



UNIVERSITY OF
SAN FRANCISCO

Master of Science
in Analytics

Algorithms

Interview Skills



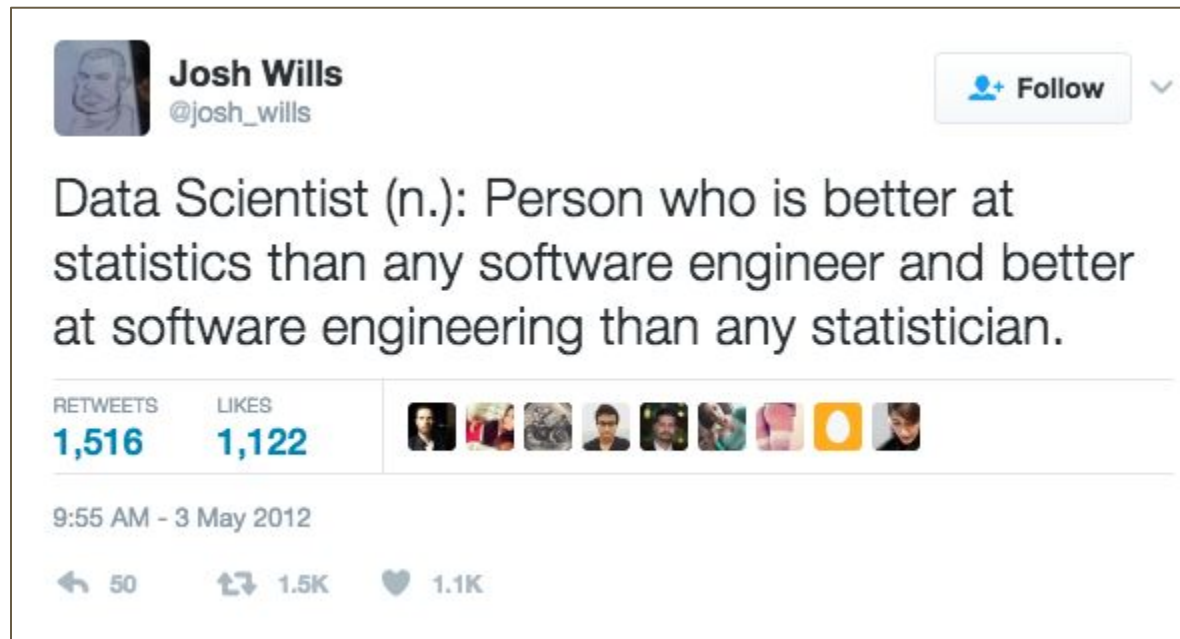
Agenda

- Whiteboarding
- Algorithms (on array / arraylist instances)
 - Searching
 - Sorting
- Data structures
 - Object Oriented Programming
 - Hashtables
 - Stacks
 - Queues
 - Trees
 - Graphs



Why Study Algorithms?

- Many people interviewing for data science are computer scientists
 - Until data science comes into its own as a field, many interviewers will be software engineers, database engineers, etc.
 - Algorithms sits at the core of computer science
- Many topics from Algorithms apply to data science
 - Ideas from computational complexity apply to databases (SQL, etc.)
 - The expectation is that you can do both well





Whiteboard Interviewing

- What is whiteboarding?
 - Candidate is given a blank writing surface (chalkboard, markerboard, paper, etc.)
 - Candidate is given a technical question and expected to write a working solution
 - Common in technical interviews
 - Measures candidate's problem-solving and communication skills
- How to succeed at whiteboarding:
 - Write the question and any details you might forget (examples, guidelines, etc.)
 - Write legibly (i.e. large letters, even spacing, predictable indentation, etc.)
 - Use the space on the whiteboard efficiently — eg. leave space between lines of code — but do not exceed the size of the canvas
 - Ask questions and talk through your solution while you write
- If you don't do well:
 - Don't take it personally — highly qualified people have tried and failed — it is a completely artificial exercise
 - Try, try again (practice with mock whiteboarding interviews)



All These Things and More...

