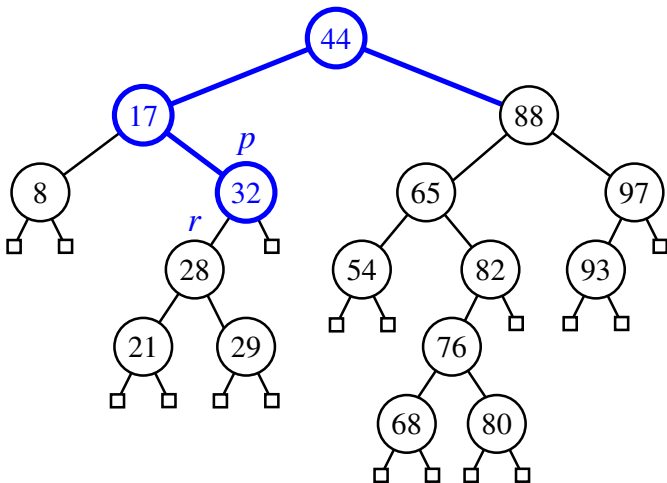
- **Case 1:** Node  $p$  is **external**.  
 $k$  is not an existing key  $\Rightarrow$  Nothing to remove
- **Case 2:** Both of node  $p$ 's child nodes are **external**.  
 No "orphan" subtrees to be handled  $\Rightarrow$  Remove  $p$  [ Still BST? ]
- **Case 3:** One of the node  $p$ 's children, say  $r$ , is **internal**.  
 •  $r$ 's sibling is **external**  $\Rightarrow$  Replace node  $p$  by node  $r$  [ Still BST? ]
- **Case 4:** Both of node  $p$ 's children are **internal**.  
 • Let  $r$  be the **right-most internal node**  $p$ 's **LST**.  
 $\Rightarrow r$  contains the **largest key s.t.  $\text{key}(r) < \text{key}(p)$** .  
**Exercise:** Can  $r$  contain the **smallest key s.t.  $\text{key}(r) > \text{key}(p)$** ?  
 • Overwrite node  $p$ 's entry by node  $r$ 's entry. [ Still BST? ]  
 •  $r$  being the **right-most internal node** may have:  
   ◊ Two **external child nodes**  $\Rightarrow$  Remove  $r$  as in **Case 2**.  
   ◊ An **external, RC** & an **internal LC**  $\Rightarrow$  Remove  $r$  as in **Case 3**.

Running time?

[  $O(h)$  ]

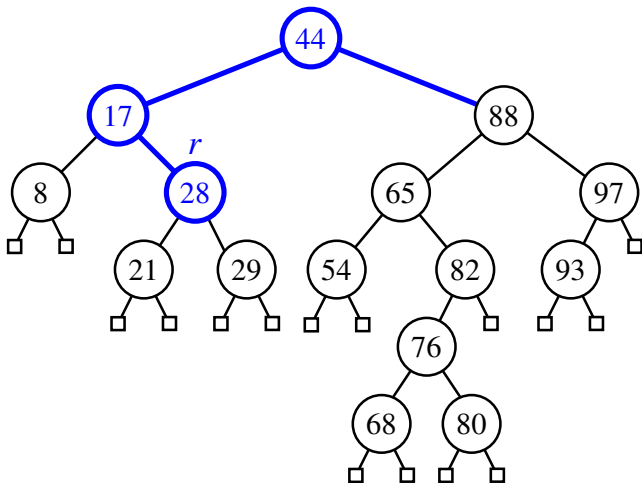
# Visualizing BST Operation: Deletion (1.1)

(Case 3) Before *deleting* the node storing *key 32*:



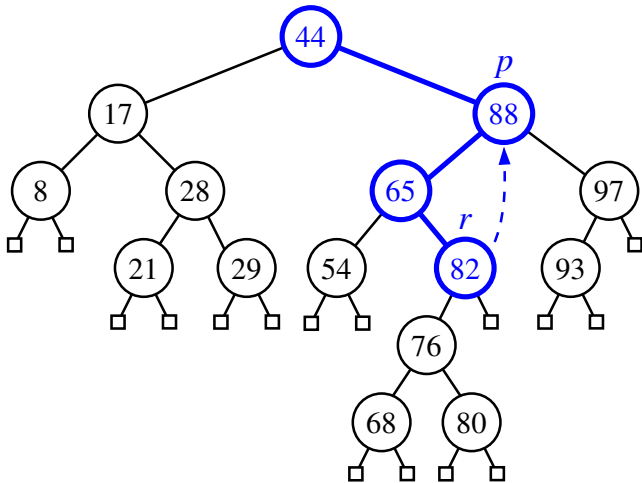
## Visualizing BST Operation: Deletion (1.2)

(Case 3) After **deleting** the node storing **key 32**:



