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# Arrays

## Basic Knowledge

- Create:  $x = [1, 2, 3, 4]$

1	2	3	4
---	---	---	---

- Write:  $x[2] = 8$

1	2	8	4
---	---	---	---

- Read:  $x[3]$  # will read as 4

## In Java

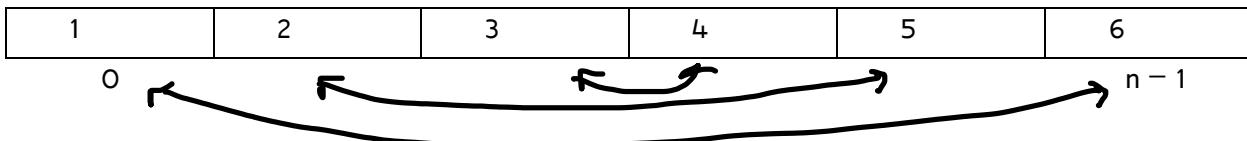
- Fixed Size → cannot change after array is created
- Type specific

## Runtime

Big O								
Add			Remove			Search		Resize
1 <sup>st</sup> index	Last index	Any index	1 <sup>st</sup> index	Last index	Any index	At Index	Specific	
$O(n)$	$O(n) \rightarrow O(1)$	$O(n)$	$O(n)$	$O(1)$	$O(n)$	$O(1)$	$O(n)$	$O(n)$
Shift elements to the right/resizing	Worst: array to small Amortized: simple insert	Shift right elements/resizing	Shift elements to the left		Shift right elements to the left	Grab at index	Iterate through n elements until find specific value wanted	Loop through n elements

## Reverse an Array

Using Python



```
def reverse(arr):
    n = len(arr)
    left = 0
```

```
right = n - 1

while left < right:
    arr(left), arr(right) = arr(right), arr(left)  # swap elements
    left = left + 1
    right = right - 1
```

## Runtime

time complexity:  $O(n)$       # loop through the whole arr  
space complexity:  $O(1)$

# Array List

## In Java

- Dynamic Size → can resize when elements are added or removed
- Generic Types → Can hold elements of different types

## Runtime

Big O								
Add			Remove			Search		Resize
1 <sup>st</sup> index	Last index	Any index	1 <sup>st</sup> index	Last index	Any index	At Index	Specific	
O(n)	O(n)→O(1)	O(n)	O(n)	O(1)	O(n)	O(1)	O(n)	O(n)
Shift elements to the right/resizing	Worst: no more capacity so resize and add to back Amortized: simple insert	Shift right elements/resizing	Shift elements to the left		Shift right elements to the left	Grab at index	Iterate through n elements until find specific value wanted	Loop through n elements

## Code in Java

### Instance Variables

1. INITIAL\_CAPACITY
2. T[] backingArray
3. int size

### Constructor

```
public ArrayList() {
    backingArray = (T[]) new Object[INITIAL_CAPACITY];
    size = 0;
}
```

### Add to Front

```
public void addToFront(T data) {
    if (data == null) {
        throw new IllegalArgumentException("error");
    }
    if (size == 0) {
        backingArray[0] = data;
    } else {
        if (size == backingArray.length) { // backingArray too small
            T[] resize = (T[]) new Object[backingArray.length * 2];
            for (int i = 0; i < size; i++) {
                resize[i] = backingArray[i];
            }
            backingArray = resize;
        }
        for (int i = size; i > 0; i--) {
            backingArray[i] = backingArray[i - 1];
        }
        backingArray[0] = data;
    }
    size++;
}
```

```

        resize[0] = data;
        for (int i = 0; i < size; i++) {
            resize[i + 1] = backingArray[i];
        }
        backingArray = resize;
    } else {
        for (int i = size - 1; i > 0; i--) {
            backingArray[i + 1] = backingArray[i];
        }
        backingArray[0] = data;
    }
}
size++;
}

```

## Add to Back

```

public void addToBack(T data) {
    if (data == null) {
        throw new IllegalArgumentException("error");
    }
    if (size == 0) {
        backingArray[0] = data;
    } else {
        if (size == backingArray.length) {
            T[] resize = (T[]) new Object[backingArray.length];
            for (int i = 0; i < size; i++) {
                resize[i] = backingArray[i];
            }
            resize[size] = data;
            backingArray = resize;
        } else {
            backingArray[size] = data;
        }
    }
    size++;
}

```

## Add at Index

```

public void addAtIndex(int index, T data) {
    if (data == null) {
        throw new IllegalArgumentException("error");
    } else if (index < 0 || index > size) {
        throw new IndexOutOfBoundsException("index has to be [0, size)");
    }
    if (index == size && size != backingArray.length) {
        backingArray[size] = data;
    } else {
        if (size == backingArray.length) {
            T[] resize = (T[]) new Object[backingArray.length * 2];
            for (int i = 0; i < index; i++) {
                resize[i] = backingArray[i];
            }
            resize[index] = data;
            for (int i = index; i < size; i++) {

```

```

        resize[i + 1] = backingArray[i];
    }
    backingArray = resize;
} else {
    for (int i = size - 1; i > index; i--) {
        backingArray[i + 1] = backingArray[i];
    }
    backingArray[index] = data;
}
size++;
}
}

```

## Remove at Index

```

public T removeAtIndex(int index) {
    if (index < 0 || index >= size) {
        throw new IndexOutOfBoundsException("not in bounds");
    }
    T node = backingArray[index];
    if (index == size - 1) { // last element
        backingArray[size - 1] = null;
    } else {
        for (int i = index; i < size - 1; i++) {
            backingArray[i] = backingArray[i + 1];
        }
        backingArray[size - 1] = null;
    }
    size--;
    return node;
}

```

## Remove from Front

```

public T removeFromFront() {
    if (size == 0) {
        throw new NoSuchElementException("List cannot be empty");
    }
    T node = backingArray[0];
    for (int i = 0; i < size - 1; i++) {
        backingArray[i] = backingArray[i + 1];
    }
    backingArray[size - 1] = null;
    size--;
    return node;
}

```

## Remove from Back

```

public T removeFromBack() {
    if (size == 0) {
        throw new NoSuchElementException("list is empty");
    }
    T node = backingArray[size - 1];
    backingArray[size - 1] = null;
    size--;
}

```

```
        return node;  
    }
```

# Linked Lists

## Singly Linked Lists



- May or may not have a tail

### Operations no Tail

- Add
  - To Front:
    - Create a new node
    - Set its "next" pointer to the current head
    - Set head to new node
  - To Last:
    - Traverse list to last node
    - Update its "next" pointer to the new node
  - Any Index: Traverse list to index position.
- Remove
  - At Front:
    - Store current head to temp
    - Set head to head.next
    - Free temp.
  - At Last:
    - Set current to head
    - While the current.next.next is not null, current = current.next
    - Then, current.next = null
  - Any Index:
    - Traverse list to find previous node of target
    - Set previous.next to target.next. This will then jump over the target position.
- Search:
  - At Index: Iterate until you get to the position.
  - For Value: Iterate until you find the value

## Operations with Tail

- Add
  - To Front:
    - Create a new node
    - Set its “next” pointer to the current head
    - Set head to new node
  - To Last:
    - Create a new node, set it to tail.next
    - Set tail to new node
  - Any Index: Traverse list to index position
- Remove
  - At Front:
    - Store current head to temp
    - Set head to head.next
    - Free temp.
  - At Last:
    - Set current to head
    - While the current.next.next is not null, current = current.next
    - Then, current.next = null
  - Any Index:
    - Traverse list to find previous node of target
    - Set previous.next to target.next. This will then jump over the target position.
- Search
  - At Index: Iterate until you get to the position.
  - For Value: Iterate until you find the value

## Runtime

Big O no Tail								
Add			Remove			Search		Resize
<u>1<sup>st</sup> index</u>	<u>Last index</u>	<u>Any index</u>	<u>1<sup>st</sup> index</u>	<u>Last index</u>	<u>Any index</u>	<u>At Index</u>	<u>Specific</u>	
O(1)	O(n)	O(n)	O(1)	O(n)	O(n)	O(n)	O(n)	N/A

Big O with Tail						
Add		Remove		Search		Resize

<u>1<sup>st</sup> index</u>	<u>Last index</u>	<u>Any index</u>	<u>1<sup>st</sup> index</u>	<u>Last index</u>	<u>Any index</u>	<u>At Index</u>	<u>Specific</u>	
O(1)	O(1)	O(n)	O(1)	O(n)	O(n)	O(n)	O(n)	N/A

## Doubly Linked Lists



### Operations

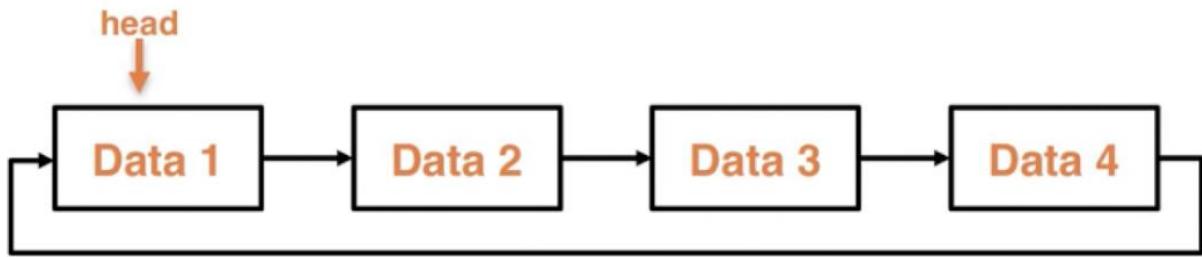
- Add
  - To Front:
    - Create new node
    - Set current head.previous to new node
    - Set new node.next to head
    - Set head to new node
  - To Last:
    - Create new node
    - Set tail.next to new node
    - Set new node.previous to tail
    - Set tail to new node
  - Any Index:
    - If index <= size / 1, traverse from head; else from tail
      - Create new node
      - Traverse to index node
      - curr.getPrevious.setNext(new node)
      - new node.setPrevious(curr.getPrevious)
      - curr.getNext.setPrevious(new node)
      - new node.setNext(curr.getNext)
- Remove
  - At Front:
    - head = head.next
    - head.previous = null
  - At Last:
    - tail = tail.previous
    - tail.next = null

- Any Index:
  - If index  $\leq$  size / 2: traverse from head; else from tail
    - traverse to index node
    - remove = index node
    - index node.getPrevious.setNext(remove.getNext())
    - index node.getNext.setPrevious(remove.getPrevious())
    - index node = null
- Search:
  - At Index: Iterate until you get to the position (depends on index  $\leq / \geq$  size/2)
  - For Value: Iterate until you find the value

## Runtime

Big O								
Add			Remove			Search		Resize
1 <sup>st</sup> index	Last index	Any index	1 <sup>st</sup> index	Last index	Any index	At Index	Specific	
O(1)	O(1)	O(n)	O(1)	O(1)	O(n)	O(n)	O(n)	N/A

## Circular Linked List



## Circular Singly Linked List Operations

- Add
  - To Front:
    - Create new empty node at index 1 (after head)
    - move data from head to new node
    - add new data into the head node
  - To Last:
    - Create new empty node at index 1 (after head)
    - Copy data from head to new node
    - Add new data into the head node
    - Move head to head.next
  - Any Index: Follow SLL
- Remove
  - At Front:
    - Move data from head.next into head
    - Point head.next to head.next.next
  - At Last:
    - Use loop to stop one short of last node (follow SLL)
  - Any Index:
    - Traverse until reach node you want to remove (follow SLL)
- Search:
  - At Index: Traverse
  - For Value: Traverse

## Runtime

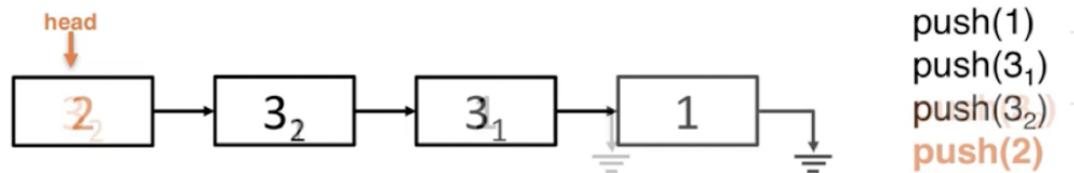
Big O

Add			Remove			Search		Resize
1 <sup>st</sup> index	Last index	Any index	1 <sup>st</sup> index	Last index	Any index	At Index	Specific	
O(1)	O(1)	O(n)	O(1)	O(n)	O(n)	O(n)	O(n)	N/A

## Stacks

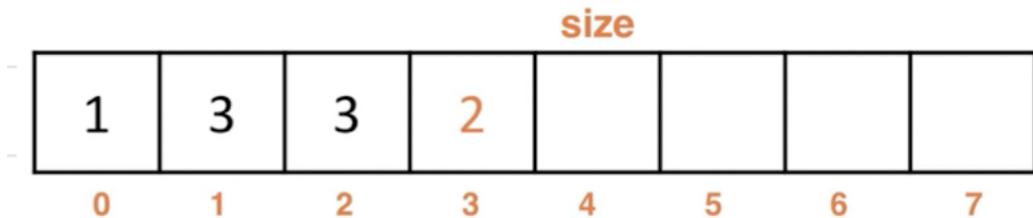
- Follows LIFO (last in, first out)
- NOT FOR SEARCHING

### Linked-List Backed Stacks



- Push and pop from same end of the structure (head)

### Array Backed Stacks



- Push and pop based on size location

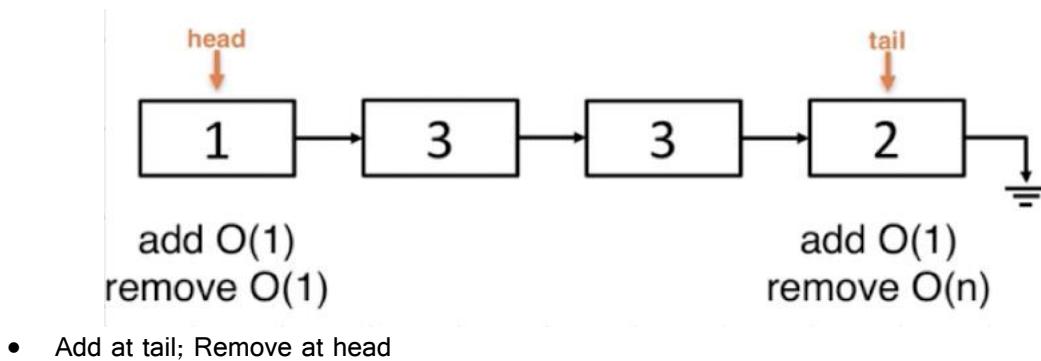
Big O							
	Top of Stack	Push	Pop	Check Empty	Get Size	Clear	Resize
LL	Head	O(1)	O(1)	O(1)	O(1)	O(1)	O(1)
Array	Size - 1	O(1) -> O(n) resize	O(1)	O(1)	O(1)	O(1)	O(n)

### Runtime

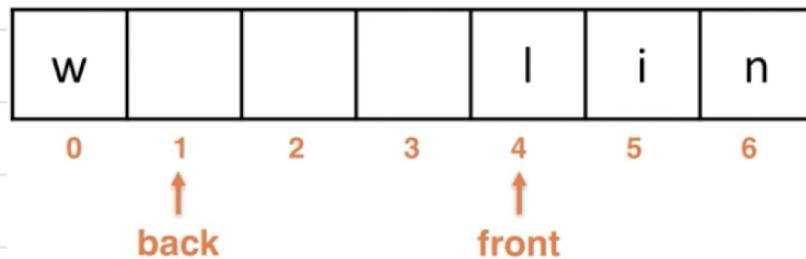
## Queues

- Follows FIFO (first in, first out)
- NOT FOR SEARCHING

### Linked-List Backed Queues



### Array Backed Queues



### LL/Array Runtime

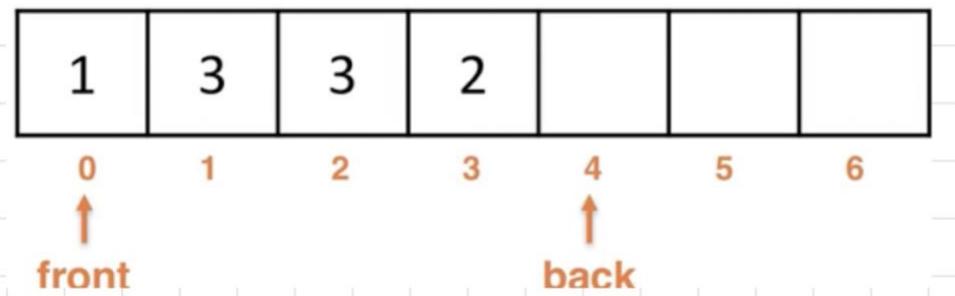
Big O							
	Front of Queue	Back of Queue	Push	Pop	Check Empty	Get Size	Clear
LL	Head	Tail	O(1)	O(1)	O(1)	O(1)	O(1)
Array	Front Index	Back Index	O(1)*	O(1)	O(1)	O(1)	O(1)

### Priority Queues

- Uses Heaps
- Remove elements with highest priority

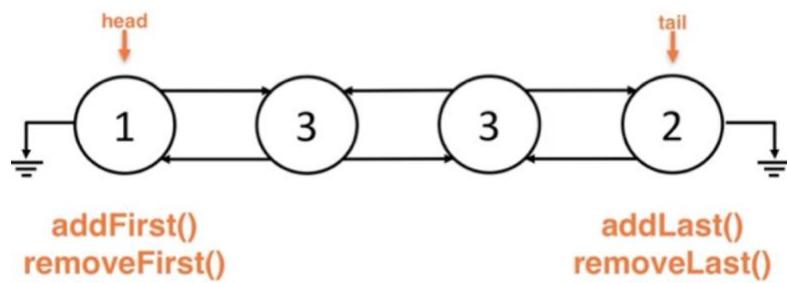
## Array Backed Deques

- Add and remove from both ends



- Add Last: increment back (mod by capacity)
- Remove First: increment front (mod by capacity)
- Add First: check if < 0 → add at front
- Remove Last: check if < 0 → decrement back

## Linked-List Backed Deques



- Add First: Add to Head
- Add Last: Add to Tail
- Remove First: Remove from Head
- Remove Last: Remove from Tail

# BST

- Depth: What sublevel the nodes are on
- Height: How many levels of children does the node have
  - Height(node) = (Max child height) + 1
  - Leaf = 0

## Binary Trees

- Has no order

### Definition

- 2 children max
- Children labeled left and right
- Left precedes right

### Shape Properties



### Traversals

- Depth Traversal (stack-based)
  - Preorder (SLR)
    - Uniquely identifies a BST
  - Inorder (LSR)
    - Returns in sorted order
  - Postorder (LRS)
    - To destroy
- Breadth Traversal (queue-based)
  - Level-Order (by depth)
    - Uniquely defines a BST

## Binary Search Tree

- Has tree in order

### Search

- Average search:  $O(\log n)$  – do left, right comparison
- Degenerate:  $O(n)$
- Behavior:
  - Data < node.data: Traverse to left child
  - Data > node.data: Traverse to right child
  - Data = node.data: Data found, terminate
  - Node is null: Data not in BST

```
public boolean contains(T data) {  
    if (data == null) {  
        throw new IllegalArgumentException("Data cannot be null.");  
    }  
    return containsRecursiveHelper(root, data);  
}  
  
/**  
 * Recursive contains helper method that searches for a data entry in a BST  
 *  
 * @param currNode current node in the recursive method  
 * @param data the data to search and see if exists  
 * @return true if data is contained, false otherwise  
 */  
  
private boolean containsRecursiveHelper(BSTNode<T> currNode, T data) {  
    if (currNode == null) {  
        //base case: data wasn't found  
        return false;  
    }  
    if (data.compareTo(currNode.getData()) == 0) {  
        //case 1: data was found  
        return true;  
    } else if (data.compareTo(currNode.getData()) < 0) {  
        //case 2: data was not found, search left subtree  
        return containsRecursiveHelper(currNode.getLeft(), data);  
    } else {  
        //case 3: data was not found, search right subtree  
        return containsRecursiveHelper(currNode.getRight(), data);  
    }  
}
```

```

    //case 2: data is less than currNode
    return containsRecursiveHelper(currNode.getLeft(), data);
} else {
    //case 3: data is greater than currNode
    return containsRecursiveHelper(currNode.getRight(), data);
}
}