- All algorithms and data structures are standard for a CS curriculum
 - All are available in standard CS references; all are available online
 - Some are variations
 - See [Problem Solving with Algorithms and Data Structures using Python](#)
- What good coders do
 - Understand the concept, memorise the steps
 - Avoid the urge to memorise code



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Arrays

- An array is a (memory-contiguous) series of objects sharing the same type
 - Possibly in 1 dimension (vector), 2D (matrix), or N-dimensions (tensor)
 - In Python, there are no arrays, but there is a List — i.e. ArrayList — which is an array of pointers to objects
- Access
 - Can access any element by name and offset (for example A[5])
 - First element is usually at index 0 (though there are some [1-based languages](#))
- Assumptions
 - Access to any part of the array can be performed in one read
 - Read operations and write operations have equal cost



Task: Search

- Task
 - Given: an array (A), a target value (target)
 - Return the index of target in A
 - Return -1 if target is not in A
- Algorithm: Linear Search
 - Iterate from the first element to the last, keeping track of the index
 - At each iteration, if the element is equal to the target, return the index
 - If no element is found at the end of the iteration, return -1
 - Used to implement "in" ala "if letter in word"
- Python implementation:

```
def linear_search (A, target):  
    for i in range(len(A)):  
        if A[i] == target:  
            return i  
    return -1
```



Another Search Solution

- Algorithm: Binary Search
 - Same task for search, but assume A is sorted in non-decreasing order
 - Look for the target at the middle of the array; if the middle is equal to the target, return it
 - Recalculate the range where the target may be
 - If the target is greater than the middle element, ignore the lower half of the array
 - If the target is less, ignore the upper half
 - If “upper” and “lower” cross, return -1
- Implementations:
 - There are recursive and iterative implementations
 - The recursive implementation may be more common, but the iterative implementation is more efficient



Iterative Binary Search (Python)

```
def binary_search (A, target):  
    lower = 0  
    upper = len(A) - 1  
    while (lower <= upper):  
        middle = (lower + upper) / 2  
        if A[middle] == target:  
            return middle  
        else:  
            if target < A[middle]:  
                upper = middle - 1  
            else:  
                lower = middle + 1  
    return -1
```



Computational Complexity

- Compare algorithms primarily by number of array reads and writes
 - Using the number of items in the array (n), the worst case for search (item not in array)
 - Linear search reads all n items
 - Binary search reads about $\log_2(n)$ items
 - Use Big-O notation
 - Formally, if f and g are two functions, we can say $f(x) = O(g(x))$ if $f(x) \leq M(g(x))$ for all $x > x_0$, given a constant M
 - Informally: throw out constants, lower-order terms
- Complexity
 - Common to use “worst case” running time
 - It is also possible to use best case, average case
 - Always use “worst case” unless it is possible to prove that it is uncommon
 - Search for our search algorithms (worst case):
 - Linear Search = $O(n)$
 - Binary Search = $O(\log n)$



Differences in Speed

- When the size of the array is small, speed differences are negligible
- With large values of n (thousands, millions, etc.), differences are stark

