



# GRAPHS

School of Artificial Intelligence



# PREVIOUSLY ON DS&A

- Data structures
  - Array-based
  - Stacks and queues
  - Linked Lists
  - Trees
  - Priority Queues
  - Maps and Hash tables
  - Search Trees
- Algorithms
  - Sorting and Selection



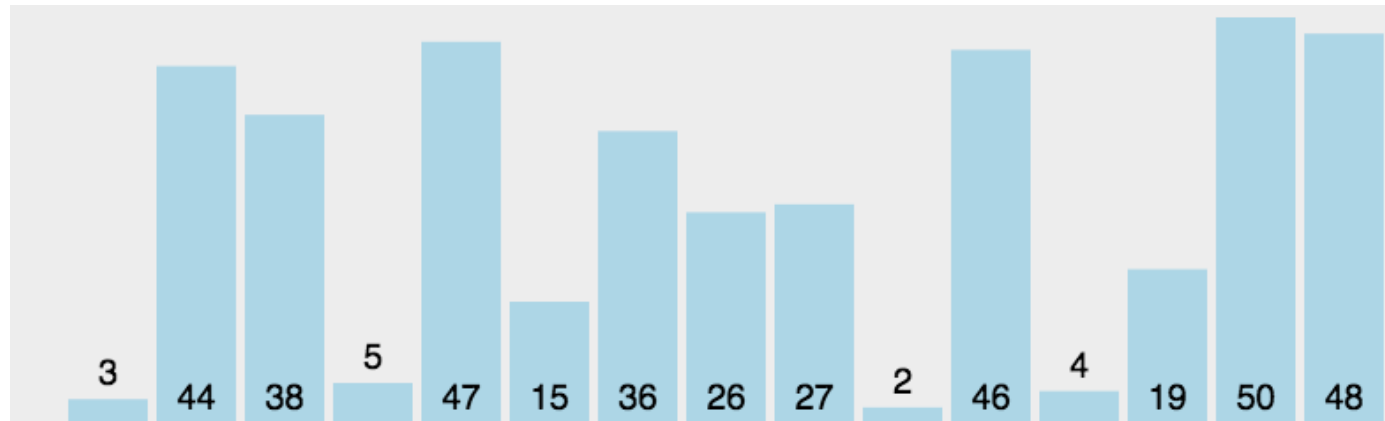
# INTERVIEW QUESTIONS

- From Facebook
- Given an array of  $n$  elements, ranging from 0 to  $N$ , determine if the array contains at least 1 element from 0 to  $N$ . What would be the space efficiency and the time efficiency?
- You are given a collection of (city, population) entries, provide an algorithm, that randomly selects a pair, so that its population is proportional to the probability that the entry is selected.

# INSERTION SORT

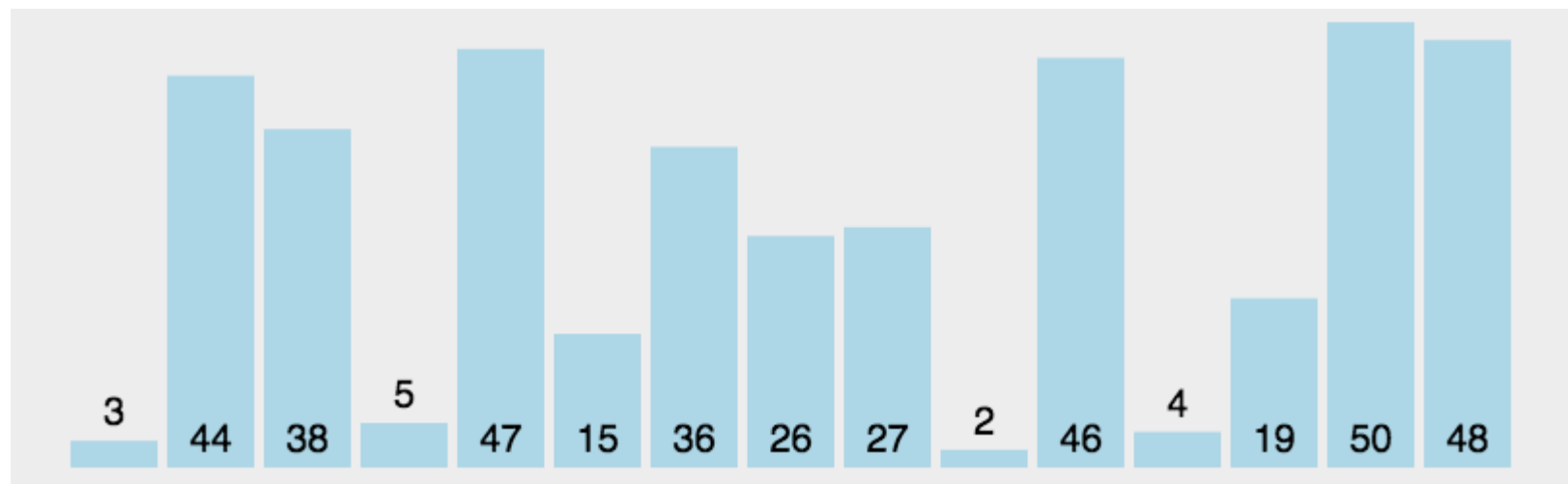
- Insertion sort

```
1 def insertion_sort(A):
2     """ Sort list of comparable elements
3     for k in range(1, len(A)):
4         cur = A[k]
5         j = k
6         while j > 0 and A[j-1] > cur:
7             A[j] = A[j-1]
8             j -= 1
9         A[j] = cur
```



# SELECTION SORT

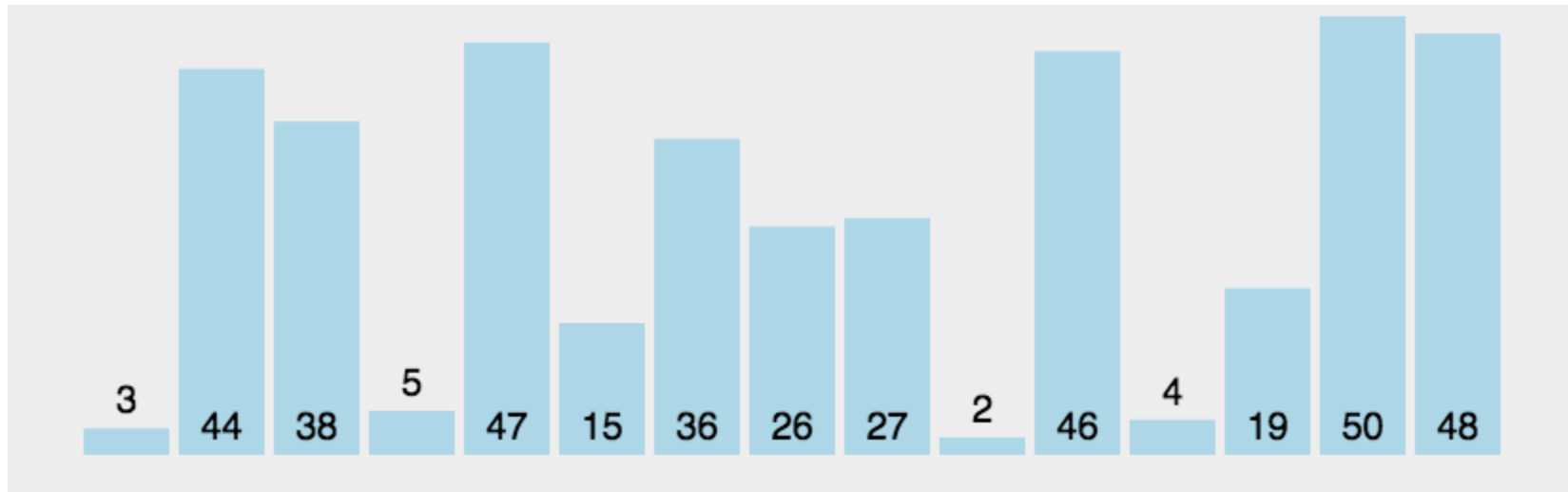
- Selection sort
- We have seen this
  - In Priority Queue



```
def selectionSort(arr):  
    for i in range(len(arr) - 1):  
        # 记录最小数的索引  
        minIndex = i  
        for j in range(i + 1, len(arr)):  
            if arr[j] < arr[minIndex]:  
                minIndex = j  
        # i 不是最小数时，将 i 和最小数进行交换  
        if i != minIndex:  
            arr[i], arr[minIndex] = arr[minIndex], arr[i]  
    return arr
```