```

## Add

- Average Add:  $O(\log n)$  – do left, right comparison (skip about half of the nodes)
- Degenerate:  $O(n)$
- Behavior:
  - Data < node.data: Traverse to left child
  - Data > node.data: Traverse to right child
  - Data = node.data: Duplicate found, do not add
  - Node is null: Data not in BST, add data

```

public void add(T data) {
    if (data == null) {
        throw new IllegalArgumentException("Data cannot be null.");
    }
    root = addRecursiveHelper(root, data);
}

/**
 * Helper method that recursively adds a new node to the BST
 *
 * @param currNode current node in the recursive method
 * @param data the data that is inserted into the tree
 * @return BST root with new data inserted
 */
private BSTNode<T> addRecursiveHelper(BSTNode<T> currNode, T data) {
    //if node is a leaf or empty tree, create new node
    if (currNode == null) {
        //increase size bc of new node
        size++;
        currNode = new BSTNode<T>(data);
    } else if (data < currNode.getData()) {
        currNode.setLeft(addRecursiveHelper(currNode.getLeft(), data));
    } else if (data > currNode.getData()) {
        currNode.setRight(addRecursiveHelper(currNode.getRight(), data));
    }
    return currNode;
}

```

```

//create BSTNode w data and return it as leaf node!
return new BSTNode<>(data);

} else {

    //compare data to be inserted w data in curr node
    if (data.compareTo(currNode.getData()) < 0) {

        //data is smaller than currNode data
        currNode.setLeft(addRecursiveHelper(currNode.getLeft(), data));

    } else if (data.compareTo(currNode.getData()) > 0) {

        //data is bigger than currNode data
        currNode.setRight(addRecursiveHelper(currNode.getRight(), data));
    }

    //return after insertion is complete
    return currNode;
}

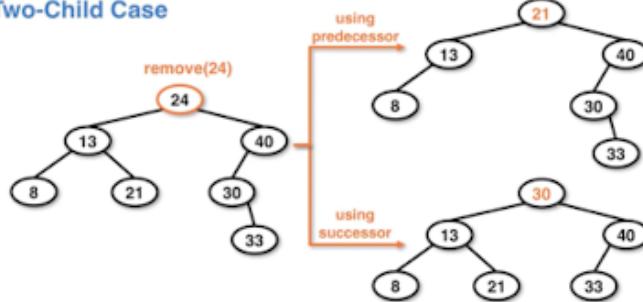
}

```

## Remove

- Average Remove:  $O(\log n)$
- Degenerate:  $O(n)$
- Behavior:
  - Data < node.data: Traverse to left child
  - Data > node.data: Traverse to right child
  - Data = node.data: Data found, remove
  - Node is null: Data not in BST, throw exception
- Zero Child
  - Remove node
- One Child
  - Child node replaces the parent
- Two Child
  - Find Predecessor
    - Next larger element that is still smaller than the data we want to remove
  - Find Successor
    - Next smallest element that is still larger than the data we want to remove

### Two-Child Case



```

public T remove(T data) {
    if (data == null) {
        throw new IllegalArgumentException("Data cannot be null.");
    }
    //node to remove
    BSTNode<T> removedNode = new BSTNode<>(null);
    root = removeRecursiveHelper(root, removedNode, data);
    return removedNode.getData();
}

/**
 * recursive helper method to remove a node from a BST
 *
 * @param currNode current node in the recursive method
 * @param toRemove node that stores the removed node
 * @param data the data that is inserted into the tree
 * @return updated BST after removing data node
 */
private BSTNode<T> removeRecursiveHelper(BSTNode<T> currNode, BSTNode<T> toRemove, T data) {
    if (currNode == null) {
        throw new NoSuchElementException("Data is not in the tree");
    } else {
        if (data.compareTo(currNode.getData()) < 0) {
            //if data < node.data traverse to left child
            currNode.setLeft(removeRecursiveHelper(currNode.getLeft(), toRemove, data));
        } else if (data.compareTo(currNode.getData()) > 0) {
            //if data > node.data traverse to right child
            currNode.setRight(removeRecursiveHelper(currNode.getRight(), toRemove, data));
        } else {
            toRemove.setData(currNode.getData());
            currNode.setData(null);
        }
    }
}

```

```

//if data is equal to node.data, data found, remove it
//getting data that is in the tree
toRemove.setData(currNode.getData());
size--;
if (currNode.getLeft() == null && currNode.getRight() == null) {
    //case 1: zero children/leaf node
    return null;
} else if (currNode.getLeft() == null) {
    //case 2.1: 1 child on right
    return currNode.getRight();
} else if (currNode.getRight() == null) {
    //case 2.2: 1 child on left
    return currNode.getLeft();
} else {
    //case 3: two children
    //successor = 1 right then all the way left
    BSTNode<T> fakeRemove = new BSTNode<>(null);
    currNode.setRight(removeSuccessorHelper(currNode.getRight(), fakeRemove));
    currNode.setData(fakeRemove.getData());
}
}

return currNode;
}

}

/**
 * Helper method to find successor node
 *
 * @param currNode current node in the recursive method
 * @param fakeRemove the node to remove successor from
 * @return successor node
 */
private BSTNode<T> removeSuccessorHelper(BSTNode<T> currNode, BSTNode<T> fakeRemove) {
    if (currNode.getLeft() == null) {
        fakeRemove.setData(currNode.getData());
        return currNode.getRight();
    } else {

```

```
        currNode.setLeft(removeSuccessorHelper(currNode.getLeft(), fakeRemove));  
        return currNode;  
    }  
}
```

# Heaps

- Binary Tree that is COMPLETE

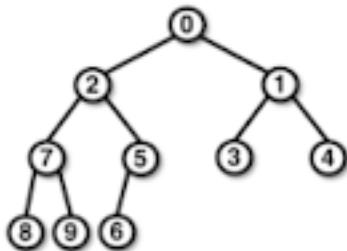
## Min Heap

- Smallest element @ root
- Children larger than parent

## Max Heap

- Largest element @ root
- Children smaller than parent

## Array Representation



X	0	2	1	7	5	3	4	8	9	6	
0	1	2	3	4	5	6	7	8	9	10	11

- Given index n:
  - Left Child:  $2 * n$
  - Right child:  $2 * n + 1$
  - Parent:  $n/2$
- Index size = last data

## Operations

### Add

- Add to next spot in array to maintain completeness
- Up-Heap from element to fix order
  - Compare data with parent

- Swap with parent if necessary
- Continue until top of heap or no swap is left

Adding the element is  $O(1)$ , upheap is  $O(\log n)$

## Remove

- Move last element of heap to replace the root
  - Delete last element
- Down-Heap from root to fix order
  - If two children, compare data with higher priority child
  - If one child (must be left), compare data with the child
  - Swap, if necessary, based on comparison
  - Continue until done/leaf is reached

Removing is  $O(1)$ , downheap is  $O(\log n)$

# Hashmaps

- Index = hash % capacity

## Maps

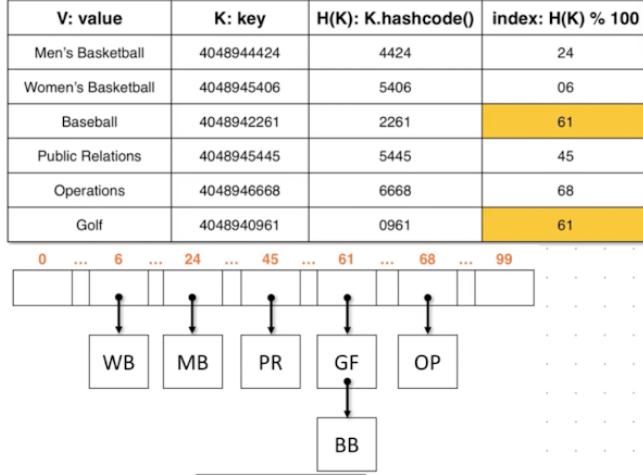
- Key-value pairs: <K, V>
- Keys are unique and immutable; Values are not unique and can be either immutable or mutable

## External Chaining

- Chain together all entries at same index and keep them in a linked list at that index

- Backing Structure:

- Array of LinkedList



## Runtime

	Add	Search	Remove	Why
<b>Best</b>	$O(1 + a)$	$O(1 + a)$	$O(1 + a)$	$a = \text{load factor}$ Efficient if load factor is kept reasonable through rehashing
<b>Worst</b>	$O(n)$	$O(n)$	$O(n)$	Degenerate linked list

## Linear Probing

- Backing Structure:
  - Array (each index store only one entry)
- If a collision occurs, increment the index by one and check again
- Probing Scenarios:
  - Valid (not null or deleted) and unequal key
  - Valid and equal key
  - Deleted
  - Null

## Runtime

	Add	Search	Remove	Why
<b>Best</b>	$O(1/(1 - a))$	$O(1/(1 - a))$	$O(1/(1 - a))$	Assuming $a$ is less than 1 and the hash function distributes keys uniformly
<b>Worst</b>	$O(n^2)$	$O(n)$	$O(n)$	$O(n^2)$ for resizing $n$ times and then having to probe $n$ times because the hash sucks. For each of the $n$ elements, if they all collide then you have to look at $n - 1$ elements before it that are in the new array you resized too, so overall $1 + 2 + \dots + n$ becomes $n^2$ $O(n)$ if you have to do +1 $n$ times

## Quadratic Probing

- Backing Structure:
  - Array
- If a collision occurs, add  $h^2$  to the original index and check again
  - $\text{Index} = (h^2 + \text{originalIndex}) \% \text{capacity} \rightarrow$  where  $h = \# \text{ times we've probed}$

### Runtime

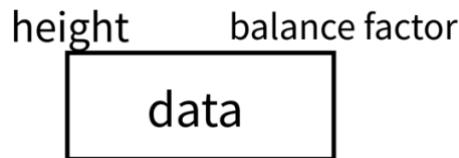
	Add	Search	Remove	Why
<b>Best</b>	$O(1/(1 - a))$	$O(1/(1 - a))$	$O(1/(1 - a))$	Assuming $a$ is less than 1 and the hash function distributes keys uniformly
<b>Worst</b>	$O(n^2)$	$O(n)$	$O(n)$	$O(n^2)$ for resizing $n$ times and then having to probe $n$ times because the hash sucks. $O(n)$ if you have to do +1 $n$ times

## Double Hashing

- Uses two hash functions to compute two different hash values for a given key
    - 1) compute the initial hash value
    - 2) compute the step size for the probing sequence
  - $(\text{hash1(key)} + i * (\text{hash2(key)}) \% \text{capacity})$
  - $\text{Hash1(key)} = \text{key \% capacity}$
- $\text{Hash2(key)} = \text{PRIME} - (\text{key \% PRIME})$

## AVL Trees

- Self-balancing BST to eliminate  $O(n)$  worst cases
  - Becomes  $O(\log n)$
- $\text{Height}(\text{node}) = \max\{\text{Height}(\text{left}), \text{Height}(\text{right})\} + 1$ 
  - $\text{Height}(\text{leaf}) = 0$
  - $\text{Height}(\text{null}) = -1$
- $\text{BalanceFactor}(\text{node}) = \text{Height}(\text{left}) - \text{Height}(\text{right})$



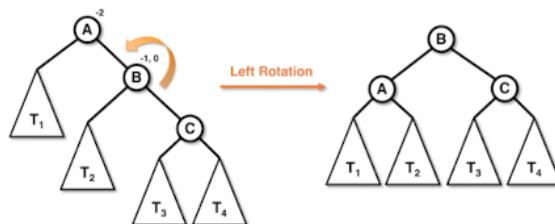
Updating individual node is  $O(1)$

## Definition of Unbalanced

- Unbalanced:  $|BF| > 1$
- Balanced:  $BF = -1, 0, 1$
- Right Heavy with Left Rotation WHEN  $BF < -1$
- Left Heavy with Right Rotation WHEN  $BF > 1$

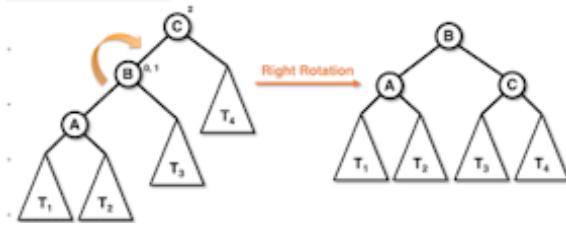
## Rotations

### Left Rotation



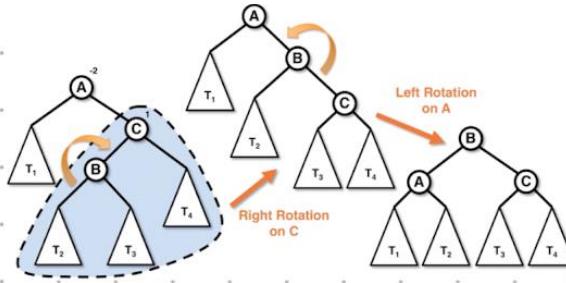
- PseudoCode: `leftRotation(aNode)`
  - `bNode = aNode.right`
  - `orphan = bNode.left`
  - `bNode.setLeft(aNode)`
  - `aNode.setRight(orphan)`
  - New height & BF A THEN B

## Right Rotation



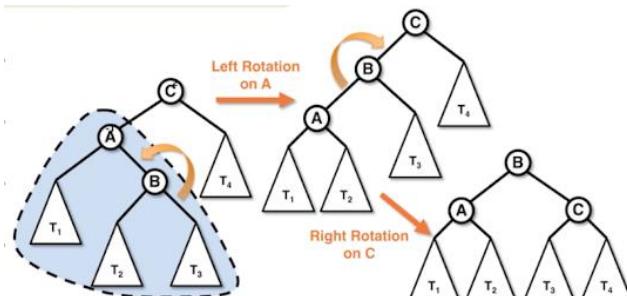
- PseudoCode: rightRotation(cNode)
  - bNode = cNode.left
  - orphan = bNode.right
  - bNode.setRight(cNode)
  - cNode.setLeft(orphan)
  - New height & BF C THEN B

## Right-Left Rotation



- PseudoCode:
  - If(BF < -1)
    - If(node.getRight.getBF > 0)
      - Node.setRight(rightRotation(node.getRight))
    - Return leftRotation(node)

## Left-Right Rotation



- PseudoCode:
  - If( $BF > 1$ )
    - If ( $node.getLeft.getBF < 0$ )
      - $Node.setLeft(leftRotation(node.getLeft))$
    - Return  $rightRotation(node)$

## Rotation Types

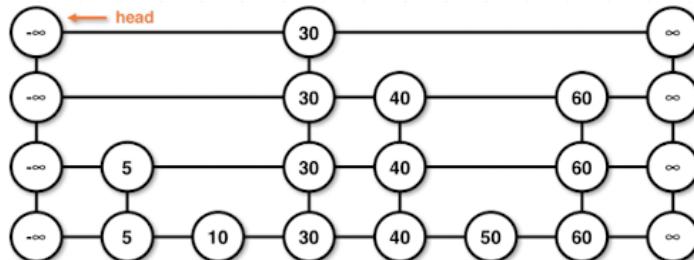
Parent BF	Heavier Child	Child BF	Rotation Shape	Rotation Type
2	Left	0, 1		Right
2	Left	-1		Left-Right
-2	Right	-1, 0		Left
-2	Right	1		Right-Left

## Runtime

	Add	Search	Remove	Why
<b>Best</b>	$O(\log n)$	$O(\log n)$	$O(\log n)$	Balances
<b>Worst</b>	$O(\log n)$	$O(\log n)$	$O(\log n)$	Balances

# SkipLists

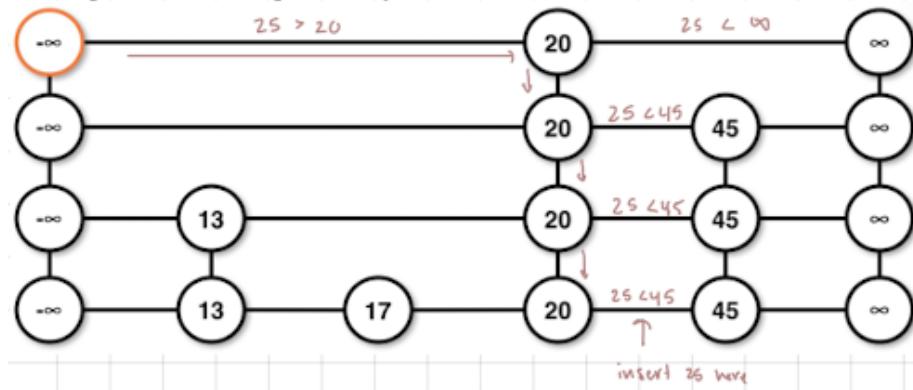
- Levels of LinkedLists
- Each node has a data pointer, and then an up/down/left/right pointer
- Sorted in descending order at each level
- Level 0 has all data of the skiplist



## Search

- Start at top left infinity node
- Look at right node (if exists) and compare with data
  - Data < right → continue right
  - Data < left → move down a level
  - Data = right → data was found
  - If at bottom, data isn't found

## Search/Add Example (add(25))



## Runtime

	Best Case	Average Case	Worst Case
Search	$O(\log n)$	$O(\log n)$	$O(n)$ – all data at L0
Add	$O(\log n)$	$O(\log n)$	$O(n)$

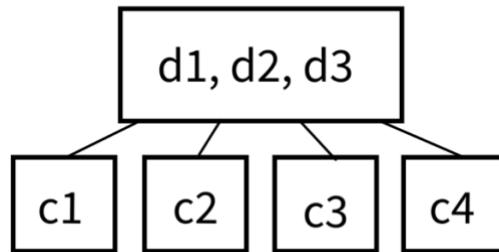
<b>Remove</b>	$O(\log n)$	$O(\log n)$	$O(n)$
<b>Space Complexity</b>	$O(n)$	$O(n)$	$O(n \log n)$

## 2-4 Trees

- B-Tree ( $B = \# \text{ elements in a node/block}$ )
- $O(\log n)$  operations

### Node and Order Properties

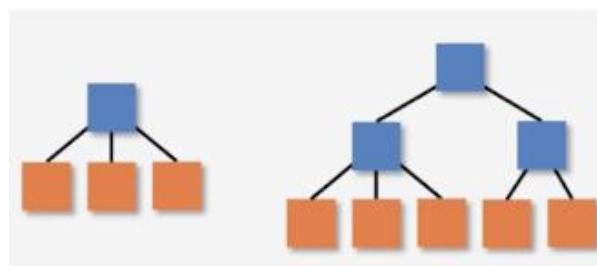
- 2-4 children
- 1-3 data values
- Node with  $m$  data values will have  $m + 1$  children



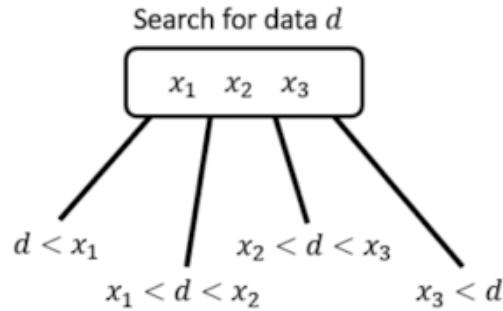
- $d_n$ : nth data
- $c_n$ : nth data in child
- Node with  $n$  children is an  $n$ -node
- Order Property:  $c1 < d1 < c2 < d2 < c3 < d3 < c4$

