

CS 261 Data Structures

Spring 2025

Priority Queue, Graph

Tianshuo Zhou

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Roadmap

- priority queue
 - concept of priority
 - binary heap
 - heap sort
 - assignment: heap implementation
- graph
 - concept, types, representations
 - BFS
 - DFS
 - topological sort

Priority Queue

- examples of priority?

Priority Queue

- examples of priority?
 - emergency room
 - process niceness
 - email responses
 - queue

Priority Queue

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- priority queue is an ADT
- a data structure is an instance of ADT

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 - $\text{push}(P, x)$ / $\text{insert}(P, x)$
 - $\text{pop}(P)$ / $\text{dequeue}(P)$
 - build heap / heapify
 - $\text{peek}(P)$ / $\text{max}(P)$

Priority Queue

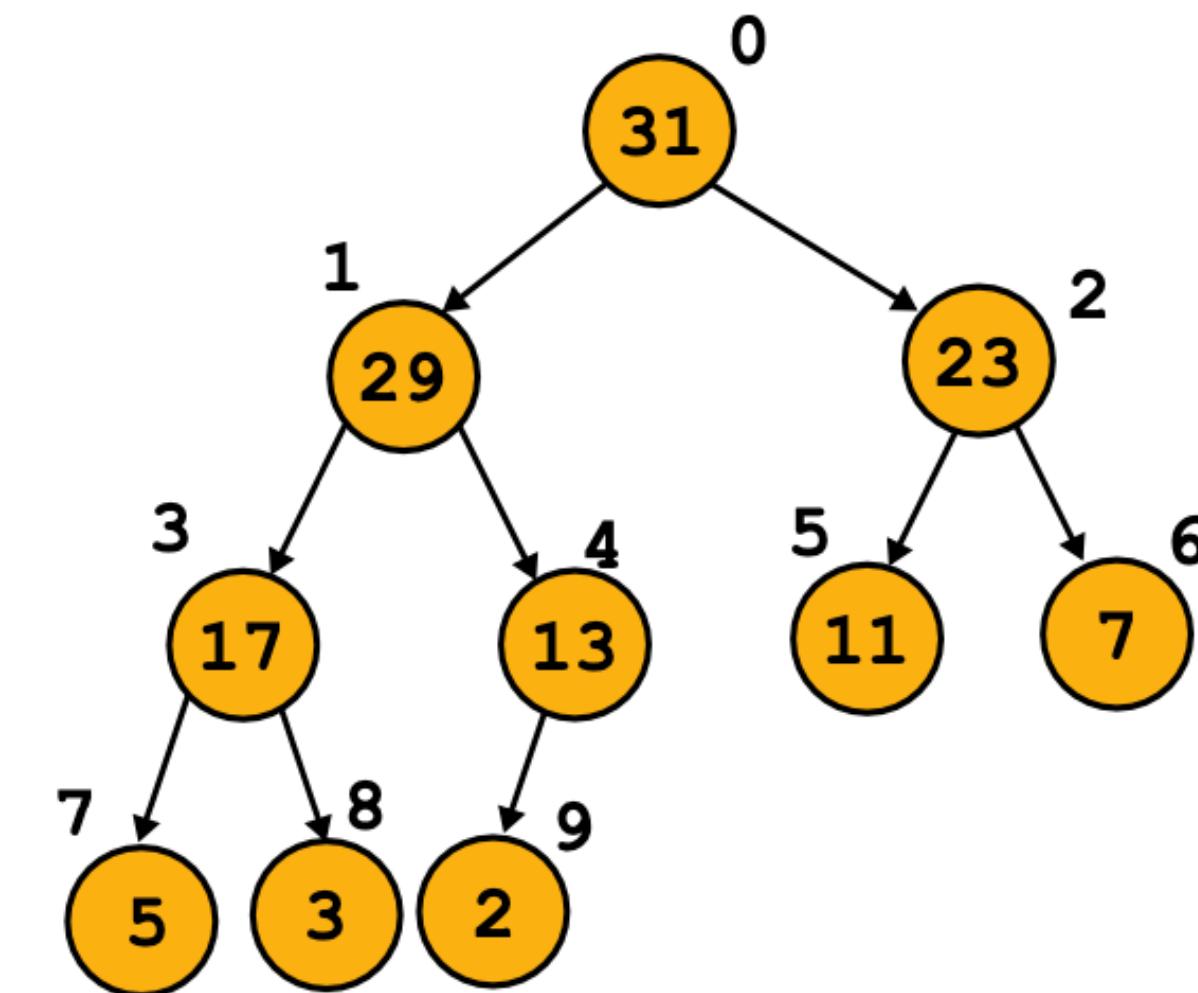
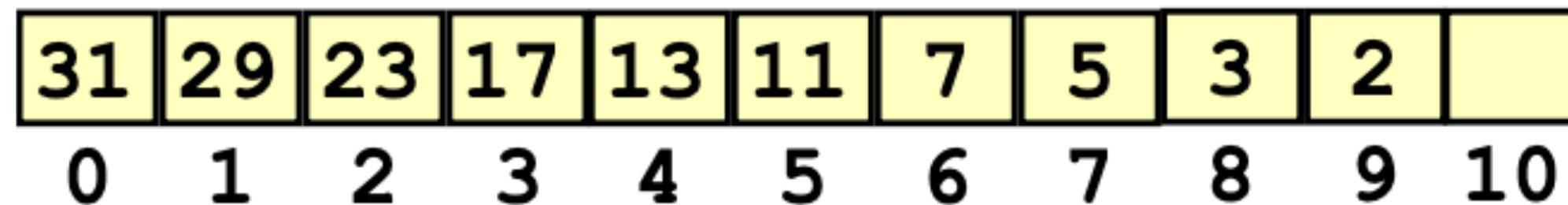
- priority queue is an ADT
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 - build heap / heapify
 - $\text{peek}(P)$ / $\text{max}(P)$
- implementation?

Priority Queue Implementations

	sorted array	reversely sorted array	sorted doubly linked list	balanced BST	binary heap
push	O(n)	O(n)	O(n)	O(log n)	O(log n)
pop-max	O(1)	O(n)	O(1)	O(log n)	O(log n)
peek-max	O(1)	O(1)	O(1)	O(log n)	O(1)
build	O($n \log n$)	O($n \log n$)	O($n \log n$)	O($n \log n$)	O(n)

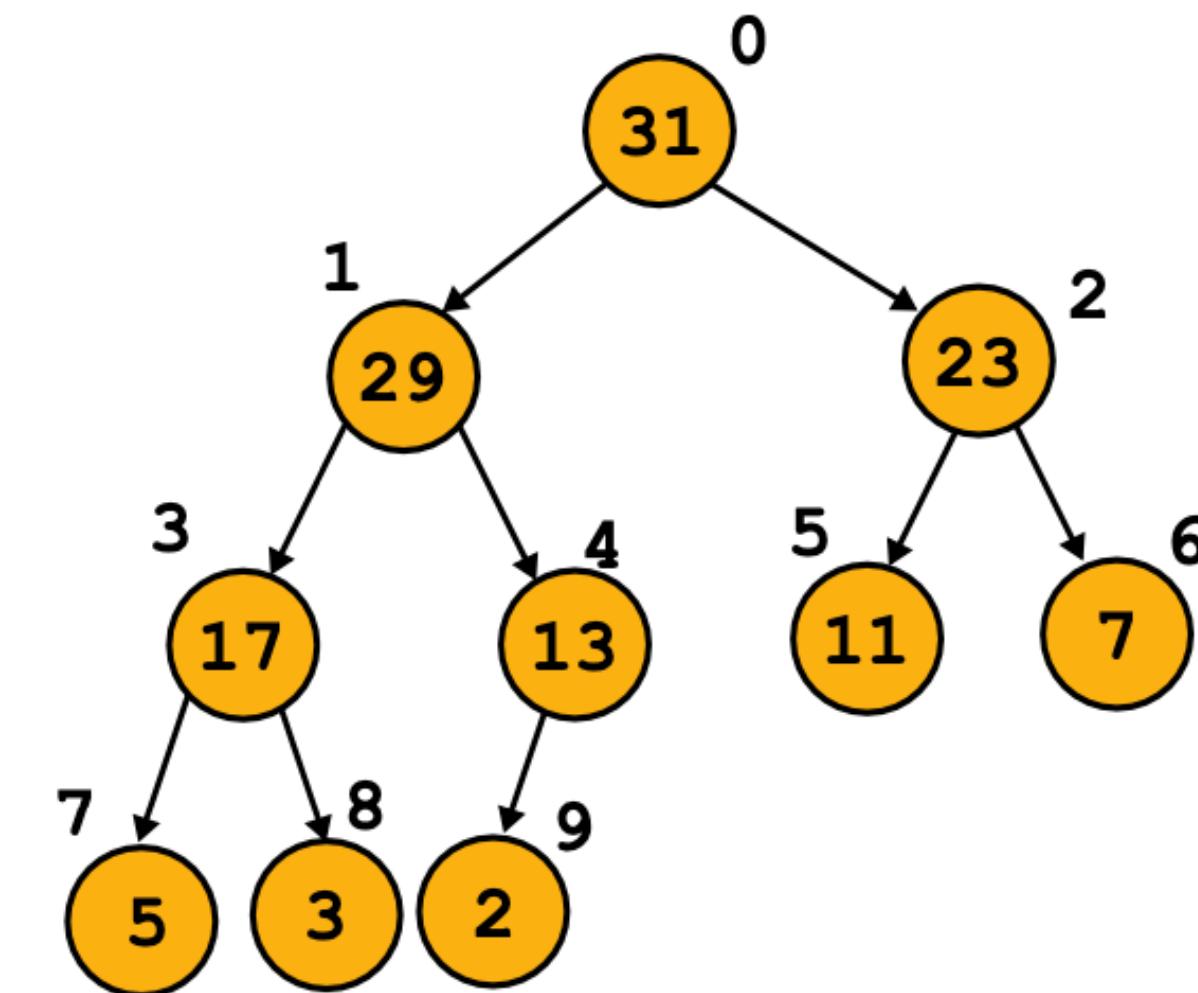
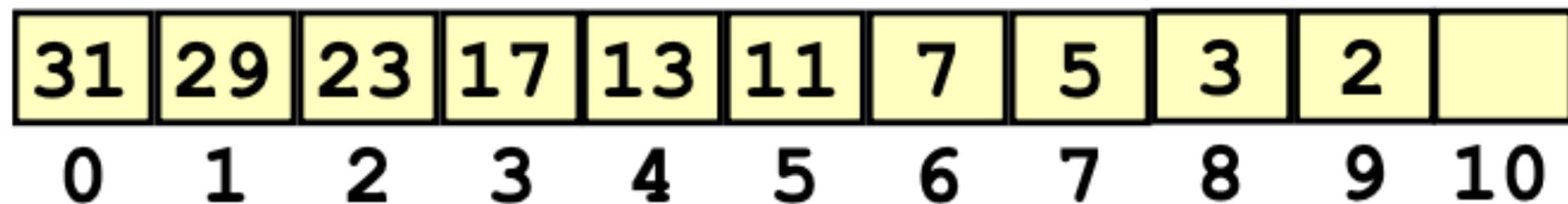
Binary Heap

- structure or topology
 - logically: complete tree
 - physically: linear array



Binary Heap

- structure or topology
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- complexity
 - push → bubble up: $O(h) = O(\log n)$
 - pop → bubble down: $O(h) = O(\log n)$
 - build_heap or heapify: $O(n)$

Binary Heap

- build a heap or heapify

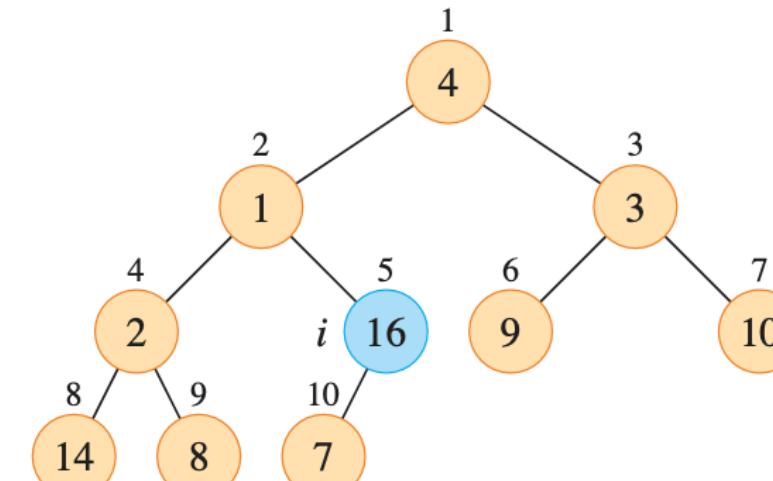
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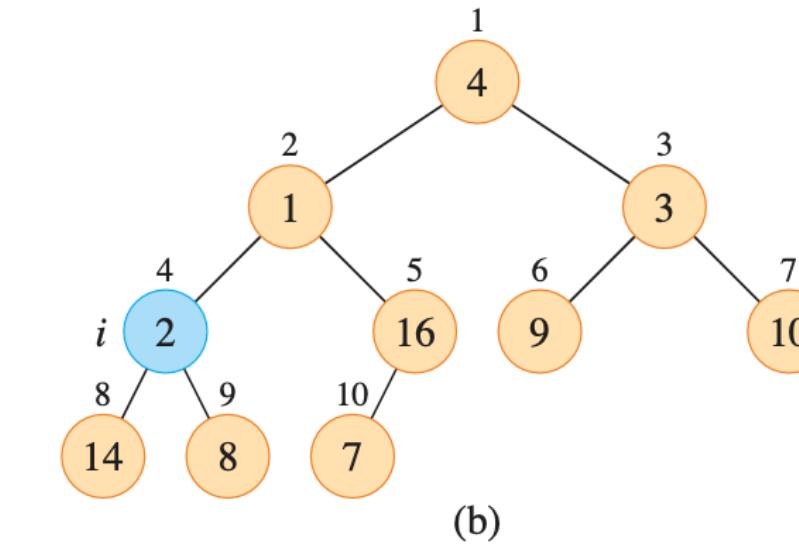
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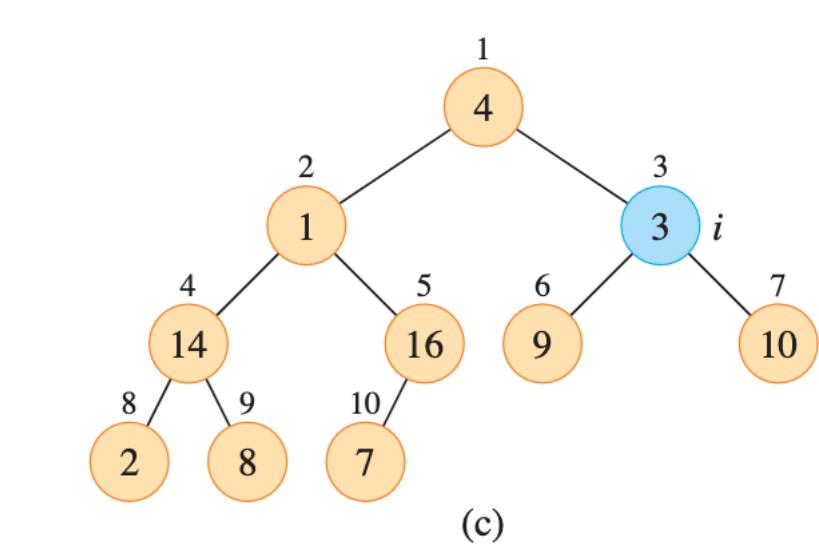
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-----	---	---	---	---	----	---	----	----	---	---



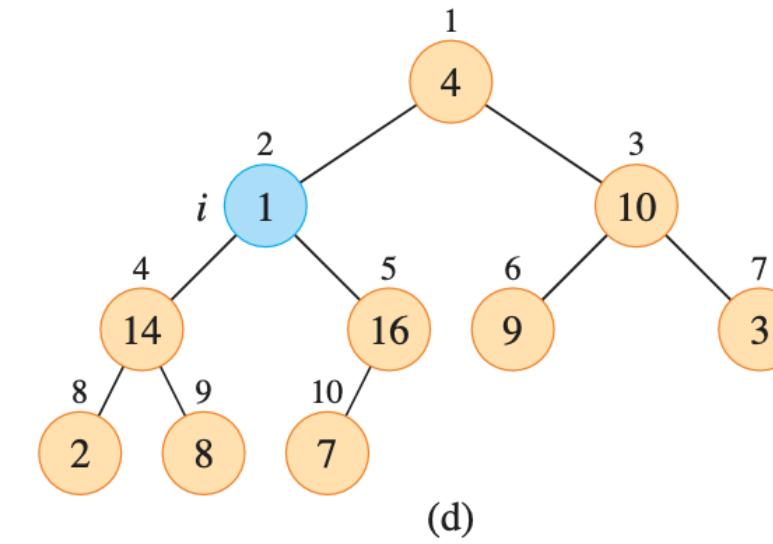
(a)



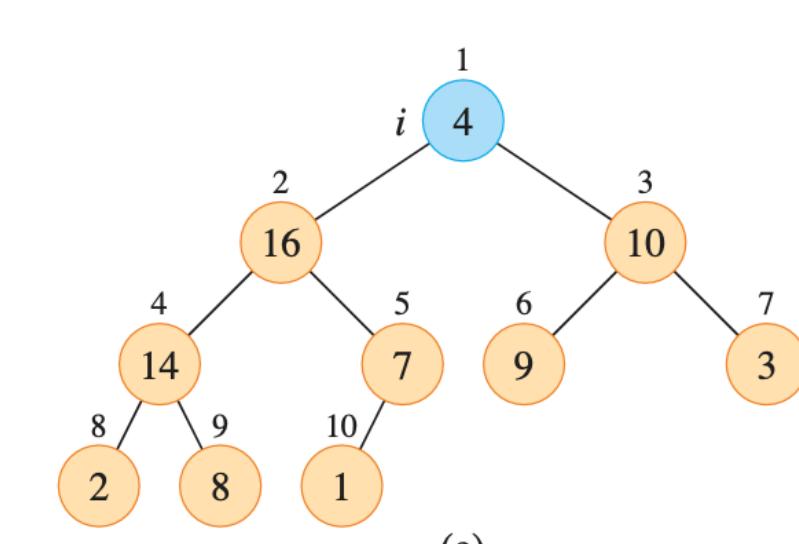
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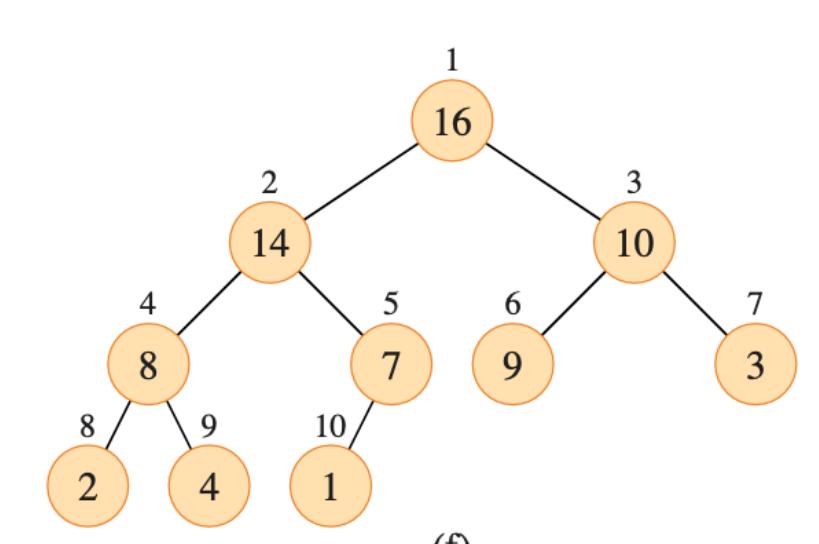
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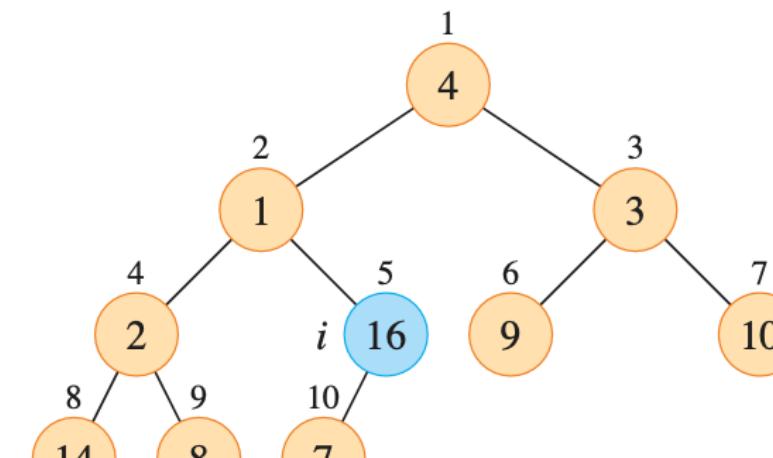
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- count how many bubble-down steps