# BUBBLE SORT

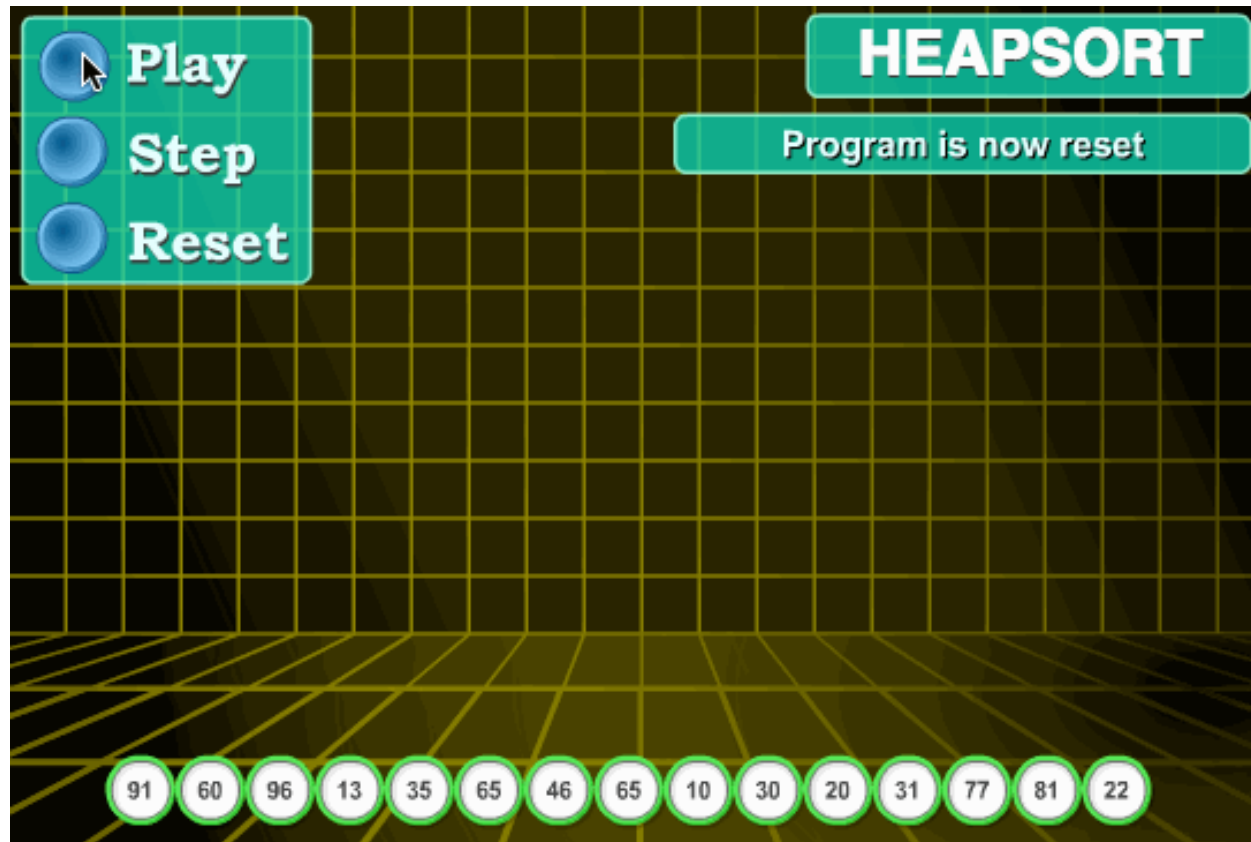
- Bubble sort



```
def bubbleSort(arr):  
    for i in range(1, len(arr)):  
        for j in range(0, len(arr)-i):  
            if arr[j] > arr[j+1]:  
                arr[j], arr[j + 1] = arr[j + 1], arr[j]  
    return arr
```

# HEAP SORT

- Heap sort





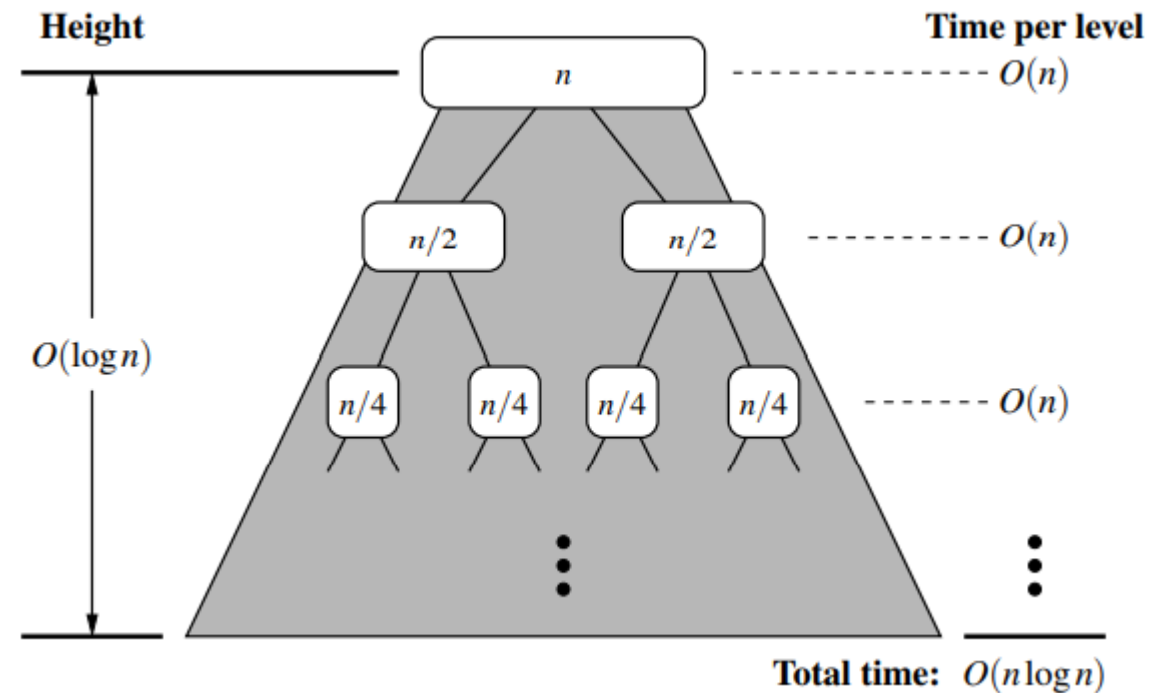
# MERGE SORT

- Divide and conquer for sorting
  - **Divide**: if  $S$  has zero or one element, return  $S$ . Otherwise, remove all the elements from  $S$  and put them into two sequences,  $S1$  and  $S2$ , each containing half of the elements of  $S$
  - **Conquer**: recursively sort sequence  $S1$  and  $S2$
  - **Combine**: put back the elements into  $S$  by merging the sorted sequences  $S1$  and  $S2$  into a sorted sequence