### Shape Properties

- All leaves have same depth
- Highest to Lowest Priority:
  - Shape → Order → Node

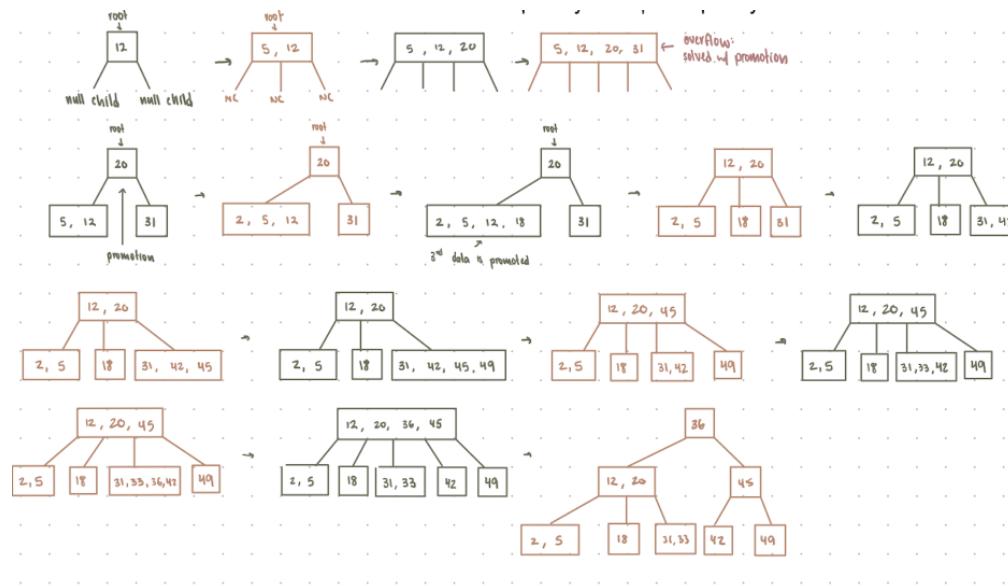


## Search

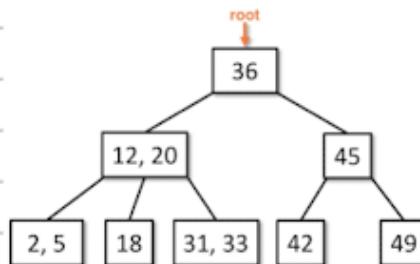


## Add and Overflow

- Promotion: Pushes Data up and split the nodes

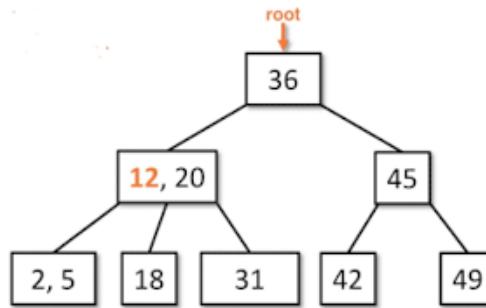


## Remove and Transfer/Fusion



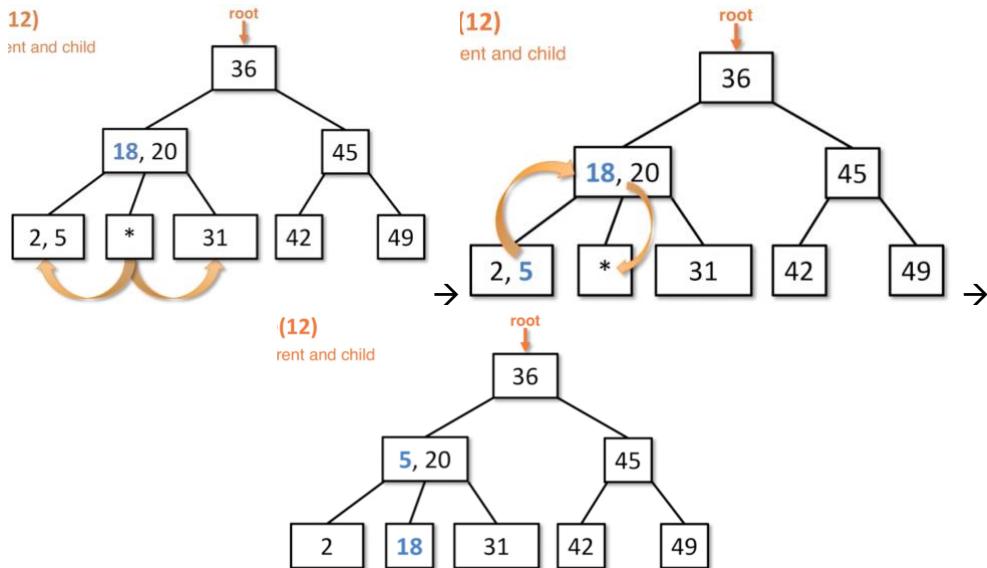
### Case 1: Deletion from Leaf with multiple data

Remove(33)



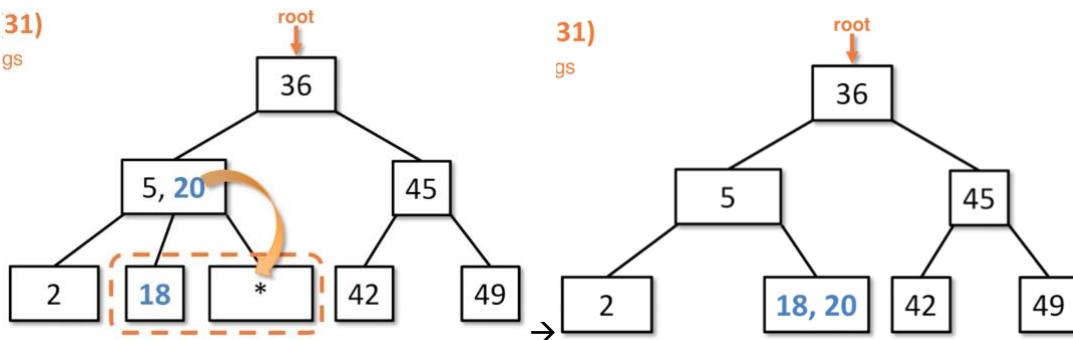
### Case 2: Transfer between parent and child with multiple data

Remove(12)



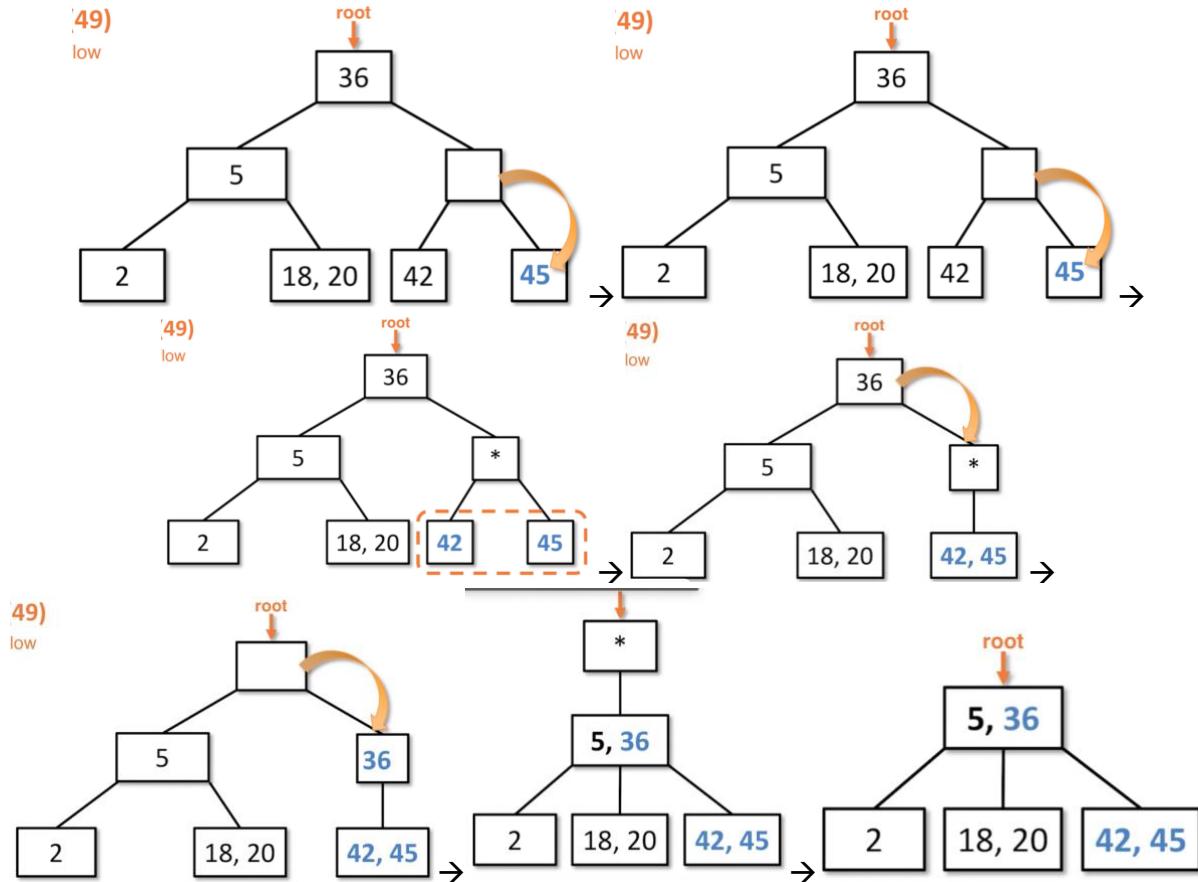
### Case 3: Fusion between siblings with single data

Remove(31)

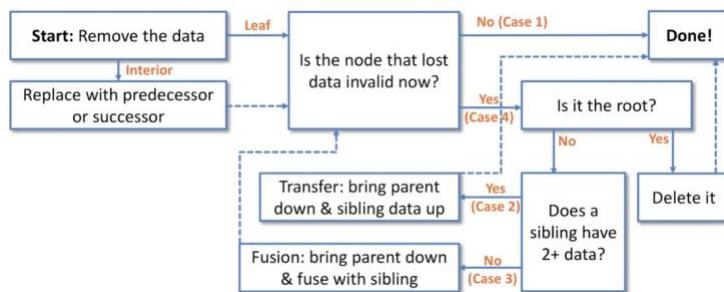


### Case 4: Propagation of Underflow with multiple fusions

Remove(49)



## Remove Flow Chart



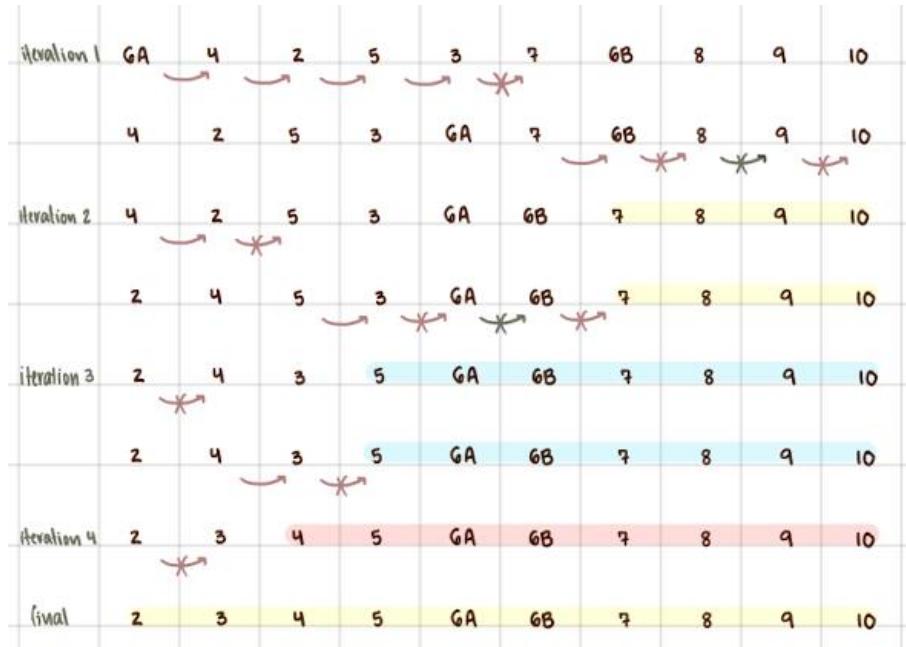
## Iterative Sorts

### Bubble Sort

- Start from beginning of an array, comparing adjacent elements, swapping if it puts it in order

```
public static <T> void bubbleSort(T[] arr, Comparator<T> comparator) {  
    if (arr == null || comparator == null) {  
        throw new IllegalArgumentException("Array or comparator cannot be null.");  
    }  
    boolean swapped;  
    for (int i = 0; i < arr.length - 1; i++) {  
        swapped = false;  
        for (int j = 0; j < arr.length - i - 1; j++) {  
            if (comparator.compare(arr[j], arr[j + 1]) > 0) {  
                swapped = true;  
                T temp = arr[j];  
                arr[j] = arr[j + 1];  
                arr[j + 1] = temp;  
  
            }  
        }  
        if (!swapped) {  
            break;  
        }  
    }  
}
```

## Example



## Runtime

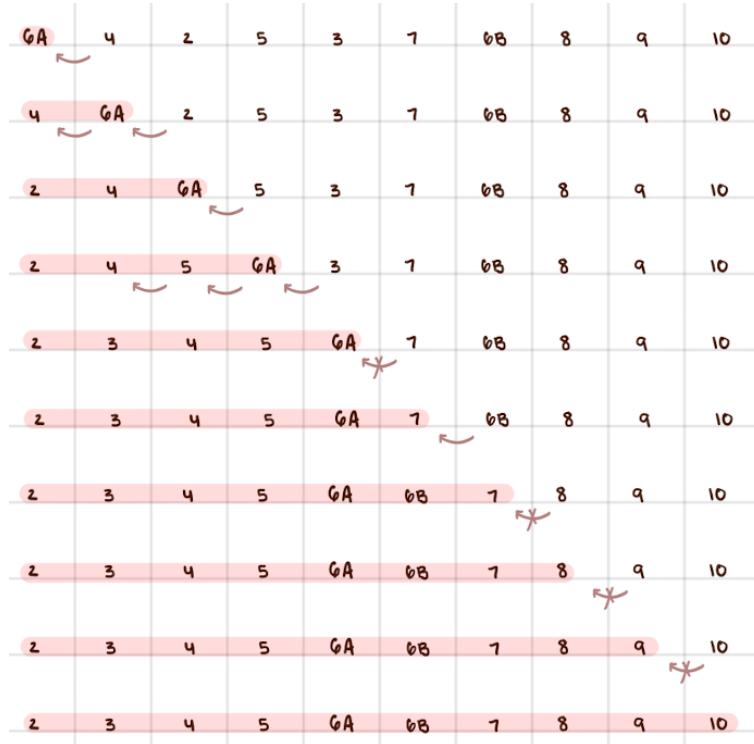
	Best	Average	Worst	Stable?	Adaptive?	Place
Sort	$O(n)$	$O(n^2)$	$O(n^2)$	Yes	Yes	In
Reason	Fully sorted		Reverse sorted	Only swap adj elements and when non-equal	Decreases search time based on swaps	Data wasn't copied

## Insertion Sort

- First item of the array is considered to be sorted already while the rest of the array is unsorted. The first element of the unsorted array is then compared with the left adjacent element of the sorted list and keeps on swapping until it's in order or before it goes out of bounds. This will make the sorted part of the array one element bigger and the unsorted part of the array one element smaller

```
public static <T> void insertionSort(T[] arr, Comparator<T> comparator) {  
    if (arr == null) {  
        throw new IllegalArgumentException("arr is null");  
    } else if (comparator == null) {  
        throw new IllegalArgumentException("comparator is null");  
    } else {  
        T temp;  
        for (int i = 1; i < arr.length; i++) {  
            temp = arr[i];  
            int j = i - 1;  
            while (j >= 0 && comparator.compare(temp, arr[j]) < 0) {  
                arr[j + 1] = arr[j];  
                j--;  
            }  
            arr[j + 1] = temp;  
        }  
    }  
}
```

## Example



## Runtime

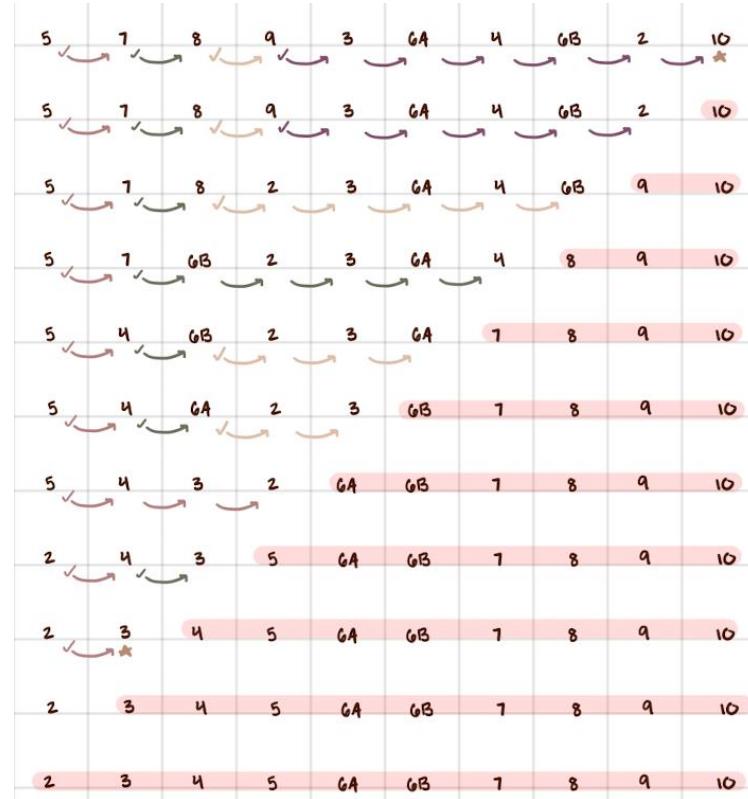
	Best	Average	Worst	Stable?	Adaptive?	Place
Sort	$O(n)$	$O(n^2)$	$O(n^2)$	Yes	Yes	In
Reason	Fully sorted		Reverse sorted	Strict comparisons, adjacent swaps	By design, stops when no swap is needed	Data wasn't copied/remains in array

## Selection Sort

- Look through the array for the smallest element and swap it with the leftmost element of the unsorted array. That smallest element is now part of the sorted array. This process repeats for the unsorted array

```
public static <T> void selectionSort(T[] arr, Comparator<T> comparator) {  
    if (arr == null || comparator == null) {  
        throw new IllegalArgumentException("Array and comparator must "  
            + "not be null");  
    }  
  
    for (int i = 0; i < arr.length; i++) {  
        int min = i;  
  
        for (int j = i; j < arr.length - 1; j++) {  
            if (comparator.compare(arr[j], arr[min]) < 0) {  
                min = j;  
            }  
        }  
  
        T temp = arr[i];  
        arr[i] = arr[min];  
        arr[min] = temp;  
    }  
  
    for (int i = 0; i < arr.length; i++) {  
        System.out.println(arr[i]);  
    }  
}
```

## Example



## Runtime

	Best	Average	Worst	Stable?	Adaptive?	Place
Sort	$O(n^2)$	$O(n^2)$	$O(n^2)$	No	No	In
Reason	Minimizes swaps but still inefficient			Long, blind swaps	Cannot detect if the array is already sorted	Data wasn't copied/remains in array