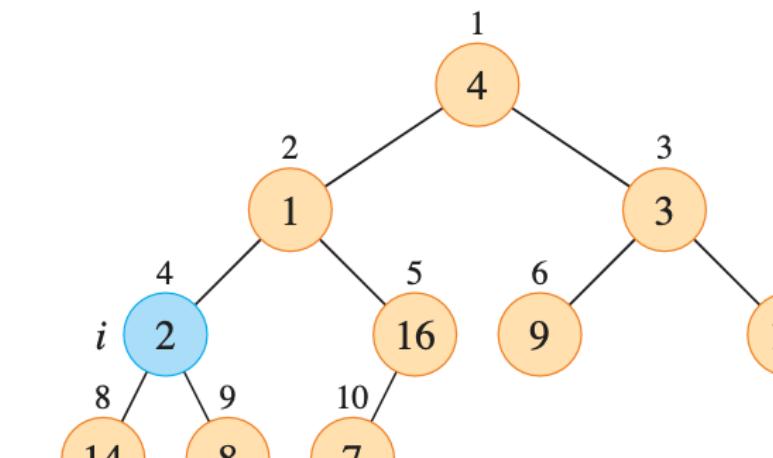
$$\bullet \frac{n}{2} \times 0 + \frac{n}{4} \times 1 + \frac{n}{8} \times 2 + \frac{n}{16} \times 3 + \dots$$

- arithmetico-geometric sequence

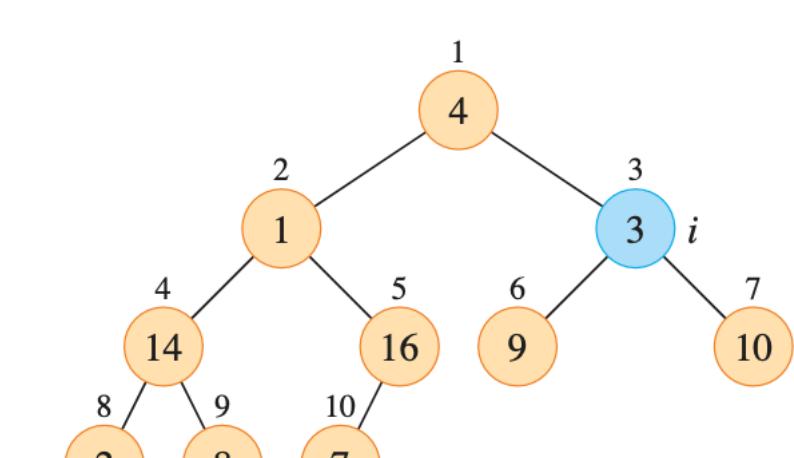
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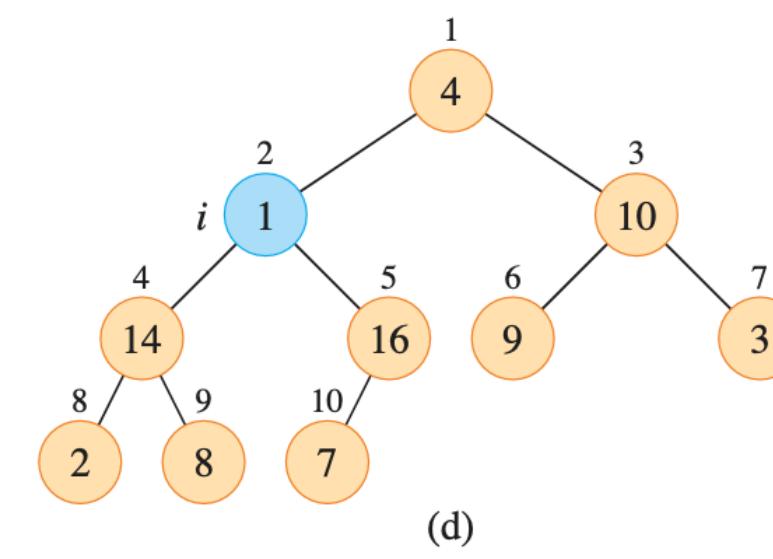
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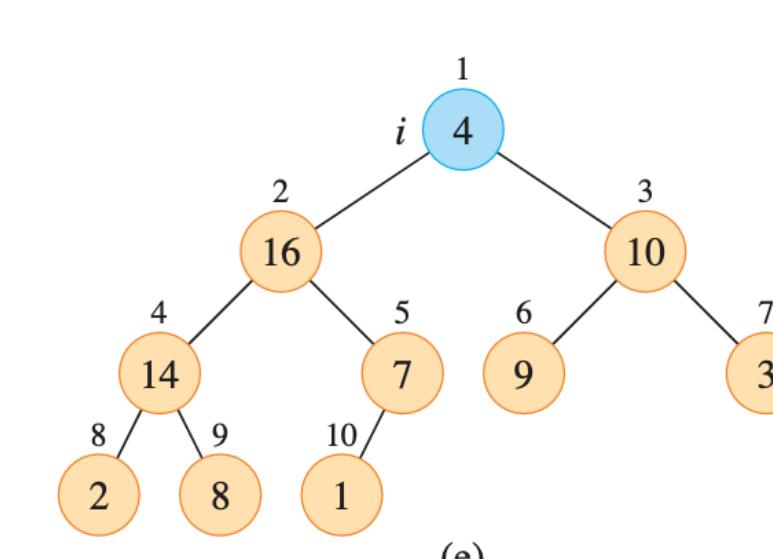
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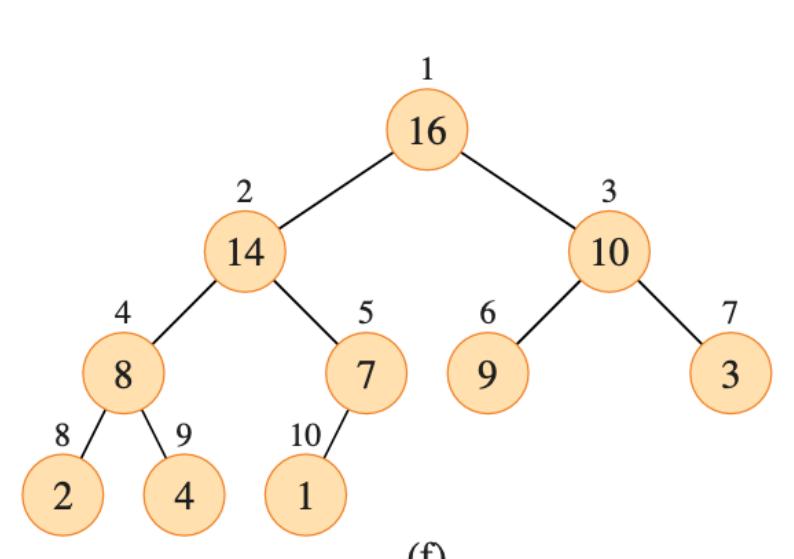
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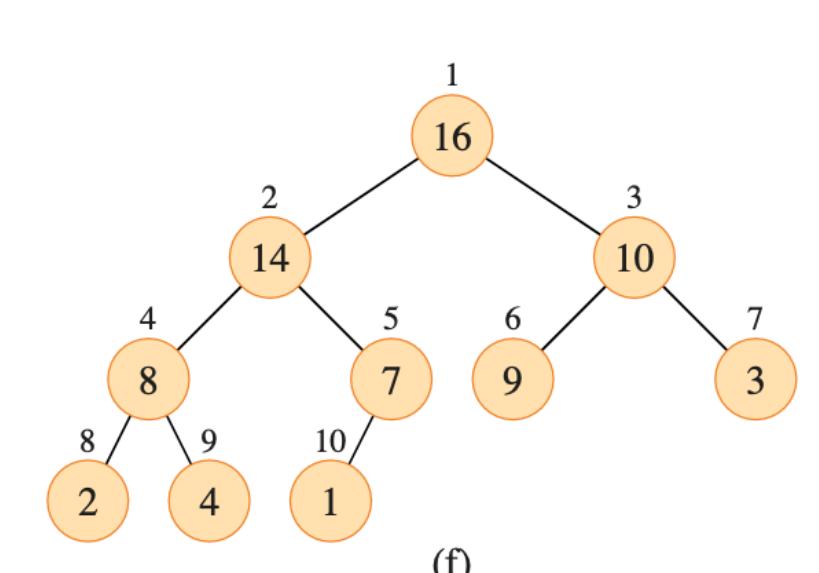
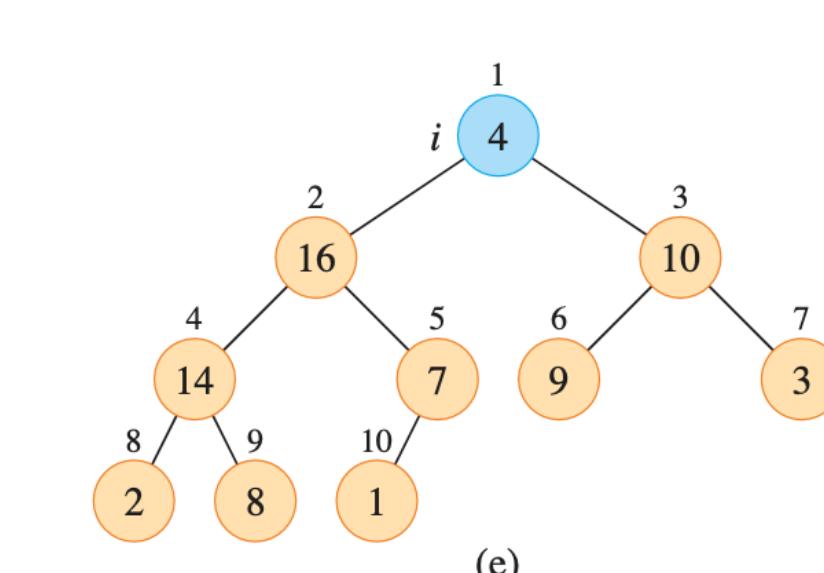
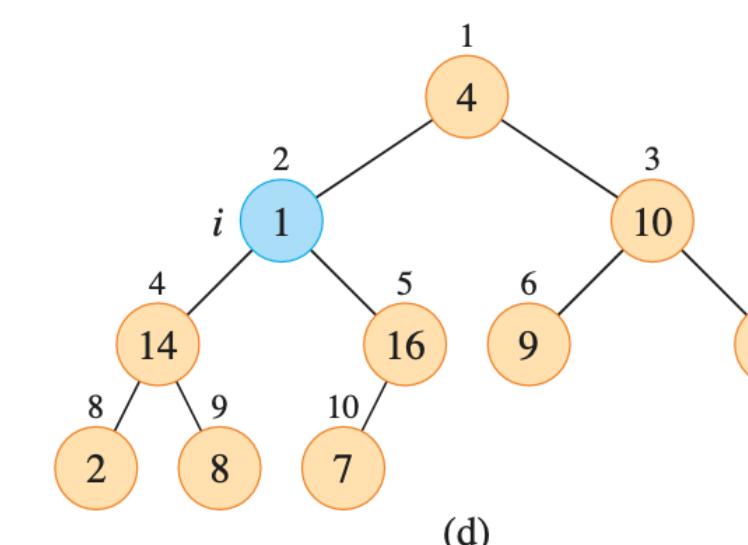
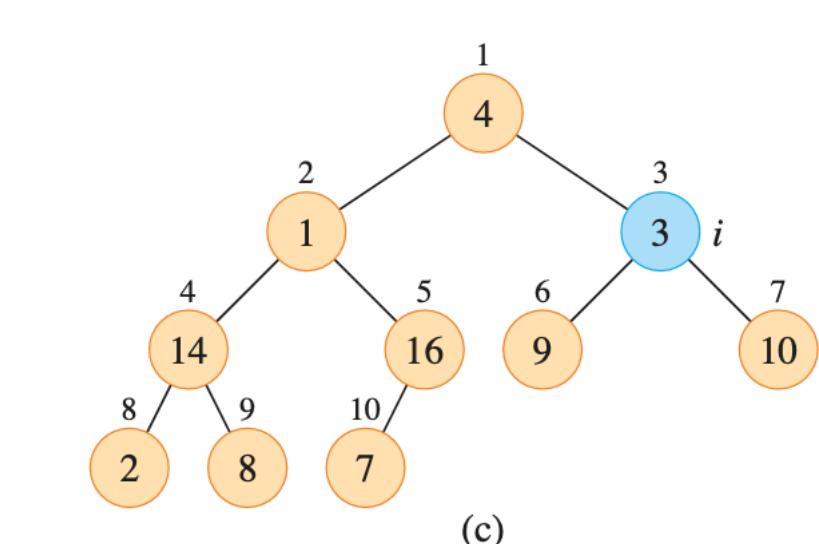
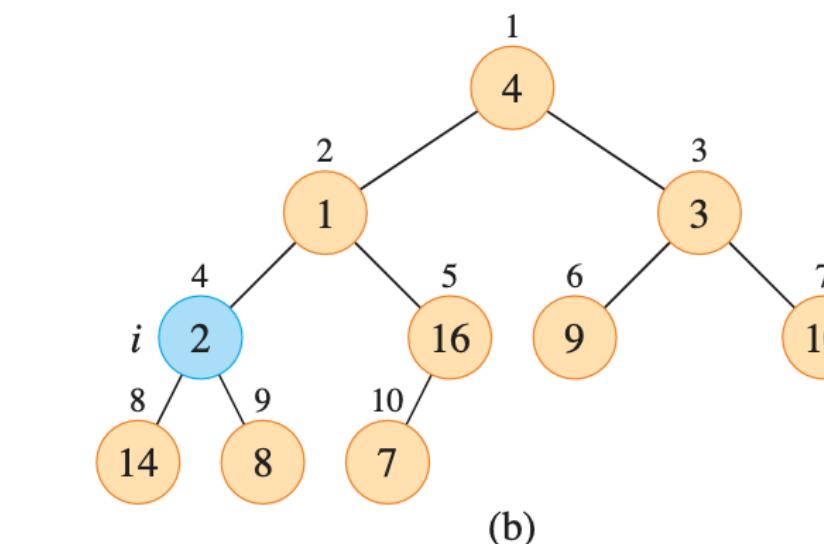
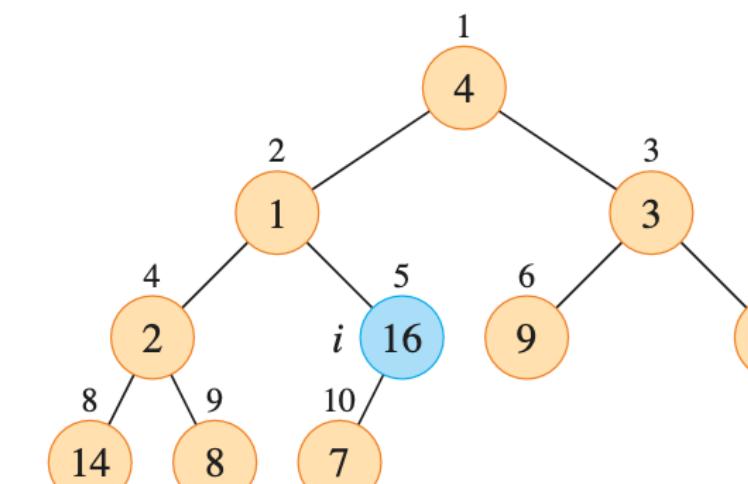
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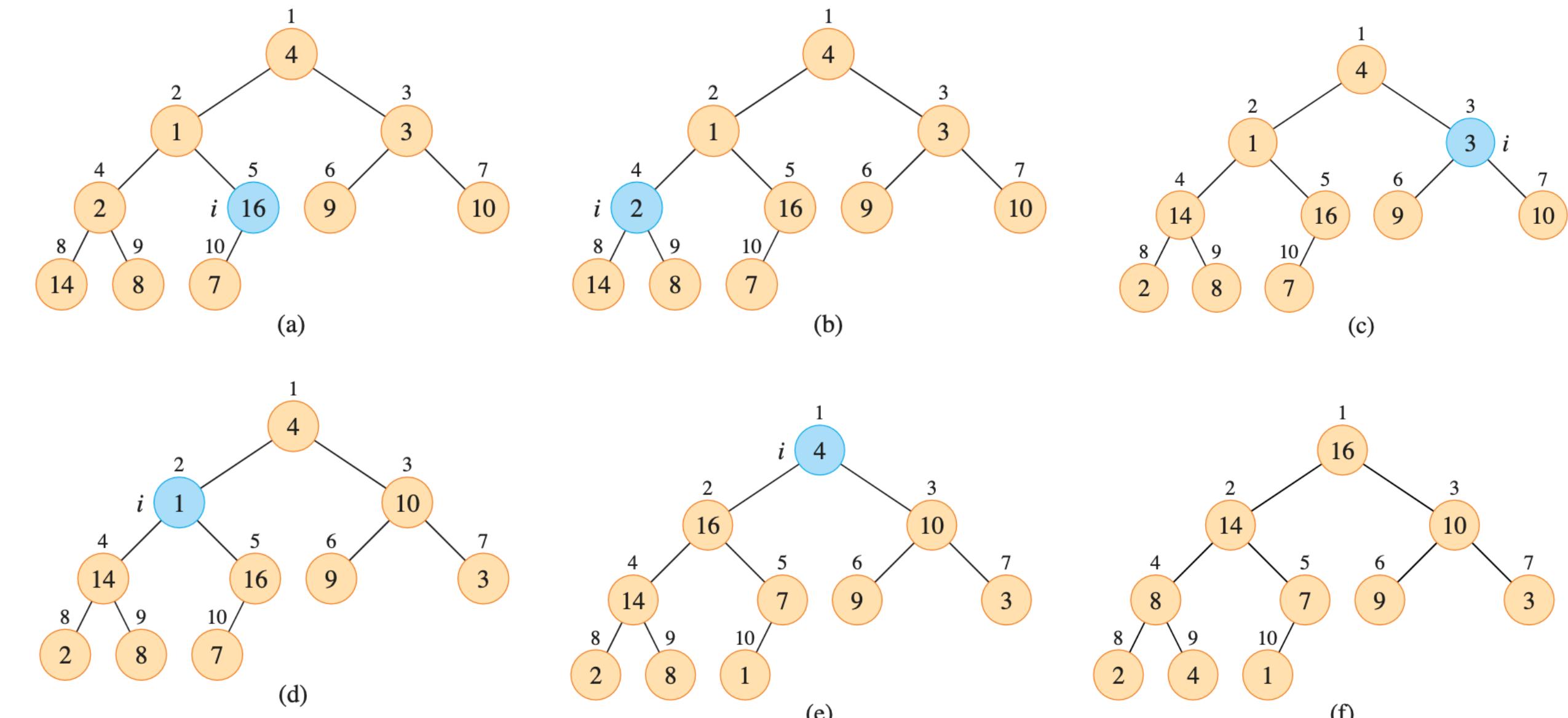
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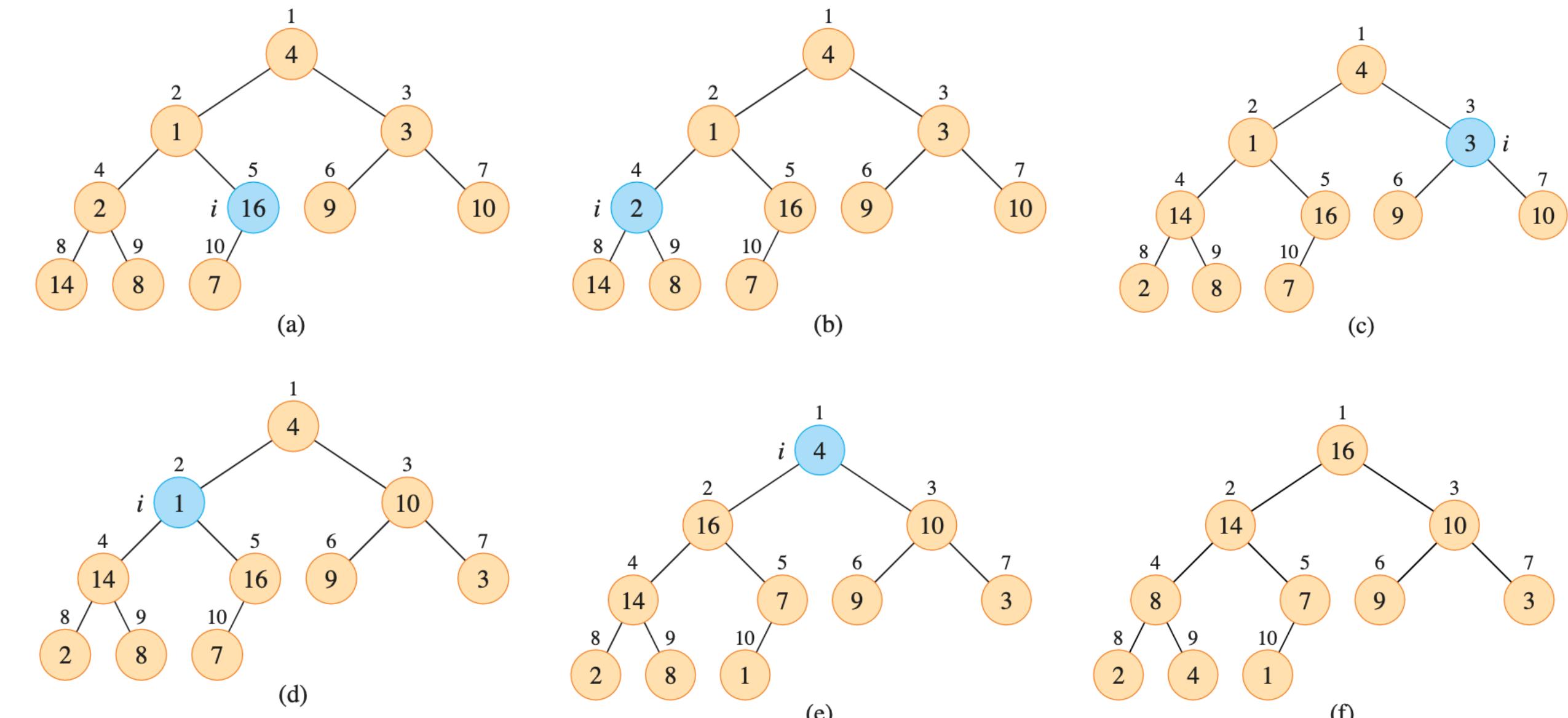
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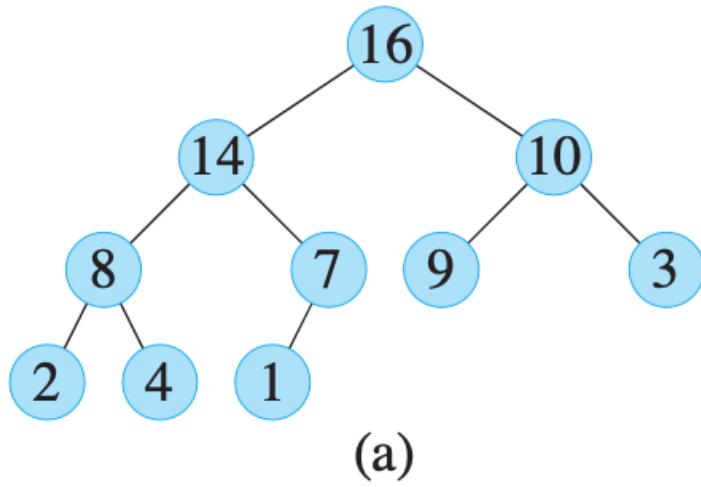
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HeapSort