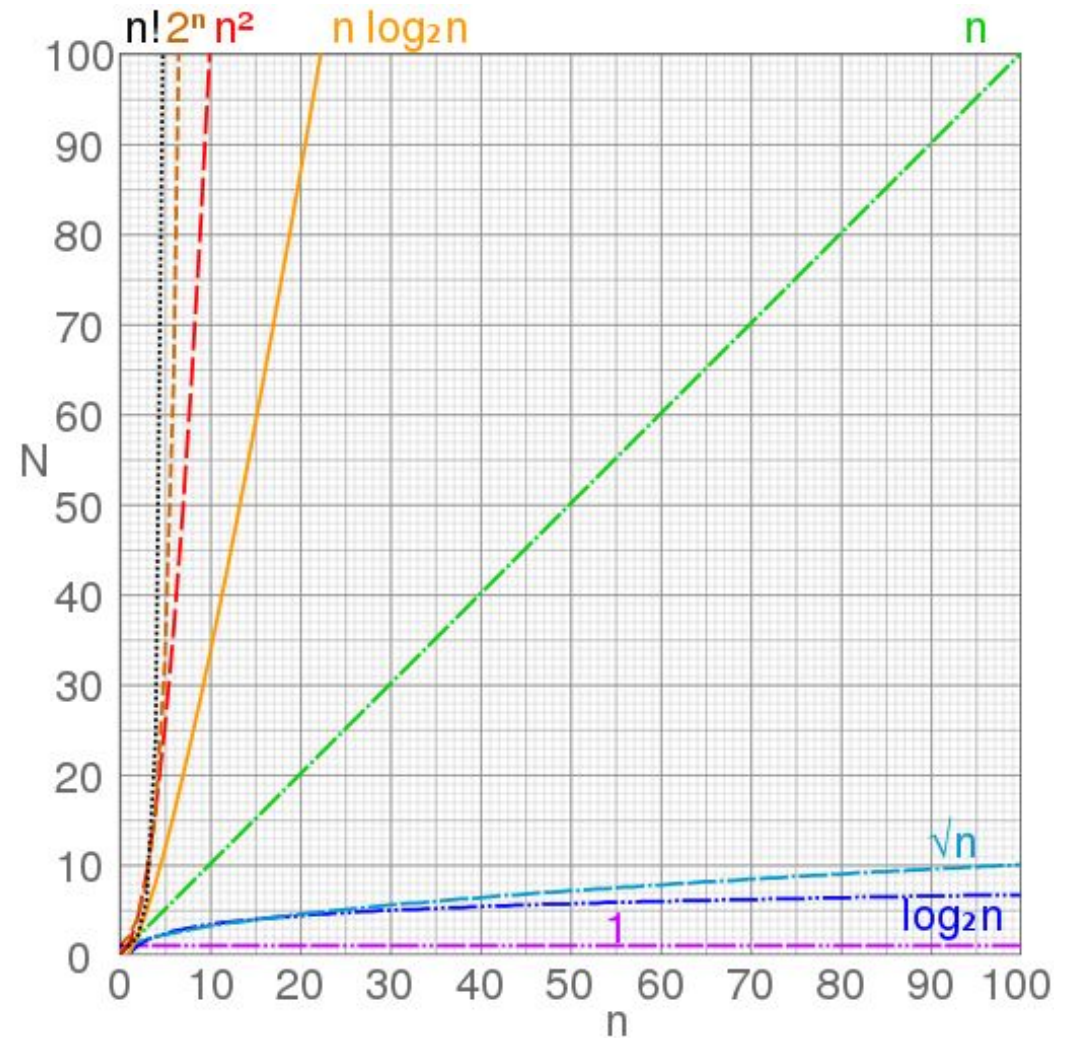


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<https://commons.wikimedia.org/w/index.php?curid=50321072>



Task: Sort

- Task
 - Given: an array (A)
 - Reorder elements in A in non-decreasing order
- In-place algorithms:
 - Selection Sort
 - Bubble Sort
 - Insertion Sort
 - Merge Sort
 - Quick Sort
- See [VisuAlgo](#) for animations
- In Python, we can swap in $O(1)$ time:

```
def swap (A, pos1, pos2):  
    A[pos1], A[pos2] = A[pos2], A[pos1]
```

- ... but it's really 3 operations
- ... and it's



Selection Sort

- Algorithm:
 - Find the highest-valued item and place it in the last position
 - Eliminate last array position and fill second-to-last position with highest-valued item
 - Fill all subsequent positions similarly
- Python implementation

```
def selection_sort (A):  
    for fillslot in range(len(A)-1, 0, -1):  
        max = 0  
        for location in range(1, fillslot+1):  
            if A[location] > A[max]:  
                max = location  
        A[fillslot], A[max] = A[max], A[fillslot]
```

- Running time: $O(n^2)$



Bubble Sort

- Algorithm:
 - Look at each pair of items; swap if in not in order
 - Repeat $n-1$ times
- Python implementation

```
def bubble_sort (A):  
    for j in range(len(A)-1, 0, -1):  
        for i in range(j):  
            if A[i] > A[i+1]:  
                swap(A, i, i+1)
```

- Running time: $O(n^2)$
 - Efficiency: stop sorting when there are no swaps
 - Running time is unchanged



Insertion Sort

- Algorithm:
 - Assume part (first item) of the array is sorted
 - Insert one (the next) item from the unsorted part and shift (if necessary) to maintain sorted order
 - Repeat for all subsequent positions similarly
- Python implementation

```
def insertion_sort (A):  
    for index in range(1, len(A)):  
        currval = A[index]  
        position = index  
        while position>0 and A[position-1]>currval:  
            A[position] = A[position-1]  
            position -= 1  
        A[position] = currval
```