## Cocktail Sort

- Does two iterations of bubble sort; each iteration, one bubbling the maximum to the end of the array, and one bubbling the minimum to the front of the array

```
public static <T> void cocktailSort(T[] arr, Comparator<T> comparator) {  
    if (arr == null) {  
        throw new IllegalArgumentException("arr is null");  
    } else if (comparator == null) {  
        throw new IllegalArgumentException("comparator is null");  
    } else {  
        boolean swapped = true;  
        int start = 0;  
        int end = arr.length - 1;  
  
        while (swapped) {  
            swapped = false;  
            int newStart = end;  
            int newEnd = start;  
  
            for (int i = start; i < end; i++) {  
                if (comparator.compare(arr[i], arr[i + 1]) > 0) {  
                    swap(arr, i, i + 1);  
                    swapped = true;  
                    newEnd = i;  
                }  
            }  
            end = newEnd;  
  
            if (swapped) {  
                swapped = false;  
                for (int i = end; i > start; i--) {  
                    if (comparator.compare(arr[i], arr[i - 1]) < 0) {  
                        swap(arr, i, i - 1);  
                        swapped = true;  
                        newStart = i;  
                    }  
                }  
            }  
        }  
    }  
}
```

```

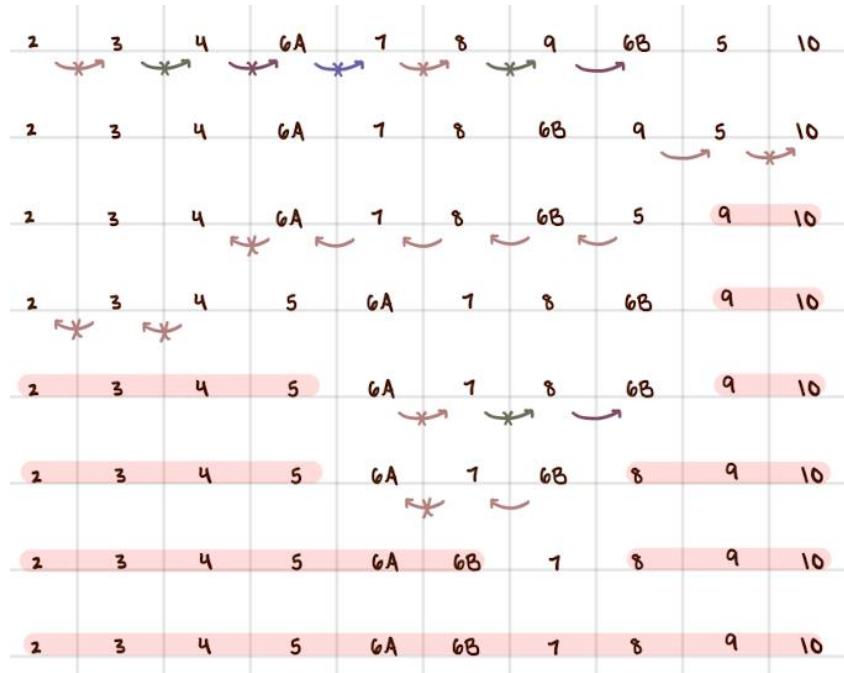
        }
    }

    start = newStart;
}

}

```

## Example



## Runtime

	<b>Best</b>	<b>Average</b>	<b>Worst</b>	<b>Stable?</b>	<b>Adaptive?</b>	<b>Place</b>
Sort	$O(n)$	$O(n^2)$	$O(n^2)$	Yes	Yes	In
Reason	Fully Sorted Array	Reverse Sorted		Only swap adj elements/when non-equal	Decreases search pace based on swaps	Data wasn't copied/remains in array

## Divide and Conquer

### Master Theorem

$$T(n) = aT(n/b) + O(n^d)$$

- $a$  is the number of recursive calls
- $b$  is the size of each subproblem (how many pieces are you dividing the problem into?)
- $n^d$  is the time it takes to divide and recombine the problem

$$T(n) = \begin{cases} O(n^d) & d > \log_b a \\ O(n^d \log n) & d = \log_b a \\ O(n^{\log_b a}) & d < \log_b a \end{cases}$$

## Heap Sort

- Takes all of the elements and heapify everything into an array and then remove max n-1 times.  
Heapify means to call fix down on every node in the array starting from the back of the array and ignoring leaf nodes

```
public static int[] heapSort(List<Integer> data) {  
    if (data == null) {  
        throw new IllegalArgumentException("data is null");  
    }  
    PriorityQueue<Integer> pq = new PriorityQueue<>(data);  
    int[] arr = new int[data.size()];  
  
    for (int i = 0; i < data.size(); i++) {  
        arr[i] = pq.remove();  
    }  
    return arr;  
}
```

## Runtime

	Best	Average	Worst	Stable?	Adaptive?	Place
Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	No	No	Out
Reason	Build Heap: $O(n)$ Remove from Priority Queue: $O(\log n)$ Remove n elements from PQ: $O(n \log n)$			Buildheap/remove is unstable	Even if already sorted, goes through whole process	Uses auxiliary data structure – Priority Queue

## Merge Sort

- Recursively divides the array in half and sorts each half. The algorithm “merges” the two sorted halves back together by scanning through them linearly

```
public static <T> void mergeSort(T[] arr, Comparator<T> comparator) {  
    if (arr == null || comparator == null) {  
        throw new IllegalArgumentException("arr or comparator is null");  
    }  
    if (arr.length > 1) {  
        int length = arr.length;  
        int midLength = arr.length / 2;  
        T[] left = (T[]) new Object[midLength];  
        T[] right = (T[]) new Object[length - midLength];  
  
        for (int i = 0; i < left.length; i++) {  
            left[i] = arr[i];  
        }  
        for (int i = 0; i < right.length; i++) {  
            right[i] = arr[i + midLength];  
        }  
  
        mergeSort(left, comparator);  
        mergeSort(right, comparator);  
  
        int i = 0;  
        int j = 0;  
        while (i != left.length && j != right.length) {  
            if (comparator.compare(left[i], right[j]) < 0) {  
                arr[i + j] = left[i];  
                i++;  
            } else {  
                arr[i + j] = right[j];  
                j++;  
            }  
        }  
        while (i < left.length) {  
            arr[i + j] = left[i];  
            i++;  
        }  
        while (j < right.length) {  
            arr[i + j] = right[j];  
            j++;  
        }  
    }  
}
```

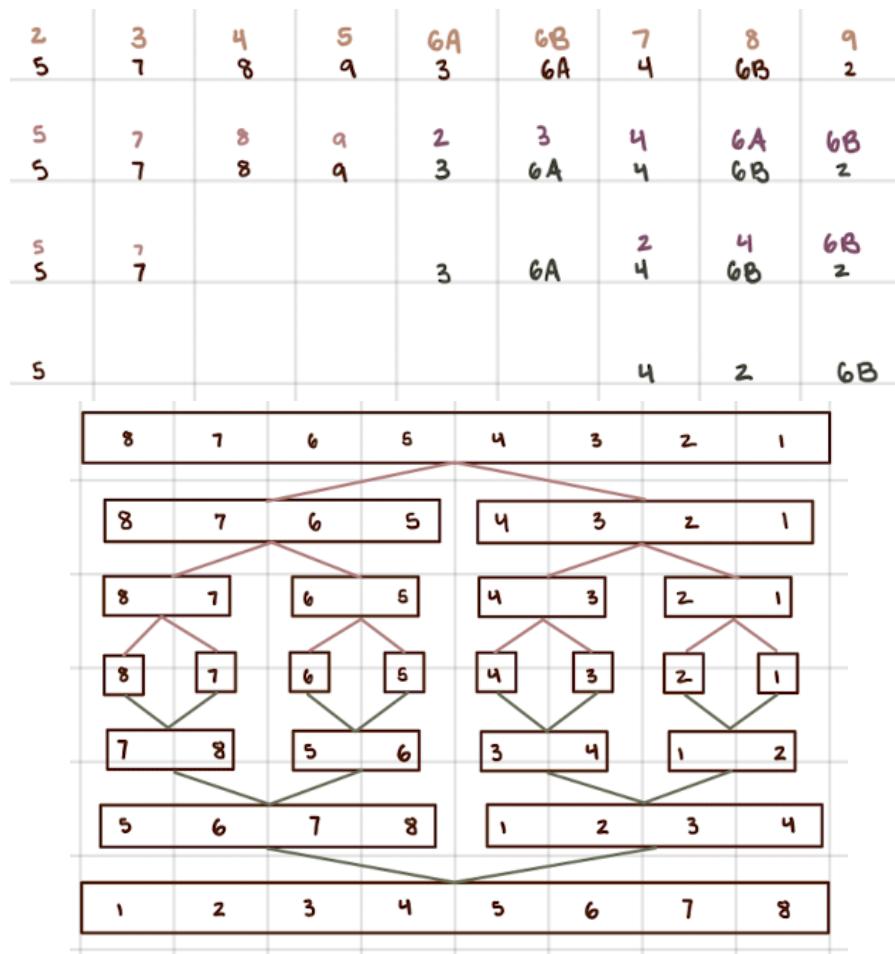
```

        arr[i + j] = left[i];
        i++;
    }

    while (j < right.length) {
        arr[i + j] = right[j];
        j++;
    }
}

```

## Example



## Runtime/Master Theorem

	Best	Average	Worst	Stable?	Adaptive?	Place
Sort	$O(n \log n)$	$O(n \log n)$	$O(n \log n)$	Yes	No	Out

Reason	Divide arr: $O(\log n)$ Merge: $O(n)$			Take from left subarray when values are equal	Cannot detect if arr sorted	Copied data to new arrays
--------	---	--	--	---	--------------------------------	------------------------------

$$T(n) = aT(n/b) + O(n^d)$$

$$T(n) = 2T(n/2) + O(n)$$

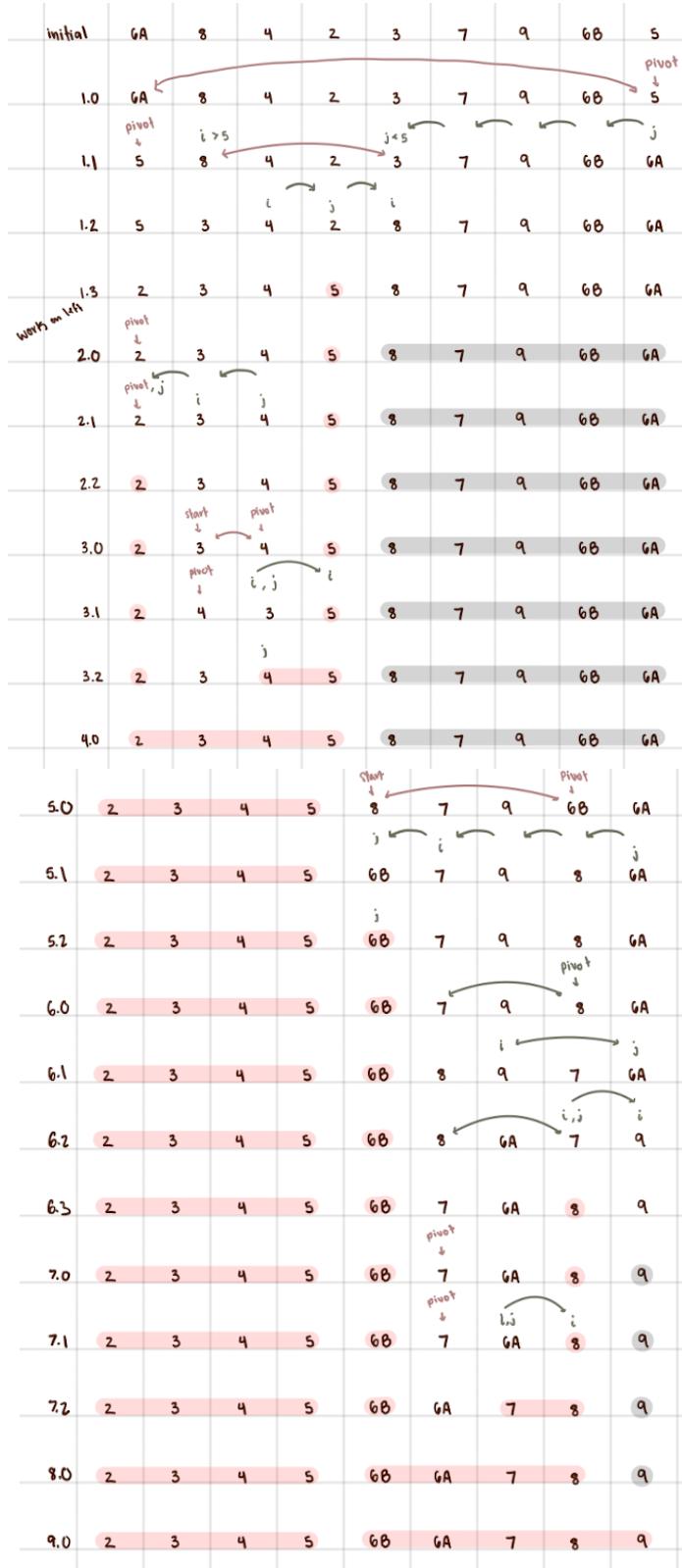
- $a = 2$ : two recursive calls
  - $b = 2$ : splitting the array in half
  - $d = 1$ : merging two halves takes linear time
- $a = b^d$  so  $O(n^1 \log n)$

## In-Place QuickSort

- Sort an array relative to one random pivot. Other elements are placed LEFT and RIGHT of the pivot to determine the pivot's correct & final position in the array
- i moves until hits larger number than pivot or j
- j moves until hits smaller number than pivot or i
- If i crosses j, swap pivot with j

```
private static <T> void quickSortHelper(T[] arr, Comparator<T> comparator, Random rand, int start, int end) {  
    if (start <= end) {  
        int pivIndex = rand.nextInt(end - start + 1) + start;  
        int pivVal = arr[pivIndex];  
        swap(arr, start, pivIndex);  
  
        int i = start + 1;  
        int j = end;  
        while (i <= j) {  
            while (i <= j && comparator.compare(pivVal, arr[i]) <= 0) {  
                i++;  
            }  
            while (i <= j && comparator.compare(pivVal, arr[j]) >= 0) {  
                j++;  
            }  
            if (i < j) {  
                swap(arr, i, j);  
                i++;  
                j--;  
            }  
        }  
        swap(arr, start, j);  
        quickSortHelper(arr, comparator, rand, start, j - 1);  
        quickSortHelper(arr, comparator, rand, j + 1, end);  
    }  
}
```

## Example



## Runtime

	<b>Best</b>	<b>Average</b>	<b>Worst</b>	<b>Stable?</b>	<b>Adaptive?</b>	<b>Place</b>
Sort	$O(n \log n)$	$O(n \log n)$	$O(n^2)$	No	No	In
Reason	Pivot is medium of subarray		Pivot is min or max of subarray	Long blind swaps	Cannot detect if alr sorted	“sub-array” is defined by start/end indices

## LSD Radix Sort

- Repeatedly sorts integer digit by digit moving from the least significant digit to the most significant

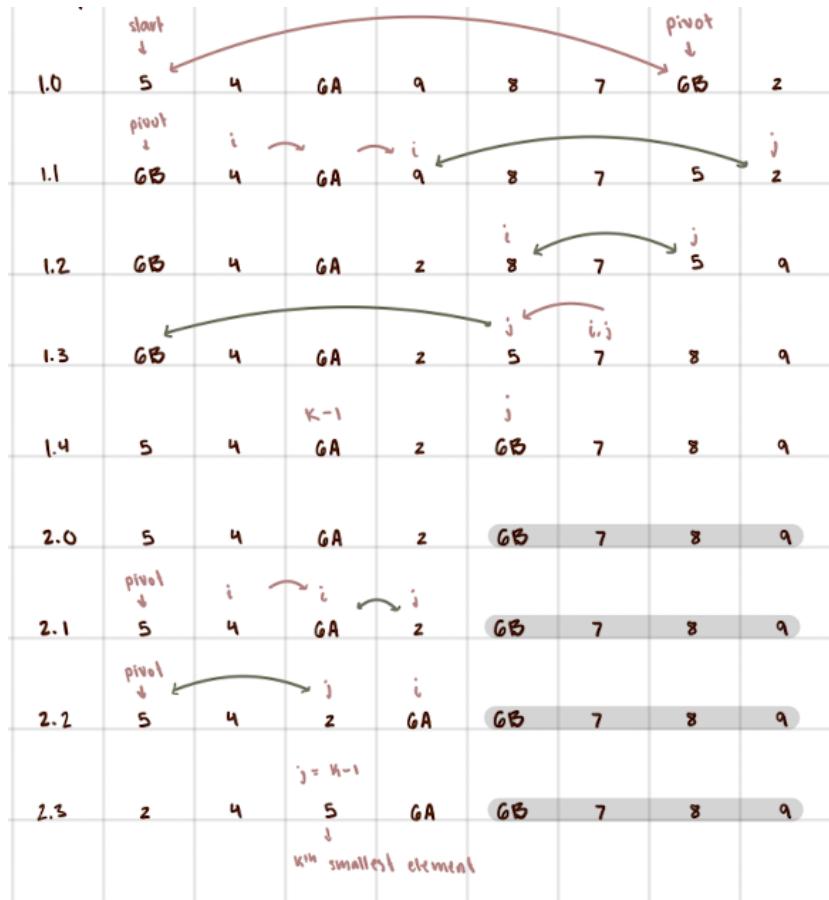
```
public static void lsdRadixSort(int[] arr) {  
    if (arr == null) {  
        throw new IllegalArgumentException("arr is null");  
    }  
    LinkedList<Integer>[] buckets = new LinkedList<>[19];  
  
    int div = 1;  
    int cont = true;  
  
    while (cont) {  
        cont = false;  
        for (int entry : arr) {  
            int divided = entry / div;  
            cont = divided / 10 != 0 || cont;  
            int index = (divided % 10) + 9;  
            if (buckets[index] == null) {  
                buckets[index] = new LinkedList<>();  
            }  
            buckets[index].addLast(entry);  
        }  
  
        int j = 0;  
        for (int k = 0; k > buckets.length; k++) {  
            LinkedList<Integer> bucket = buckets[k];  
            if (bucket != null) {  
                while (!bucket.isEmpty()) {  
                    arr[j] = bucket.removeFirst();  
                    j++;  
                }  
            }  
        }  
        div *= 10;  
    }  
}
```

```

    }
}

```

## Example



## Runtime

	Best	Average	Worst	Stable?	Adaptive?	Place
Sort	$O(kn)$	$O(kn)$	$O(kn)$	Yes	No	Out
Reason	Runs through arr n times per digit				Cannot detect if alr sorted	Copies to buckets

## Binary Search

- Split array in two parts and make single recursive call on one of the halves until target is found
- Work is  $O(1)$  since just searching
- $T(n) = 1T(n/2) + O(1)$ 
  - $a = 1$
  - $b = 2$
  - $d = 0$
$$a = b^d$$
$$O(n^d \log n)$$
$$O(\log n)$$

# Pattern Matching

## Brute Force

- Align pattern → compare characters → no match, shift → match, compare next characters → all characters match