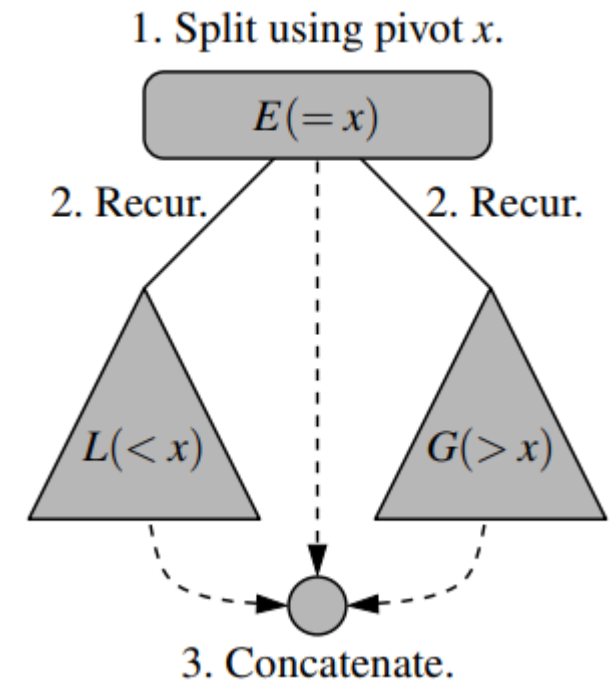
# MERGE SORT

- `merge_sort(A[1..n])`
  1. If  $n = 1$ , done
  2. recursively sort  $A[1..n//2]$ ,  $A[n//2+1..n]$
  3. Merge the two sorted list



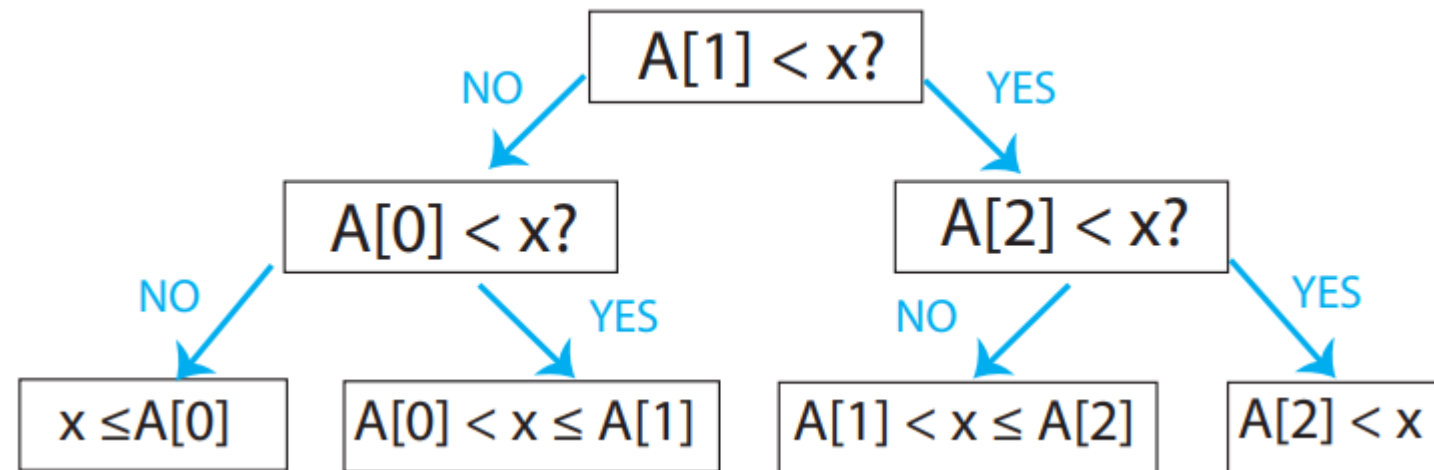
# QUICK SORT

- **Divide:** if  $S$  has at least two elements, select a specific element  $x$  from  $S$ , which is called the pivot(枢纽). The common practice is to choose the last element of  $S$ . Remove all the elements from  $S$  and put them into three sequences:
  - $L$ , storing elements in  $S$  less than  $x$
  - $E$ , storing elements in  $S$  equal to  $x$
  - $G$ , storing elements in  $S$  greater than  $x$
- **Conquer:** recursively sort sequences  $L$  and  $G$
- **Combine:** Put back the elements into  $S$  in order by first inserting the elements of  $L$ , then  $E$ , then  $G$



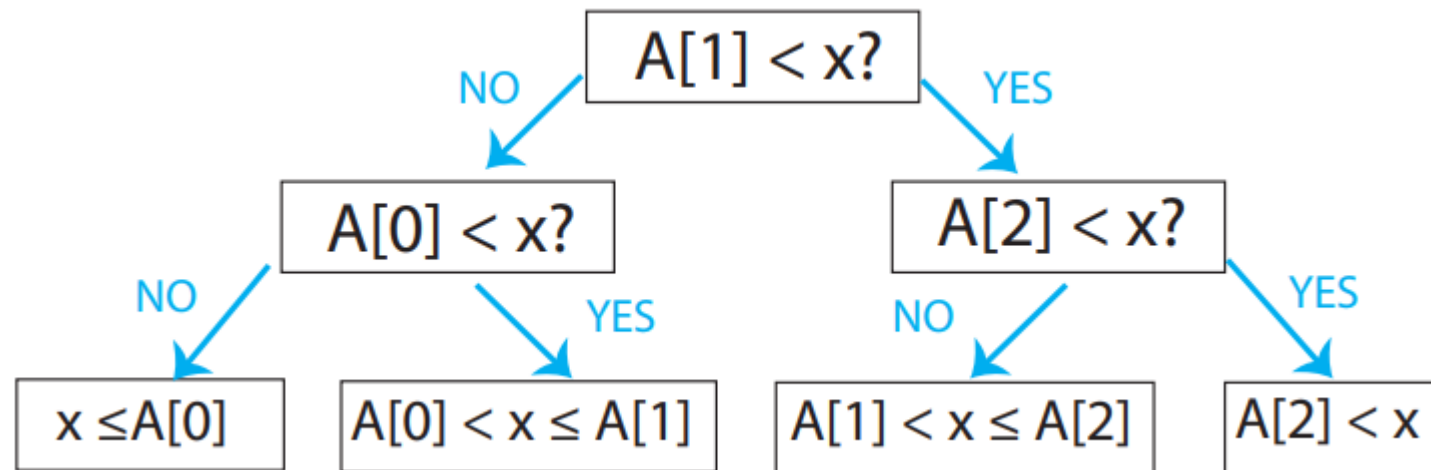
# SEARCH LOWER BOUND

- N preprocessed items
- Finding a given item among them in comparison model requires  $\Omega(\log n)$  in worst case
  - Why?
- Decision tree is binary, and therefore must have at least n leaves for each answer
  - Height  $\geq \log n$