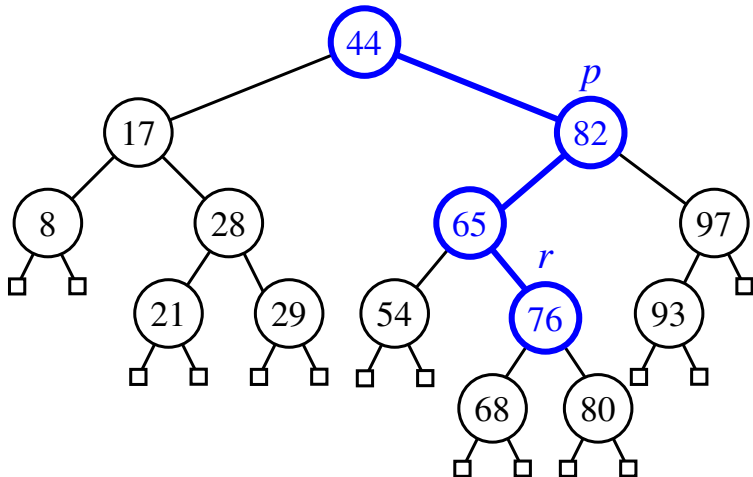
## Visualizing BST Operation: Deletion (2.1)

(Case 4) Before *deleting* the node storing *key 88*:



## Visualizing BST Operation: Deletion (2.2)

(Case 4) After **deleting** the node storing **key 88**:



# Exercise on BST Operation: Deletion

Exercise: In `BSTUtilities` class, *implement* and *test* the  
`void delete(BSTNode<E> p, int k)` method.

# Index (1)

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**BST: Internal Nodes vs. External Nodes**

**BST: Sorting Property**

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**Implementation: BST Utilities – Traversal**

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**Visualizing BST Operation: Deletion (2.2)**

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## Exercise on BST Operation: Deletion

# Priority Queues, Heaps, and Heap Sort



EECS2101 X & Z:  
Fundamentals of Data Structures  
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# Learning Outcomes of this Lecture

This module is designed to help you understand:

- When the **Worst-Case RT** of a **BST Search** Occurs
- **Height-Balance** Property
- The **Priority Queue** (**PQ**) ADT
- Time Complexities of **List**-Based **PQ**
- The **Heap** Data Structure (Properties & Operations)
- **Heap Sort**
- Time Complexities of **Heap**-Based **PQ**
- **Heap** Construction Methods: Top-Down vs. Bottom-Up
- **Array**-Based Representation of a **Heap**

# Balanced Binary Search Trees: Motivation

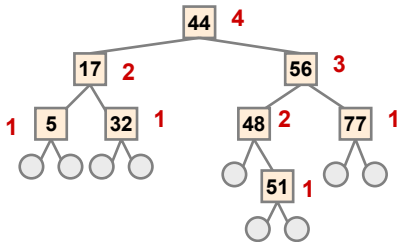
- After *insertions* into a BST, the **worst-case RT** of a *search* occurs when the *height*  $h$  is at its **maximum**:  **$O(n)$** :
  - e.g., Entries were inserted in an decreasing order of their keys  
 $\langle 100, 75, 68, 60, 50, 1 \rangle$   
 $\Rightarrow$  **One-path**, **left-slanted** BST
  - e.g., Entries were inserted in an increasing order of their keys  
 $\langle 1, 50, 60, 68, 75, 100 \rangle$   
 $\Rightarrow$  **One-path**, **right-slanted** BST
  - e.g., Last entry's key is in-between keys of the previous two entries  
 $\langle 1, 100, 50, 75, 60, 68 \rangle$   
 $\Rightarrow$  **One-path**, **side-alternating** BST
- To avoid the worst-case RT ( $\because$  a **ill-balanced tree**), we need to take actions **as soon as** the tree becomes **unbalanced**.

# Balanced Binary Search Trees: Definition

- Given a node  $p$ , the **height** of the subtree rooted at  $p$  is:

$$\text{height}(p) = \begin{cases} 0 & \text{if } p \text{ is external} \\ 1 + \text{MAX} (\{ \text{height}(c) \mid \text{parent}(c) = p \}) & \text{if } p \text{ is internal} \end{cases}$$

- A **balanced** BST  $T$  satisfies the **height-balance property**:  
For every **internal node**  $n$ , **heights** of  $n$ 's child nodes differ  $\leq 1$ .



Q: Is the above tree a **balanced BST**?



Q: Will the tree remain **balanced** after inserting 55?

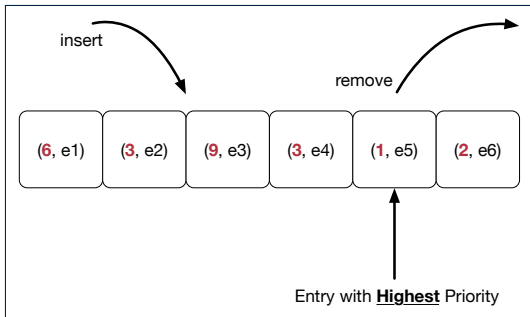


Q: Will the tree remain **balanced** after inserting 63?