- Running time: $O(n^2)$



Merge Sort — Overview

- Algorithm:
 - Recursively split A in half into subarrays until there are 0 or 1 items (which is sorted)
 - Merge elements in adjacent subarrays by selecting lowest item and placing it in the first position
 - Add remaining elements from left half or right half
- Algorithm is recursive
 - Divide-and-conquer
 - Easy split; difficult merge



Merge Sort — Split

- Algorithm:
 - Divide-and-conquer algorithm
 - Step 1: Recursively split A in half into subarrays until there are 0 or 1 items
- Python implementation

```
def merge_sort (A):  
    if len(A)>1:  
        mid = len(A)//2  
        lefthalf = A[:mid]  
        righthalf = A[mid:]  
  
    merge_sort(lefthalf)  
    merge_sort(righthalf)
```



Merge Sort — Merge

- Algorithm:
 - Step 2: merge
 - Merge elements in adjacent subarrays by selecting lowest item and placing it in the first position
- Python implementation

```
# Array is now split
i=j=k=0
while i < len(lefthalf) and j < len(righthalf):
    if lefthalf[i] < righthalf[j]:
        A[k] = lefthalf[i]
        i += 1
    else:
        A[k] = righthalf[j]
        j += 1
    k += 1
```



Merge Sort — Add Remaining

- Algorithm:
 - Step 3: add remaining elements not already merged in step 2
 - Add remaining elements from left half or right half
- Python implementation

```
# Array is now split and (mostly) merged
while i < len(lefthalf):
    A[k] = lefthalf[i]
    i += 1
    k += 1
while j < len(righthalf):
    A[k] = righthalf[j]
    j += 1
    k += 1
```



Merge Sort — Review

- Algorithm:
 - Recursively split A in half into subarrays until there are 0 or 1 items (which is sorted)
 - Merge elements in adjacent subarrays by selecting lowest item and placing it in the first position
 - Add remaining elements from left half or right half
- Running time: $O(n \log_2 n)$
 - Each step divides array exactly in half and creates a new level of recursion
 - Each level of recursion perform $O(n)$ reads / writes



Quick Sort — Overview

- Algorithm:
 - Pick a pivot value from A
 - Partition: move all elements less than pivot to left part of (sub-)array; all elements greater than pivot to right portion
 - Recursively quick sort left part and right part
 - Difficult split; easy merge
- Python implementation

```
def quick_sort (A):  
    qs(A, 0, len(A)-1)  
  
def qs(A, first, last):  
    if first<last:  
        split = partition(A, first, last)  
        qs(A, first, split-1)  
        qs(A, split+1, last)
```



Quick Sort — Partition

```
def partition (A, first, last):  
    pivot = A[first]  # Other ways to select pivot?  
    left = first+1  
    right = last  
    done = False  
    while not done:  
        while left <= right and A[left] < pivot  
            left += 1  
        while A[right] >= pivot and right >= left:  
            right -= 1  
        if right < left:  
            done = True  
        else:  
            swap(A, left, right)  
    swap(A, first, right)  
    return right
```




Quick Sort — Review

- Algorithm:
 - Difficult split; easy merge
 - Partition array and recursively quick sort the left and right halves
- Running time: $O(n \log_2 n)$
 - Complicated analysis because the pivot does not appear in a stable place in the array
 - Worst case: pivot is consistently on left side or right side of array: n levels of recursion, each $\sim O(n)$ reads/writes $\rightarrow O(n^2)$
 - The worst case may be rare $\sim (1/n^2)$?
 - The average case is $O(n \log n)$



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Object Oriented Programming

- OOP is a framework for building programs / applications / software
- Elements of object oriented languages:
 - **Encapsulation** (variables and functions in one place)
 - **Inheritance** (classes, special classes)
 - **Polymorphism** (a subclass can act as a member of its superclass)
- In Python (also: Java, C++, ...), you can:

Do this	... for example (in Python):
Create an object	<code>my_file = open("my_file.txt", "wb")</code>
Use an object	<code>my_file.write("I want to learn OOP.")</code> <code>my_file.close()</code>
Destroy an object	<code>my_file = None</code>



Example: person

- What makes a person?
 - Name (first names, surname)
 - DOB
 - Biography
 - [...]
- What can a person do?
 - Get married
 - Add to biography
 - Change names
 - Print to a file
 - [...]
- Example fields:
 - Guido van Rossum
 - 1956-01-31
 - Created python
- Example actions:
 - Married Kim Knapp
 - Biography additions: Working for Dropbox (2012)