# SORTING LOWER BOUND

- N preprocessed items
- Finding a given item among them in comparison model requires  $\Omega(n \log n)$  in worst case
  - Why?
- Decision tree is binary
- # leaves  $\geq$  # possible answers
- Possible answers?
  - $n!$



# LINEAR TIME SORTING

- Counting sort

L = array of k empty lists

For j in range n:

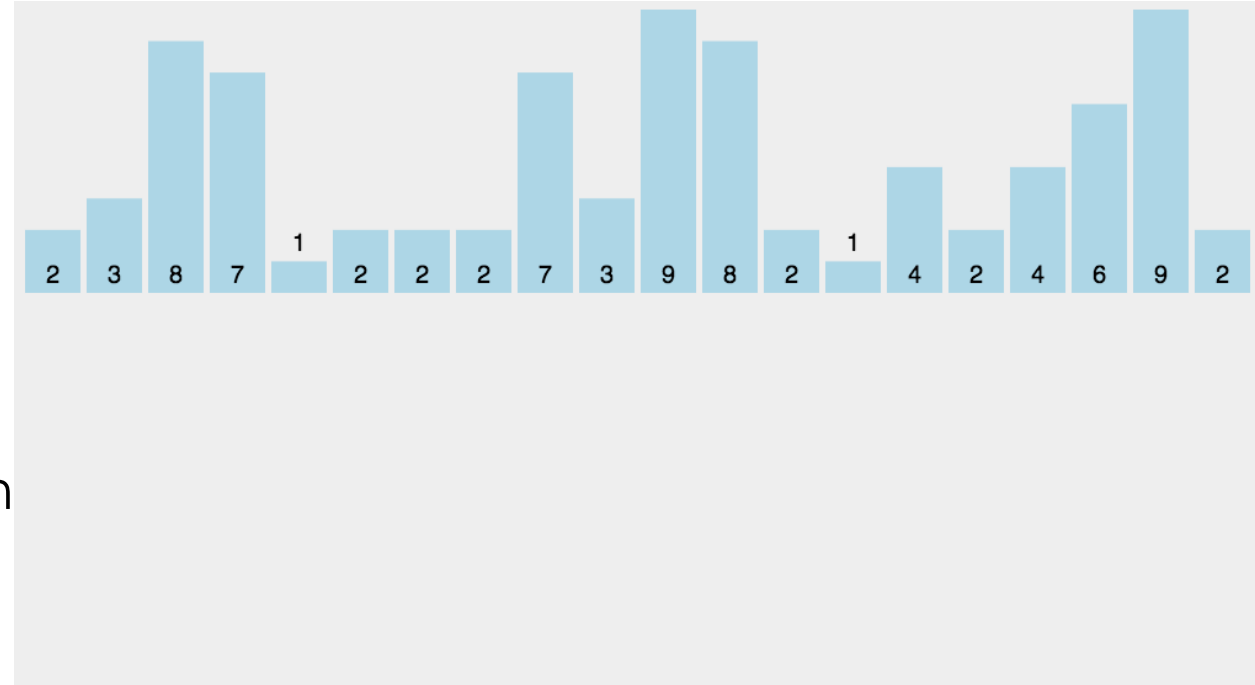
    L[key(A[j])].append(A[j])

Output = []

For i in range k:

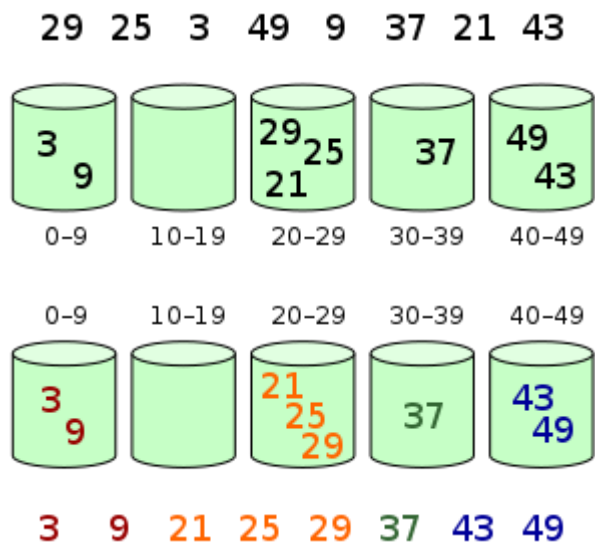
    Output.extend(L[i])

- Time:  $O(n + k)$
- Count key occurrences using RAM  
output <count> copies of each key in order
- But item is more than just a key



# LINEAR TIME SORTING

- Bucket sort
  - Sequence  $S$  of  $n$  entries
  - Keys are integers in  $[0, N-1]$ ,  $N \geq 2$
  - It is possible to sort  $S$  in  $O(n+N)$
  - $O(n)$  if  $N$  is  $O(n)$



**Algorithm** bucketSort( $S$ ):

*Input:* Sequence  $S$  of entries with integer keys in the range  $[0, N - 1]$

*Output:* Sequence  $S$  sorted in nondecreasing order of the keys

let  $B$  be an array of  $N$  sequences, each of which is initially empty

**for** each entry  $e$  in  $S$  **do**

$k$  = the key of  $e$

    remove  $e$  from  $S$  and insert it at the end of bucket (sequence)  $B[k]$

**for**  $i = 0$  to  $N-1$  **do**

**for** each entry  $e$  in sequence  $B[i]$  **do**

        remove  $e$  from  $B[i]$  and insert it at the end of  $S$

# LINEAR TIME SORTING

- Radix sort
  - Imagine each integer as base  $b$
  - # digits =  $d = \log_b k + 1$
  - Sort integers by least significant digit
  - ...
  - Sort by most significant digit
- Sort using counting sort
  - Each iteration:  $O(n + b)$
  - Total:  $O((n+b)*d)$
  - =  $O((n+b)*\log_b k)$
  - When  $b = \Theta(n)$
  - $O(n*\log_n k)$
  - If  $k = n^c$  then  $O(nc)$

3	44	38	5	47	15	36	26	27	2	46	4	19	50	48
---	----	----	---	----	----	----	----	----	---	----	---	----	----	----



# SORTING ALGORITHMS

- $O(n^2)$  algorithms
  - Selection sort, insertion sort
- $O(n \log n)$  algorithms
  - Heap sort, merge sort, quick sort
- Linear time algorithms
  - Counting sort, bucket sort, radix sort
- Poorest choice of algorithms?
  - Selection sort
  - Best case  $O(n^2)$



# SORTING ALGORITHMS

- Insertion sort
  - $O(n + m)$ , where  $m$  is number of inversions
  - Good when  $n$  is small
  - Good when  $m$  is small
  - $O(n^2)$  when  $m$  is large and  $n$  is large
- Heap sort
  - $O(n \log n)$  time worst case
  - In-place easy to implement
  - Slower than quick sort and merge sort on larger sequences