# What is a Priority Queue?

- A **Priority Queue (PQ)** stores a collection of **entries**.



- Each **entry** is a pair: an **element** and its **key**.
- The **key** of each **entry** denotes its **element's** "priority".
- Keys** in a Priority Queue (PQ) are **not** used for uniquely identifying an entry.

- In a PQ, the next entry to remove has the "**highest**" priority.
  - e.g., In the stand-by queue of a fully-booked flight, **frequent flyers** get the higher priority to replace any cancelled seats.
  - e.g., A network router, faced with insufficient bandwidth, may only handle **real-time tasks** (e.g., streaming) with highest priorities.

# The Priority Queue (PQ) ADT

- *min*

[ *precondition*: PQ is not empty ]

[ *postcondition*: return entry with highest priority in PQ ]

- *size*

[ *precondition*: none ]

[ *postcondition*: return number of entries inserted to PQ ]

- *isEmpty*

[ *precondition*: none ]

[ *postcondition*: return whether there is no entry in PQ ]

- *insert(k, v)*

[ *precondition*: PQ is not full ]

[ *postcondition*: insert the input entry into PQ ]

- *removeMin*

[ *precondition*: PQ is not empty ]

[ *postcondition*: remove and return a min entry in PQ ]

# Two List-Based Implementations of a PQ

Consider two strategies for implementing a **PQ**, where we maintain:

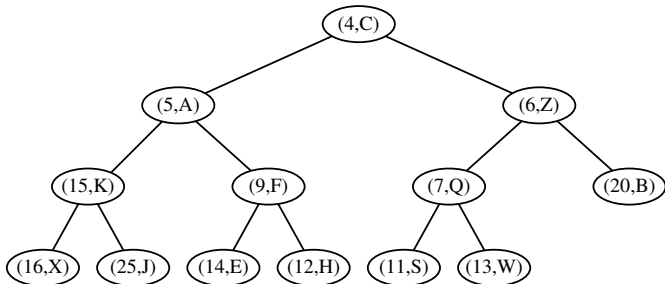
1. A list always sorted in a non-descending order [  $\approx$  **INSERTIONSORT** ]
2. An unsorted list [  $\approx$  **SELECTIONSORT** ]

PQ Method	List Method			
	SORTED LIST		UNSORTED LIST	
size	list.size $O(1)$			
isEmpty	list.isEmpty $O(1)$			
min	list.first	$O(1)$	search min	$O(n)$
insert	insert to “right” spot	$O(n)$	insert to front	$O(1)$
removeMin	list.removeFirst	$O(1)$	search min and remove	$O(n)$

# Heaps

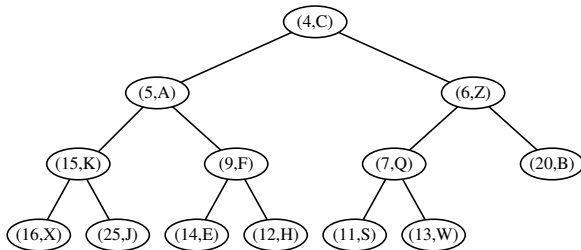
A **heap** is a *binary tree* which:

1. Stores in each node an *entry* (i.e., *key* and *value*).



2. Satisfies a *relational* property of stored keys
3. Satisfies a *structural* property of tree organization

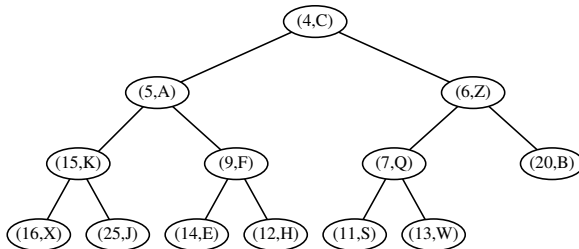
# Heap Property 1: Relational



**Keys** in a **heap** satisfy the **Heap-Order Property** :

- Every node  $n$  (other than the root) is s.t.  $\text{key}(n) \geq \text{key}(\text{parent}(n))$   
 $\Rightarrow$  **Keys** in a **root-to-leaf path** are sorted in a non-descending order.  
 e.g., Keys in entry path  $\langle (4, C), (5, A), (9, F), (14, E) \rangle$  are sorted.  
 $\Rightarrow$  The **minimal key** is stored in the **root**.  
 e.g., Root  $(4, C)$  stores the minimal key 4.
- Keys** of nodes from **different subtrees** are **not** constrained at all.  
 e.g., For node  $(5, A)$ , key of its **LST**'s root (15) is not minimal for its **RST**.

# Heap Property 2: Structural



A **heap** with **height  $h$**  satisfies the **Complete BT Property** :

- Nodes with **depth  $\leq h - 2$**  has two child nodes.
- Nodes with **depth  $h - 1$**  may have zero, one, or two child nodes.
- Nodes with **depth  $h$**  are filled from left to right.