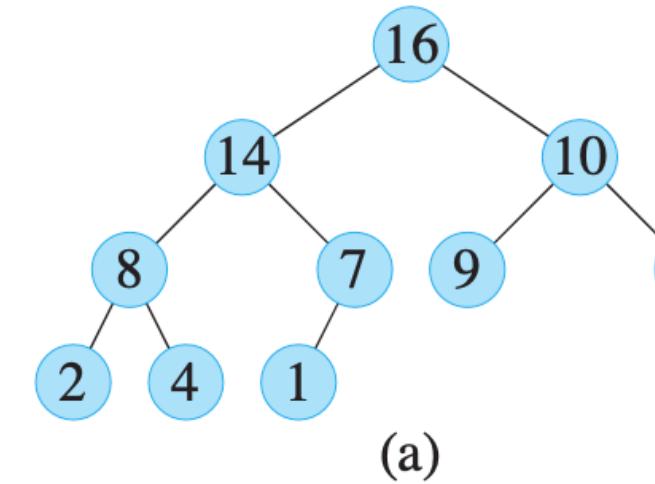


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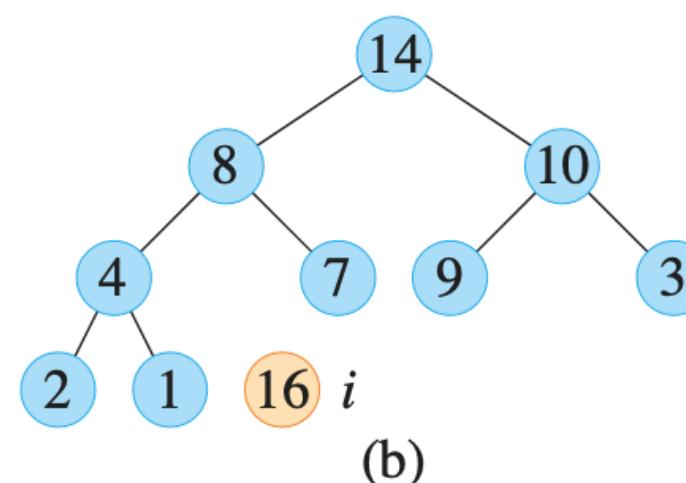
- two-stage algorithm

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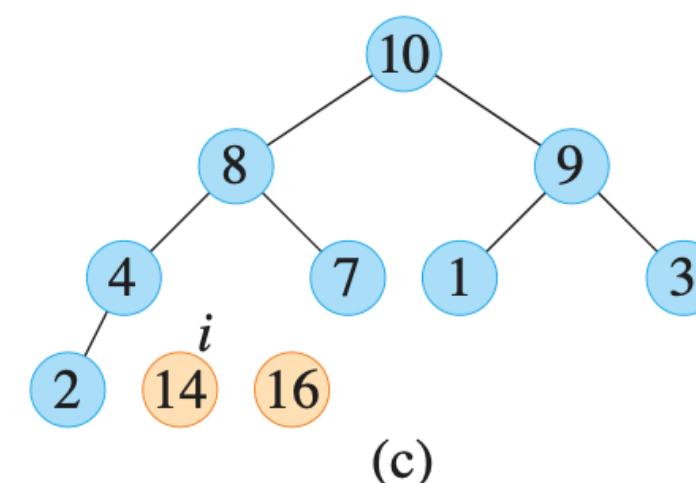
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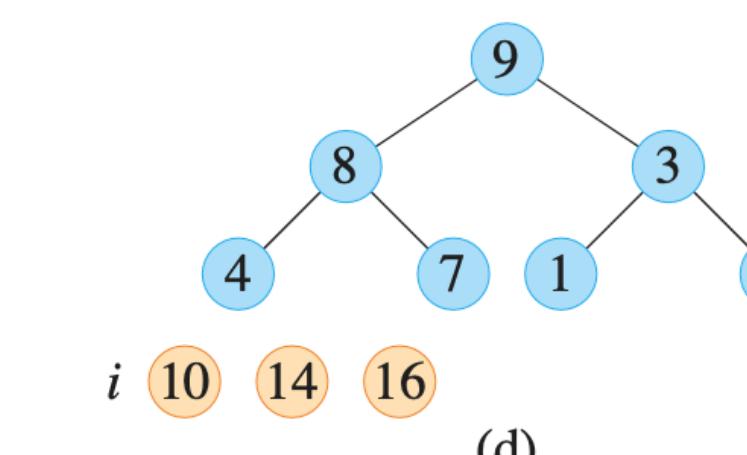
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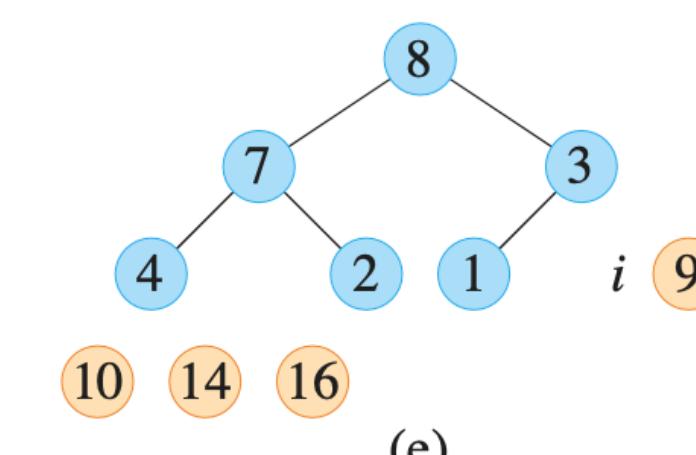
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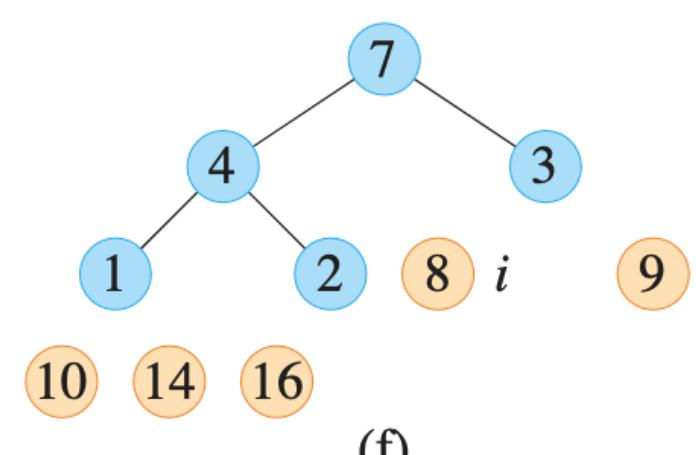
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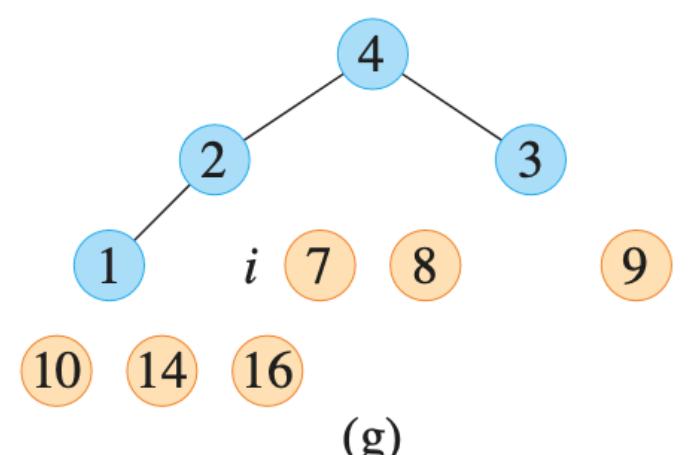
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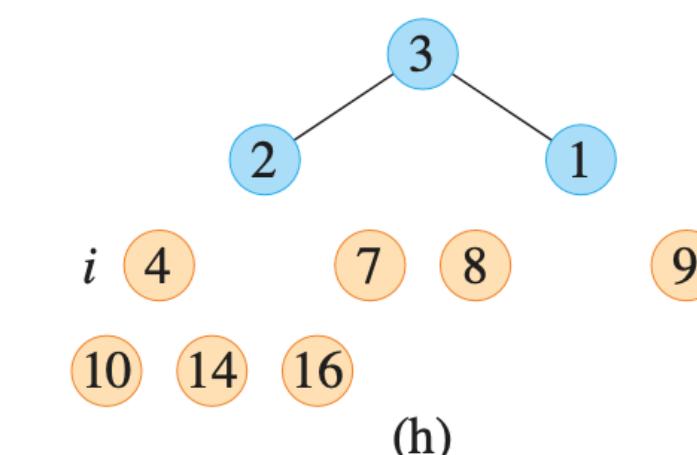
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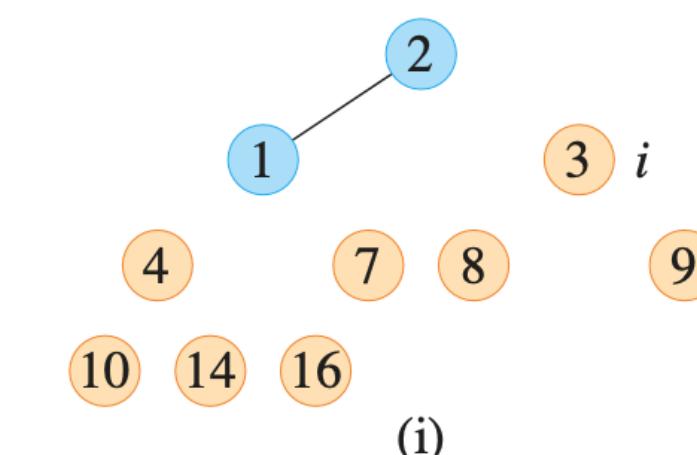
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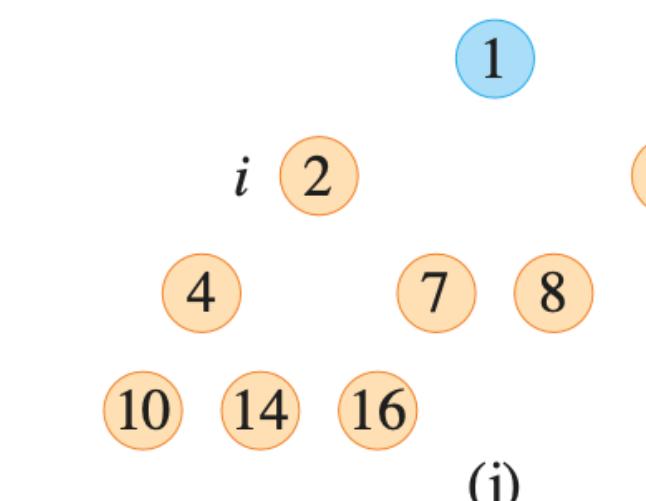
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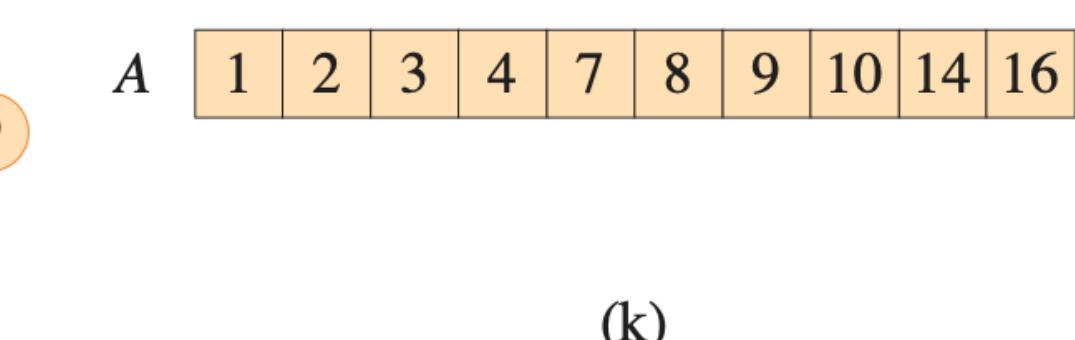
(h)



(i)



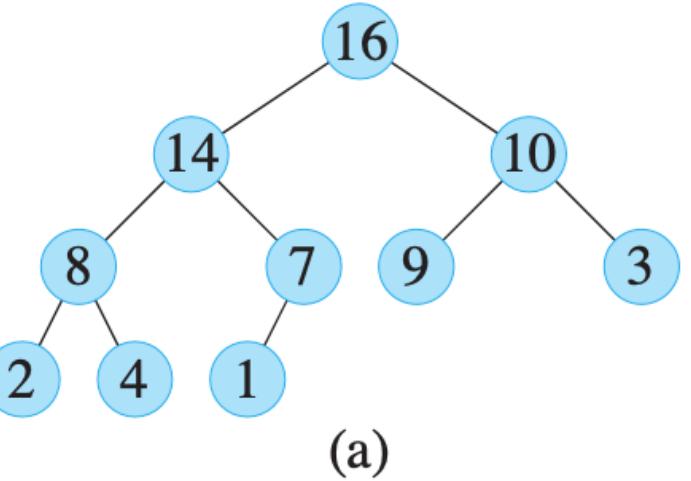
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HeapSort

- two-stage algorithm

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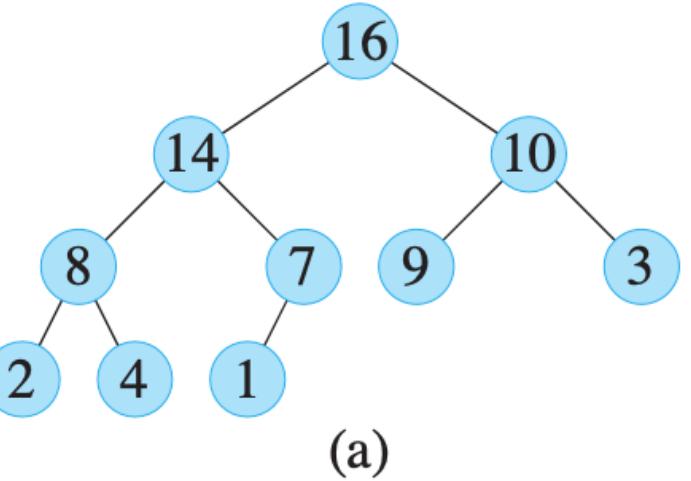
- complexity analysis

- build heap: $O(n)$
- $n - 1$ pops (bubble-down): $O(n \log(n))$
- total cost: $O(n \log(n))$

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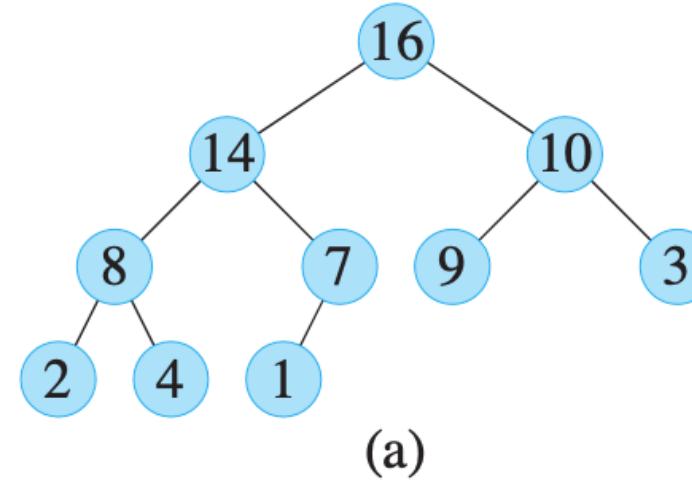
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- asymptotic notations
 - upper bound: $T(n) = O(f(n))$
 - $T(n) \leq c \cdot f(n)$
 - lower bound: $T(n) = \Omega(f(n))$
 - $T(n) \geq c \cdot f(n)$
 - tight bounds: $T(n) = \Theta(f(n))$
 - $c_1 \cdot f(n) \leq T(n) \leq c_2 \cdot f(n)$