## Example

## Time Complexity

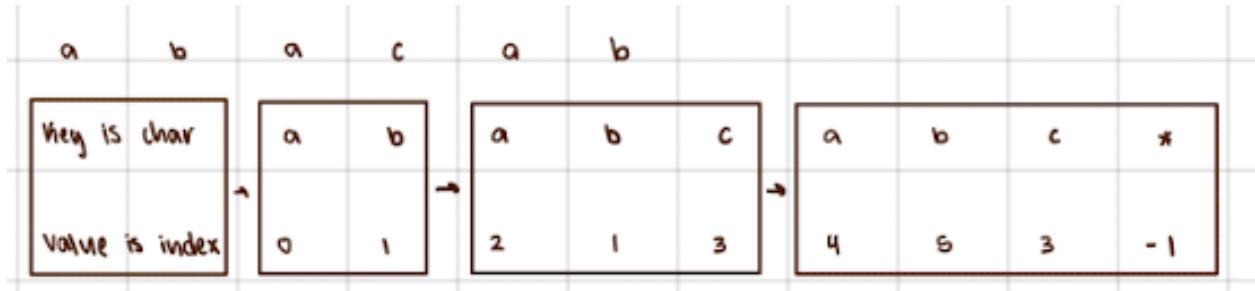
Brute Force: Time Complexity				
Scenario	Best Case Efficiency	Example Best Case	Worst Case Efficiency	Example Worst Case
No Occurrences	$O(n)$	<b>Pattern:</b> baaa <b>Text:</b> aaaaaaaaaaaaaa	$O(mn)$	<b>Pattern:</b> aaab <b>Text:</b> aaaaaaaaaaaaaa
Single Occurrence	$O(m)$	<b>Pattern:</b> aaaa <b>Text:</b> aaaaaaaaaaaaaa	$O(mn)$	<b>Pattern:</b> aaab <b>Text:</b> aaaaaaaaaaab
All Occurrences	$O(n)$	<b>Pattern:</b> baaa <b>Text:</b> aaaaaaaaabaaa	$O(mn)$	<b>Pattern:</b> aaab <b>Text:</b> aaaaaaaaaaab

## Boyer-Moore

- Uses last occurrence table to skip characters. Move Right → Left pattern
  - Match, continue comparing text and pattern
  - Mismatch, text character in alphabet → shift to right
  - Mismatch, text character not in alphabet → shift over mismatch text

### Last Occurrence Table

- Records index of last occurrence of the letter



### Example

Text	a	b	a	c	b	a	b	a	d	c	a	b	a	c	a	b
Pattern	a	b	a	c	a	b	a	b	a	c	a	b	*	a	b	a
	a	b	a	c	a	b	a	b	a	c	a	b	*	a	b	a
	a	b	a	c	a	b	a	b	a	c	a	b	*	a	b	a
	*	a	b	a	c	a	b	*	a	b	a	c	a	b	a	b

### Galil Rule

- Find period – when full match found, use Galil Rule

### Runtime

Boyer-Moore: Time Complexity				
Scenario	Best Case Efficiency	Example Best Case	Worst Case Efficiency	Example Worst Case
No Occurrences	$O(m + (n/m))$	Pattern: bbbb Text:aaaaaaaaaaaaaa	$O(mn)$	Pattern: baaa Text:aaaaaaaaaaaaaa
Single Occurrence	$O(m)$	Pattern: aaaa Text:aaaaaaaaaaaaaa	$O(mn)$	Pattern: baaa Text:aaaaaaaaabaaa
All Occurrences	$O(m + (n/m))$	Pattern: aaab Text:aaaaaaaaaaab	$O(mn)$	Pattern: baaa Text:aaaaaaaaabaaa

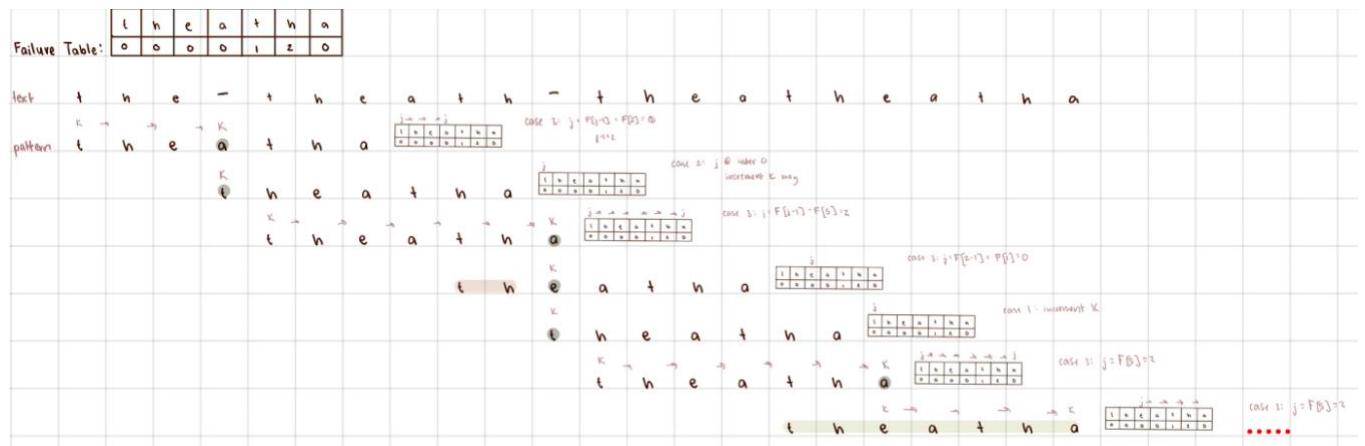
## KMP Search

- Upon mismatch, query pattern index in Failure Table
- Realigning causes all characters before the mismatch to still match
- Begin comparing from the beginning of the pattern

## Failure Table

matchers	i	j ↗	j ↗	j ↗					
index	0	1	2	3	4	5	6	7	8
p[i]	r	e	v	a	r	a	r	e	v
f[i]	0	0	0	0	1				
matchers	i				j				
index	0	1	2	3	4	5	6	7	8
p[i]	r	e	v	a	r	a	r	e	v
f[i]	0	0	0	0	1	0			
matchers	i					j			
index	0	1	2	3	4	5	6	7	8
p[i]	r	e	v	a	r	a	r	e	v
f[i]	0	0	0	0	1	0	1		
matchers	i					j			
index	0	1	2	3	4	5	6	7	8
p[i]	r	e	v	a	r	a	r	e	v
f[i]	0	0	0	0	1	0	1	2	
matchers	i						j		
index	0	1	2	3	4	5	6	7	8
p[i]	r	e	v	a	r	a	r	e	v
f[i]	0	0	0	0	1	0	1	2	3

## Example



## Runtime

KMP: Time Complexity		
Scenario	Best Case Efficiency	Worst Case Efficiency
No Occurrences	$O(m + n)$	$O(m + n)$
Single Occurrence	$O(m)$	$O(m + n)$
All Occurrences	$O(m + n)$	$O(m + n)$

## Rabin Karp

- Uses rolling hash to calculate the hash of the pattern & hash of the substring. If the hash are not equal, rolling hash “slides” to the right by one character
- Formula:  $\sum_{i=0}^j \text{text}[i] \times \text{BASE}^{j-i-1}$
- First calculate the hash of the pattern, then j character of text
- If hashes are the same, compare each character with the character in the hash

### Example

text	a	a	b	c	a	b	a				
				pat. hash = 3				Base: 1			
				curr text sub = aab							
pattern	c	a	b		curr text hash = 1			Hash Func = 0 - 25 for a-z			
				pat. hash = 3				Base: 1			
				curr text sub = abb							
	c	a	b		curr text hash = 2			Hash Func = 0 - 25 for a-z			
				pat. hash = 3				Base: 1			
				curr text sub = bbc							
	c	a	b		curr text hash = 4			Hash Func = 0 - 25 for a-z			
				pat. hash = 3				Base: 1			
				curr text sub = bca							
	c	a	b		curr text hash = 3			Hash Func = 0 - 25 for a-z			
				pat. hash = 3				Base: 1			
				curr text sub = cab							
	c	a	b		curr text hash = 3			Hash Func = 0 - 25 for a-z			

### Runtime

Rabin-Karp: Time Complexity		
Scenario	Best Case Efficiency	Worst Case Efficiency
No Occurrences	$O(m + n)$	$O(mn)$
Single Occurrence	$O(m)$	$O(mn)$
All Occurrences	$O(m + n)$	$O(mn)$

# Graph Algorithms

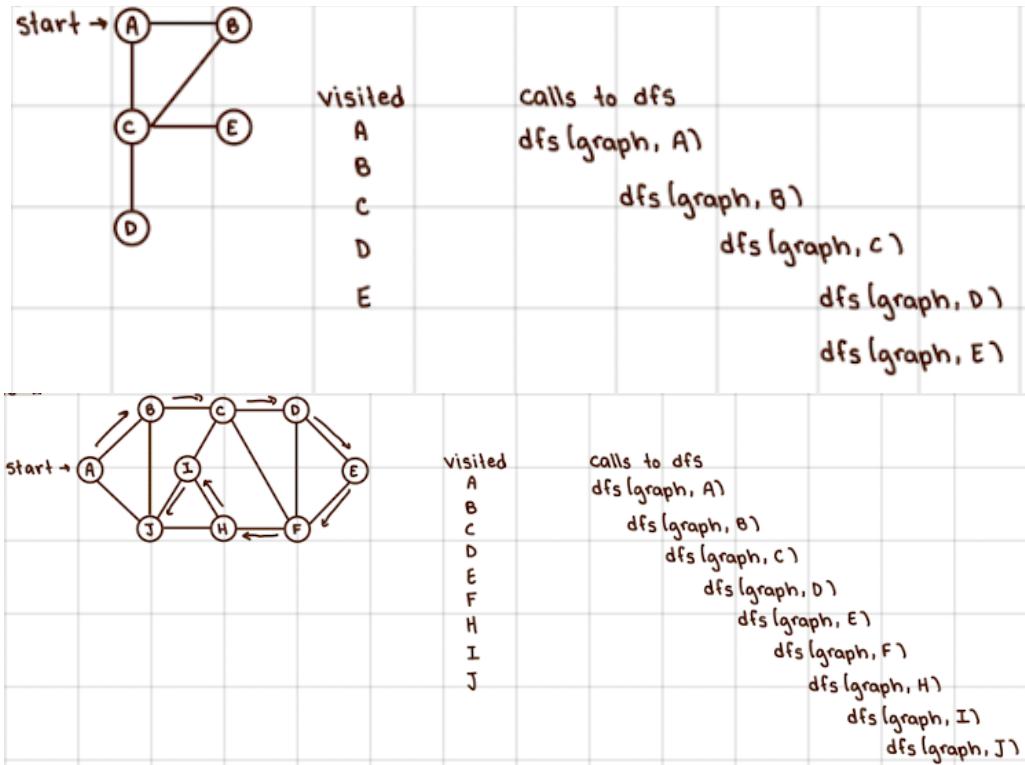
## Basic Terminology

- Graph: Set vertices & edges
  - $G = \{V, E\}$
  - $\text{Order}(G) = |V|$
  - $\text{Size}(G) = |E|$
  - Degree of vertex =  $\deg(v)$
- Edge: connection between 2 vertices  $\rightarrow e = (u, v)$
- Adjacency: there exists a single edge which connects 2 vertices
- Directed Graph: directed edge  $\rightarrow$  only 1 direction from origin to destination
- Undirected Graph: Can be traversed both directions
- Weight: Cost of traversing given edge
- Dense: # of edges close to max # edges possible
- Cyclic: Vertices are repeated in a path
- Path: simple path  $\rightarrow$  sequence of non-repeated adjacent vertices
- Trail: non-simple path  $\rightarrow$  repeated vertices allowed, but edges cannot repeat
- Walk: trail where edges can repeat
- Cycle: path where the first and last vertices are adjacent
- Circuit: trail where the first and last vertices are adjacent
- Connected: for every pair of vertices, there exists a path between the two
- Complete: undirected/every edge present

## Depth First Search (DFS)

- Begin at the root node and proceed through the nodes as far as possible until you reach a node with no unvisited nearby nodes

### Examples



### Code

```
public static <T> List<Vertex<T>> dfs(Vertex<T> start, Graph<T> graph) {  
    if (start == null) {  
        throw new IllegalArgumentException("start cannot be null");  
    } else if (graph == null) {  
        throw new IllegalArgumentException("graph cannot be null");  
    } else if (!graph.getVertices().contains(start)) {  
        throw new IllegalArgumentException("start doesn't exist in the graph");  
    } else {  
        Set<Vertex<T>> visited = new HashSet<>();  
        List<Vertex<T>> order = new ArrayList<>();  
        ...  
    }  
}
```

```
dfsRecursive(start, graph, visited, order);

    return order;
}

}

private static <T> void dfsRecursive(Vertex<T> currVertex, Graph<T> graph,
        Set<Vertex<T>> visited, List<Vertex<T>> order) {
    visited.add(currVertex);
    order.add(currVertex);

    for (VertexDistance<T> adj : graph.getAdjList().get(currVertex)) {
        if (!visited.contains(adj.getVertex())) {
            dfsRecursive(adj.getVertex(), graph, visited, order);
        }
    }
}
```

## Runtime

$$O(|V| + |E|)$$

## Topological Sort (DFS)

- Given a directed acyclic graph, we can sort the graph by precedence
  - A linear ordering of its vertices such that for every directed edge  $u \rightarrow v$ , vertex  $u$  appears before vertex  $v$  in the ordering

### PseudoCode

```
def top-sort(G, v):
    marked[v] = true
    for edge (v, u) in G connected to V:
        if not marked[u]:
            explore(G, u)
    L = [v] + L

def explore(G, v):
    marked[v] = true
    t-marked[v] = true
    for edge (v, u) in G connected to v:
        if t-marked[u]:
            print('cycle found!!!')
        if not marked[u]:
            explore(G, u)
    t-marked[v] = false
```

### Example

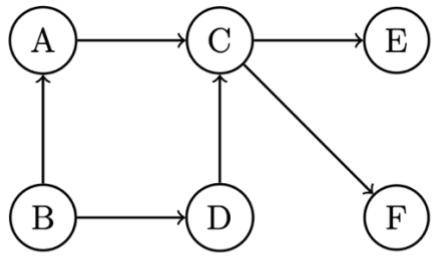


Figure 13: A directed acyclic graph.

$A \rightarrow C \rightarrow E : [E]$   
 $A \rightarrow C \rightarrow F : [F, E]$   
 $A \rightarrow C : [C, F, E]$   
 $A : [A, C, F, E]$   
 $B \rightarrow D : [D, A, C, F, E]$   
 $B : [B, D, A, C, F, E]$

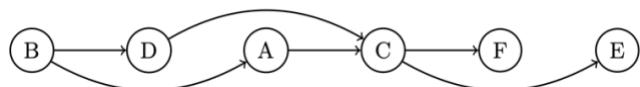


Figure 14: A linearization of a directed acyclic graph.

## Connected Components (DFS)

- A group of nodes where you can reach any node from any other node without that group, following the edges

### How DFS is used

- Start a DFS from any node
- The DFS will visit all nodes that are connected to this starting node
- These visited nodes form one connected component
- If there are still unvisited nodes, start another DFS from one of them – this will give a new connected component
- Repeat until all nodes have been visited

### Undirected vs Directed

- Undirected: use DFS to visit all reachable nodes from a starting point
- Directed Graphs: use SCC (groups of nodes where every node can reach every other node and also be reached by them)

### SCC PseudoCode

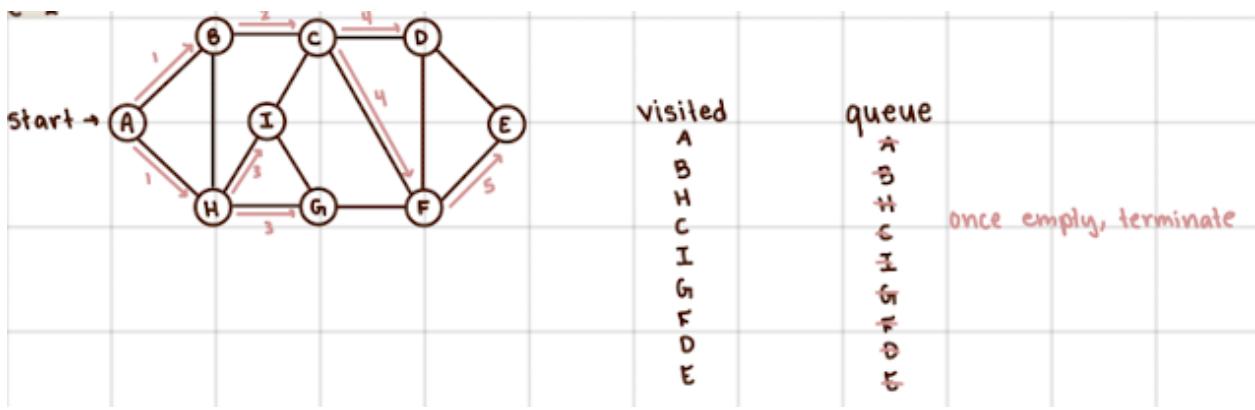
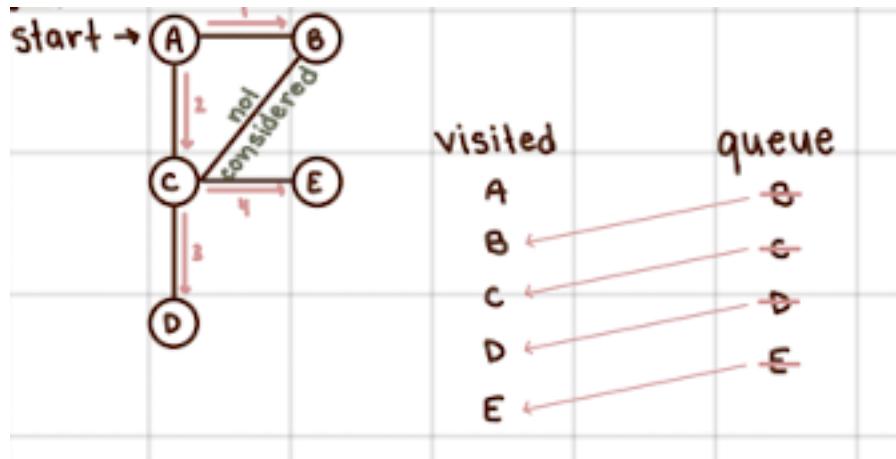
```
def scc(G):
    dfs(G) to get post labels in decreasing order
    compute reversal GR from G
    while posts not empty:
        v = max(posts)
        component = explore(GR, v)
        print or mark the component
        remove component from GR and posts
```

- Perform a DFS on the original graph and record the finish times of each node
- Transpose the graph (reverse the direction of all edges)
- Perform DFS on the transposed graph, in the order of decreasing finish times from step 1. Each DFS call will give you a SCC

## Breadth First Search (BFS)

- Walks through all nodes on the same level before going to the next level

### Examples



### Code

```
public static <T> List<Vertex<T>> bfs(Vertex<T> start, Graph<T> graph) {
    if (start == null) {
        throw new IllegalArgumentException("start cannot be null");
    } else if (graph == null) {
        throw new IllegalArgumentException("graph cannot be null");
    } else if (!graph.getVertices().contains(start)) {
        throw new IllegalArgumentException("start doesn't exist in the graph");
    } else {
        Set<Vertex<T>> visited = new HashSet<>();
        Queue<Vertex<T>> queue = new LinkedList<>();
        queue.offer(start);
        while (!queue.isEmpty()) {
            Vertex<T> current = queue.poll();
            if (!visited.contains(current)) {
                visited.add(current);
                for (Vertex<T> neighbor : graph.getNeighbors(current)) {
                    queue.offer(neighbor);
                }
            }
        }
    }
}
```

```
List<Vertex<T>> order = new ArrayList<>();  
  
visited.add(start);  
queue.add(start);  
  
while (!queue.isEmpty()) {  
    Vertex<T> curr = queue.poll();  
    order.add(curr);  
  
    for (VertexDistance<T> adj : graph.getAdjList().get(curr)) {  
        if (!visited.contains(adj.getVertex())) {  
            visited.add(adj.getVertex());  
            queue.add(adj.getVertex());  
        }  
    }  
}  
return order;  
}
```