**Q.** When the # of nodes is  $n$ , what is  $h$ ?

**Q.** # of nodes from Level 0 through Level  $h - 1$ ?

**Q.** # of nodes at Level  $h$ ?

**Q.** **Minimum** # of nodes of a complete BT?

**Q.** **Maximum** # of nodes of a complete BT?

$$\lceil \log_2 n \rceil$$

$$2^h - 1$$

$$n - (2^h - 1)$$

$$2^h$$

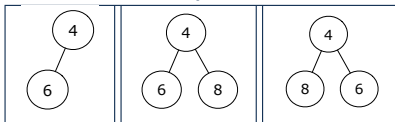
$$2^{h+1} - 1$$

# Heaps: More Examples

- The **smallest heap** is just an empty binary tree.
- The **smallest non-empty heap** is a one-node heap.  
e.g.,



- Two-node and Three-node Heaps:



- These are **not** two-node heaps:



# Heap Operations

- There are three main operations for a **heap**:
  1. **Extract the Entry with Minimal Key:**  
Return the stored entry of the **root**. [  $O(1)$  ]
  2. **Insert a New Entry:**  
A single **root-to-leaf path** is affected. [  $O(h)$  or  $O(\log n)$  ]
  3. **Delete the Entry with Minimal Key:**  
A single **root-to-leaf path** is affected. [  $O(h)$  or  $O(\log n)$  ]
- After performing each operation,  
both **relational** and **structural** properties must be maintained.

# Updating a Heap: Insertion

To insert a new entry  $(k, v)$  into a heap with *height*  $h$ :

1. Insert  $(k, v)$ , possibly temporarily breaking the *relational property*.

1.1 Create a new entry  $e = (k, v)$ .

1.2 Create a new *right-most* node  $n$  at *Level*  $h$ .

1.3 Store entry  $e$  in node  $n$ .

After steps 1.1 and 1.2, the *structural property* is maintained.

2. Restore the **heap-order property (HOP)** using *Up-Heap Bubbling* :

2.1 Let  $c = n$ .

2.2 While **HOP** is not restored and  $c$  is not the root:

2.2.1 Let  $p$  be  $c$ 's parent.

2.2.2 **If**  $\text{key}(p) \leq \text{key}(c)$ , then **HOP** is restored.

Else, swap nodes  $c$  and  $p$ . [ "upwards" along  $n$ 's *ancestor path* ]

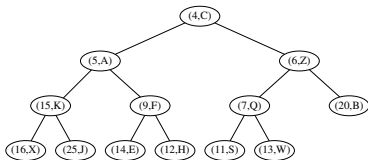
## Running Time?

- All sub-steps in 1, as well as steps 2.1, 2.2.1, and 2.2.2 take  $O(1)$ .
- Step 2.2 may be executed up to  $O(h)$  (or  $O(\log n)$ ) times.

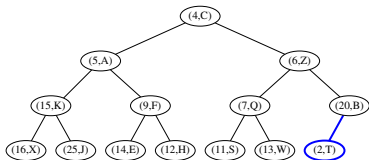
[  $O(\log n)$  ]

# Updating a Heap: Insertion Example (1.1)

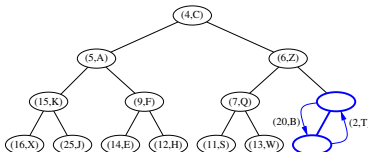
(0) A heap with height 3.



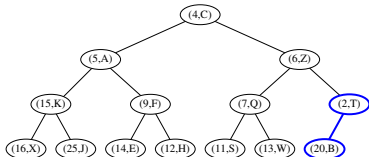
(1) Insert a new entry (2, T)  
as the **right-most** node at Level 3.  
Perform **up-heap bubbling** from here.



(2) **HOP** violated  $\because 2 < 20 \therefore$  Swap.

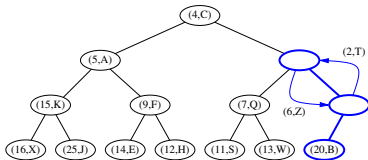


(3) After swap, entry (2, T) prompted up.

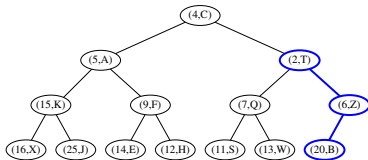


# Updating a Heap: Insertion Example (1.2)

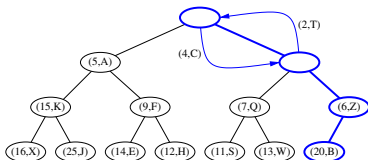
(4) **HOP** violated  $\because 2 < 6 \therefore$  Swap.



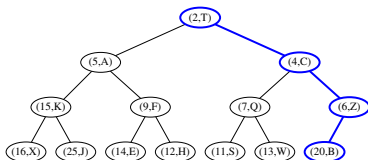
(5) After swap, entry (2, T) prompted up.