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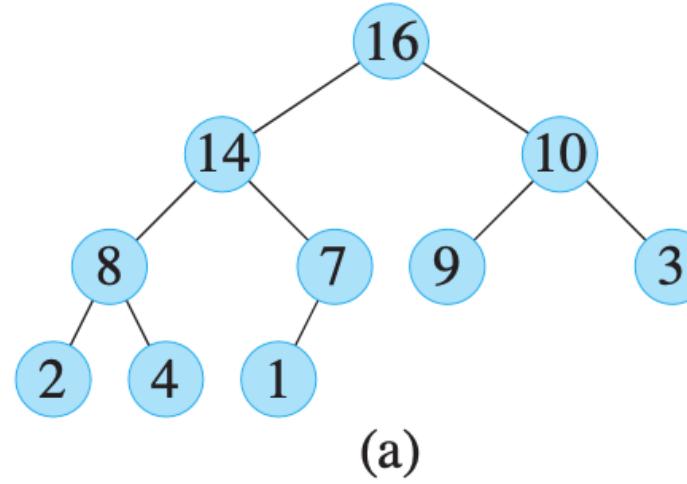
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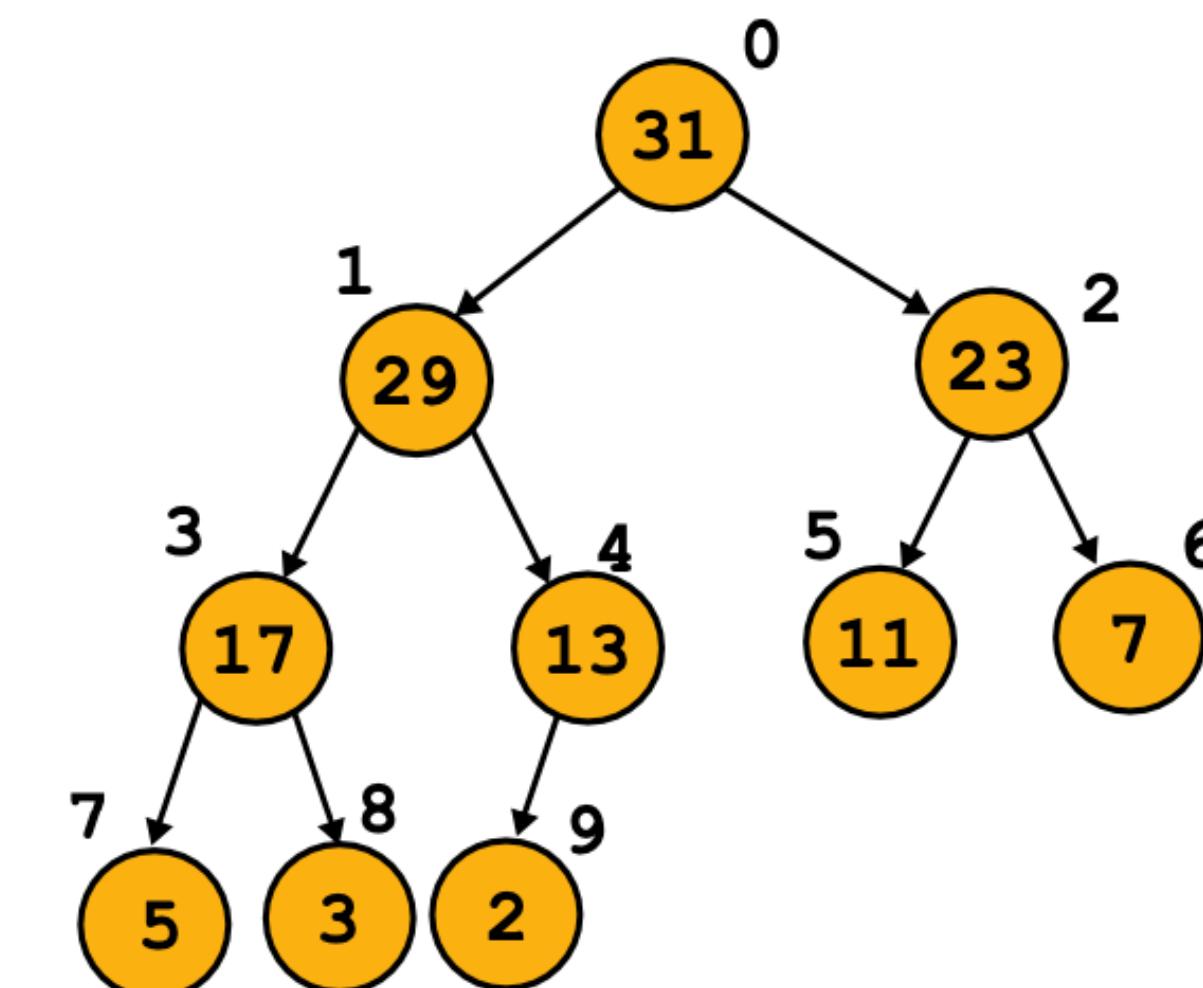
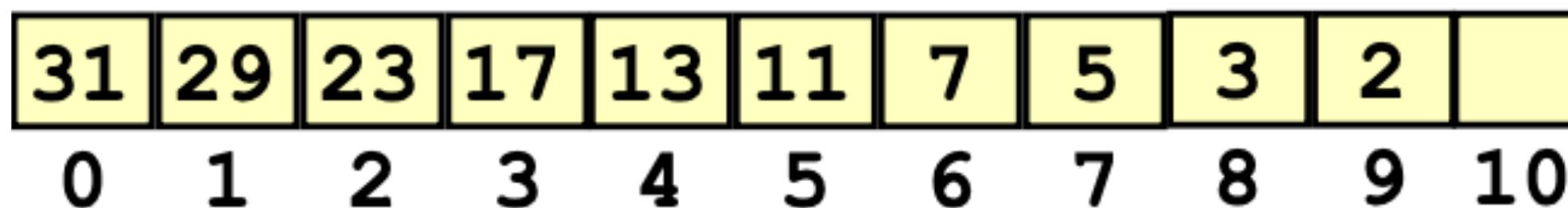
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- is $n \log(n)$ tight? cost is at least $\Theta\left(\frac{n}{2} \times \log \frac{n}{2}\right) = \Theta(n \log n)$
- yes! first half of the nodes have a height of $\log(n/2)$ when popped

Assignment 6

- objectives
 - heap data: dynamic arrays
 - heap operation: insert, percolate_up, percolate_down
- structure or topology
 - logically: complete tree
 - physically: linear array

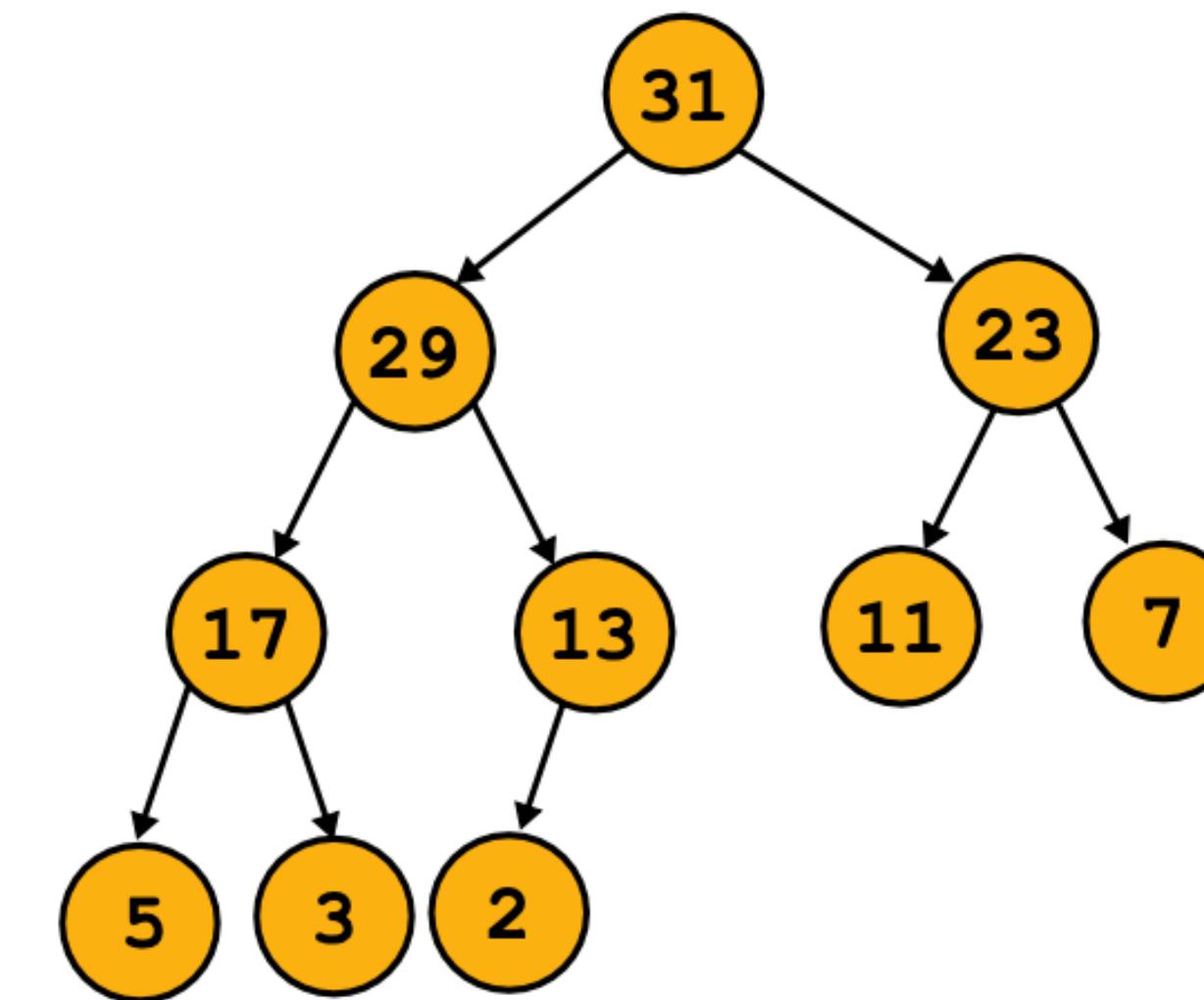
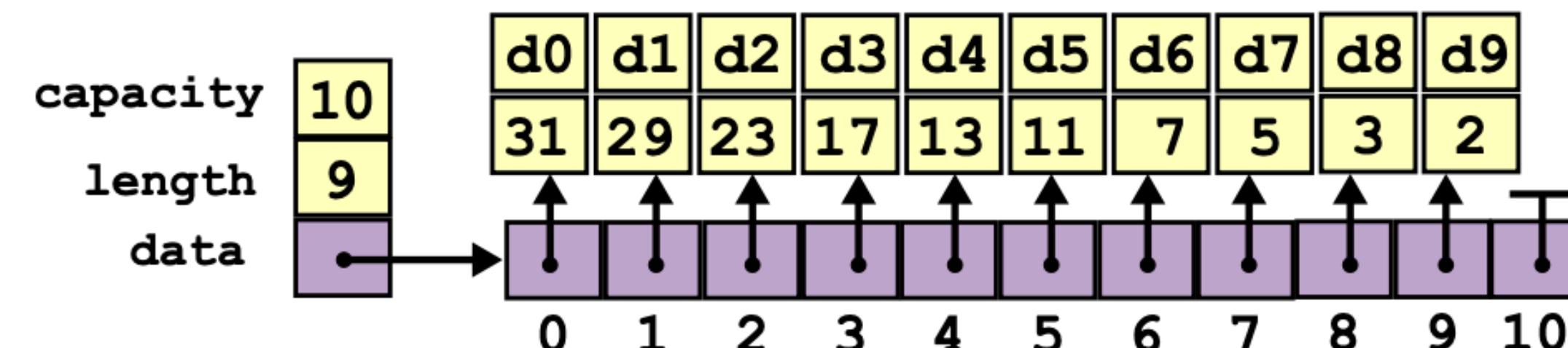


Assignment 6

- objectives
 - heap data: dynamic arrays of **key(priority)** and **value**
 - heap operation: insert, percolate_up, percolate_down

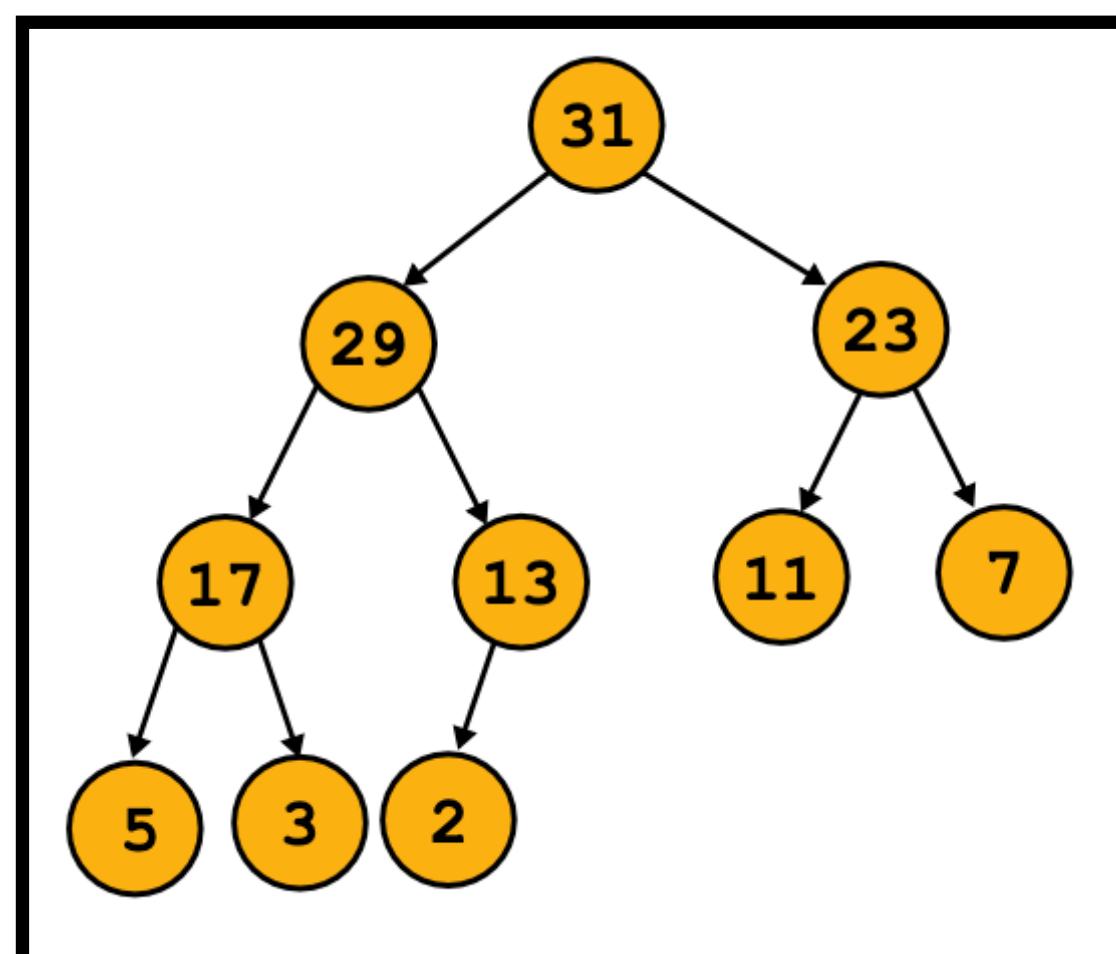
```
struct element {  
    void* data;  
    int priority;  
};
```

```
struct dynarray {  
    void** data;  
    int length;  
    int capacity;  
};
```



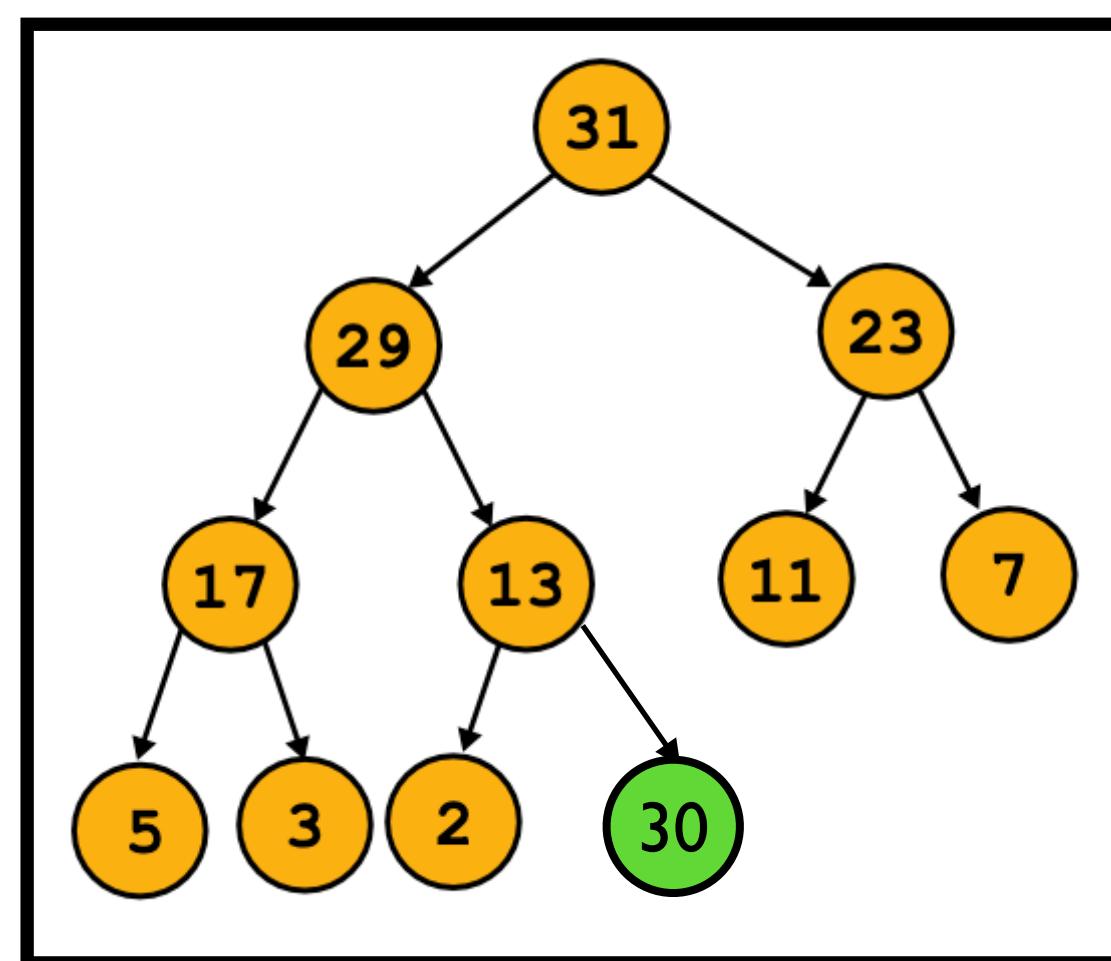
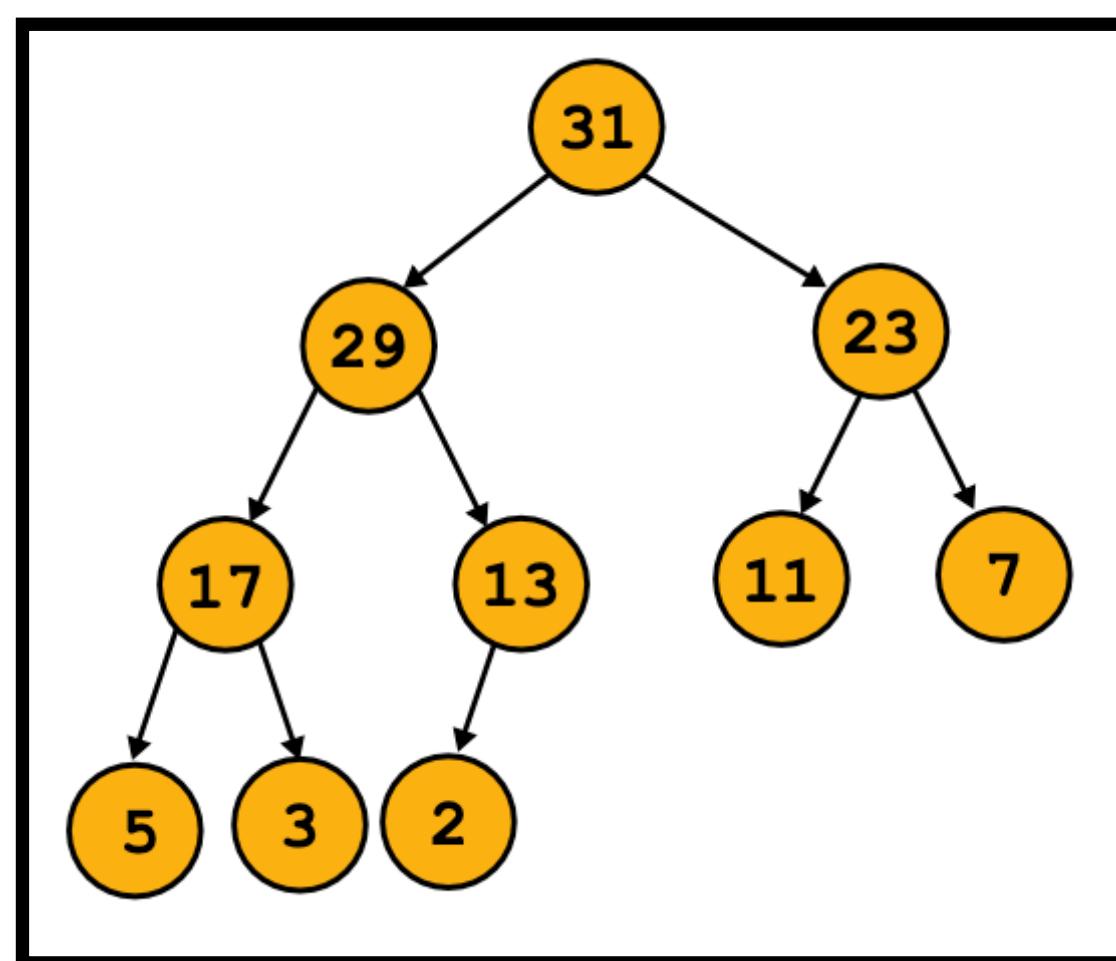
Assignment 6