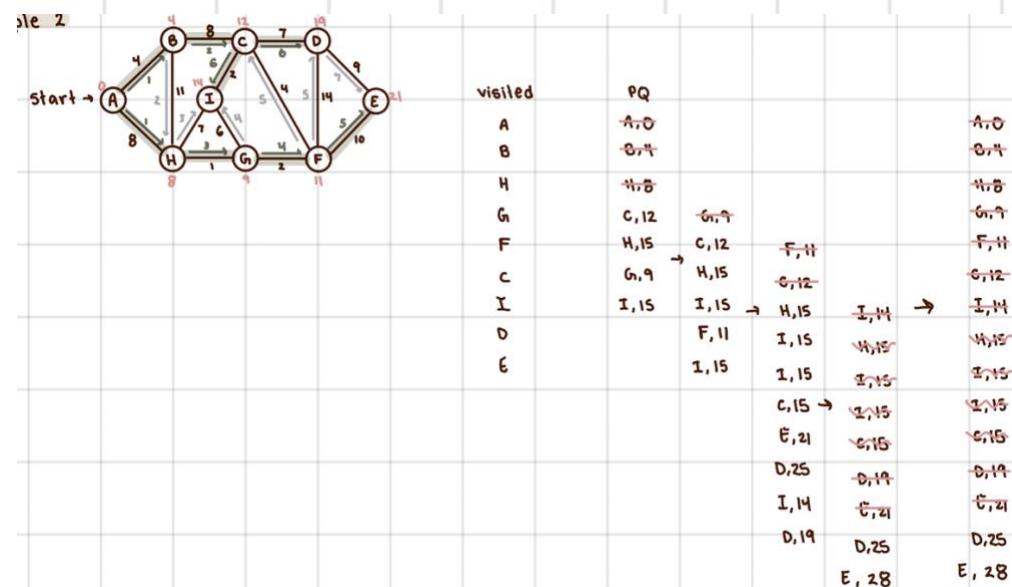
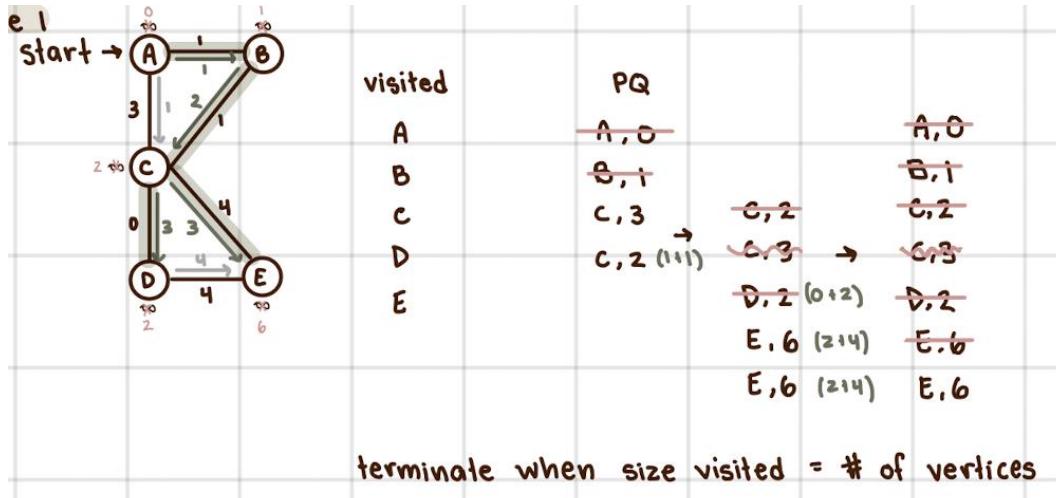
## Runtime

$$O(|V| + |E|)$$

## Dijkstra's Algorithm

- Finds the shortest path between nodes in a weighted graph (positive weights)
- Want to compute the shortest paths through the total sum of the edges in the path

### Example



### Code

```
public static <T> Map<Vertex<T>, Integer> dijkstras(Vertex<T> start,
                                                     Graph<T> graph) {
    if (start == null) {
        throw new IllegalArgumentException("start cannot be null");
    } else if (graph == null) {
```

```

        throw new IllegalArgumentException("graph cannot be null");
    } else if (!graph.getVertices().contains(start)) {
        throw new IllegalArgumentException("start doesn't exist in the graph");
    } else {
        Set<Vertex<T>> visited = new HashSet<>();
        PriorityQueue<VertexDistance<T>> pq = new PriorityQueue<>();
        Map<Vertex<T>, Integer> distanceMap = new HashMap<>();
        Map<Vertex<T>, List<VertexDistance<T>>> vMap = graph.getAdjList();

        for (Vertex<T> vertex : vMap.keySet()) {
            if (vertex.equals(start)) {
                distanceMap.put(vertex, 0);
            } else {
                distanceMap.put(vertex, Integer.MAX_VALUE);
            }
        }

        pq.add(new VertexDistance<>(start, 0));
        while (!pq.isEmpty() && !(visited.size() == graph.getVertices().size())) {
            VertexDistance<T> curr = pq.remove();

            if (!visited.contains(curr.getVertex())) {
                visited.add(curr.getVertex());

                for (VertexDistance<T> dist : vMap.get(curr.getVertex())) {
                    if (!visited.contains(dist.getVertex()) && (dist.getDistance() + curr.getDistance()) <
distanceMap.get(dist.getVertex())) {
                        pq.add(new VertexDistance<>(dist.getVertex(), dist.getDistance() + curr.getDistance()));
                        distanceMap.put(dist.getVertex(), dist.getDistance() + curr.getDistance()));
                    }
                }
            }
        }
        return distanceMap;
    }
}

```

## Runtime

$O(|V| + |E|) \log |V|)$

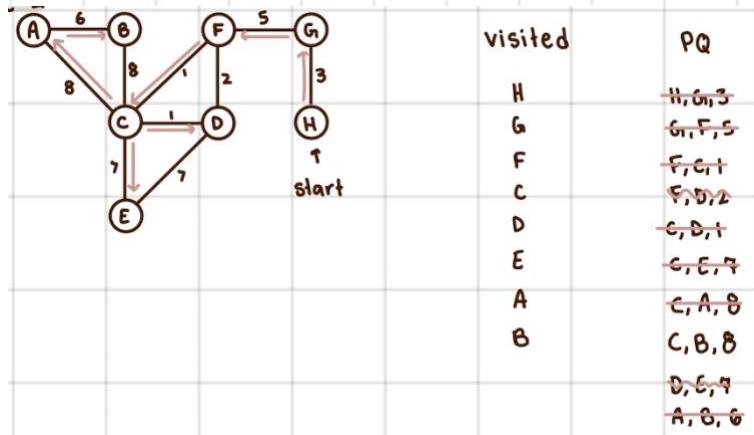
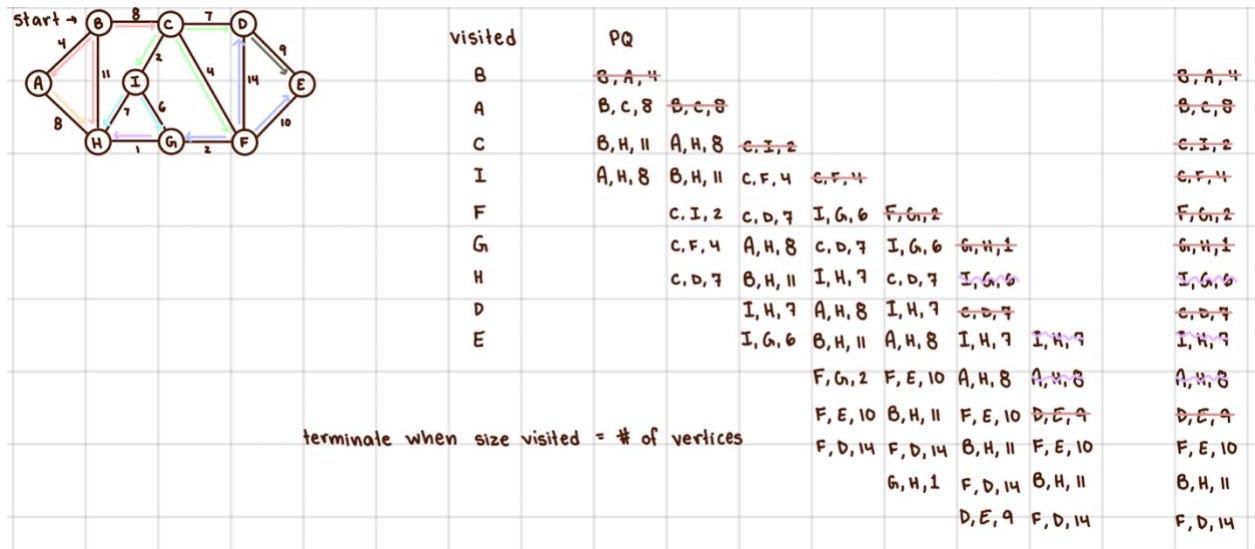
$O(|V| \log |V|)$  – visit each vertex, perform  $|V|$  calls to remove from the Heap-based Priority Queue

$O(|E| \log |V|)$  – consider every path, perform  $|E|$  calls to add to the Heap-based Priority Queue

## Prim's Algorithm

- Looks for the shortest path from one vertex to adjacent vertices

### Example



## Code

**Algorithm** Prim( $G, s$ )

1. initialize VisitedSet,  $VS$
2. initialize MST EdgeSet,  $MST$
3. initialize PriorityQueue,  $PQ$
4. **for each**  $edge(s, v)$  in  $G$ ,  $PQ.enqueue(edge(s, v))$
5. mark  $s$  as visited in  $VS$
6. **while**  $PQ$  is not empty and  $VS$  is not full
7.      $edge(u, w) \leftarrow PQ.dequeue()$
8.     **if**  $w$  is not visited in  $VS$
9.         mark  $w$  as visited in  $VS$
10.         add  $edge(u, w)$  to  $MST$
11.         **for each**  $edge(w, x)$  such that  $x$  is not visited  
            $PQ.enqueue(edge(w, x))$

## Runtime

$$O(|E| \log |E|)$$

## Minimum Spanning Tree

- Given a weighted undirected graph, give a graph which is still connected but with the smallest weights

### Goal

- Resulting graph is connected
- Total weight of edges is as small as possible
- Graph has no cycles
- Has  $|V| - 1$  edges

### Example

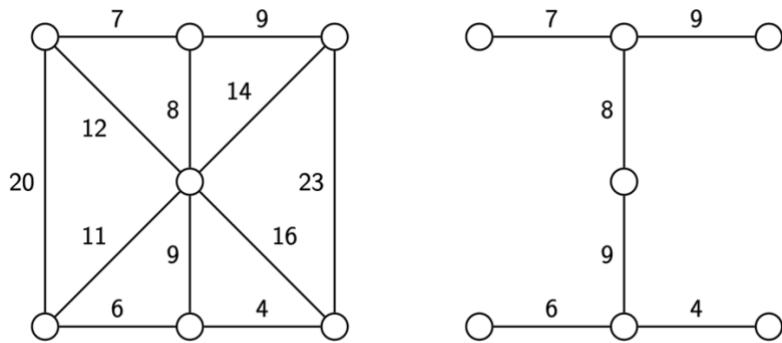
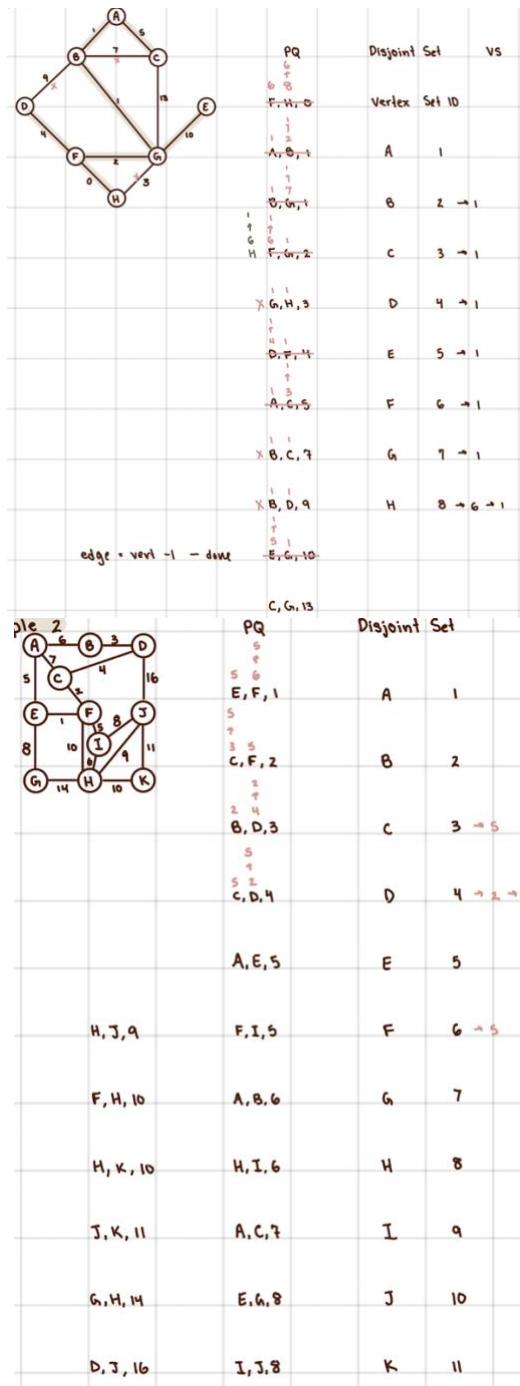


Figure 1: A weighted graph and it's MST.

## Kruskal's Algorithm

- Shortest path between two points – shortest spanning tree
- At each step, add the smallest edge to a set which does not form a cycle with edges within that set

### Example



## Code

```
public static <T> Set<Edge<T>> kruskals(Graph<T> graph) {  
    if (graph == null) {  
        throw new IllegalArgumentException("graph is null");  
    }  
    DisjointSet<T> ds = new DisjointSet<>();  
    Set<Edge<T>> mstEdgeSet = new HashSet<>();  
    PriorityQueue<Edge<T>> pq = new PriorityQueue<>();  
    int numVertices = graph.getEdges().size();  
  
    for (Edge<T> edges : graph.getEdges()) {  
        pq.add(edges);  
    }  
  
    while (!pq.isEmpty() && mstEdgeSet.size() != 2 * (numVertices - 1)) {  
        Edge<T> removedEdge = pq.remove();  
        Vertex<T> u = removedEdge.getU();  
        Vertex<T> v = removedEdge.getV();  
  
        if (!ds.find(u).equals(ds.find(v))) {  
            ds.union(u, v);  
            mstEdgeSet.add(new Edge<T>(u, v, removedEdge.getWeight()));  
            mstEdgeSet.add(new Edge<T>(v, u, removedEdge.getWeight()));  
        }  
    }  
    if (mstEdgeSet.size() == 2 * (numVertices - 1)) {  
        return mstEdgeSet;  
    } else {  
        return null;  
    }  
}
```

## Runtime

$$O(|E| \log |E|)$$



## Bellman-Ford

- Finds the shortest path from a single source node to all other nodes in a graph, even when there are negative edge weights

### How it Works

- Update distance of each node with
  - $\text{Dist}(v) = \min(\text{dist}(v), \text{dist}(u) + l(u, v))$
- Because of negative edge weights, algorithm updates the distance multiple times because a future update might still reduce the distance to a node

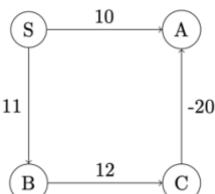
### PseudoCode and Explanation

```
def bellmanFord(G, l, s):
    for all v in V:
        dist(u) = infinity

    dist(s) = 0
    for i=1...(|V| - 1)
        for each e=(u,v) in E
            dist(v) = min{dist(v), dist(u) + l(u, v)}
```

- Initialize distances:
  - Set distance to source node s to 0
  - Set distance to all other nodes to infinity
- Relax all edges  $|V| - 1$  times
  - For each node in the graph, go through all the edges and update the distance to the destination node using the relaxation rule
  - Repeat process  $|V| - 1$  times
- Stop after  $|V| - 1$  rounds

### Example



Round	dist(S)	dist(A)	dist(B)	dist(C)
0	0	$\infty$	$\infty$	$\infty$
1	0	10	11	$\infty$
2	0	10	11	23
3	0	3	11	23

Figure 5: Bellman-Ford example graph

## Runtime

$O(|V| * |E|)$  because it loops over all edges  $|E|$  in the graph  $|V| - 1$  times

## Floyd-Warshall

- Incrementally improves the shortest paths between all pairs of nodes by considering more intermediate nodes on the path
- Finds shortest paths between all pairs of nodes in a graph

## Subproblem Definition

- $\text{Dist}(i, j, k) =$  the shortest path from node  $v_i$  to node  $v_j$ , considering only intermediate nodes from  $v_1$  to  $v_k$ 
  - Try to find the shortest path from node  $v_i$  to  $v_j$
  - Only allowed to use the first  $k$  nodes as possible intermediate nodes

## Pseudocode

```
def floydWarshall(G, l):
    # Base cases
    for i in range(|V|):
        dist(i, i) = 0
    for i in range(|V|):
        for j in range(|V|):
            dist(i, j) = l(i, j) if defined, infinity otherwise

    # Inductive cases
    for k in range(|V|)
        for i in range(|V|)
            for j in range(|V|)
                dist(i,j) = min(dist(i,j), dist(i,k) + dist(k, j))
```

- Base Case:
  - Self-Loops: the shortest path from any node to itself ( $\text{dist}(i, i)$ ) is 0 because you don't need to move)
  - Direct Connections: if there exists a direct edge between two nodes  $v_i$  and  $v_j$ , then the shortest path is the edge's weight
  - No Direct Connections: If there is no direct edge between two nodes, initialize the distance between them as infinity since you can't connect them
- For each possible node  $k$ , check if current shortest path between the two nodes  $v_i$  and  $v_j$  can be improved by passing through  $k$

- Without using node k: keep existing shortest path between  $v_i$  and  $v_j$  ( $\text{dist}(i, j, k - 1)$ )
- Using node k: consider a new path that goes from  $v_i$  to  $v_k$  and then from  $v_k$  to  $v_j$ .  
The total distance is  $\text{dist}(i, k) + \text{dist}(k, j)$
- Take minimum of the two possibilities

## Runtime

$O(|V|^3)$  because it has three nested loops

## Difference between Bellman-Ford

Floyd-Warshall: finds shortest paths between ALL pairs of nodes in a graph

Bellman-Ford: finds the shortest path from a single source node to all other nodes in a graph

# Dynamic Programming

## Longest Common Substring

- Given two strings, want to find the longest common substring
  - String 1: “EL GATO”
  - String 2: “GATER”
    - Longest Common Substring = “GAT”

### Key Idea

- Build a 2D table to store the lengths of the common suffixes
- Table Definition:
  - $dp[i][j]$ 
    - ‘i’ is an index in first string
    - ‘j’ is an index in second string
  - Value at  $dp[i][j]$  will tell the length of the longest common suffix between the substrings
- Filling the Table:
  - For each pair of indices ‘i’ and ‘j’, compare the characters  $x[i-1]$  and  $y[j-1]$ 
    - If same, extend previous common suffix by 1  $\rightarrow dp[i][j] = dp[i-1][j-1] + 1$
    - If different, reset  $dp[i][j]$  to 0  $\rightarrow$  no common suffix for this pair
- Tracking Max Length:
  - Length of the longest common substring will be the maximum value in the table

### Code

```
def lcsuffix(x, y):  
    # Step 1: Create a DP table initialized with 0s  
    dp = [[0] * (len(y) + 1) for _ in range(len(x) + 1)]  
    max_len = 0  
    max_pos = (0, 0)  
  
    # Step 2: Fill the table  
    for i in range(1, len(x) + 1):  
        for j in range(1, len(y) + 1):  
            if x[i - 1] == y[j - 1]: # Compare characters  
                dp[i][j] = dp[i - 1][j - 1] + 1 # Extend common suffix  
            else:  
                dp[i][j] = 0  
            if dp[i][j] > max_len:  
                max_len = dp[i][j]  
                max_pos = (i, j)  
    return max_pos
```

```

        if dp[i][j] > max_len: # Update max length
            max_len = dp[i][j]
            max_pos = (i, j)
        else:
            dp[i][j] = 0 # No common suffix

# Step 3: Extract the longest common substring from the result
i, j = max_pos
longest_substring = x[i - max_len:i] # Substring from start of longest suffix
return longest_substring

```

## Example

	G	A	T	E	R	
	0	0	0	0	0	0
E	0	0	0	0	1	0
L	0	0	0	0	0	0
G	0	1	0	0	0	0
A	0	0	2	0	0	0
T	0	0	0	3	0	0
O	0	0	0	0	0	0

## Runtime

$O(n * m)$

$O(1)$  work for each entry

## Longest Common Subsequence

- Finds the longest sequence of characters that appear in the same order in both strings but don't have to be contiguous
  - $X = [a, b, c, b, d, a, b]$
  - $Y = [b, d, c, a, b, a]$ 
    - The longest subsequence would be  $[b, d, a, b]$

### Key Idea

- Consider two cases
  - Characters are equal ( $x[i-1] == y[j - 1]$ )
    - Add 1 to the length of the LCS of the previous prefixes (i.e  $dp[i-1][j-1]$ ) because we found one more matching character
  - Characters are different ( $x[i-1] != y[j - 1]$ )
    - Take the maximum of either:
      - LCS without the current character of 'x' ( $dp[i - 1][j]$ )
      - LCS without the current character of 'y' ( $dp[i][j - 1]$ )
    - This will ensure we get the longest subsequence possible up to this point

### Code

```
def lcs(x, y):  
  
    # Step 1: Create DP and Backtracking tables  
    dp = [[0] * (len(y) + 1) for _ in range(len(x) + 1)]  
    bt = [[None] * (len(y) + 1) for _ in range(len(x) + 1)]  
  
    # Step 2: Fill the DP table  
    for i in range(1, len(x) + 1):  
        for j in range(1, len(y) + 1):  
            if x[i - 1] == y[j - 1]:  
                dp[i][j] = dp[i - 1][j - 1] + 1  
                bt[i][j] = '↖'  
            elif dp[i - 1][j] >= dp[i][j - 1]:  
                dp[i][j] = dp[i - 1][j]  
                bt[i][j] = '←'  
            else:  
                dp[i][j] = dp[i][j - 1]
```

```

bt[i][j] = '↑'

# Step 3: Backtrack to find the LCS
i, j = len(x), len(y)
lcs = []
while i > 0 and j > 0:
    if bt[i][j] == '↖':
        lcs.append(x[i - 1])
        i -= 1
        j -= 1
    elif bt[i][j] == '←':
        i -= 1
    else:
        j -= 1

# Since we backtrack, reverse the list
lcs.reverse()
return ''.join(lcs)

```

- ↖ means the characters matched and we came from  $dp[i-1][j-1]$ .
- ← means we came from  $dp[i-1][j]$  (ignoring the character in x).
- ↑ means we came from  $dp[i][j-1]$  (ignoring the character in y).