(6) **HOP** violated  $\because 2 < 4 \therefore$  Swap.



(7) Entry (2, T) becomes root  $\therefore$  Done.



# Updating a Heap: Deletion

To delete the **root** (with the *minimal* key) from a heap with *height*  $h$ :

1. Delete the **root**, possibly temporarily breaking **HOP**.

1.1 Let the *right-most* node at *Level*  $h$  be  $n$ .

1.2 Replace the **root**'s entry by  $n$ 's entry.

1.3 Delete  $n$ .

After steps 1.1 – 1.3, the *structural property* is maintained.

2. Restore **HOP** using *Down-Heap Bubbling* :

2.1 Let  $p$  be the **root**.

2.2 While **HOP** is not restored and  $p$  is not external:

2.2.1 **IF**  $p$  has no **right child**, let  $c$  be  $p$ 's *left child*.

**Else**, let  $c$  be  $p$ 's child with a *smaller key value*.

2.2.2 **If**  $\text{key}(p) \leq \text{key}(c)$ , then **HOP** is restored.

**Else**, swap nodes  $p$  and  $c$ . [ “downwards” along a *root-to-leaf path* ]

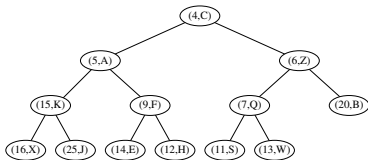
## Running Time?

- All sub-steps in 1, as well as steps 2.1, 2.2.1, and 2.2.2 take  $O(1)$ .
- Step 2.2 may be executed up to  $O(h)$  (or  $O(\log n)$ ) times.

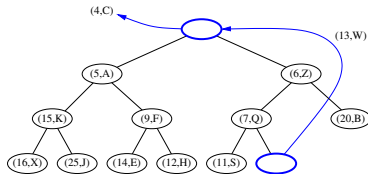
[  $O(\log n)$  ]

# Updating a Heap: Deletion Example (1.1)

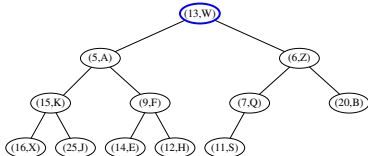
(0) Start with a heap with height 3.



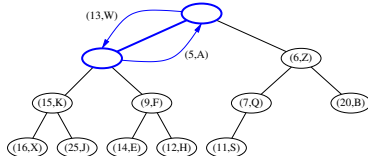
(1) Replace root with (13, W) and delete **right-most** node from Level 3.



(2) (13, W) becomes the root. Perform **down-heap bubbling** from here.

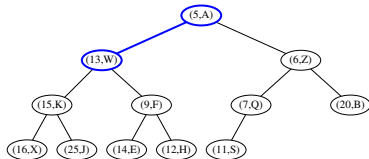


(3) Child with smaller key is (5, A). **HOP** violated  $\because 13 > 5 \therefore$  Swap.

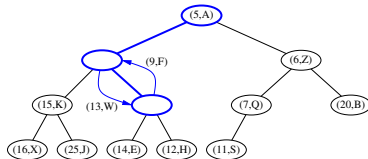


# Updating a Heap: Deletion Example (1.2)

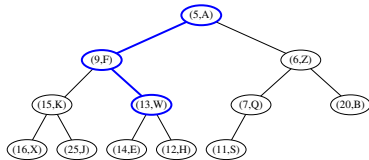
(4) After swap, entry (13, W) demoted down.



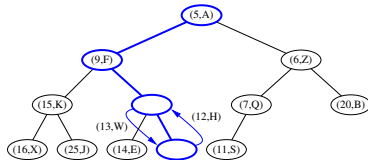
(5) Child with smaller key is (9, F).  
**HOP** violated  $\because 13 > 9 \therefore$  Swap.



(6) After swap, entry (13, W) demoted down.

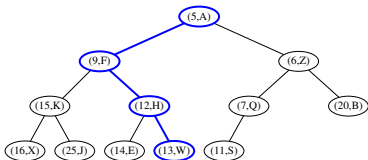


(7) Child with smaller key is (12, H).  
**HOP** violated  $\because 13 > 12 \therefore$  Swap.



# Updating a Heap: Deletion Example (1.3)

(8) After swap, entry (13, W) becomes an external node  $\therefore$  Done.



# Heap-Based Implementation of a PQ

PQ Method	Heap Operation	RT
min	root	$O(1)$
insert	insert then up-heap bubbling	$O(\log n)$
removeMin	delete then down-heap bubbling	$O(\log n)$

# Top-Down Heap Construction: List of Entries is Not Known in Advance

**Problem:** Build a heap out of  $N$  entries, supplied one at a time.

- Initialize an *empty heap*  $h$ . [  $O(1)$  ]
- As each new entry  $e = (k, v)$  is supplied, insert  $e$  into  $h$ .
  - Each insertion triggers an *up-heap bubbling* step, which takes  $O(\log n)$  time. [  $n = 0, 1, 2, \dots, N - 1$  ]
  - There are  $N$  insertions.