- push / insert
 - append new element
 - percolate / bubble up



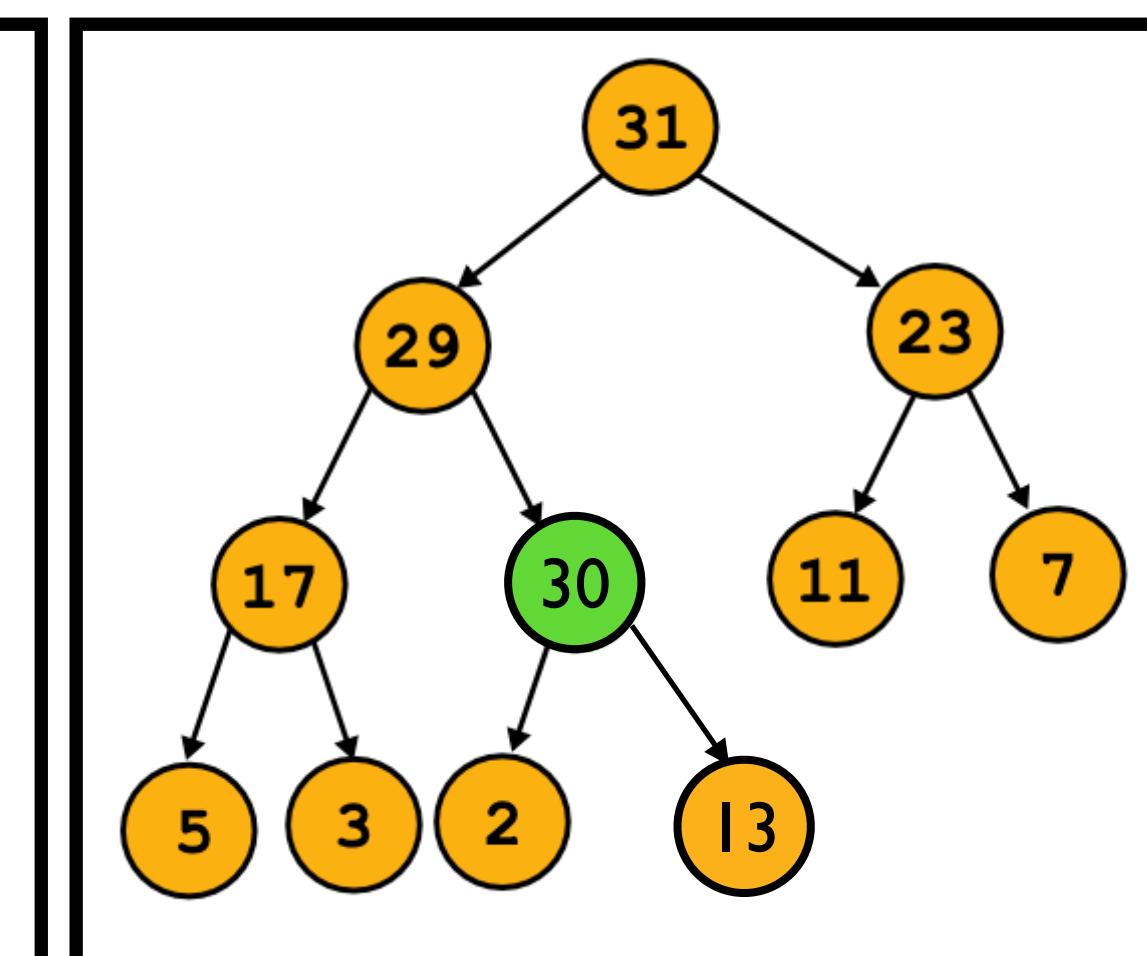
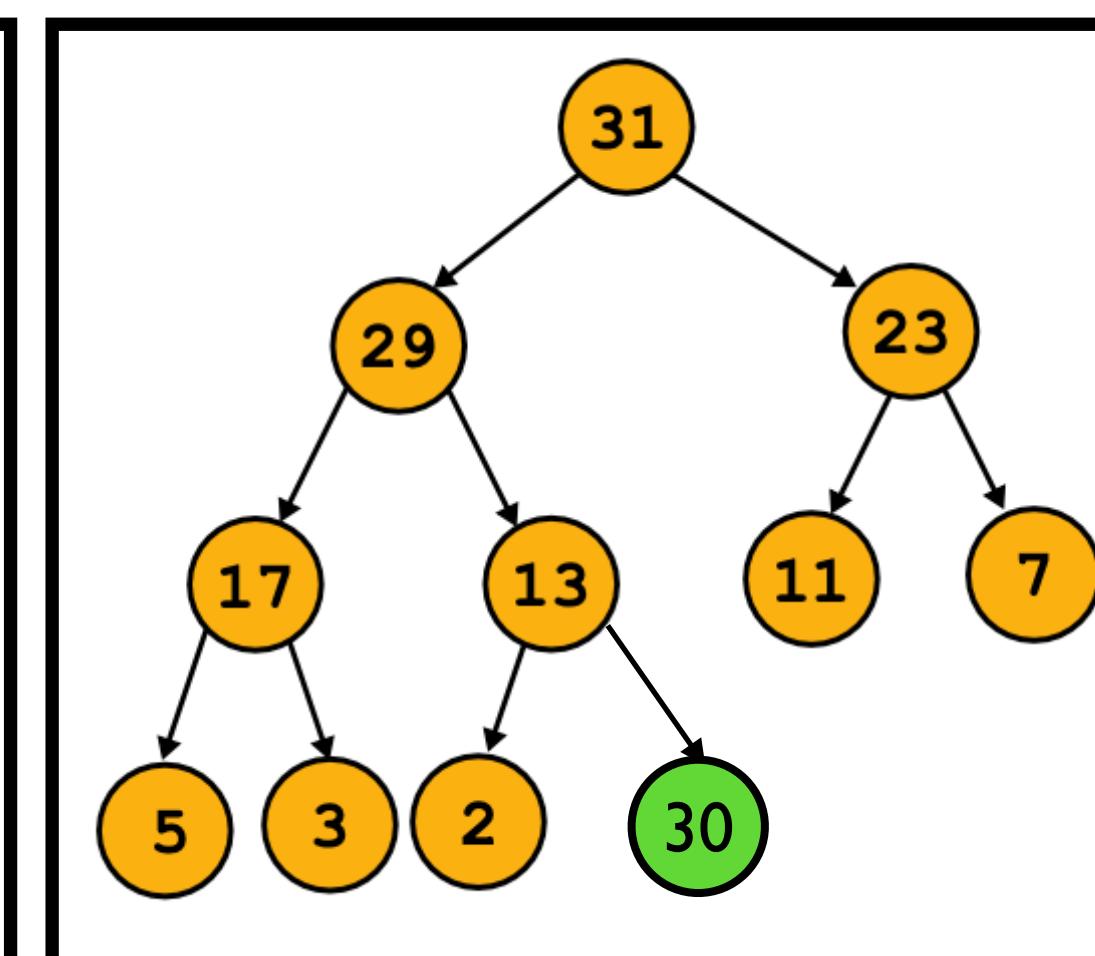
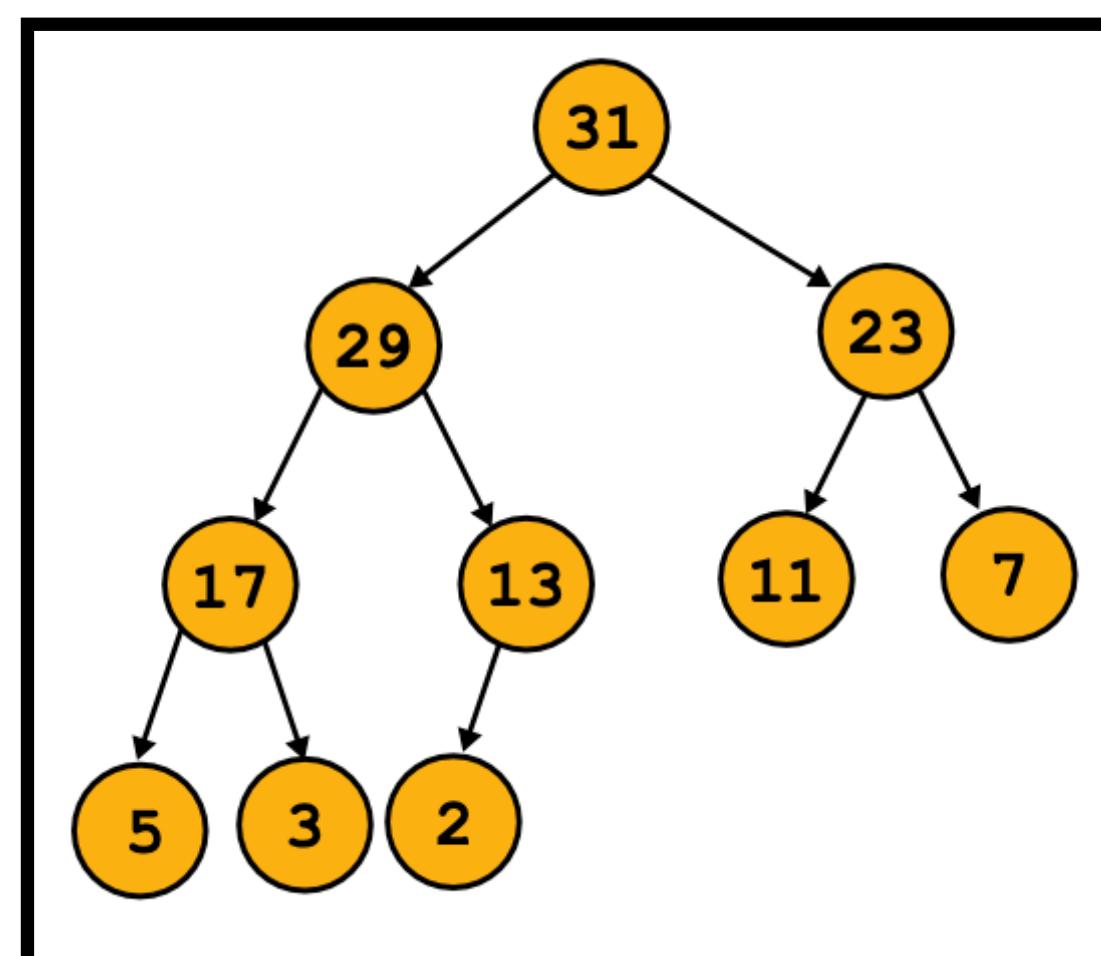
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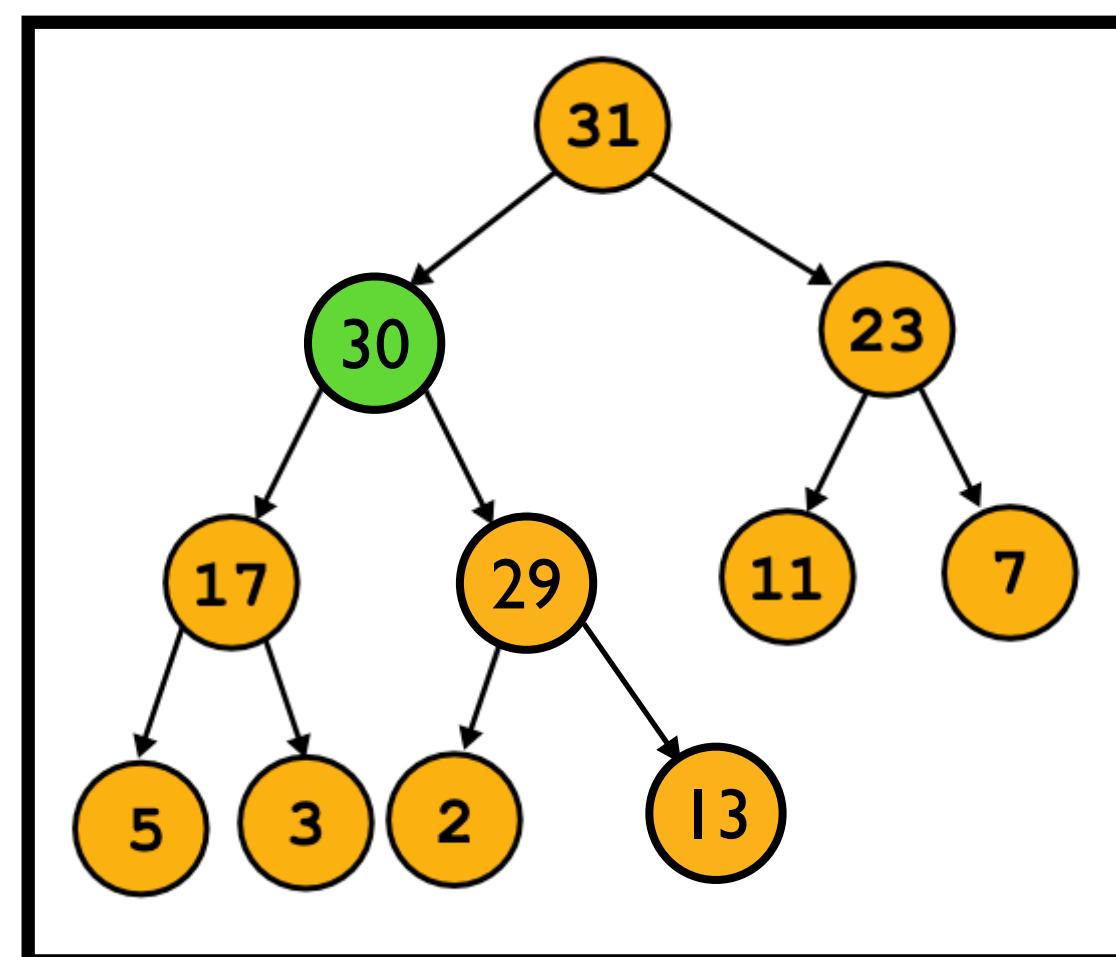
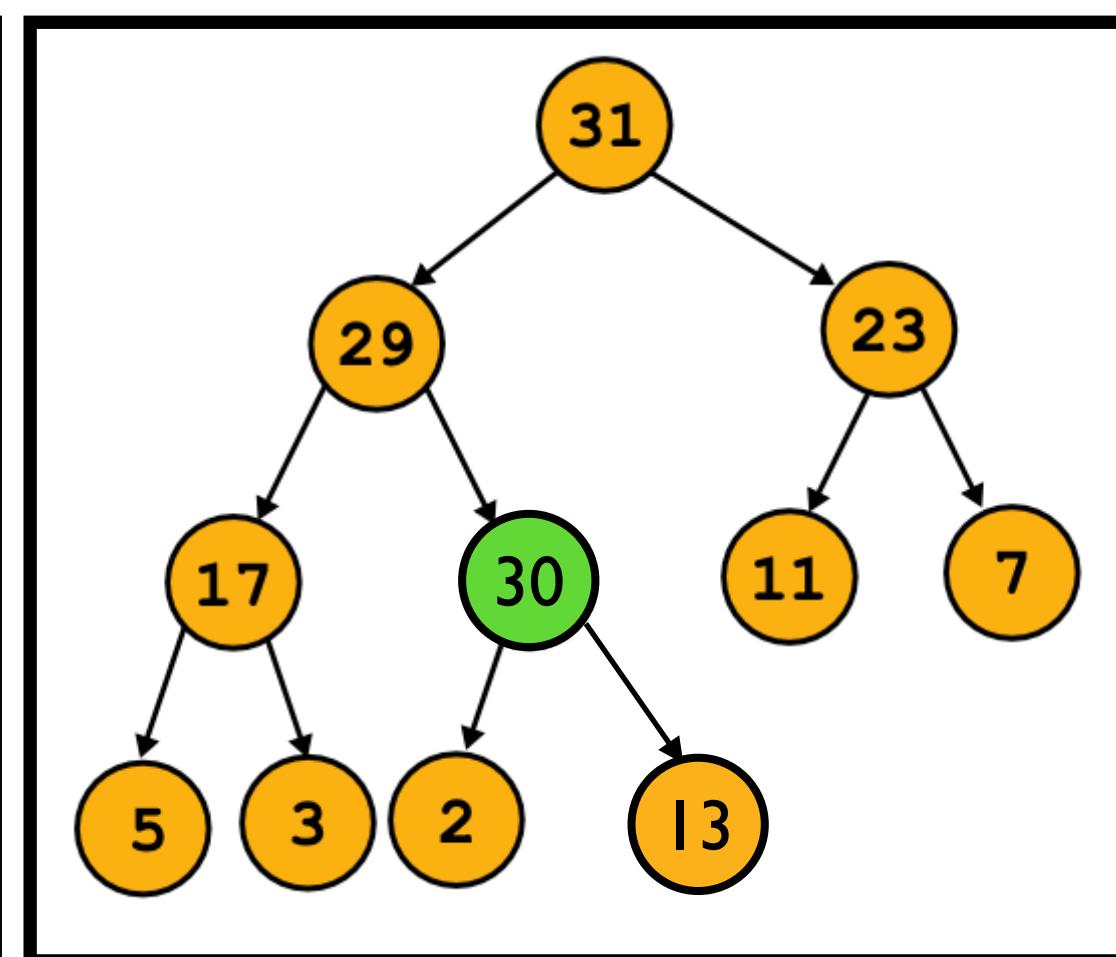
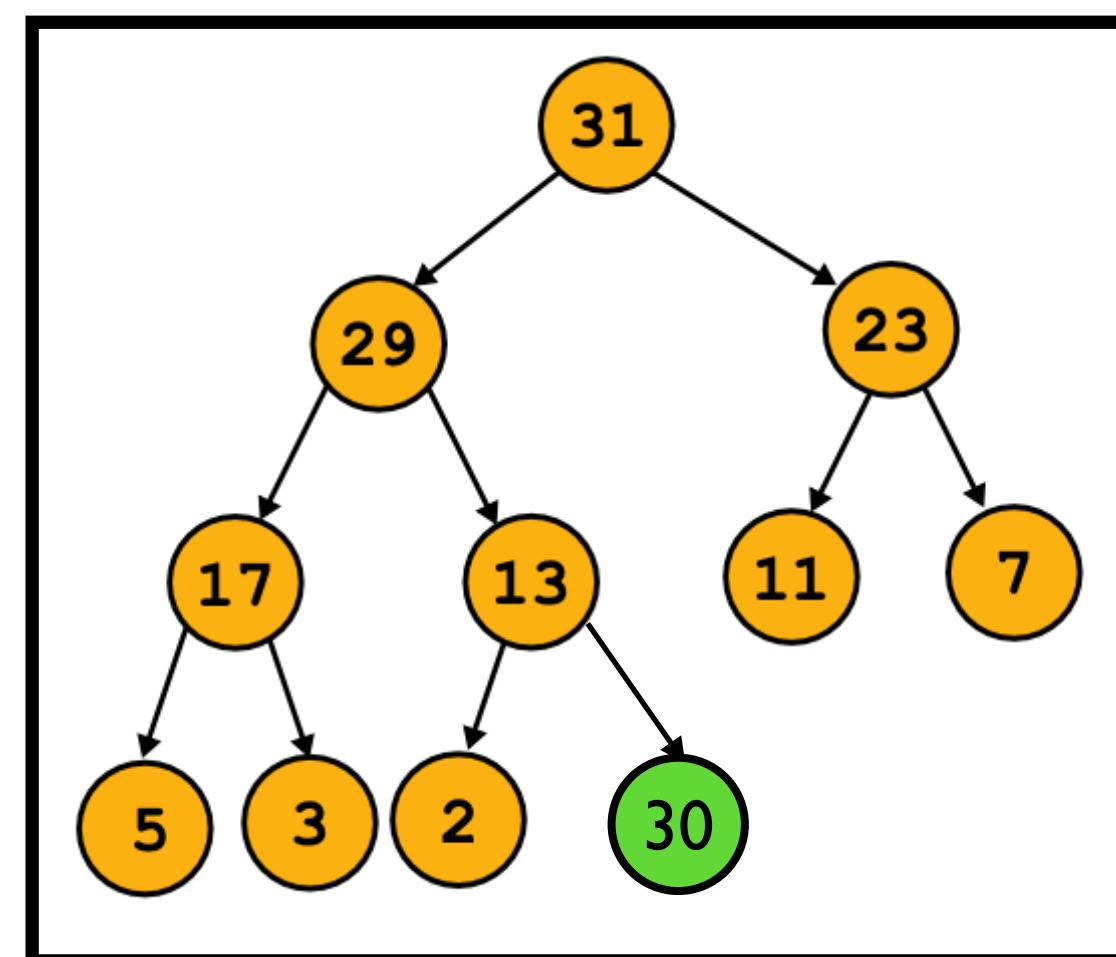
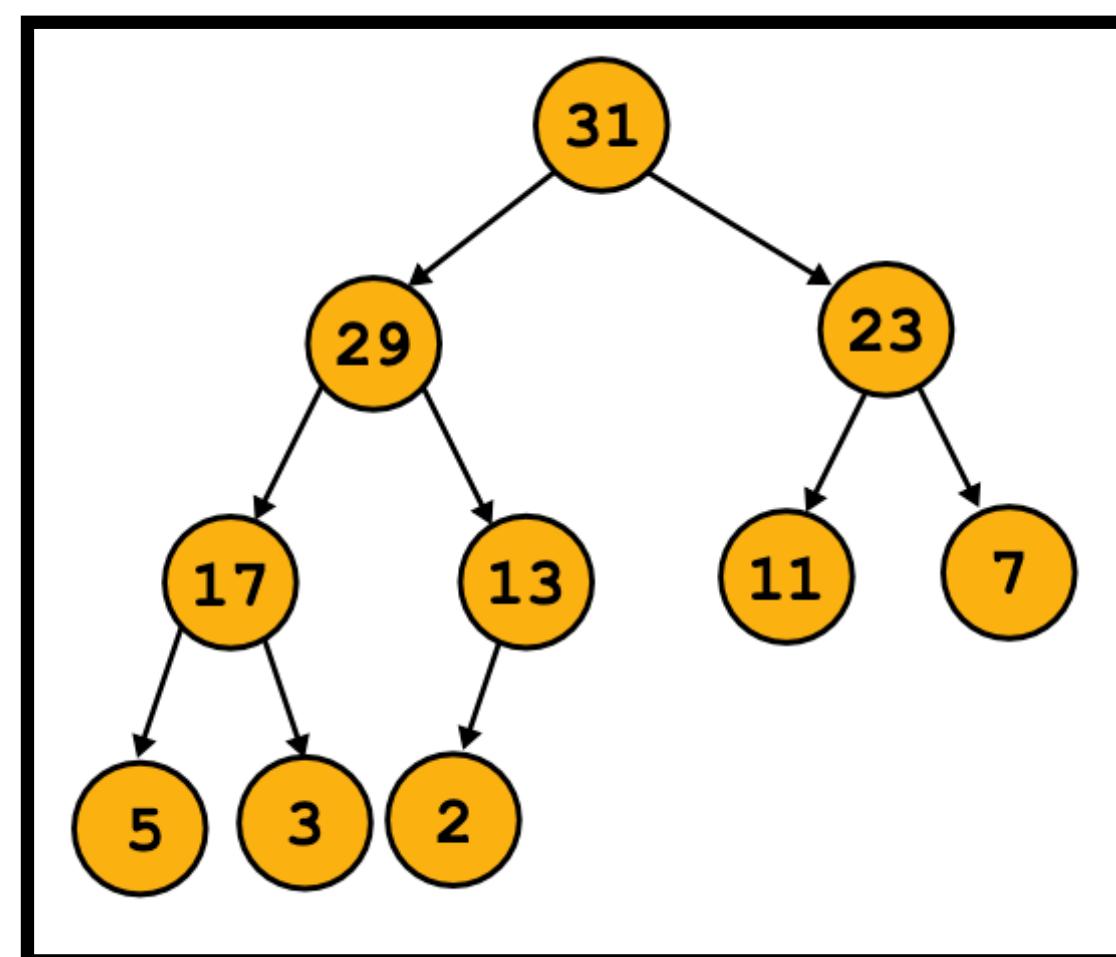
Assignment 6