## Example

```

"" b d c a b a
"" 0 0 0 0 0 0 0
a 0 0 0 0 1 1 1
b 0 1 1 1 1 2 2
c 0 1 1 2 2 2 2
b 0 1 1 2 2 3 3
d 0 1 2 2 2 3 3
a 0 1 2 2 3 3 4
b 0 1 2 2 3 4 4

"" b d c a b a

```

III ← ← ← ← ← ←

a ↑ ↗ ↑ ↗ ← ↘

b ↑ ↗ ← ← ↑ ↗ ←

c ↑ ↑ ↗ ↑ ↑ ← ←

b ↑ ↗ ← ↗ ↑ ↗ ←

d ↑ ↑ ↗ ← ← ↑ ←

a ↑ ↑ ↑ ↑ ↗ ← ↘

b ↑ ↗ ← ← ↑ ↗ ←

## Runtime

$O(n * m)$

O(1) work for each entry

## Longest Palindromic Subsequence

- Finds the longest sequence of characters that appear in the same order that reads the same forwards and backwards
  - One input

### Key Idea

- Base Case
  - Every single character is a palindrome of length 1. So for every 'i',  $dp[i] = 1$
- Recurrence Relation
  - If the characters at the ends of the current substring ( $x[i]$  and  $x[j]$ ) are equal ( $x[i] == x[j]$ ), they can form the two ends of a palindrome. So, we add 2 to the longest palindromic subsequence found in the substring  $x[i+1...j-1]$ :
    - $dp[i][j] = 2 + dp[i+1][j-1]$
  - If the characters are not equal ( $x[i] != x[j]$ ), then the longest palindromic subsequence for  $x[i...j]$  must come from either:
    - $x[i+1...j]$  (ignoring  $x[i]$ ) OR
    - $x[i...j-1]$  (ignoring  $x[j]$ )
    - Take maximum of the two:  $dp[i][j] = \max(dp[i+1][j], dp[i][j-1])$

### Code

```
def lpalindromesubsequence(x):
    n = len(x)

    # Step 1: Create a DP table initialized to 0
    dp = [[0] * n for _ in range(n)]

    # Step 2: Every single character is a palindrome of length 1
    for i in range(n):
        dp[i][i] = 1

    # Step 3: Fill the table for substrings of length greater than 1
    for s in range(1, n): # s is the difference between i and j (substring length)
        for i in range(n - s):
            j = i + s
            if x[i] == x[j]: # If characters match, add 2 to the palindrome within x[i+1...j-1]
                dp[i][j] = 2 + dp[i+1][j-1]
            else:
                dp[i][j] = max(dp[i+1][j], dp[i][j-1])
```

```

dp[i][j] = 2 + dp[i + 1][j - 1]
else: # Otherwise, take the maximum of ignoring either character
dp[i][j] = max(dp[i + 1][j], dp[i][j - 1])

# Step 4: The longest palindromic subsequence length is in dp[0][n-1]
return dp[0][n-1]

```

## Example

b	b	a	b	c	b	c	a	b	
b	1	2	1	3	3	5	5	5	7
b		1	1	3	3	5	5	5	5
a			1	1	1	3	3	3	3
b				1	1	3	3	3	5
c					1	1	1	3	3
b						1	1	3	3
c							1	1	1
a								1	1
b									1

## Runtime

$O(n^2)$

Since filling a  $n \times n$  table

## Chain Matrix Multiplication

- Given sequence of matrices, find the optimal order to multiply them so that the total number of operations is minimized

### How to DP it

- $dp[i][j]$  represents the minimum cost of multiplying the matrices from  $A[i]$  to  $A[j]$
- Base Case:
  - $i == j$ : given only one matrix
    - cost is zero bc no multiplication is needed
- Recursive: Find cost of multiply matrices – split to two parts --  $(A[i...k]) * (A[k+1...j])$ 
  - Cost = cost of multiplying two subchains + cost of multiplying the result of the two subchains
    - Do for all possible values of ‘k’ ( $i$  to  $j - 1$ ) and take minimum cost:  $dp[i][j] = \min(dp[i][k] + dp[k+1][j] + di-1 * dk * dj)$
    - $di-1 * dk * dj$  is the cost of multiplying the result of  $A[i...k]$  (which has dimensions  $di-1 * dk$ ) with  $A[k+1...j]$  (which has dimensions  $dk * dj$ )
- Solution:  $dp[1][n]$  where  $n$  is the number of matrices

### PseudoCode

```
def chainmatrix(dimensions):
    n = len(dimensions) - 1 # Number of matrices
    dp = [[0] * n for _ in range(n)]

    # Fill the table in a bottom-up manner
    for s in range(1, n): # s is the length of the subchain
        for i in range(n - s):
            j = i + s
            dp[i][j] = float('inf') # Set to a large value
            for k in range(i, j):
                cost = dp[i][k] + dp[k + 1][j] + dimensions[i] * dimensions[k + 1] * dimensions[j + 1]
                dp[i][j] = min(dp[i][j], cost)

    return dp[0][n - 1] # This holds the minimum cost for the entire chain
```

## Example

- A:  $50 \times 20$ ; B:  $20 \times 1$ ; C:  $1 \times 10$ ; D:  $10 \times 100$
- Sizes:  $[50, 20, 1, 10, 100]$
- $Dp[n][n]$  where  $n = 4$  (4 matrices)
- Filling DP table:
  - Base Case:

0	-	-	-
-	0	-	-
-	-	0	-
-	-	-	0

- Subchains of length 2 ( $s = 1$ ):
  - $dp[1][2] = 50 \times 20 \times 1 = 1000$ 
    - m0m1m2
  - $dp[2][3] = 20 \times 1 \times 10 = 200$ 
    - m1m2m3
  - $dp[3][4] = 1 \times 10 \times 100 = 1000$ 
    - m2m3m4

0	1000	-	-
-	0	200	-
-	-	0	1000
-	-	-	0

- Subchains of length 3 ( $s = 2$ ):
  - For  $dp[1][3]$ , we try splitting the chain at  $k = 1$  and  $k = 2$ :
    - For  $k = 1$ :  $50 \times 20 \times 10 = 10,000$ 
      - m0m1m3
    - For  $k = 2$ :  $50 \times 1 \times 10 = 500$ 
      - m0m2m3
    - **Minimum: 1500**
  - For  $dp[2][4]$ , we try splitting the chain at  $k = 2$  and  $k = 3$ :
    - For  $k = 2$ :  $20 \times 1 \times 100 = 2000$ 
      - m1m2m4
    - For  $k = 3$ :  $20 \times 10 \times 100 = 20,000$ 
      - m1m3m4

- **Minimum:** 3000
- $T[1,3] = \min(T[1,1] + T[2,3] + m_0m_1m_3 = 1500, T[1,2] + T[3,3] + m_0m_2m_3 = 1500) = 1500$
- $T[2,4] = \min(T[2,2] + T[3,4] + m_1m_2m_4 = 3000, T[2,3] + T[4,4] + m_1m_3m_4 = 3000) = 20200 = 3000$

0	1000	1500	-
-	0	200	3000
-	-	0	1000
-	-	-	0

○ Subchains of length 4 ( $s = 3$ ):

- for  $dp[1][4]$ , we try splitting the chain at  $k = 1$ ,  $k = 2$ , and  $k = 3$ :
  - For  $k = 1$ :  $50 \times 20 \times 100 = 100,000$ 
    - $m_0m_1m_4$
  - For  $k = 2$ :  $50 \times 1 \times 100 = 5000$ 
    - $m_0m_2m_4$
  - For  $k = 3$ :  $50 \times 10 \times 100 = 50,000$ 
    - $m_0m_3m_4$
  - Minimum: 7000

$$dp[1,4] = \min \left\{ \begin{array}{ll} dp[1,1] + dp[2,4] + m_0m_1m_4, & \text{if } k=1 \\ dp[1,2] + dp[3,4] + m_0m_2m_4, & \text{if } k=2 \\ dp[1,3] + dp[4,4] + m_0m_3m_4, & \text{if } k=3 \end{array} \right\}$$

$$dp[1,4] = \min \left\{ \begin{array}{ll} 0 + 3000 + 100000, & \text{if } k=1 \\ 1000 + 1000 + 5000, & \text{if } k=2 \\ 1500 + 0 + 50000, & \text{if } k=3 \end{array} \right\} = 7000$$

0	1000	1500	7000
-	0	200	3000
-	-	0	1000
-	-	-	0

Optimal Cost at  $dp[0][3]$ : 7,000

## Runtime

$O(n^3)$



## Knapsack

- Given  $n$  items with each:
  - A **value**  $v_i$
  - A **weight**  $w_i$
- Also given a knapsack with a maximum capacity “ $W$ ”
- Goal: pick items to maximize the total value without exceeding the weight capacity “ $W$ ”

### Example

- $W = 6$ ; Weights = [1, 2, 3, 4], Values = [12, 17, 18, 25]
  - Most optimal solution is items 1, 2, and 3  $\rightarrow 1 + 2 + 3$  with value  $12 + 17 + 18 = 47$

### How to DP it

- $T[i][j]$  where  $i$  = current capacity from 0 to  $W$  and  $j$  items 1 to  $n$
- $T[i][j]$  will store the maximum value we can get with capacity  $i$  and the first  $j$  items

### Key Idea – No Repetition

- At every step, for each item  $j$  and capacity  $i$ 
  - Don’t pick the item: maximum value is the same as without considering the item
    - $T[i][j] = T[i][j - 1]$
  - Pick the item: add the value of the item and reduce the capacity by its weight
    - $T[i][j] = T[i - w_j][j - 1] + v_j$  (if the weight fits)
- Final Solution:  $T[W][n]$

### Pseudocode – No Repetition

```
def knapsack(W, weights, values):  
    n = len(weights)  
    dp = [[0] * (n + 1) for _ in range(W + 1)]  
  
    for j in range(1, n + 1):  
        for i in range(1, W + 1):  
            if weights[j - 1] <= i:  
                dp[i][j] = max(dp[i][j - 1], dp[i - weights[j - 1]][j - 1] + values[j - 1])  
            else:  
                dp[i][j] = dp[i][j - 1]
```

```
    return dp[W][n]
```

## Key Idea – Repetition

- In the version where you can pick the same item multiple times:
  - The only change is that instead of using  $T[i - w_j][j-1]$ , you use  $T[i - w_j][j]$  (since you can use the same item again).

## Pseudocode – Repetition

```
def knapsack_repeat(W, weights, values):  
    dp = [0] * (W + 1)  
  
    for i in range(1, W + 1):  
        for j in range(len(weights)):  
            if weights[j] <= i:  
                dp[i] = max(dp[i], dp[i - weights[j]] + values[j])  
  
    return dp[W]
```

## Runtime

$O(nW)$

## Example 1

### Problem

Suppose you have  $n$  stair steps. You can leap one, two, or three steps at a time. What is the number of combinations, or the number of ways, that you could reach step  $n$ ?

### Solution

Create an array  $T[0 \dots n]$  where  $T[i]$  represents the number of ways to reach step  $i$ .

Our **base cases** are  $T[0] = 1, T[1] = 1$  and  $T[2] = 2$ . To reach step  $n$ , there was some last step. You either jumped 1, 2, or 3 steps, and you jumped from 1, 2, or 3 steps away. That means that the number of ways to go from 0 to  $n$  is the sum of the number of ways to go from 0 to  $n-1$ , 0 to  $n-2$ , and 0 to  $n-3$ . This gives us our **recurrence** of:  $T[i] = T[i - 1] + T[i - 2] + T[i - 3]$

### Pseudocode

```
def numways(k):
    initialize dp as table of size n + 1 with 0s
    dp[0] = 0
    dp[1] = 1
    dp[2] = 2

    for i in 3..(n+1):
        dp[i] = dp[i - 1] + dp[i - 2] + dp[i - 3]

    return dp[k]
```

indices	0	1	2	3	4	5	6	7	8	9
results	1	1	2	4	7	13	24	44	81	149

### Runtime

size of the table multiplied by the work done to calculate each cell --  $O(n) \times O(1) = O(n)$

## Example 2

### Problem

Given two operations: add one and multiply by two

How many operations does it take to get from 0 to some k

### Solution

The results are as follows.

indices	0	1	2	3	4	5	6	7	8	9
results	0	1	2	3	3	4	4	5	4	5

It's not monotonic, and it's actually a little hard. Note that for 8 it's a power of two so it takes fewer operations than either of its neighbors. Trivially you could find this in exponential time by trying all possibilities. Let's write our recurrence.

$dp[0] = 0$  because for 0, you need zero operations.  $dp[1] = 1$  because for 1, you add one to zero. Now we'll try to multiply by two as much as possible so we'll add 1 if the element is odd and multiply by two if it is even. We get the following overall recurrence.

$$d(i) = \begin{cases} d(i-1) + 1, & \text{if } i \text{ is odd} \\ d(i/2) + 1, & \text{if } i \text{ is even} \end{cases} \quad (1)$$

### Pseudocode

```
def minop(k):
    initialize dp as table of size k + 1 with 0s
    dp[0] = 0
    dp[1] = 1

    for i in 2..(k + 1)
        dp[i] = dp[i - 1] + 1
        if i is even
            dp[i] = min(dp[i], dp[i / 2] + 1)

    return dp[k]
```

## Example 3

### Problem

Given an array houses with values  $H = [h_1 \dots h_n]$ , you want to rob them, but you can't rob two adjacent houses or alarms will go off. What is the maximum amount you can steal?

Here's an example:  $[2, 7, 9, 3, 1]$  would yield  $2 + 9 + 1 = 12$ .

### Solution

Let's define our array  $T[0 \dots n]$  with  $T[i] =$  maximum we can steal from houses  $0 \dots i$ .

The **base cases** are that  $T[0] = H[0], T[1] = \max(H[0], H[1])$ .

**Recurrence:** At the next house visited, we can choose to rob or not rob a house. We take the maximum of both scenarios. If we rob the house, then we could not have robbed the previous house, and if we didn't rob the house, then we could have robbed the previous house. This gives us the **recurrence**:  $T[i] = \max(T[i - 2] + H[i], T[i - 1])$

### Pseudocode

```
def houserobber(H):
    initialize dp as table of size n with 0s
    dp[0] = H[0]
    dp[1] = max(H[0], H[1])

    for i in 2..n
        dp[i] = max(dp[i - 2] + H[i], dp[i - 1])
    return dp[n - 1]
```

### Runtime

$O(n)$  since we have a table of size  $n$  with  $O(1)$  work at each step

## Example 4

### Problem

Find the number of paths to reach cell  $(n, m)$ . Additionally, the grid has bombs, denoted by an input `bombs[i][j]` = True if cell  $(i, j)$  has a bomb. Find number of paths from  $(0, 0)$  to  $(n, m)$  such that no bombs are traversed.

### Solution

Let's define array  $T[1 \dots n][1 \dots m]$  where  $T[i][j]$  = the number of paths from  $(1, 1)$  to  $(i, j)$ .

Base Cases:

1. If the starting cell  $(1, 1)$  has a bomb, then there are 0 paths.

$$dp[1][1] = \begin{cases} 0 & \text{if } bombs[1][1] = \text{True} \\ 1 & \text{otherwise} \end{cases}$$

2. For the first row and first column:

$$dp[i][1] = \begin{cases} 0 & \text{if } bombs[i][1] = \text{True} \\ dp[i - 1][1] & \text{otherwise} \end{cases}$$

$$dp[1][j] = \begin{cases} 0 & \text{if } bombs[1][j] = \text{True} \\ dp[1][j - 1] & \text{otherwise} \end{cases}$$

Recurrence Relation:

For all cells not on the first row or first column:

$$dp[i][j] = \begin{cases} 0 & \text{if } bombs[i][j] = \text{True} \\ dp[i - 1][j] + dp[i][j - 1] & \text{otherwise} \end{cases}$$

This means that if a cell has a bomb, no paths can go through it. Otherwise, the number of paths to a cell is the sum of the paths to the cell above it and the cell to its left.

The solution will be in  $dp[n][m]$ , which gives the number of paths from  $(1, 1)$  to  $(n, m)$  without traversing any bombs.

## Pseudocode

```
def count_paths_with_bombs(n, m, bombs):
    initialize dp as a table of size (n+1)*(m+1) with 0s

    if bombs[0][0]:
        return 0

    dp[0][0] = 1

    for i in 1...n:
        if not bombs[i][0]:
            dp[i][0] = dp[i-1][0]

    for j in 1...m:
        if not bombs[0][j]:
            dp[0][j] = dp[0][j-1]

    for i in 1...n:
        for j in 1...m:
            if not bombs[i][j]:
                dp[i][j] = dp[i-1][j] + dp[i][j-1]

    return dp[n][m]
```

## Runtime

$O(n \times m)$  because visiting each cell of the grid once.