Python Class: self, __init__

- The class keyword
 - Defines a class
 - Next token (person) names the class
 - All code in class indented
- Encapsulation
 - Fields and actions contained within the class
 - Actions: called “methods” (or “behaviours” or...)
- Keywords ⇒ key concepts
 - self: a particular *instance* of an object; required as first argument to each method
 - __init__: automatically called when the object is instantiated; also defines the class' fields

```
class person:
    '''
    Comments for a person class!
    '''

    def __init__(self, name):
        '''
        First function called.
        That's 2 underscores before
        and 2 underscores after.
        '''
        self.first_names = name['f']
        self.surname = name['last']
        self.biography = []
        self.spouse = None

    def add_to_bio(self, words):
        '''
        A function to add to bio.
        '''
        self.biography.append(words)
```



Declaration of person

```
class person:

    def __init__(self, name):
        self.first_names = name['f']
        self.surname = name['last']
        self.biography = []
        self.spouse = None

    def add_to_bio(self, words):
        self.biography.append(words)

    def change_name(self, name):
        pass

    def change_spouse(self, spouse):
        pass
```

- Code needed to set up example

```
name = dict()
name['f'] = 'Guido'
name['last'] = 'van Rossum'
```

- Need an instance? Assign to a variable from a class' name

```
p = person(name)
```

- Using an instance? Use a "."

```
p.add_to_bio('2012: Dropbox')
```

- Other than the use of "self", code in methods = code in functions



Limitations & Good Practices

- Limitations of python

- All fields (or methods) are accessible:

```
p = person(name)
p.biography = ['No'] # yuck!
```

- Violation of proper encapsulation

- Good practices:

- In each class:

- Create accessors for fields
- Create mutators for fields
- Rely on methods more than on data
- Create tiny functions

- Between classes (collaborations):

- Create specialty classes — eg. for I/O
- ... but be practical

```
class person:
```

```
    def __init__(self, name):
        self.name = dict()
        self.biography = []
        self.spouse = None
        self.change_name(name)
```

```
    def get_name(self, name):
        pass
```

```
    def set_name(self, name):
        pass
```

```
    def to_string(self):
        pass
```



Nomenclature

- Class — (“person”) the blueprint / template
- Object — (“guido” or “kim”) an instance of a class
- Instantiation — the act of creating an object from a class
- “On” — (apologies to English L1) the preposition used for a function of a class. For example, “the get_name function on person...”.
- Attributes
- Constructor



Hashing

- Concept
 - Implementation of dictionary (Dict) container
 - Store values in an array of (static) size; each entry with a unique position (slot)
 - 1: Convert array index to unique value by scrambling — eg. $54 = 5 + 4 \Rightarrow 9$
 - 2: Ensure index of items being stored fit in array — eg. $54 \Rightarrow 9 \% \text{len}(A)$
- Hash function
 - Function to convert values to hash positions
 - Strings can be converted to numbers by ASCII value, etc.
 - Should be quick to calculate, result in the fewest collisions (two or more values hashing to the same location)
 - *Folding*: divide key into groups and add individual portions — eg. $415-422-5101 \Rightarrow [41 + 54 + 22 + 51 + 01] \Rightarrow 169$
 - *Mid-squaring*: square the key and take middle portion — eg. $44^2 \Rightarrow 1936 \Rightarrow 93$
- Resolving collisions
 - Rehashing — eg. $\text{new_hash} = \text{hash}(\text{old_hash_value})$
 - Open addressing — place an element in next available slot (linear, quadratic, etc.)
 - Chaining — Each slot is a structure (list?) of items — example [here](#)



HashTable implementation

Concept — need two classes

- Many instances of hashnode class to store a record (key-value pair)
- One instance of hashtable to manage hashnode instances

```
class hashnode:
```

```
    def __init__(self, key, value):  
        self.key = key  
        self.val = value  
        self.next = None # This is a low-cost linked list
```

```
class hashtable:
```

```
    def __init__(self):  
        self.size = 101  
        self.slots = [None] * self.size
```