- push / insert
 - append new element
 - percolate / bubble up



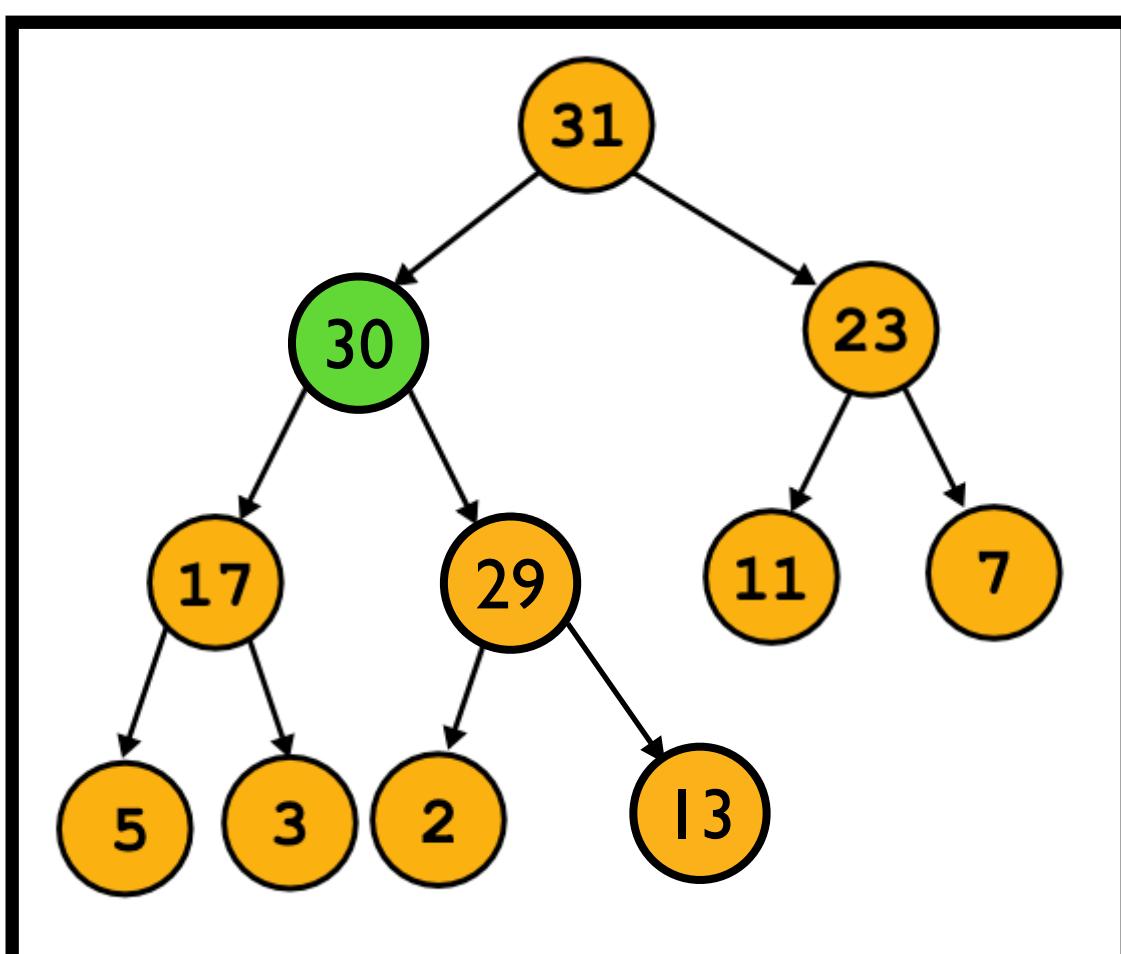
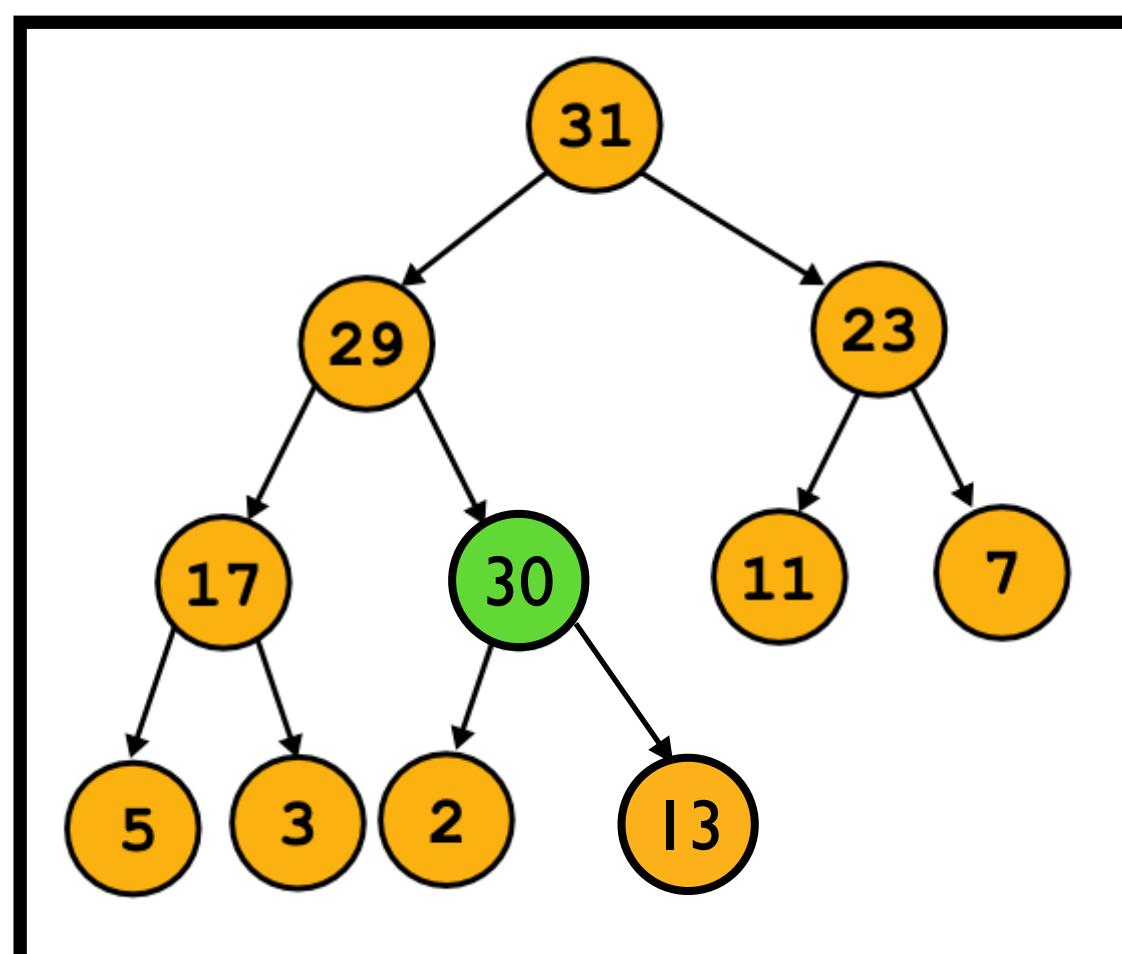
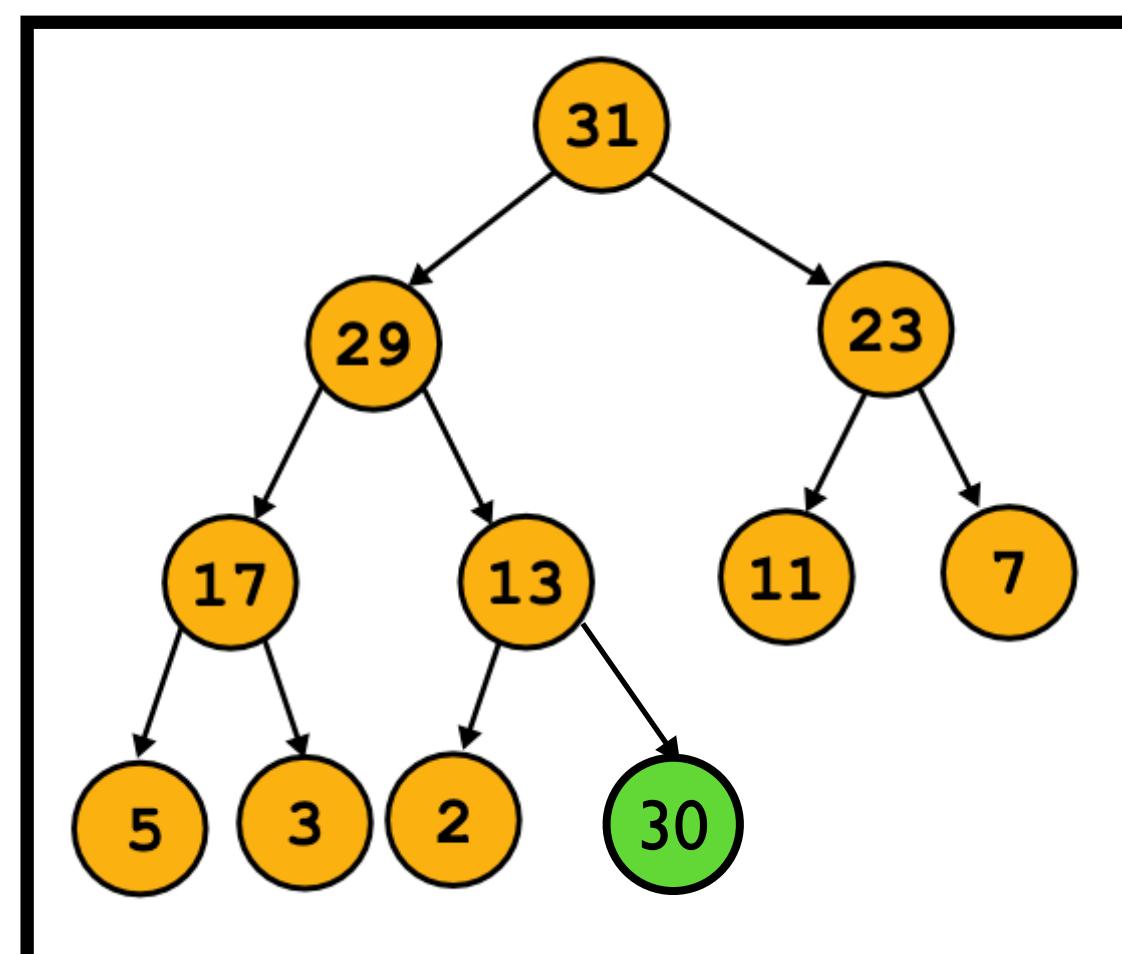
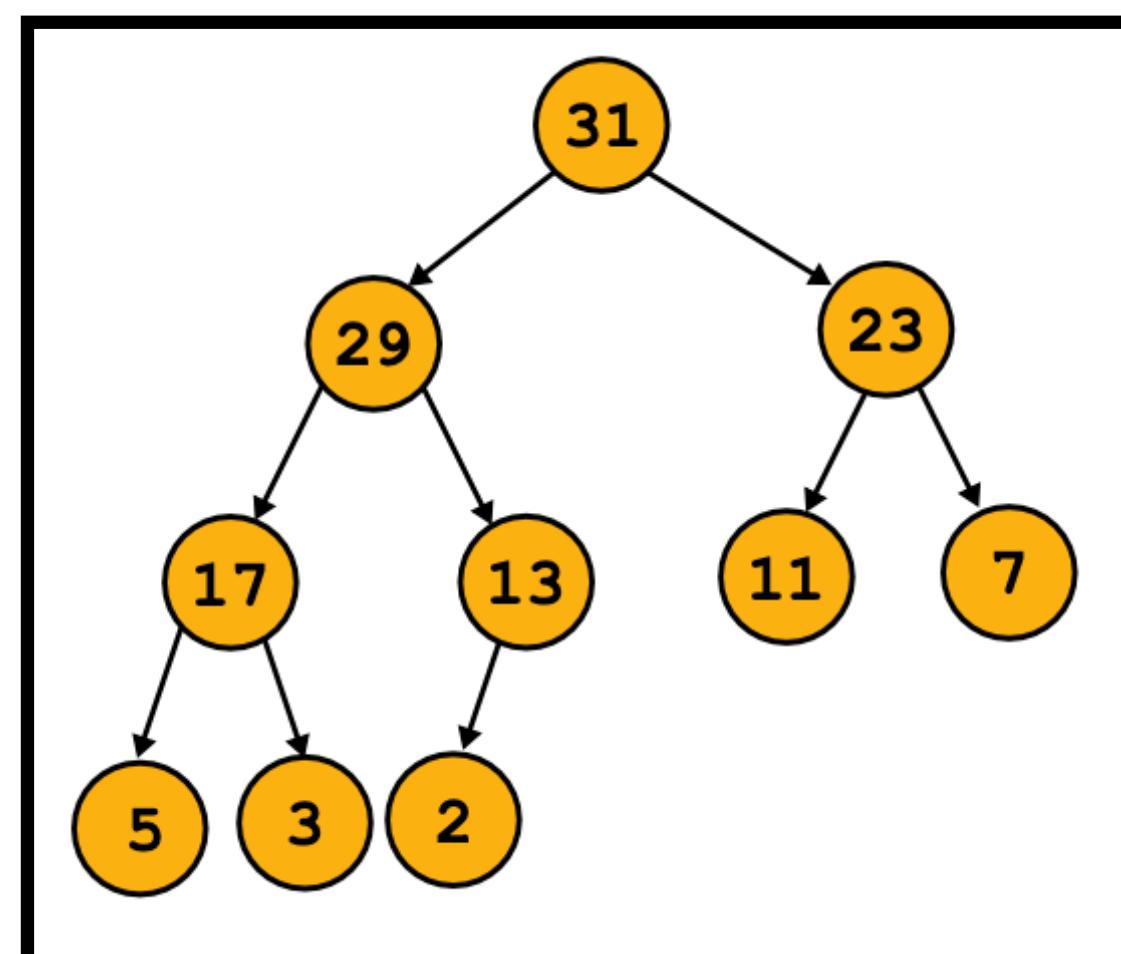
Assignment 6

- push / insert
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 - percolate / bubble up