$\therefore$  Running time is  $O(N \cdot \log N)$

# Bottom-Up Heap Construction: List of Entries is Known in Advance

**Problem:** Build a heap out of  $N$  entries, supplied all at once.

- **Assume:** The resulting heap will be *completely filled* at all levels.  
 $\Rightarrow N = 2^{h+1} - 1$  for some *height*  $h \geq 1$   $[h = (\log(N + 1)) - 1]$
- Perform the following steps called *Bottom-Up Heap Construction*:

**Step 1:** Treat the first  $\frac{N+1}{2^1}$  list entries as heap roots.

$\therefore \frac{N+1}{2^1}$  heaps with height 0 and size  $2^1 - 1$  constructed.

**Step 2:** Treat the next  $\frac{N+1}{2^2}$  list entries as heap roots.

◇ Each *root* sets two heaps from **Step 1** as its *LST* and *RST*.

◇ Perform *down-heap bubbling* to restore **HOP** if necessary.

$\therefore \frac{N+1}{2^2}$  heaps, each with height 1 and size  $2^2 - 1$ , constructed.

...

**Step  $h + 1$ :** Treat next  $\frac{N+1}{2^{h+1}} = \frac{(2^{h+1}-1)+1}{2^{h+1}} = 1$  list entry as heap root.

◇ Each *root* sets two heaps from **Step  $h$**  as its *LST* and *RST*.

◇ Perform *down-heap bubbling* to restore **HOP** if necessary.

$\therefore \frac{N+1}{2^{h+1}} = 1$  heap, each with height  $h$  and size  $2^{h+1} - 1$ , constructed.

# Bottom-Up Heap Construction: Example (1.1)

- Build a heap from the following list of 15 keys:

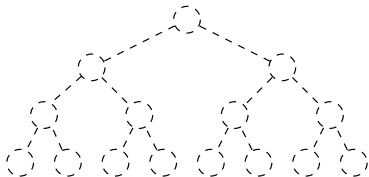
$\langle 16, 15, 4, 12, 6, 7, 23, 20, 25, 9, 11, 17, 5, 8, 14 \rangle$

- The resulting heap has:
  - Size  $N$**  is 15
  - Height  $h$**  is  $(\log(15 + 1)) - 1 = 3$
- According to the **bottom-up heap construction** technique, we will need to perform  $h + 1 = 4$  steps, utilizing 4 sublists:

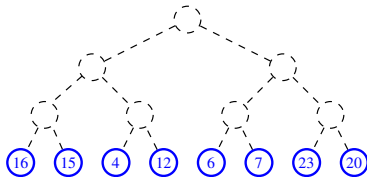
$$\langle \underbrace{16, 15, 4, 12, 6, 7, 23, 20}_{\frac{15+1}{2^1} = 8}, \underbrace{25, 9, 11, 17}_{\frac{15+1}{2^2} = 4}, \underbrace{5, 8}_{\frac{15+1}{2^3} = 2}, \underbrace{14}_{\frac{15+1}{2^4} = 1} \rangle$$

# Bottom-Up Heap Construction: Example (1.2)

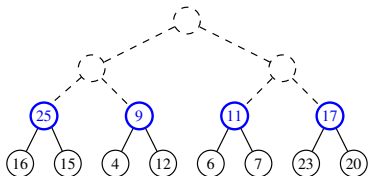
We know in advance to build a heap with height 3 and size  $2^{3+1} - 1 = 15$



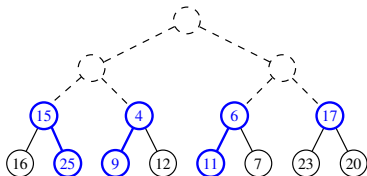
**(Step 1)** Treat first  $\frac{15+1}{2^1}$  entries as roots.  
 $\therefore$  8 one-node heaps.



**(Step 2)** Treat next  $\frac{15+1}{2^2}$  entries as roots.  
 Set LST and RST of each root.

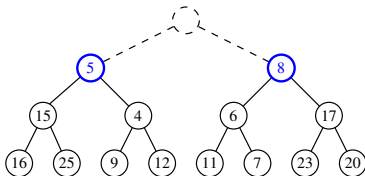


**(Step 2 cont.)** Down-heap bubbling.  
 $\therefore$  4 three-node heaps.

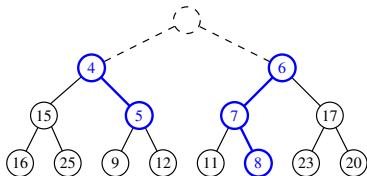


# Bottom-Up Heap Construction: Example (1.3)

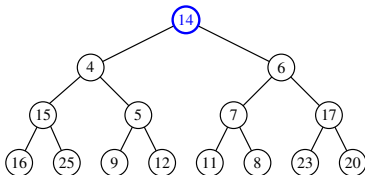
**(Step 3)** Treat next  $\frac{15+1}{2^3}$  entries as roots.  
Set LST and RST of each root.