HashTable insert and get

```
# Equivalent to: dictionary[key] = value
# This function belongs within the hashtable class
def insert(self, key, value):

    slot = hash(key) % self.size
    node = hashnode(key, value)
    node.next = self.slots[slot]
    self.slots[slot] = node

# Equivalent to: dictionary[key]
# This function also belongs within the hashtable class
def get(self, key):

    slot = hash(key) % self.size
    if self.slots[slot] is not None:
        node = self.slots[slot]
        while node is not None:
            if node.key == key:
                return node.value
            node = node.next
```



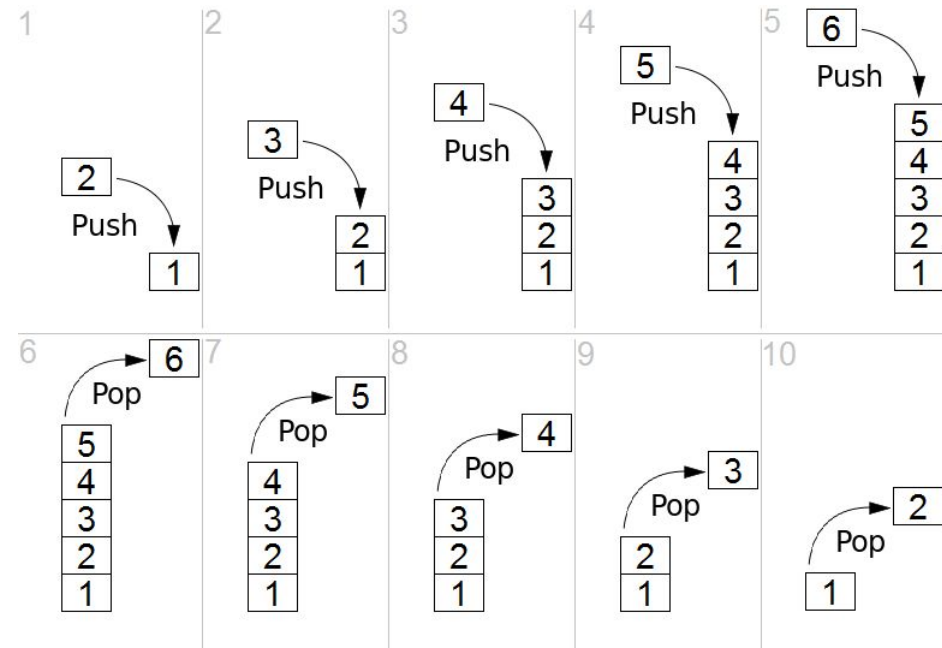
Stacks

- Concept

- LIFO data structure
- Interface:
 - `push(X)` — places an item on (the top of) the Stack
 - `pop()` — removes and returns the item from the Stack
 - `peek()` — returns the item from (the top of) the Stack
 - `empty()` — True if the Stack is empty (a/k/a “`is_empty()`”)

- Implementation

- Want to keep functions to $O(1)$ time
- In Python, common to use list



By Maxtremus - Own work, CC0, <https://commons.wikimedia.org/w/index.php?curid=44458752>

Stack Function	List Function
<code>push(X)</code>	<code>append(X)</code>
<code>pop()</code>	<code>pop()</code>
<code>peek()</code>	# Write a function
<code>empty()</code>	# Use len function