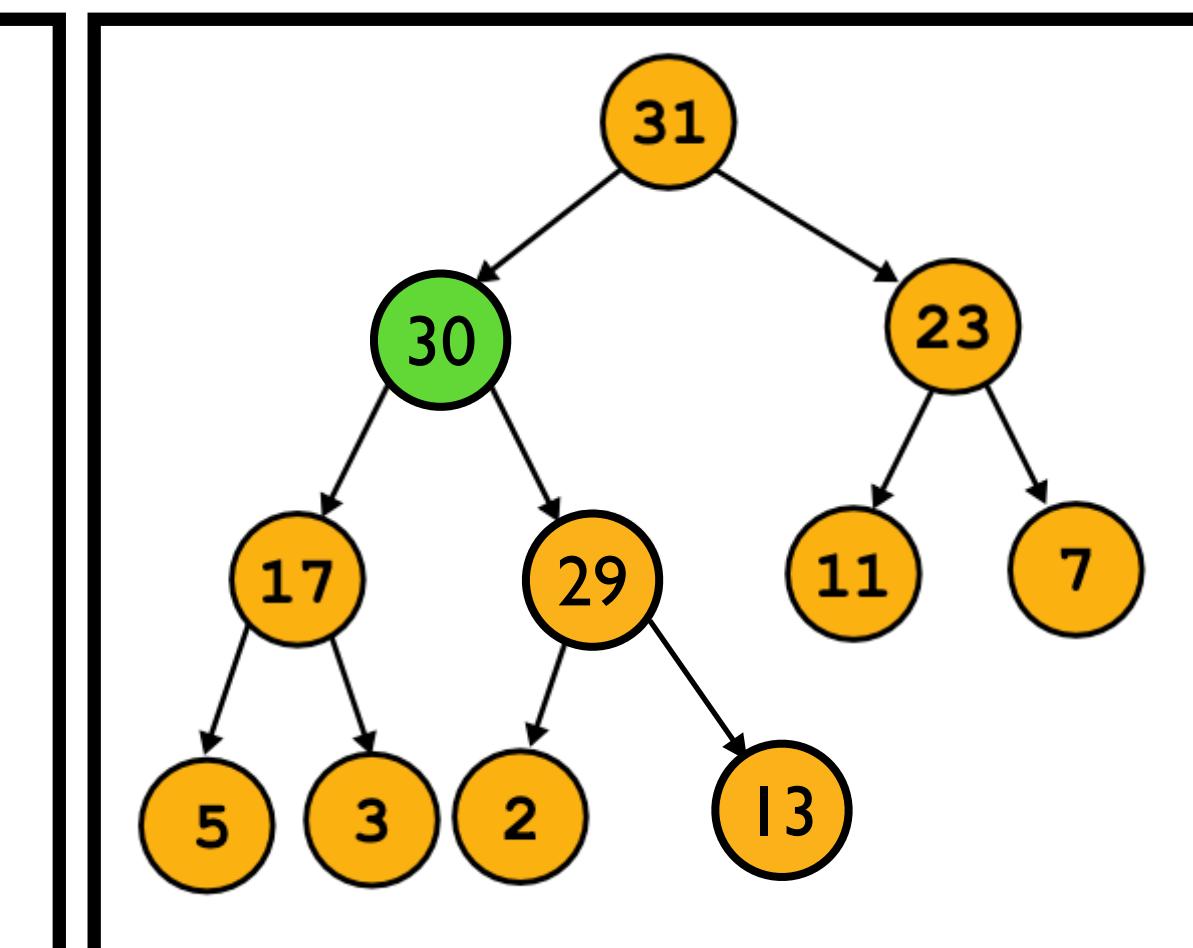
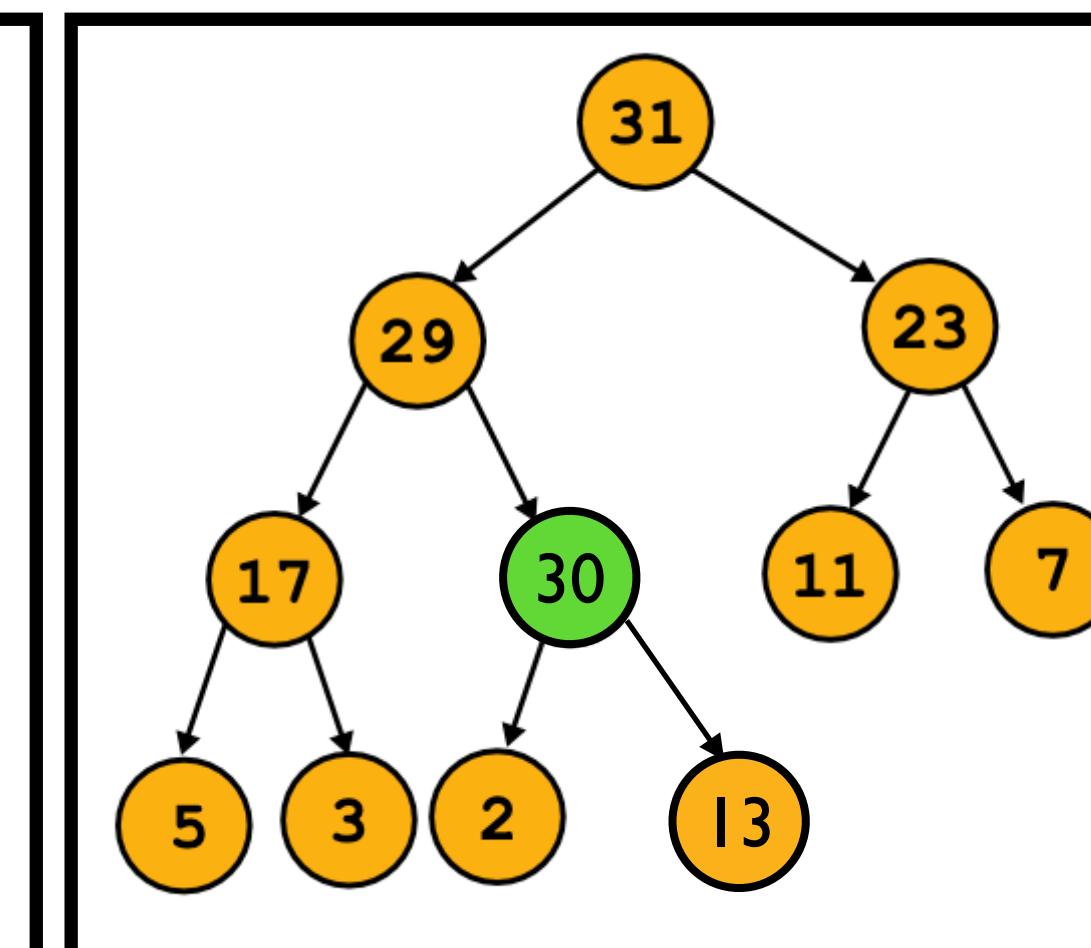
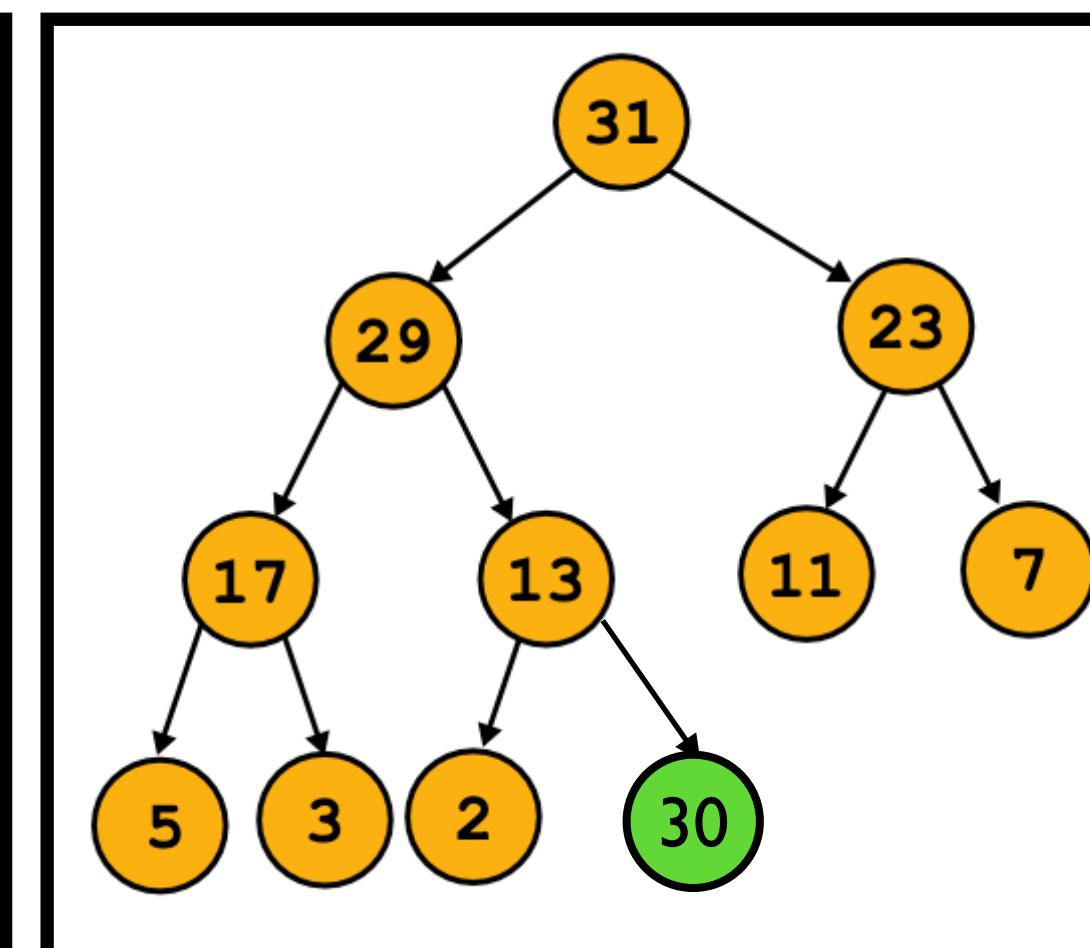
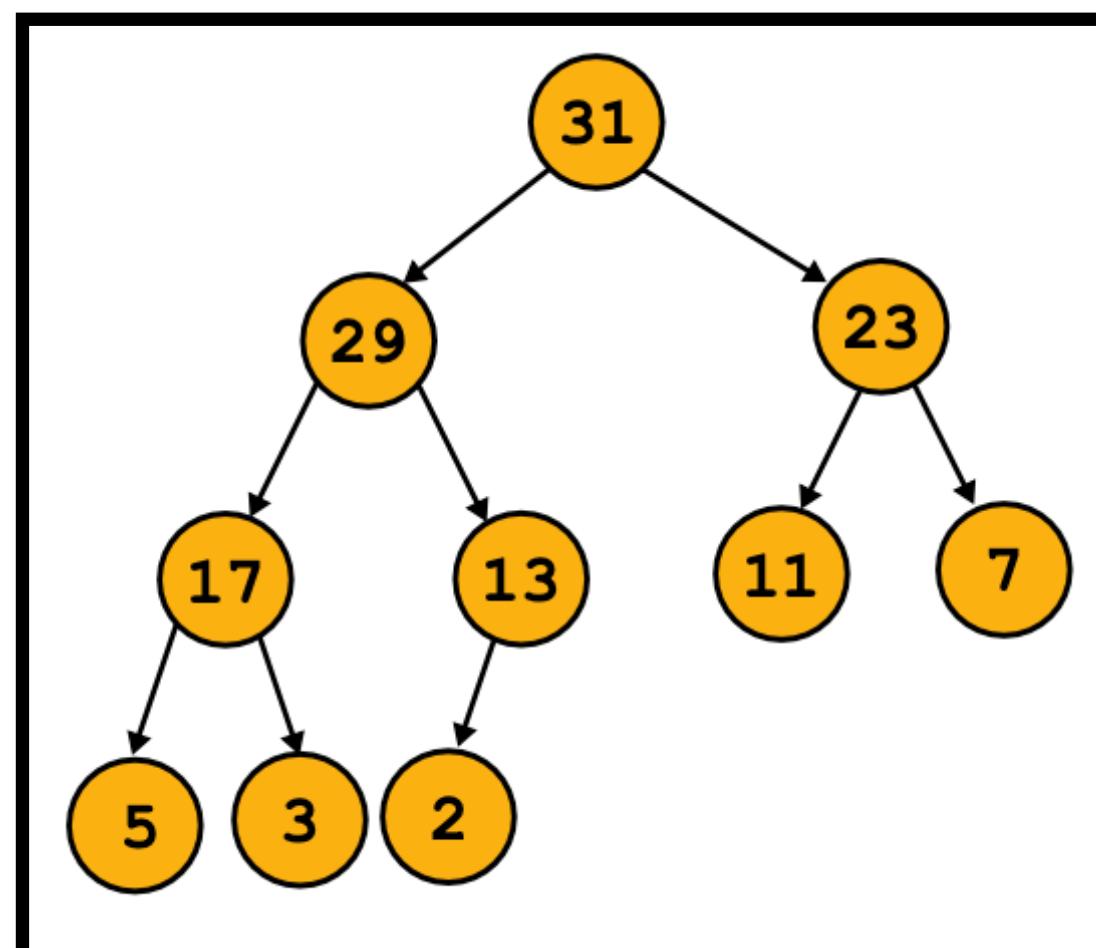
Assignment 6

- push / insert
 - append new element
 - percolate / bubble up
 - current node is not the root
 - higher priority than parent



Assignment 6

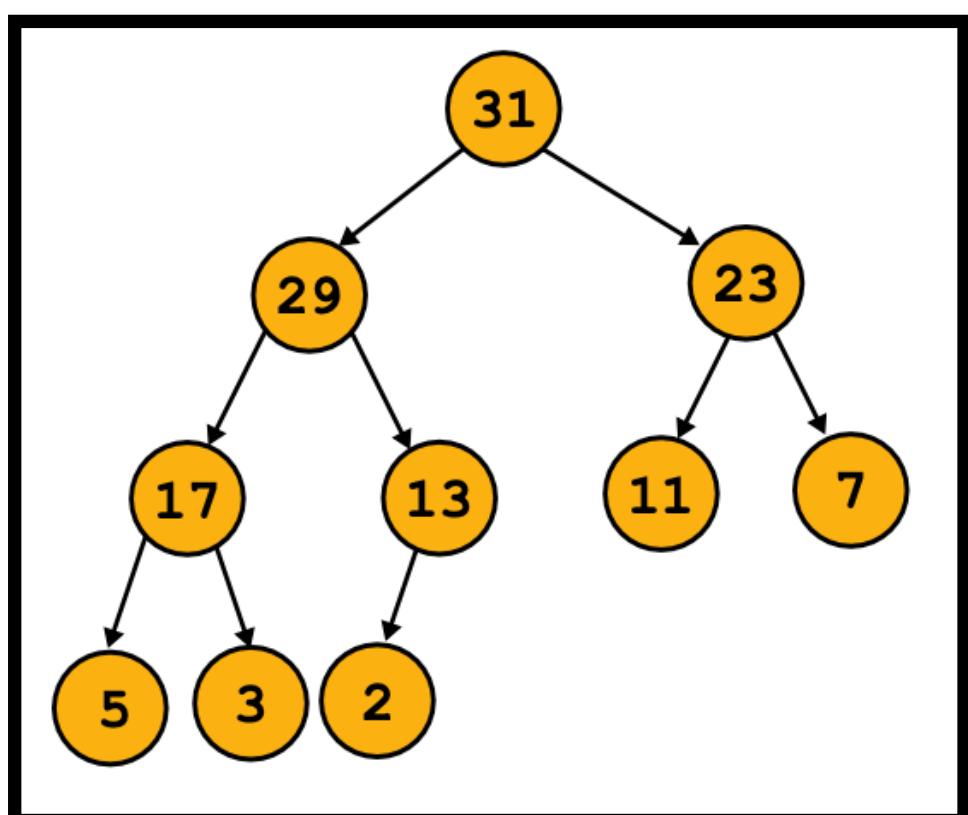
- push / insert
- append new element
- percolate / bubble up
 - current node is not the root
 - higher priority than parent



```
void pq_insert(struct pq* pq, void* data, int priority) {  
    struct element* new_element;  
  
    //Ensure pre-conditions are met  
    assert(pq);  
  
    //Create new element  
    new_element = (struct element*) malloc(sizeof(struct element));  
    new_element->priority = priority;  
    new_element->data = data;  
  
    //Add to end of priority queue  
    dynarray_insert(pq->array, _pq_length(pq), new_element);  
  
    //Percolate up from just-inserted element  
    _percolate_up(pq, _pq_length(pq) - 1);  
}
```

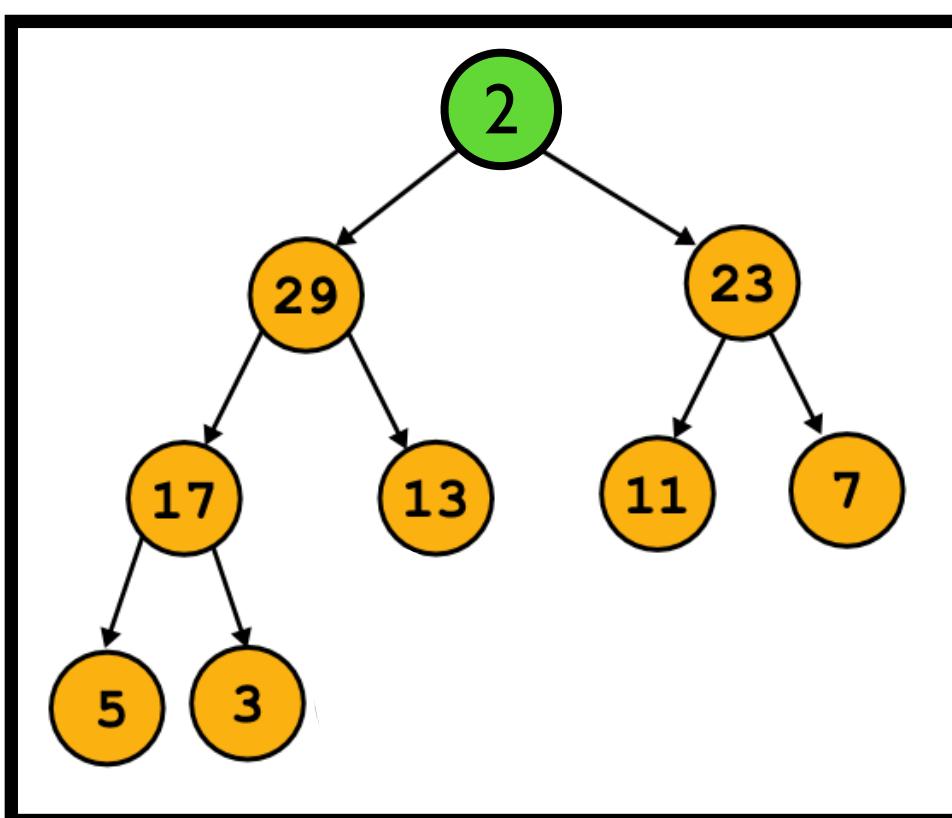
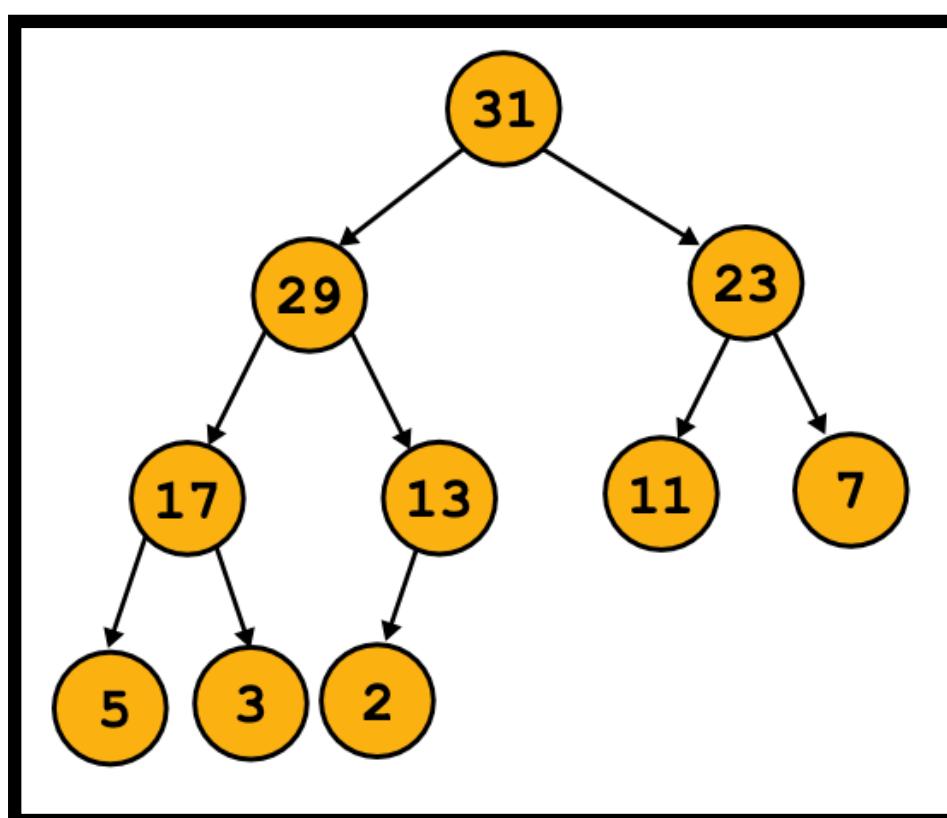
Assignment 6