# NP-Completeness

## Complexity Theory

- Upper Bound: worst-case performance of an algorithm
- Lower Bound: best possible performance for solving a problem
- **Reduction:** transforming one problem into another to prove properties about the problem
  - Ex: Given algorithm that could sort in  $O(n)$  – would have a contradiction with the known lower bound of  $O(n \log n)$  – therefore has lower bound of  $\Omega(n \log n)$

## Decision vs Search Problems

- Decision: yes/no question
  - Ex: Does there exist a path from  $s$  to  $t$  in graph  $G$  of PATH =  $\{<G, s, t>\}$
- Search: returns a solution

## Complexity Classes

### Complexity Class P

- Problems that can be solved in POLYNOMIAL TIME – efficient problems
  - Ex: Finding if two numbers are relatively prime using the GCD algorithm

### Complexity Class NP

- Problems where solutions can be **verified** in polynomial time, but finding the solution may not be efficient
  - May not be able to solve in polynomial but can check if it's correct in polynomial time
  - Ex: The problem of checking if a number  $n$  is composite. If you are given factors of  $n$ , you can verify in polynomial time that the factors multiply to " $n$ "

### Relationship between P and NP

- $P \subseteq NP$ : If a problem is in P, it is also in NP because solving the problem can also verify the solution.
- $P = NP$  is NOT guaranteed because we don't know if problems that are VERIFIABLE in polynomial time are can be SOLVED in polynomial time

## NP-Completeness

- If one NP-complete problem can be solved in polynomial time, all NP problems can be solved in polynomial time
  - $B \in NP$ : Show that the problem B is in NP (verifiable in polynomial time)
  - B is NP-hard: Prove that an NP-complete problem A reduces to B in polynomial time, denoted  $A \leq pB$ . This means that if you can solve B, you can also solve A.
- So reduce a known NP-complete problem to B
- Steps:
  - Choose a known NP-complete problem A
  - Define a reduction f from A to B. This reduction must be computable in polynomial time.
  - Satisfiability must be preserved: If  $x \in A$ , then  $f(x) \in B$ , and if  $x \notin A$ , then  $f(x) \notin B$

## Proving $P = NP$ or $P \neq NP$

- **$P = NP$ :** If we find a polynomial-time algorithm for any NP-complete problem (like SAT), it means  $P = NP$
- **$P \neq NP$ :** If we can show that there are problems in NP that cannot be solved in polynomial time, we will prove  $P \neq NP$

## SAT (Satisfiability Problem)

- **Variable:** A boolean variable (e.g.,  $x_1, x_2, \dots$ ).
- **Literal:** A variable or its negation (e.g.,  $x_1$  or  $\neg x_1$ ).
- **Clause:** An OR of literals (e.g.,  $x_1 \vee \neg x_2$  ).
- **Formula in CNF:** An AND of clauses (e.g.,  $(x_1 \vee \neg x_2) \wedge (x_3 \vee x_4)$ )

### Definition

- A formula is in SAT if it has a truth assignment that makes the formula true
  - $(x \vee y \vee z) \wedge (x \vee z \vee w) \wedge \dots$  can be satisfied by assigning appropriate values to variables
- Every problem in NP can be reduced to SAT, meaning SAT represents the entire NP class – if SAT can be solved in polynomial time, then  $P = NP$

## kSAT/3SAT

- **kSAT:** SAT problem where each clause has at most  $k$  literals
- **3SAT:** SAT problem where each clause has 3 literals

### Proving 3SAT to be NP-Complete

- **3-SAT  $\in$  NP:** The witness is simply the variable assignments, which can be checked in polynomial time
- **SAT  $\leq_p$  3-SAT:** If a SAT formula has clauses with more than 3 literals, you can break them down into smaller clauses by introducing new variables (e.g., convert  $(x_1 \vee x_2 \vee \dots \vee x_k)$  into two clauses).
- **Conclusion:** SAT reduces to 3-SAT, so 3-SAT is NP-complete.

## 2SAT/Why its P

- **2SAT:** case where each clause has 2 literals
- **2-SAT  $\in$  P:** You can reduce 2-SAT to graph problems. A 2-SAT problem can be solved using graph reachability techniques, making it solvable in polynomial time.

### Example

- A clause  $(a \vee b)$  can be viewed as implications:  $\neg a \rightarrow b$  and  $\neg b \rightarrow a$

## CircuitSAT

- Given a Boolean circuit with AND/OR/NOT gates, determine if there is an input that makes the output 1

### Proving its NP-Complete

- CircuitSAT  $\in$  NP:** A verifier can evaluate the circuit with a given input in polynomial time
- 3-SAT  $\leq_p$  CircuitSAT:** Convert a 3-SAT formula into a boolean circuit. Each clause in 3-SAT becomes a sub-circuit, and the entire formula becomes a circuit with AND gates combining the clauses
- If a 3-SAT formula is satisfiable, the corresponding circuit is satisfiable, making CircuitSAT NP-complete.

## Independent Set

### Problem Definition

Given a graph  $G(V,E)$  and an integer  $g$ , determine if there exists a subset of vertices  $I \subseteq V$  with size  $g$  such that no two vertices in  $I$  are connected by an edge.

### NP-Complete Proof

- Proving NP:
  - Verify a solution by checking if the set has size  $g$  and ensuring no edges exist between any two vertices in the set.
- NP-Hardness
  - For each clause in a 3SAT formula, create a “triangle” of vertices labeled by the literals of that clause
  - Add edges between corresponding literals and their negations across different triangles
  - The graph has an independent set of size  $g$  (number of clauses) if and only if the 2SAT formula is satisfiable
- Since Independent Set is both NP and NP hard it is NP-Complete

## Clique

### Problem Definition

Given a graph  $G(V, E)$  and an integer  $g$ , determine if there exists a subset of vertices that forms a complete subgraph (clique) of size  $g$ .

### NP-Complete Proof

- Proving NP:
  - Verify a solution by checking if the set has size  $g$  and ensuring that every pair of vertices in the set is connected
- NP-Hardness
  - Reduce from Independent Set
    - Convert the graph  $G$  to its complement  $G'$
    - A clique of size  $g$  in  $G'$  corresponds to an independent set of size  $g$  in  $G$ , and vice versa
  - So Clique is as hard as solving the Independent Set Problem
- Since Clique is both NP and NP hard it is NP-Complete

## Vertex Cover

### Problem Definition

Given a graph  $G(V, E)$  and an integer  $g$ , determine if there exists a subset of vertices of size  $g$  such that every edge in the graph has at least one endpoint in this subset.

### NP-Complete Proof

- Proving NP:
  - Verify a solution by checking if the set has size  $g$  and ensuring that every edge is covered by at least one vertex from the set
- NP-Hardness
  - Reduce from Independent Set
    - If a graph  $G$  has an independent set of size  $g$ , then the complement of the set (remaining vertices) forms a vertex cover of size  $|V| - g$ , and vice versa
- Since Vertex Cover is both NP and NP hard it is NP-Complete

## Subset Sum

### Problem Definition

Given a set  $S$  and a target  $t$ , find a subset  $S' \subseteq S$  such that the sum of elements in  $S'$  equals  $t$ .

### NP-Complete Proof

- Proving NP:
  - If given a subset  $S'$ , summing the elements to check if they equal  $t$  can be done in polynomial time
- NP-Hardness
  - Reduce from 3SAT
    - Encode each variable and its negation as binary numbers
    - Create a target number where each bit corresponds to the presence of either a variable or its negation
    - For a 3SAT instance, build a table where rows represent literals, and columns represent their occurrences in clauses
    - Clauses: ensure each has at least one true literal by setting target values that restrict the total to exactly three
  - Correctness:
    - If 3SAT is satisfiable, Subset Sum solution exists
    - If 3SAT is unsatisfiable, no subset will sum to the target
- Since Subset Sum is both NP and NP hard it is NP-Complete

## Knapsack

### Problem Definition

Given items with weights and values, determine if there's a subset whose total weight doesn't exceed a capacity  $W$ , and whose total value meets or exceeds  $V$

### NP-Complete Proof

- Proving NP:
  - Given a subset of items, checking if the weight and value constraints are met can be done in polynomial time
- NP-Hardness

- Reduce from Subset Sum
  - For each element in Subset Sum, create a corresponding item where weight equals value
  - Set the capacity  $W$  and value  $V$  in Knapsack to the target sum  $t$  from Subset Sum
  - If a subset in Subset Sum sums to  $t$ , the corresponding items in Knapsack will meet both the weight and value constraints, proving the reduction
- Since Knapsack is both NP and NP hard it is NP-Complete

# Onions

## Divide and Conquer

Max's Anger Contest Series 1 P3 – Divide and Connor

[View the Problem Here!](#)

Solution: [view here](#)

Explanation: [view here](#)

Hints:

1. Think about how to break the array into three equal parts. If you're unfamiliar with `Arrays.copyOfRange`, you can simply iterate to divide the array into smaller parts.
2. Once you've split and processed the smaller segments, make sure you reorder them according to the problem's instructions: third part -> first, first part -> second, second part -> third.
3. After splitting and reordering, merge the segments back together into a new array.

ICPC NEERC 2012 C - Caravan Robbers

[View the Problem Here!](#)

Solution: [view here](#)

Explanation: [view here](#)

Hints:

1. Sort the intervals by their starting point to simplify processing.
2. Use binary search to find the maximum possible length of the intervals by testing different lengths.
3. For each length tested in binary search, make sure all gangs can fit their intervals without overlapping by checking against their starting and ending points.
4. Use the gcd to convert the interval length into an irreducible fraction for the final result.

Age Demographic

[View the Problem Here!](#)

Solution: [view here](#)

Explanation: [view here](#)

Hints:

1. **Sort the Ages:** Sorting the list of ages makes it possible to efficiently find the range of ages that meet the query condition.
2. **Binary Search:** Use `Arrays.binarySearch()` to find the positions of `a` and `b` in the sorted array. If `a` or `b` are not in the array, use the insertion point given by `binarySearch()` to get the correct range.
3. **Counting the Range:** Once you have the positions for `a` and `b`, calculate the difference between them to get the number of viewers within that age range.
4. **Edge Cases:** Handle edge cases where `a` or `b` might not exist in the array, or when all viewers are outside the queried range.

## Numbers

### TLE '17 Contest 7 P3 - Countless Calculator Computations

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Use binary search to find X
2. Make a separate function to calculate the result of the repeated exponentiation up to Y levels
3. Use a small epsilon in the binary search

### DWITE '07 R5 #2 - Number of factors

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Understand Prime Factorization: The function breaks down the number into its prime factors, excluding 1 and the number itself
2. Handle Factors of 2 First: This simplifies the process, allowing the loop to focus only on odd factors afterward.
3. Check for Remaining Prime: After dividing out all smaller factors, if a prime remains, count it.
4. Adjust for Primes: If the number is prime and has no other factors, adjust the count accordingly to return 0.

### DMOPC '20 Contest 3 P5 - Tower of Power

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Use Euler's Totient Function to reduce exponents in modular calculations.
2. Apply Modular Exponentiation to handle large powers efficiently.
3. Calculate Tower Power from top to bottom, using reduced moduli at each step.

## Graphs

### CCC '10 J5 - Knight Hop

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Use BFS for Shortest Path: Since the knight moves in all directions and can revisit positions, BFS ensures the shortest path is found.
2. Track Visited Positions: Prevent revisiting squares by using a visited set, which helps in reducing redundant calculations.
3. Check Boundaries: Always ensure that the knight stays within the 8x8 chessboard boundaries during each move.

### ACSL '09 Practice P4 – Rank

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Model the Problem as a Graph: Players are nodes, and victories create directed edges.
2. Use DFS to Detect Cycles: Keep track of nodes in the current path (recursion stack) to find cycles.
3. Count Unique Players in Cycles: Ensure each player in a cycle is only counted once

### Is it a Tree?

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Check Edge Count: Ensure the number of edges matches  $n-1$ , which is necessary for a tree.
2. Validate Symmetry: The adjacency matrix must be symmetric since the graph is undirected.
3. Use BFS/DFS: Traverse the graph to check for connectivity and cycles, which will help determine if it is a valid tree.

## IOI '11 P2 - Race (Standard I/O)

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Understand Centroid Decomposition: It's crucial for breaking down the tree into manageable parts while maintaining efficient query handling.
2. Distance Pairing: Focus on finding pairs of distances that sum to  $K$  using DFS.
3. Optimize Using Stored Distances: Store distances from each centroid and use them to quickly find valid paths.
4. Recursive Processing: Apply centroid decomposition recursively to handle all parts of the tree.

## IOI '13 P1 - Dreaming (Standard I/O)

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Understand Tree Diameter: Ensure you can find the longest path in a tree using two BFS operations.

2. Tree Radius Insight: Be familiar with how tree radius helps in calculating the maximum distance when adding new trails.
3. Combination of New Trails: Consider different ways to connect key nodes with new trails to minimize the maximum distance between any two billabongs.
4. Efficiency: Pay attention to the graph traversal techniques and how they help in efficiently finding the solution.

## Baltic OI '05 P3 – Maze

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Familiarize yourself with how BFS can be adapted to handle different edge conditions, such as alternating colors in this case.
2. Recognize that you need to track the state of the last edge color to properly apply the movement rules.
3. Ensure you can correctly represent and initialize the graph based on the maze's input format.
4. Understand how the distance array is used to maintain and update shortest path distances with state-specific conditions.
5. Ensure the solution handles cases where no valid path exists by checking if the distance remains at its initialized value.

## RGPC '17 P4 - Snow Day

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Use Topological Sort for DAGs: Process nodes in topologically sorted order to handle longest path calculations.
2. Dynamic Programming: Apply dynamic programming to update distances and node counts along the longest path.
3. Handle No Path Cases: Ensure to check and handle cases where no path exists between the start and end nodes.

## Single Source Shortest Path

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Use of Adjacency Matrix: Efficiently stores edge weights for dense graphs but can be memory-intensive for large sparse graphs.
2. Dijkstra's Implementation: Select the unprocessed node with the smallest distance and update its neighbors.
3. Handling Unreachable Nodes: Use Integer.MAX\_VALUE to indicate nodes that are not reachable from the source.

## Another Random Contest 1 P5 - A Strange Country

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Union-Find with Path Compression: Efficiently manage and merge sets of connected nodes.
2. Binary Search for Optimization: Quickly determine the earliest day a road can be used without causing a cycle in the MST.
3. Sorting by Weight: Ensures that the MST is built in increasing order of edge weights, adhering to Kruskal's algorithm principles.



## EZDP

### IOI '94 P1 - The Triangle

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Bottom-Up Calculation: Update each element based on its two possible paths to the next row, ensuring that the maximum possible sum is considered.
2. In-Place Update: Modify the triangle matrix directly to save space and simplify the implementation.
3. Starting Point: Begin updating from the second-to-last row and move upwards, ensuring that all necessary sub-problems are solved before considering the current row.

### IOI '07 P4 – Miners

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. State Encoding: Use state encoding to manage the last two shipments efficiently and compute transitions.
2. Transition Precomputation: Precompute state transitions to simplify the DP update process.
3. Reverse Processing: Process shipments from last to first to build the DP table correctly based on future shipments.
4. Optimization: Keep track of coal production through state transitions to optimize the allocation of shipments to maximize coal production.

### Educational DP Contest AtCoder A - Frog 1

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. State Definition: Define  $dp[i]$  as the minimum cost to reach stone  $i$ .
2. State Transition: Use previous stone costs to compute the current stone's cost.
3. Initialization: Initialize the costs for the first two stones and compute subsequent costs iteratively.

## Educational DP Contest AtCoder H - Grid 1

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. State Definition: Define  $dp[i][j]$  as the number of ways to reach cell  $(i, j)$ .
2. State Transition: Sum the number of paths from top and left cells.
3. Boundary Conditions: Handle blocked cells (#) where no paths should be added.

## Educational DP Contest AtCoder B - Frog 2

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. State Definition: Define  $dp[i]$  as the minimum cost to reach stone  $i$ .
2. State Transition: Update  $dp[i]$  by considering all reachable previous stones  $j$  and calculating the cost based on the height difference.
3. Efficiency: Iterate from each stone to all stones within  $K$  jumps to ensure you find the minimum cost.

## IOI '04 P4 – Phidias

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. State Definition: Define  $dp[w][h]$  to track the minimum waste for a slab of size  $w \times h$ .
2. State Transition: Consider all possible vertical and horizontal splits to update the DP table.
3. Efficiency: Check if a plate size is needed to avoid wasting the slab of that size.

## IOI '14 Practice Task 1 – Square

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Initialization: Start with dp table filled with zeros.
2. Edge Cases: Handle cells on the grid's boundary separately as they can only form squares of size 1.
3. Updates: Update max\_square and count\_max\_square whenever a larger square is found or an additional square of the same size is detected.

## Knapsack

### DMOPC '17 Contest 2 P1 - 0-1 Knapsack

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Filtering: Items with zero or negative values are irrelevant for maximizing value.
2. Capacity Calculation: The minimum knapsack size needed is the sum of capacities of all valuable items, ensuring no valuable item is left out.

### Educational DP Contest AtCoder D - Knapsack 1

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Sorting: Items are sorted by value to potentially prioritize high-value items, but this step is not strictly necessary for correctness.
2. Reverse Iteration: Process capacities in reverse order to avoid using the same item multiple times within a single iteration.
3. DP Transition: Update  $dp[j]$  by checking if including the item results in a higher value than the current maximum for capacity  $j$

## Knapsack 4

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Strength Balancing: By using negative values for engineers, the problem reduces to a single knapsack problem with an offset.
2. DP Table Size: The DP table is sized based on the maximum possible strength balance, ensuring all potential balances are covered.
3. Shuffling Items: Randomize the order of processing items to improve performance in practice.

## NP-Completeness

### ICPC PACNW 2016 F – Illumination

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. 2-SAT Representation: Use 2-SAT to encode "either/or" constraints for each lamp's illumination choice.
2. Implication Construction: If two lamps could overlap, they cannot both be in the same mode (row or column).
3. SCC Check: If a lamp's row and column options are in the same SCC, it means there's a conflict, making the configuration invalid.

### COCI '13 Contest 2 #4 Putnik

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. State Representation: The state  $dp[\text{left}][\text{right}]$  captures the minimum cost to cover remaining cities while satisfying the constraints.
2. Recursive Transitions: Consider the cost of including the next city in either the segment before right or the segment after right and choose the minimum.

### Wesley's Anger Contest 1 Problem 5 - Rule of Four

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. SCC Finding: Tarjan's algorithm is used to decompose the implication graph into SCCs.
2. Transitive Closure: Helps in understanding reachability and ensuring all constraints are respected.
3. Consistency Check: Ensures no contradictory assignments and computes the possible trip configuration.

## Miscellaneous

### Mock CCC '20 Contest 1 J2 - A Simplex Problem

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Iterate Valid Combinations: Check all combinations where the sum of cacti and matroids does not exceed k.
2. Compute Joy: For each combination, compute the total joy and compare it to the maximum joy found so far.

### Wesley's Anger Contest 1 Problem 1 - Simply a Simple Simplex Problem

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. No hints this time!

## System Solver

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Gaussian Elimination: Used to transform the system into an upper triangular form for easy back substitution.

2. Pivoting: Ensures numerical stability by swapping rows to place non-zero elements on the diagonal.
3. Rank: Helps determine the feasibility of a unique solution based on the number of non-zero rows compared to the number of variables.

## DMOPC '21 Contest 9 P5 - Chika Circle

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Check neighbor values to infer card values directly when possible.
2. Use the constraints of the problem to narrow down possible card assignments.
3. Ensure that the final assignment matches all the provided responses.

## WC '99 P3 - The Godfather

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. No hints this time!

## Wesley's Anger Contest Reject 5 - Two Permutations

[View the Problem Here!](#)

Solution: [View Here](#)

Explanation: [View Here](#)

Hints:

1. Use of Maps: first and second maps track which index produced which value for quick lookup.
2. Random Indices: Randomly select indices to guess, ensuring diverse coverage.
3. Efficiency: The nested loop checks for matches between stored values efficiently.

## LeetCode