Queues

- Concept

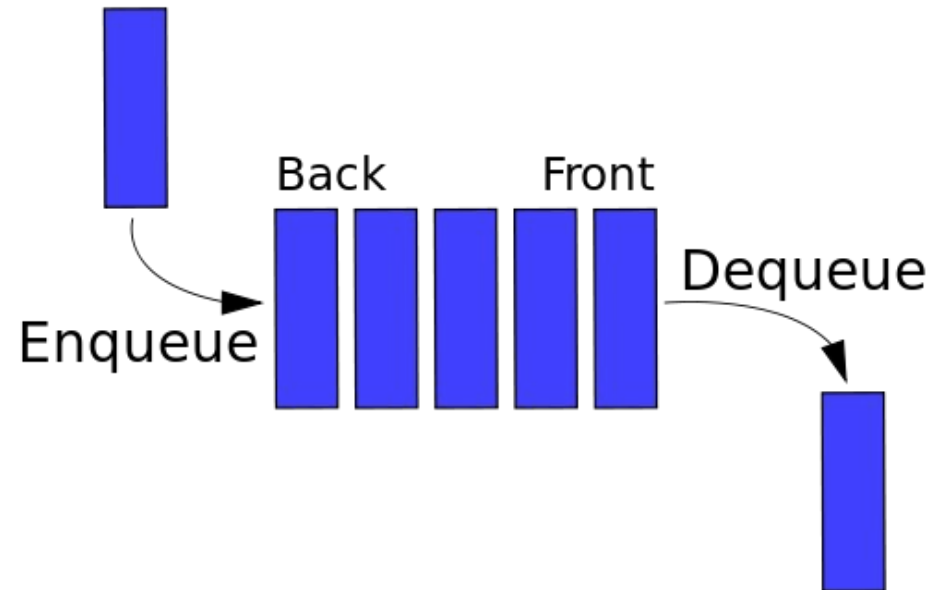
- FIFO data structure
- Interface:
 - `enqueue(X)` — places an item on (the end of) the Queue
 - `dequeue()` — removes and returns item from the Queue
 - `size()` — returns the number of items on the Queue
- Variation: dequeue allows insertion & removal from both front & back

- Implementation

- Common to use a dequeue (double-ended queue), constructor:

```
queue = dequeue()
```

- May also use a list



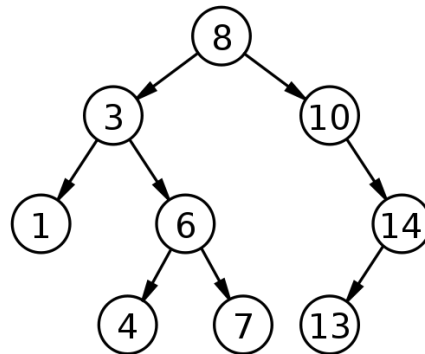
Queue Function	dequeue Function
<code>enqueue(X)</code>	<code>append(X)</code>
<code>dequeue()</code>	<code>popleft()</code>
<code>size()</code>	# Use <code>len()</code> function



Binary Search Trees (BST)

- Concept

- Dynamic data structure with principles from binary search
- Entire structure needs reference to one element of the tree (root)
- Interface:
 - `find(key)` — recursive find key in BST; return associated value
 - `insert(key, val)` — recursive find to add {key, val} to BST
 - `delete(key)` — delete node with key from BST
 - `traverse()` — list all items in BST; 3 possible traversals
- Visualisations usually only show keys, no associated values:



- Implementation uses two classes:

- `bst` — can operate on a tree even if it's empty
- `node` — does most of the work



BST — node implementation

```
class node:

    def __init__(self, key, val):
        self.key = key
        self.val = val
        self.left = None
        self.right = None

    def get(self):
        return self.val

    def set(self, val):
        self.val = val
```



BST — public implementation

```
class bst:

    def __init__(self):
        self.root = None

    def insert(self, key, val):
        if self.root is None:
            self.root = node(key, val)
        else:
            self.insert_node(self.root, key, val)

    def find(self, key):
        return self.find_node(self.root, key)

    def traverse_in_order(self):
        return self.in_order(self.root, key)
```




BST — find_node

Concept:

- Compare current node's key to target
- Recursively look left or right depending on value compared to target

```
def find_node(self, current_node, key):  
    if current_node is not None:  
        return False # Maybe return some other value  
    elif key == current_node.key:  
        return current_node.val  
    elif key < current_node.key:  
        return self.find_node(current_node.left, key)  
    else:  
        return self.find_node(current_node.right, key)
```



BST — insert_node

Concept:

- Descend the tree (going left or right) until getting to an empty node
- Create new node and attach it to the tree

```
def insert_node(self, current_node, key, val):  
    if key <= current_node.key:  
        if current_node.left is not None:  
            self.insert_node(current_node.left, key, val)  
        else:  
            current_node.left = node(key, val)  
    elif key > current_node.key:  
        if current_node.right is not None:  
            self.insert_node(current_node.right, key, val)  
        else:  
            current_node.right = node(key, val)
```