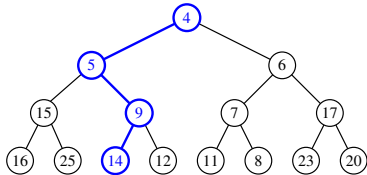
**(Step 3 cont.)** Down-heap bubbling.  
 $\therefore$  2 three-node heaps.



**(Step 4)** Treat next  $\frac{15+1}{2^4}$  entry as roots.  
Set LST and RST of each root.



**(Step 4 cont.)** Down-heap bubbling.  
 $\therefore$  1 fifteen-node heap.



# RT of Bottom-Up Heap Construction

- Intuitively, the majority of the intermediate roots from which we perform **down-heap bubbling** are of very **small height values**:
  - The first  $\frac{n+1}{2}$  **1**-node heaps with **height 0** require **no** down-heap bubbling.  
[ About 50% of the list entries processed ]
  - Next  $\frac{n+1}{4}$  **3**-node heaps with **height 1** require down-heap bubbling.  
[ Another 25% of the list entries processed ]
  - Next  $\frac{n+1}{8}$  **7**-node heaps with **height 2** require down-heap bubbling.  
[ Another 12.5% of the list entries processed ]
  - ...
  - Next two  $\frac{N-1}{2}$ -node heaps with **height (h - 1)** require down-heap
  - Final one **N**-node heaps with **height h** requires down-heap bubbling.
- Running Time of the **Bottom-Up Heap Construction** takes only  **$O(n)$** .

# The Heap Sort Algorithm

## Sorting Problem:

Given a list of  $n$  numbers  $\langle a_1, a_2, \dots, a_n \rangle$ :

Precondition: NONE

Postcondition: A permutation of the input list  $\langle a'_1, a'_2, \dots, a'_n \rangle$  sorted in a non-descending order (i.e.,  $a'_1 \leq a'_2 \leq \dots \leq a'_n$ )

The **Heap Sort** algorithm consists of two phases:

1. **Construct** a **heap** of size  $N$  out of the input array.
  - Approach 1: Top-Down “Continuous-Insertions” [  $O(N \cdot \log N)$  ]
  - Approach 2: Bottom-Up Heap Construction [  $O(N)$  ]
2. **Delete**  $N$  entries from the heap.
  - Each deletion takes  $O(\log N)$  time.
  - 1st deletion extracts the **minimum**, 2nd deletion the 2nd **minimum**, ...  
 $\Rightarrow$  Extracted **minimums** from  $N$  deletions form a **sorted** sequence.

$\therefore$  Running time of the **Heap Sort** algorithm is  $O(N \cdot \log N)$ .

# The Heap Sort Algorithm: Exercise

**Sort** the following array of integers

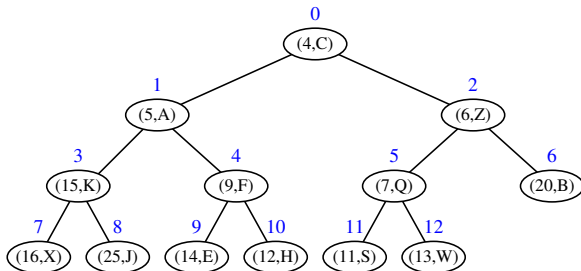
$\langle 16, 15, 4, 12, 6, 7, 23, 20, 25, 9, 11, 17, 5, 8, 14 \rangle$

into a **non-descending** order using the **Heap Sort Algorithm**.

Demonstrate:

1. Both top-down and bottom-up heap constructions in Phase 1
2. Extractions of minimums in Phase 2

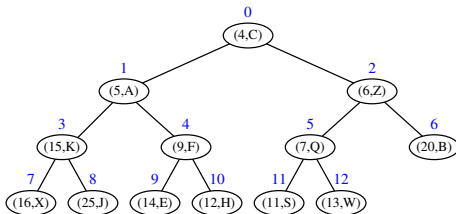
# Array-Based Representation of a CBT (1)



$$index(x) = \begin{cases} 0 & \text{if } x \text{ is the root} \\ 2 \cdot index(\text{parent}(x)) + 1 & \text{if } x \text{ is a left child} \\ 2 \cdot index(\text{parent}(x)) + 2 & \text{if } x \text{ is a right child} \end{cases}$$