# SORTING ALGORITHMS

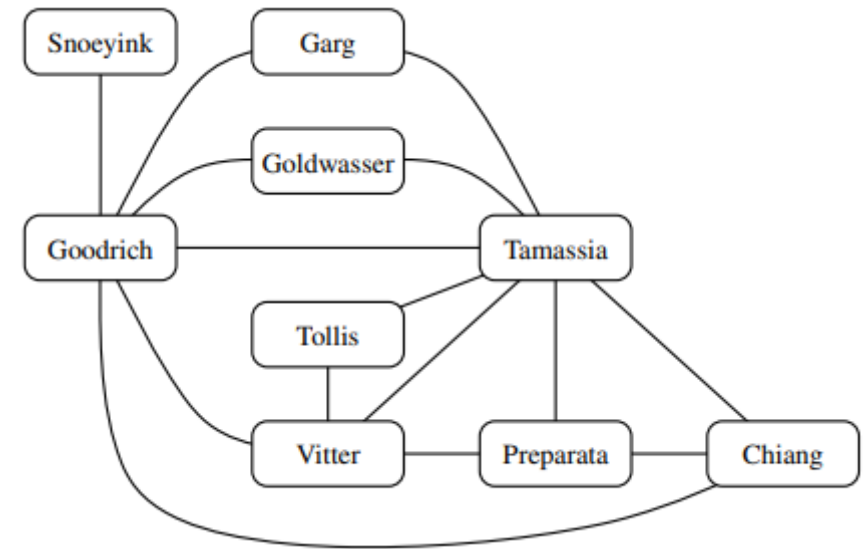
- Quick sort
  - Worst case  $O(n^2)$
  - $O(n \log n)$  expected
  - Faster than heap sort and merge sort on many tests
  - Can be implemented in-place
- Merge sort
  - $O(n \log n)$  worst case
  - In-place implementation: difficult
  - Hence slower than quick sort
- Bucket sort and radix sort
  - Limitations: small integer keys, character strings, d-tuples from a discrete range
  - $O(d(n+N))$  time, in range  $[0, N-1]$

# THIS LECTURE

- Graphs: relationships between pairs of objects: vertices together with a collection of connections between them (edges).
  - Mapping, transportation, computer networks electrical engineering
- Definition:
  - Set  $V$  of **vertices** and a collection  $E$  of pairs of vertices from  $V$ , called **edges**
- Edges
  - Directed: an edge  $(u, v)$  is directed from  $u$  to  $v$  if the pair  $(u, v)$  is ordered
  - Undirected: an edge  $(u, v)$  is not ordered, sometimes denoted as  $\{u, v\}$
- Undirected graph: the edges are all undirected
- Directed graph (digraph): edges are all directed
- Mixed graph: directed edges + undirected edges
- Undirected graph  $\Rightarrow$  directed graph

# GRAPHS

- Example
  - Collaboration graph (undirected)
  - Class relationships within an Object-Oriented program (directed)
  - City map (mixed graph)
  - Electric wiring/Schematics
    - Directed/undirected

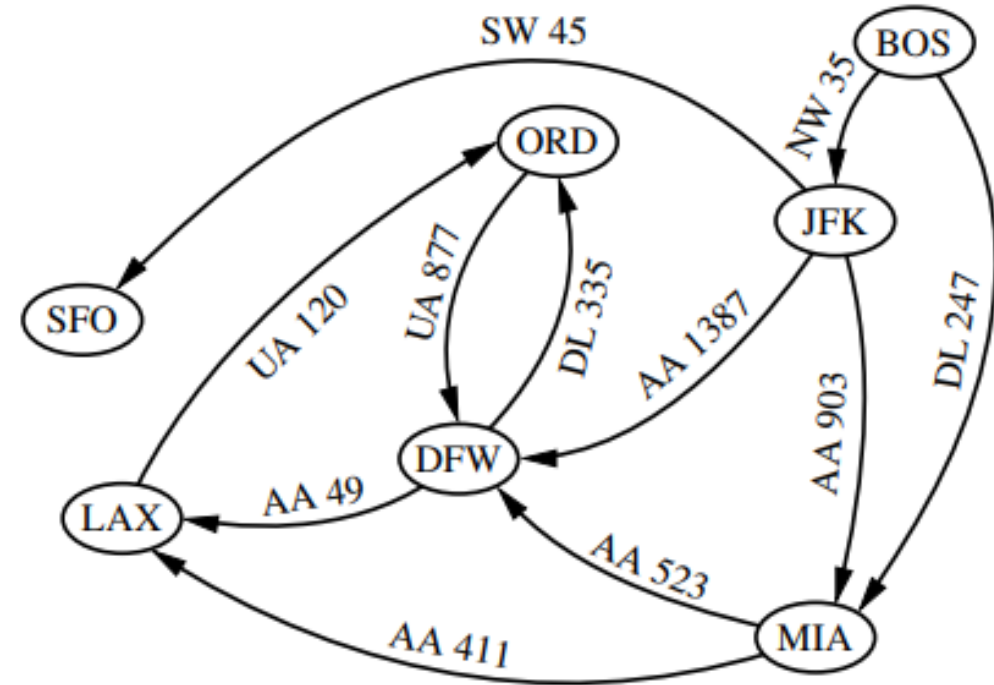


# GRAPHS

- End vertices (endpoints)
  - Two vertices joined by an edge
  - Directed edge: origin  $\rightarrow$  destination
- Adjacent vertices:  $u$  and  $v$  are adjacent if there is an edge whose end vertices are  $u$  and  $v$
- Incident: an edge is incident to a vertex if the vertex is one of the edge's endpoints
- Outgoing edges of a vertex: directed edges whose origin is the vertex
- Incoming edges of a vertex: directed edges whose destination is the vertex
- Degree of a vertex: # of incident edges of  $v$
- In-degree: # of incoming edges of  $v$
- Out-degree: # of outgoing edges of  $v$

# GRAPHS

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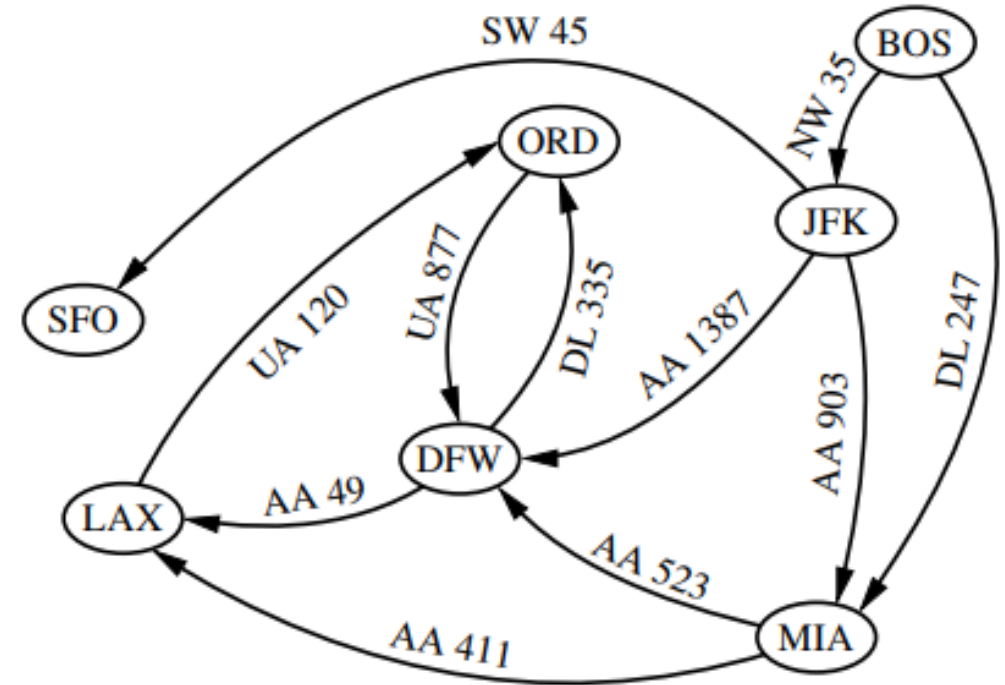