BST — traverse

Concept: process (print) everything to the left, the current, everything right

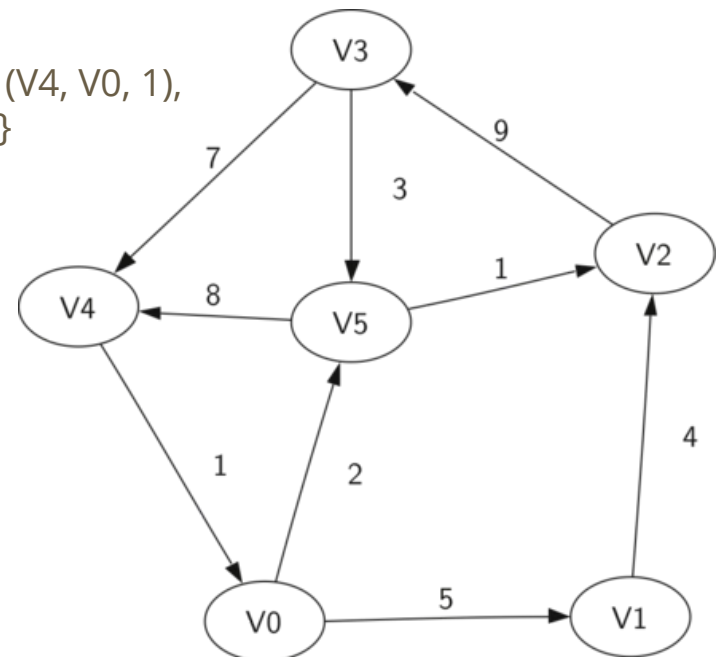
```
# Assuming the traversal action is printing
def in_order(self, current_node):
    if current_node is None:
        return # Nothing to do
    self.in_order(current_node.left)
    print(current_node.val)
    self.in_order(current_node.right)
```



Graphs — Concept and Storage

- Vocabulary: $G = (V, E)$
 - *Vertices* / nodes; may have a key / name + additional information
 - *Edges* connect vertices; may be weighted or unweighted; directed or bidirectional
 - *Path* — a sequence of vertices connected by edges (in the correct direction)
 - *Cycle* — a path which begins and ends with the same vertex
- Example
 - $V = \{V0, V1, V2, V3, V4, V5\}$
 - $E = \{ (V0, V1, 5), (V1, V2, 4), (V2, V3, 9), (V3, V4, 7), (V4, V0, 1), (V0, V5, 2), (V5, V4, 8), (V3, V5, 3), (V5, V2, 1) \}$
- May be stored as a matrix (below) or list

	V0	V1	V2	V3	V4	V5
V0		5				2
V1			4			
V2				9		
V3					7	3
V4	1					
V5			1		8	





Graphs — One Implementation

```
class vertex:
```

```
    def __init__(self, key):
        self.id = key
        self.connected_to = {}
```

```
    def add_neighbor(self, nbr, weight=0):
        self.connected_to[nbr] = weight
```

```
class graph:
```

```
    def __init__(self):
        self.vertices = {}
```

```
    def add_vertex(self, key):
        v = vertex(key)
        self.vertices[key] = v
```

```
    def get_vertex(self, key):
        if key in vertices:
            return vertices[key]
        return None
```

```
    def add_edge(self, src, tar, weight):
        if src not in self.vertices:
            self.add_vertex(src)
        if tar not in self.vertices:
            self.add_vertex(tar)
        sv = self.vertices[src]
        sv.add_neighbour([tar], weight)
```



Task: Find Shortest Path

- Task
 - Given: two vertices, (V_S and V_T) and $G = (V, E)$
 - Find and return a (the shortest) path starting at V_S and ending at V_T
- Algorithm: Breadth First Search
 - Visit adjacent vertices recursively, starting from V_S until V_T is found
 - Keep a queue of candidate nodes to visit, set of nodes already visited
 - Some implementations require additional methods to Vertex class
 - Running time = $O(V + E)$
 - Ignoring V_T , the same algorithm can create a (minimum) spanning tree from V_S



BFS in Python

```
# Implemented within graph class
def bfs(self, start):
    queue = [(start, [start])] # Keep track of path back to start
    while queue:
        vertex, path = queue.pop(0)
        for next in self.get_vertex(vertex) - set(path):
            if next == target:
                yield path + [next]
            else:
                queue.append((next, path + [next]))
```



Resources

Time complexity — <https://www.youtube.com/watch?v=8syQKTdgdzc>

- See [Problem Solving with Algorithms and Data Structures using Python](#)