(4,C)	(5,A)	(6,Z)	(15,K)	(9,F)	(7,Q)	(20,B)	(16,X)	(25,J)	(14,E)	(12,H)	(11,S)	(13,W)
0	1	2	3	4	5	6	7	8	9	10	11	12

# Array-Based Representation of a CBT (2)



(4,C)	(5,A)	(6,Z)	(15,K)	(9,F)	(7,Q)	(20,B)	(16,X)	(25,J)	(14,E)	(12,H)	(11,S)	(13,W)
0	1	2	3	4	5	6	7	8	9	10	11	12

- **Q1:** Where are nodes at **Levels 0 .. h - 1** stored in the array?  
Indices  $0 \dots (2^h - 2) \equiv 0 \dots (2^{\lfloor \log_2 N \rfloor} - 2)$  [e.g., Indices  $0 \dots 2^3 - 2$ ]
- **Q2:** Where are nodes at **Level h** stored in the array?  
Indices  $2^h - 1 \dots (N - 1) \equiv 2^{\lfloor \log_2 N \rfloor} - 1 \dots (N - 1)$  [e.g., Indices  $7 \dots 12$ ]
- **Q3:** How do we determine if a non-root node  $x$  is a **left or right child**?  
**IF**  $\text{index}(x) \% 2 == 1$  **THEN** *left* **ELSE** *right*
- **Q4:** Given a non-root node  $x$ , how do we determine the **index of  $x$ 's parent**?  
**IF**  $\text{index}(x) \% 2 == 1$  **THEN**  $\frac{\text{index}(x) - 1}{2}$  **ELSE**  $\frac{\text{index}(x) - 2}{2}$

# Index (1)

**Learning Outcomes of this Lecture**

**Balanced Binary Search Trees: Motivation**

**Balanced Binary Search Trees: Definition**

**What is a Priority Queue?**

**The Priority Queue (PQ) ADT**

**Two List-Based Implementations of a PQ**

**Heaps**

**Heap Property 1: Relational**

**Heap Property 2: Structural**

**Heaps: More Examples**

**Heap Operations**

## Index (2)

**Updating a Heap: Insertion**

**Updating a Heap: Insertion Example (1.1)**

**Updating a Heap: Insertion Example (1.2)**

**Updating a Heap: Deletion**

**Updating a Heap: Deletion Example (1.1)**

**Updating a Heap: Deletion Example (1.2)**

**Updating a Heap: Deletion Example (1.3)**

**Heap-Based Implementation of a PQ**

**Top-Down Heap Construction:**

**List of Entries is Not Known in Advance**

**Bottom-Up Heap Construction:**

**List of Entries is Known in Advance**

## Index (3)

**Bottom-up Heap Construction: Example (1.1)**

**Bottom-up Heap Construction: Example (1.2)**

**Bottom-up Heap Construction: Example (1.3)**

**RT of Bottom-up Heap Construction**

**The Heap Sort Algorithm**

**The Heap Sort Algorithm: Exercise**

**Array-Based Representation of a CBT (1)**

**Array-Based Representation of a CBT (2)**

# Wrap-Up



EECS2101 X & Z:  
Fundamentals of Data Structures  
Winter 2025

CHEN-WEI WANG

# What You Learned (1)

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- ***Java Programming***

- JUnit
- Recursion
- Generics

# What You Learned (2)

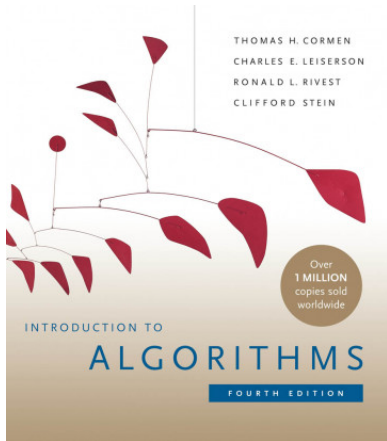
- ***Data Structures***

- Arrays
- Circular Arrays
- Singly-Linked Lists and Doubly-Linked Lists
- Stacks, Queues, Double-Ended Queues
- Trees, Binary Trees, Binary Search Trees, Balanced BSTs
- Priority Queues and Heaps

- ***Algorithms***

- Asymptotic Analysis
- Binary Search
- Insertion Sort, Selection Sort, Merge Sort, Quick Sort, Heap Sort
- Pre-order, in-order, and post-order traversals

# Beyond this course... (1)



- *Introduction to Algorithms (4th Ed.)* by Cormen, etc.
- DS by DS, Algo. by Algo.:
  - **Understand** math analysis
  - **Read** pseudo code
  - **Implement** in Java
  - **Test** in JUnit

## Beyond this course... (2)

A tutorial on building a language compiler using Java (from **EECS4302-F22**):

***Using the ANTLR4 Parser Generator to Develop a Compiler***

- Trees
- Recursion
- Composite & Visitor Design Patterns

# Wish You All the Best



- What you have learned will be **assumed** in the third year.
- Some topics we did not cover:
  - Hash table [ See Weeks 10 – 11 of EECS2030-F19 ]
  - Graphs [ EECS3101 ]
- If you're interested in taking a more advanced course with me:
  - **EECS3342** System Specification & Refinement [ F'25 ]  
Applying EECS1090 to construct & verify software systems
  - **EECS3101** Design and Analysis of Algorithm [ F'25, W'26 ]