# GRAPHS

- Definition:
  - Set  $V$  of **vertices** and a collection  $E$  of pairs of vertices from  $V$ , called **edges**
- $E$ : collection, not a set
  - Two edges can have the same end vertices
  - Such edges are called **parallel** or **multiple** edges
  - **Self-loop** edges: endpoints are the same
- **Simple** graphs: graphs do not have parallel edges or self loops
- **Path**: sequence of vertices and edges, start at a vertex and ends at a vertex
  - Simple: if each vertex in the path is distinct
- **Cycle**: path that starts and ends at the same vertex, and includes at least one edge
  - Simple: if each vertex in the cycle is distinct(except for the first and last one)
- **Acyclic**: directed graph that has no directed cycles

# GRAPHS

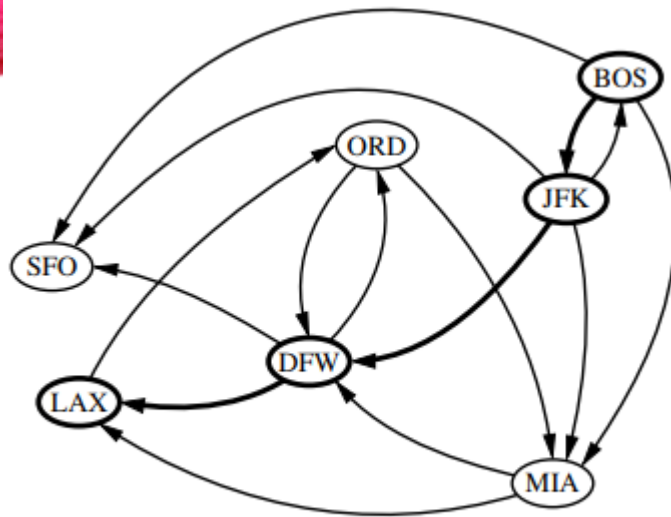
- Directed path
  - A path such that all edges are directed and are traversed along their direction
  - (BOS, NW35, JFK, AA1387, DFW)
- Directed cycle
  - A cycle such that all edges are directed and are traversed along their direction
  - (LAX, UA1200, ORD, UA877, DFW, AA49, LAX)



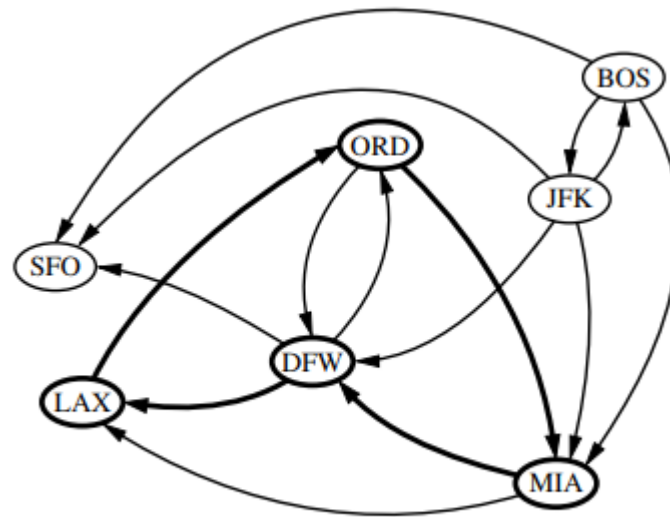
# GRAPHS

- Reachability
  - In a graph  $G$ ,  $u$  **reaches**  $v$ , and  $v$  is reachable from  $u$ , if  $G$  has a path from  $u$  to  $v$
- Connectivity
  - $G$  is **connected**, if for any two vertices, there is a path between them
  - Directed graph: **strongly connected**
- Subgraph
  - Graph  $H$  whose vertices and edges are subsets of the vertices and edges of  $G$
- Spanning subgraph
  - Subgraph of  $G$  that contains all the vertices of the graph
- Forest
  - Graph without cycles
- Tree
  - Connected forest
- Spanning tree
  - Spanning subgraph that is a tree

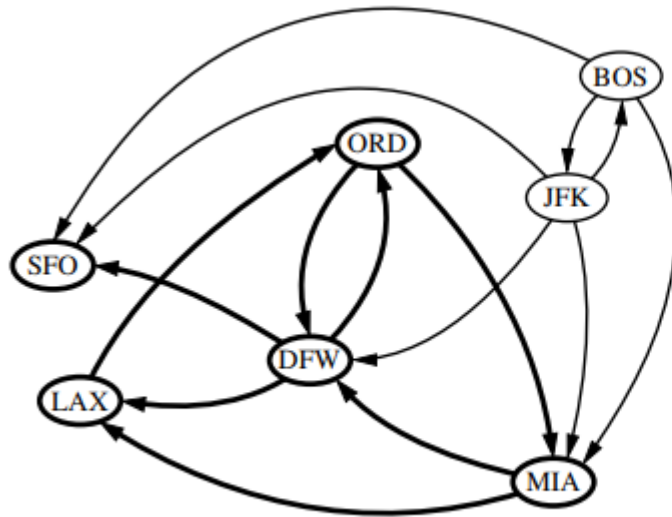
# GRAPHS



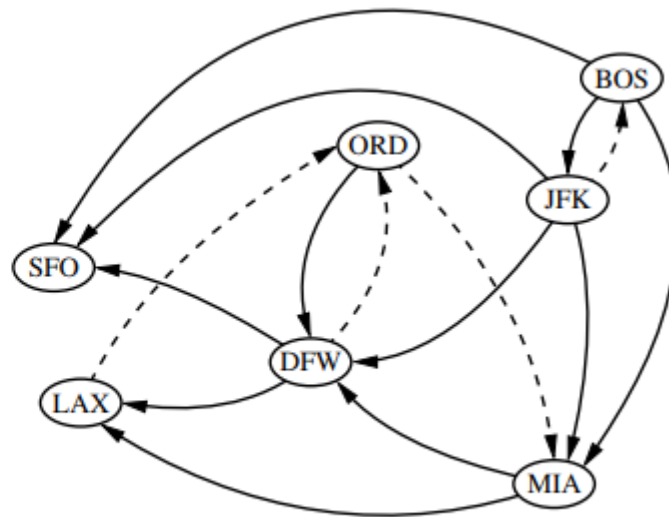
(a)



(b)



(c)



(d)

# GRAPHS

- If  $G$  is a graph with  $m$  edges and vertex set  $V$ , then
- If  $G$  is a directed graph with  $m$  edges and vertex set  $V$ , then
- Let  $G$  be a simple graph with  $n$  vertices and  $m$  edges.
  - If  $G$  is undirected, then  $m \leq n(n-1)/2$
  - If  $G$  is directed, then  $m \leq n(n-1)$
- Let  $G$  be a undirected graph with  $n$  vertices and  $m$  edges
  - If  $G$  is connected, then  $m \geq n-1$
  - If  $G$  is a tree, then  $m = n-1$
  - If  $G$  is a forest then  $m \leq n-1$

$$\sum_{v \in V} \deg(v) = 2m.$$

$$\sum_{v \in V} \text{indeg}(v) = \sum_{v \in V} \text{outdeg}(v) = m.$$

# GRAPHS

- Graph ADT
- Endpoints: return a tuple  $(u, v)$  such that vertex  $u$  is the origin and  $v$  the destination
- Opposite( $v$ ): if  $v$  is one endpoint of the edge, return the other endpoint
- Vertex\_count(): returns # of vertices
- Vertices(): returns an iteration of all vertices of the graph
- Edge\_count(): returns # of edges
- Edges(): returns an iteration of all edges
- Get\_edge( $u, v$ ): returns the edge from  $u$  to  $v$ , if one exists, otherwise return None

# GRAPHS

- Graph ADT
- `degree(v, out= True)`: returns the # of edges incident to v
- `incident_edges(v, out=True)`: returns an iteration of edges incident to v
- `insert_vertex(x = None)`: create and return a new Vertex storing element x
- `insert_edge(u, v, x = None)`: create and return a new Vertex from u to v
- `remove_vertex(v)`: remove v and all its incident edges from the graph
- `remove_edge(e)`: remove e from the graph



# DATA STRUCTURES FOR GRAPHS

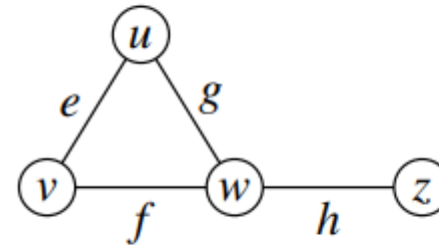
- Edge list
  - An unordered list of all edges, no efficient way to locate a particular edge  $(u, v)$ , or the set of all edges incident to a vertex  $v$
- Adjacency list
  - For each vertex, a list containing edges that are incident to the vertex
  - Complete set of edges can be determined by taking the union of the smaller sets
- Adjacency map
  - Similar to adjacency list, but secondary container for edges are maps
  - $O(1)$  expected time to access a specific edge  $(u, v)$
- Adjacency matrix
  - Worst case  $O(1)$  access to a specific edge  $(u, v)$  by maintaining a  $n \times n$  matrix for each graph with  $n$  vertices

# DATA STRUCTURES FOR GRAPHS

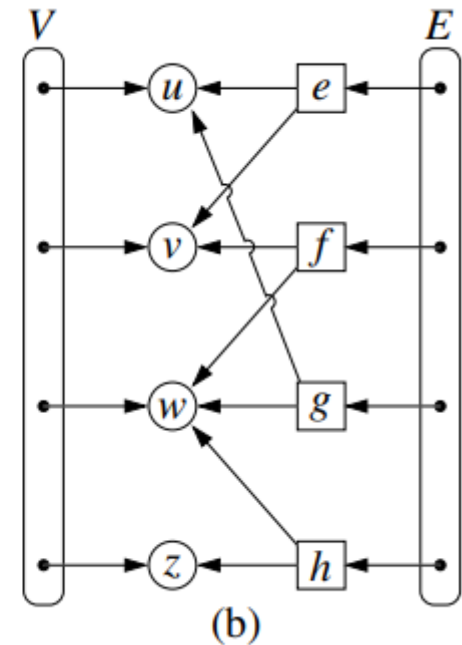
Operation	Edge List	Adj. List	Adj. Map	Adj. Matrix
vertex_count()	$O(1)$	$O(1)$	$O(1)$	$O(1)$
edge_count()	$O(1)$	$O(1)$	$O(1)$	$O(1)$
vertices()	$O(n)$	$O(n)$	$O(n)$	$O(n)$
edges()	$O(m)$	$O(m)$	$O(m)$	$O(m)$
get_edge(u,v)	$O(m)$	$O(\min(d_u, d_v))$	$O(1)$ exp.	$O(1)$
degree(v)	$O(m)$	$O(1)$	$O(1)$	$O(n)$
incident_edges(v)	$O(m)$	$O(d_v)$	$O(d_v)$	$O(n)$
insert_vertex(x)	$O(1)$	$O(1)$	$O(1)$	$O(n^2)$
remove_vertex(v)	$O(m)$	$O(d_v)$	$O(d_v)$	$O(n^2)$
insert_edge(u,v,x)	$O(1)$	$O(1)$	$O(1)$ exp.	$O(1)$
remove_edge(e)	$O(1)$	$O(1)$	$O(1)$ exp.	$O(1)$

# EDGE LIST

- Collections  $V$  and  $E$  are represented with doubly linked lists (PositionalList)
- Vertex objects
  - Reference to element  $x$
  - Reference to position of the vertex instance in the list  $V$
- Edge objects
  - Reference to element  $x$
  - Reference to the vertex objects associated with the endpoint vertices of  $e$
  - Reference to the position of the edge instance in list  $E$



(a)



(b)

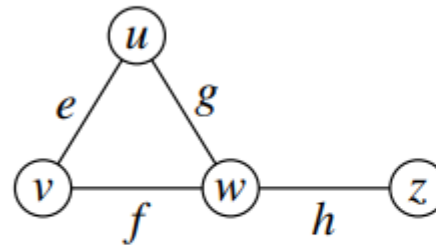
# EDGE LIST

- Performance
- Space:  $O(n + m)$ 
  - $n$  vertices and  $m$  edges
- Running time
  - `vertices()`:  $O(n)$
  - `edges()`:  $O(m)$
  - `get_edge()`:  $O(m)$ 
    - Most significant limitation
  - `remove_vertex(v)`:  $O(m)$ 
    - Why?

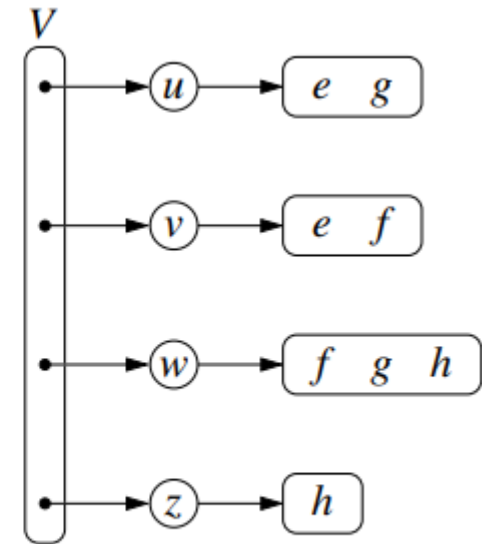
Operation	Running Time
<code>vertex_count()</code> , <code>edge_count()</code>	$O(1)$
<code>vertices()</code>	$O(n)$
<code>edges()</code>	$O(m)$
<code>get_edge(u,v)</code> , <code>degree(v)</code> , <code>incident_edges(v)</code>	$O(m)$
<code>insert_vertex(x)</code> , <code>insert_edge(u,v,x)</code> , <code>remove_edge(e)</code>	$O(1)$
<code>remove_vertex(v)</code>	$O(m)$

# ADJACENCY LIST

- Secondary containers for edges that are associated with each individual vertex
- For each  $v$ , maintain a collection  $I(v)$  called **incidence collection** of  $v$
- Primary structure: collection  $V$  of vertices
  - Positional list
- Each vertex instance
  - Direct reference to its  $I(v)$  incidence collection



(a)



(b)

# ADJACENCY LIST

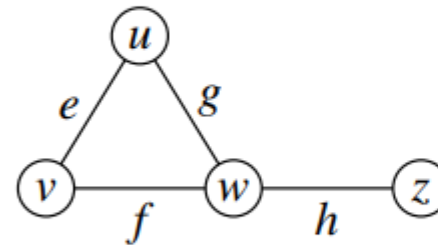
- Performance
- Space:  $O(n + m)$ 
  - $n$  vertices and  $m$  edges
- Running time
  - `vertices()`:  $O(n)$
  - `edges()`:  $O(m)$
  - `get_edge()`:  $O(\min(\deg(u), \deg(v)))$ 
    - Search through either  $I(u)$  or  $I(v)$
  - `remove_vertex(v)`:  $O(\deg(v))$

Operation	Running Time
<code>vertex_count()</code> , <code>edge_count()</code>	$O(1)$
<code>vertices()</code>	$O(n)$
<code>edges()</code>	$O(m)$
<code>get_edge(u,v)</code>	$O(\min(\deg(u), \deg(v)))$
<code>degree(v)</code>	$O(1)$
<code>incident_edges(v)</code>	$O(\deg(v))$
<code>insert_vertex(x)</code> , <code>insert_edge(u,v,x)</code>	$O(1)$
<code>remove_edge(e)</code>	$O(1)$
<code>remove_vertex(v)</code>	$O(\deg(v))$

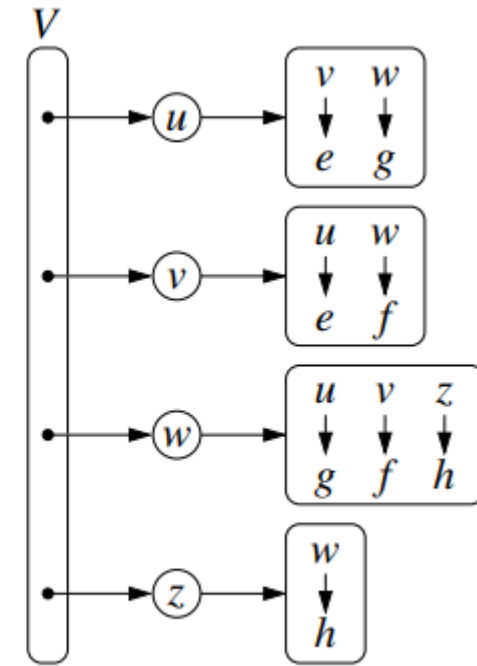


# ADJACENCY MAP

- Adjacency list
  - $I(v)$  uses  $O(\deg(v))$  space
  - $O(\deg(v))$  time
- Performance improvement
  - Hash-based map for  $I(v)$
  - $\text{get\_edge}(u,v)$  can run in expected  $O(1)$  time, worst case  $O(\min(\deg(u), \deg(v)))$



(a)

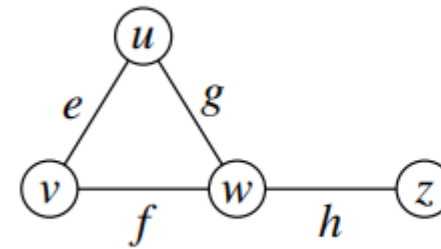


(b)



# ADJACENCY MATRIX

- Matrix  $A$  ( $n$  by  $n$ ) to locate an edge between a given pair of vertices in worst-case  $O(1)$  time
- Vertices as integers in  $\{0, 1, \dots, n-1\}$
- $A[i, j]$  holds a reference to the edge  $(u, v)$  if one exists
- Edge  $(u, v)$  can be accessed in worst-case  $O(1)$  time
- $O(n^2)$  space usage
- Matrix can be used to store only Boolean values, if edges do not store any additional data



(a)

		0	1	2	3
$u \rightarrow$	0		$e$	$g$	
$v \rightarrow$	1	$e$		$f$	
$w \rightarrow$	2	$g$	$f$		$h$
$z \rightarrow$	3			$h$	

(b)

F

```

1  #----- nested Vertex class -----
2  class Vertex:
3      """ Lightweight vertex structure for a graph. """
4      __slots__ = '_element'
5
6      def __init__(self, x):
7          """ Do not call constructor directly. Use Graph's
8              self._element = x
9
10     def element(self):
11         """ Return element associated with this vertex. """
12         return self._element
13
14     def __hash__(self):          # will allow vertex to
15         return hash(id(self))

```

```

18 class Edge:
19     """ Lightweight edge structure for a graph. """
20     __slots__ = '_origin', '_destination', '_element'
21
22     def __init__(self, u, v, x):
23         """ Do not call constructor directly. Use Graph's insert_edge(u,v,x).
24             self._origin = u
25             self._destination = v
26             self._element = x
27
28     def endpoints(self):
29         """ Return (u,v) tuple for vertices u and v. """
30         return (self._origin, self._destination)
31
32     def opposite(self, v):
33         """ Return the vertex that is opposite v on this edge. """
34         return self._destination if v is self._origin else self._origin
35
36     def element(self):
37         """ Return element associated with this edge. """
38         return self._element
39
40     def __hash__(self):          # will allow edge to be a map/set key
41         return hash( (self._origin, self._destination) )

```

# PYTHON IMPLEMENTATION

```
1 class Graph:
2     """ Representation of a simple graph using an adjacency map."""
3
4     def __init__(self, directed=False):
5         """ Create an empty graph (undirected, by default).
6
7         Graph is directed if optional paramter is set to True.
8         """
9         self._outgoing = { }
10        # only create second map for directed graph; use alias for un
11        self._incoming = { } if directed else self._outgoing
12
13    def is_directed(self):
14        """ Return True if this is a directed graph; False if undirected
15
16        Property is based on the original declaration of the graph, no
17        """
18        return self._incoming is not self._outgoing # directed if map
19
20    def vertex_count(self):
21        """ Return the number of vertices in the graph."""
22        return len(self._outgoing)
```

```
24    def vertices(self):
25        """ Return an iteration of all vertices of the graph."""
26        return self._outgoing.keys()
27
28    def edge_count(self):
29        """ Return the number of edges in the graph."""
30        total = sum(len(self._outgoing[v]) for v in self._outgoing)
31        # for undirected graphs, make sure not to double-count edges
32        return total if self.is_directed() else total // 2
33
34    def edges(self):
35        """ Return a set of all edges of the graph."""
36        result = set( ) # avoid double-reporting edges of undirected
37        for secondary_map in self._outgoing.values():
38            result.update(secondary_map.values()) # add edges of
39        return result
```

# PYTHON IMPLEMENTATION

```
40 def get_edge(self, u, v):
41     """Return the edge from u to v, or None if not adjacent
42     return self._outgoing[u].get(v)          # returns None if not adjacent
43
44 def degree(self, v, outgoing=True):
45     """Return number of (outgoing) edges incident to vertex v
46
47     If graph is directed, optional parameter used to count outgoing edges
48     """
49     adj = self._outgoing if outgoing else self._incoming
50     return len(adj[v])
51
52 def incident_edges(self, v, outgoing=True):
53     """Return all (outgoing) edges incident to vertex v in the form
54     (neighbor, edge)
55
56     If graph is directed, optional parameter used to request outgoing edges
57     """
58     adj = self._outgoing if outgoing else self._incoming
59     for edge in adj[v].values():
```

```
61 def insert_vertex(self, x=None):
62     """Insert and return a new Vertex with element x
63     v = self.Vertex(x)
64     self._outgoing[v] = {}
65     if self.is_directed():
66         self._incoming[v] = {}          # need to initialize incoming
67     return v
68
69 def insert_edge(self, u, v, x=None):
70     """Insert and return a new Edge from u to v with element x
71     e = self.Edge(u, v, x)
72     self._outgoing[u][v] = e
73     self._incoming[v][u] = e
```



# THANKS

See you in the next session!