- pop / dequeue
 - move the last element to the root
 - percolate / bubble down



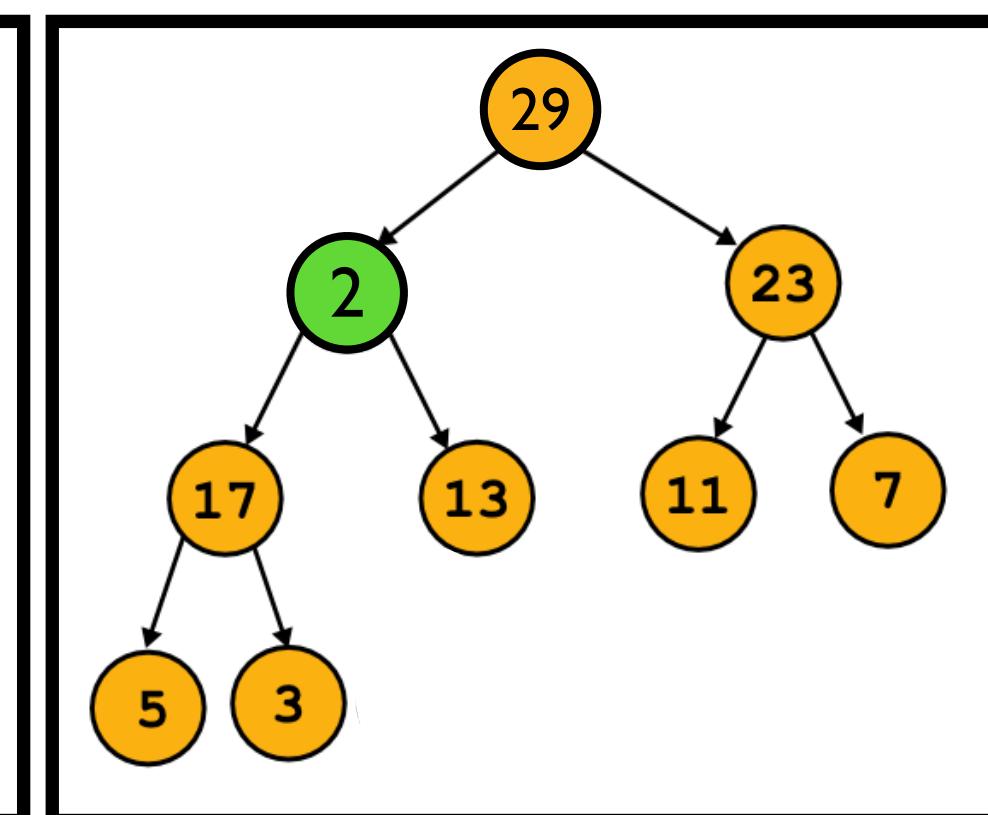
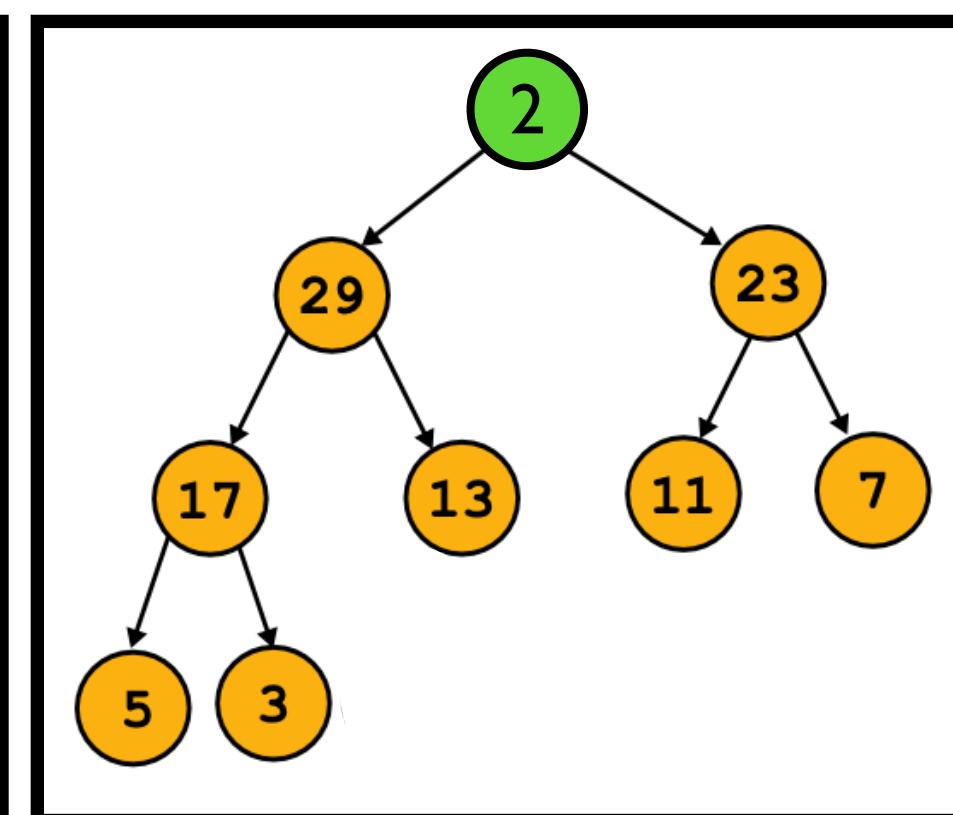
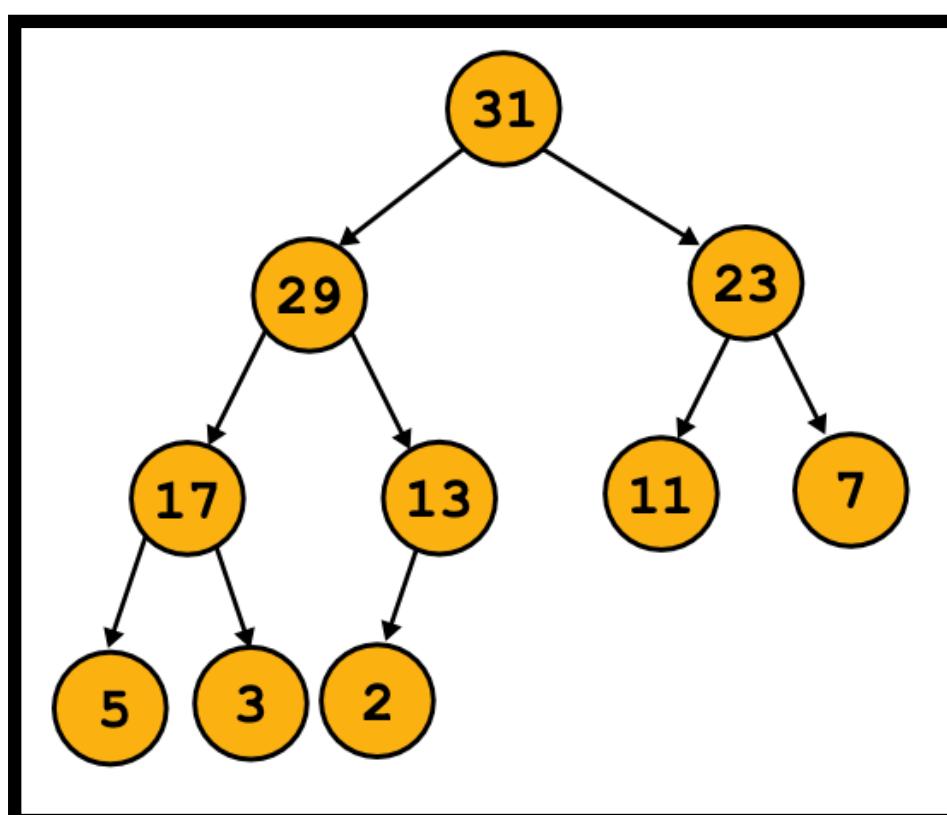
Assignment 6

- pop / dequeue
 - move the last element to the root
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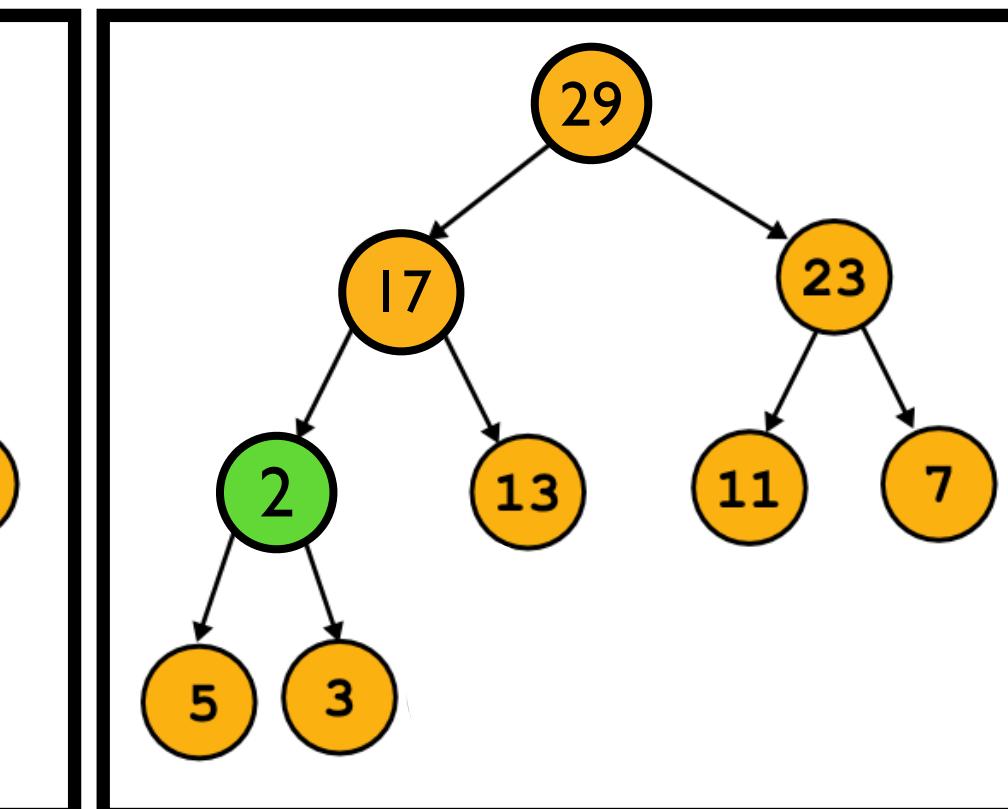
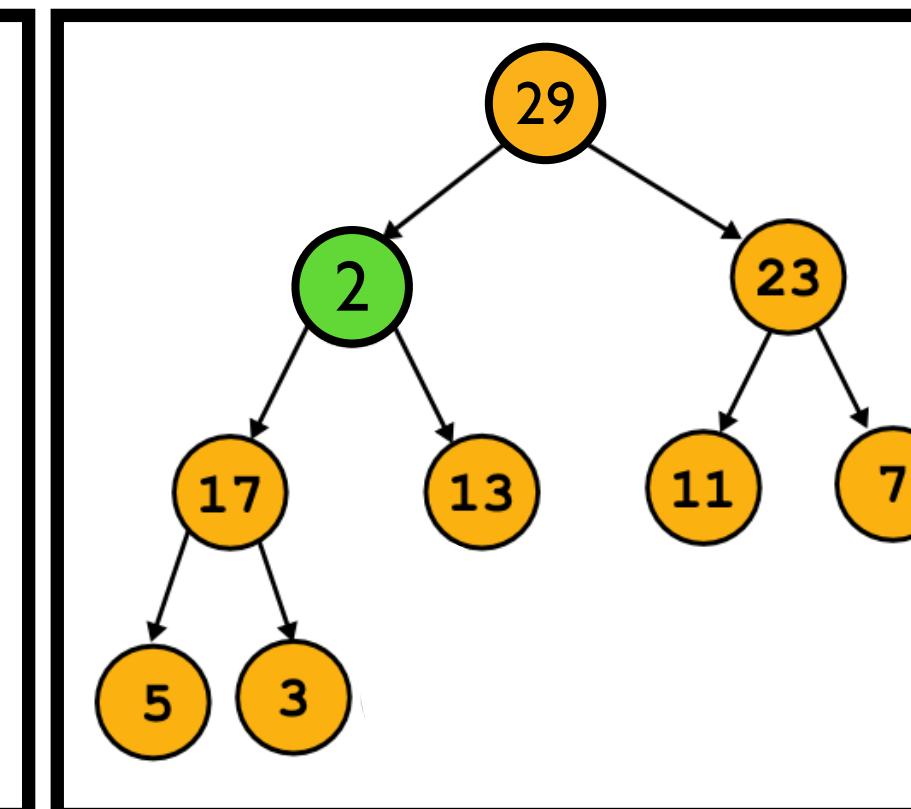
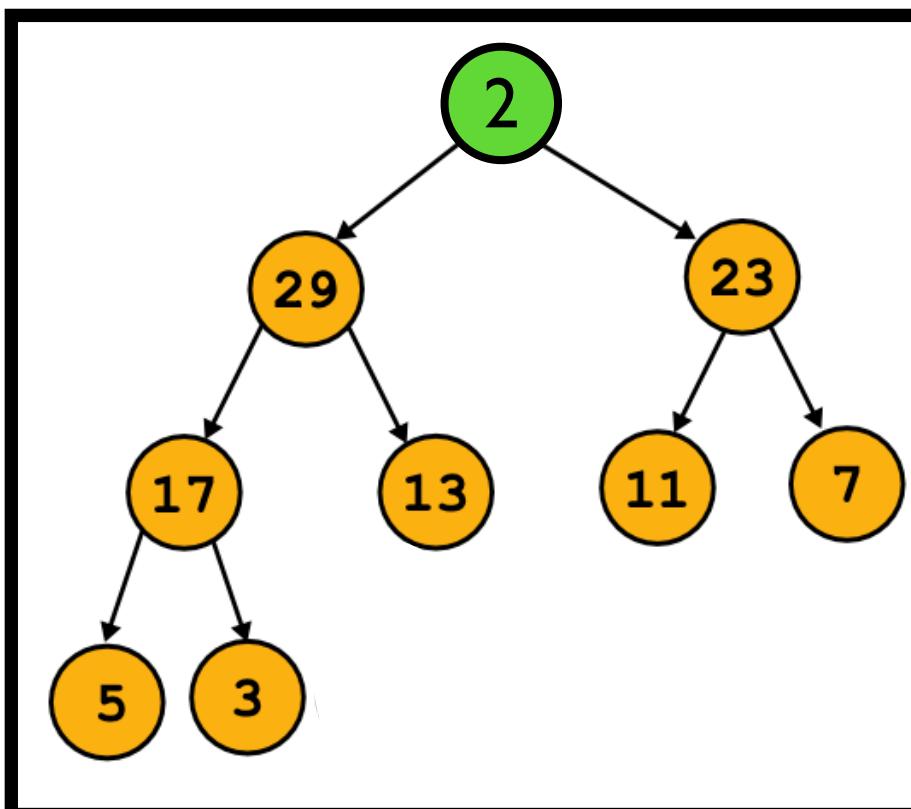
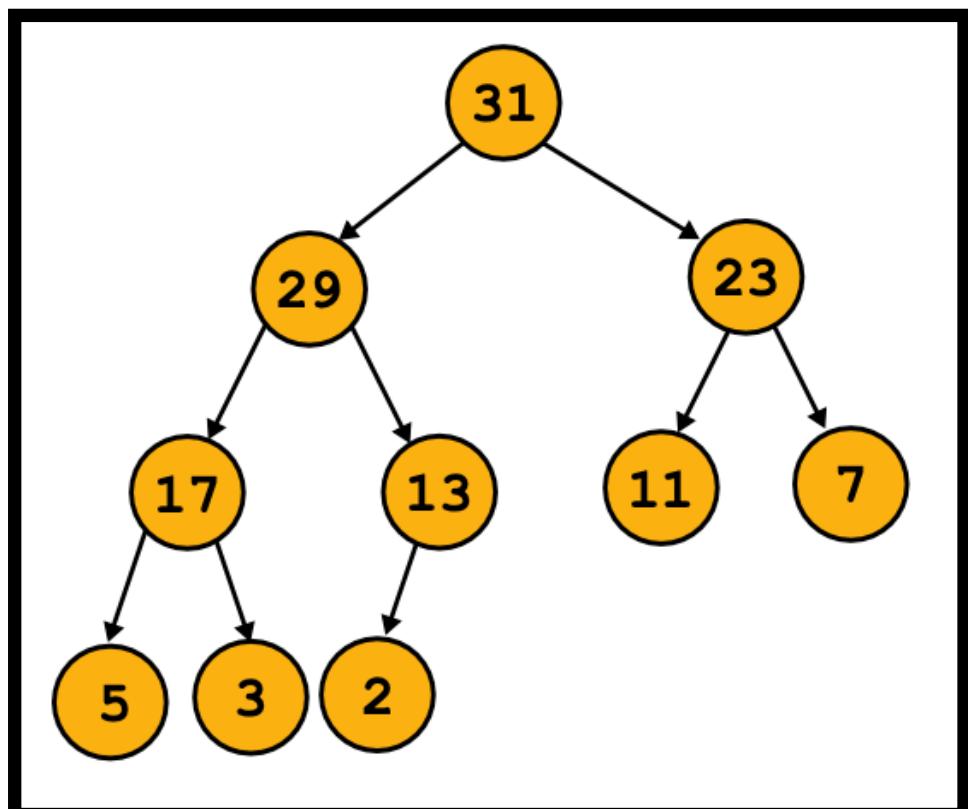
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- pop / dequeue
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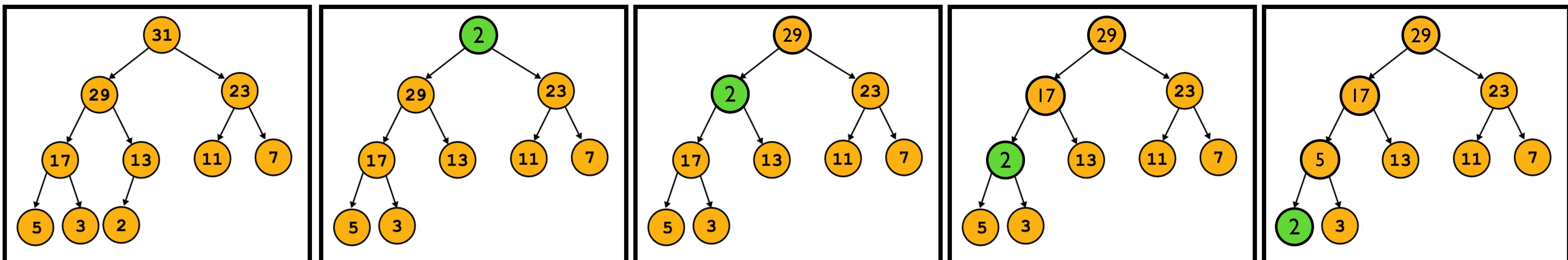
Assignment 6

- pop / dequeue
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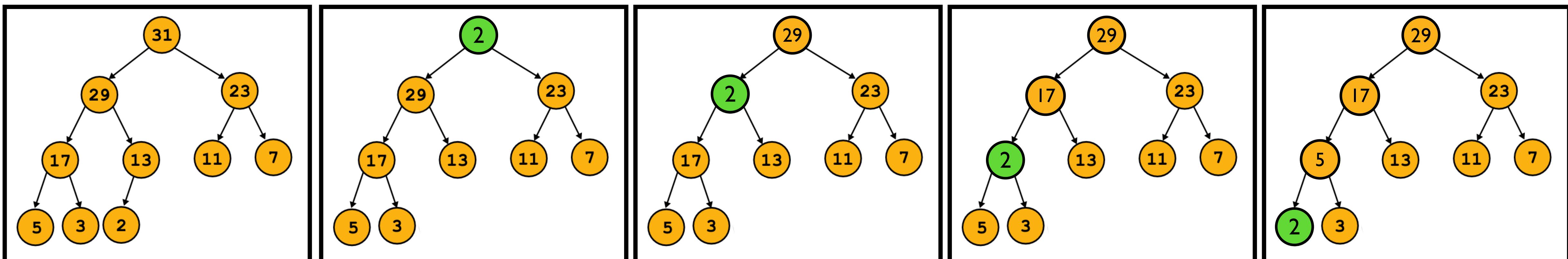
Assignment 6

- pop / dequeue
 - move the last element to the root
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Assignment 6

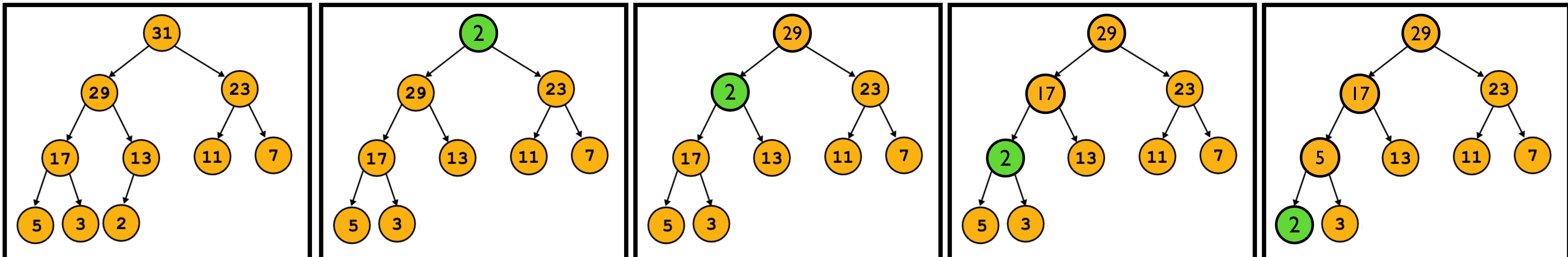
- pop / dequeue
 - move the last element to the root
 - percolate / bubble down
 - current node is not a leaf
 - lower priority than children



Assignment 6

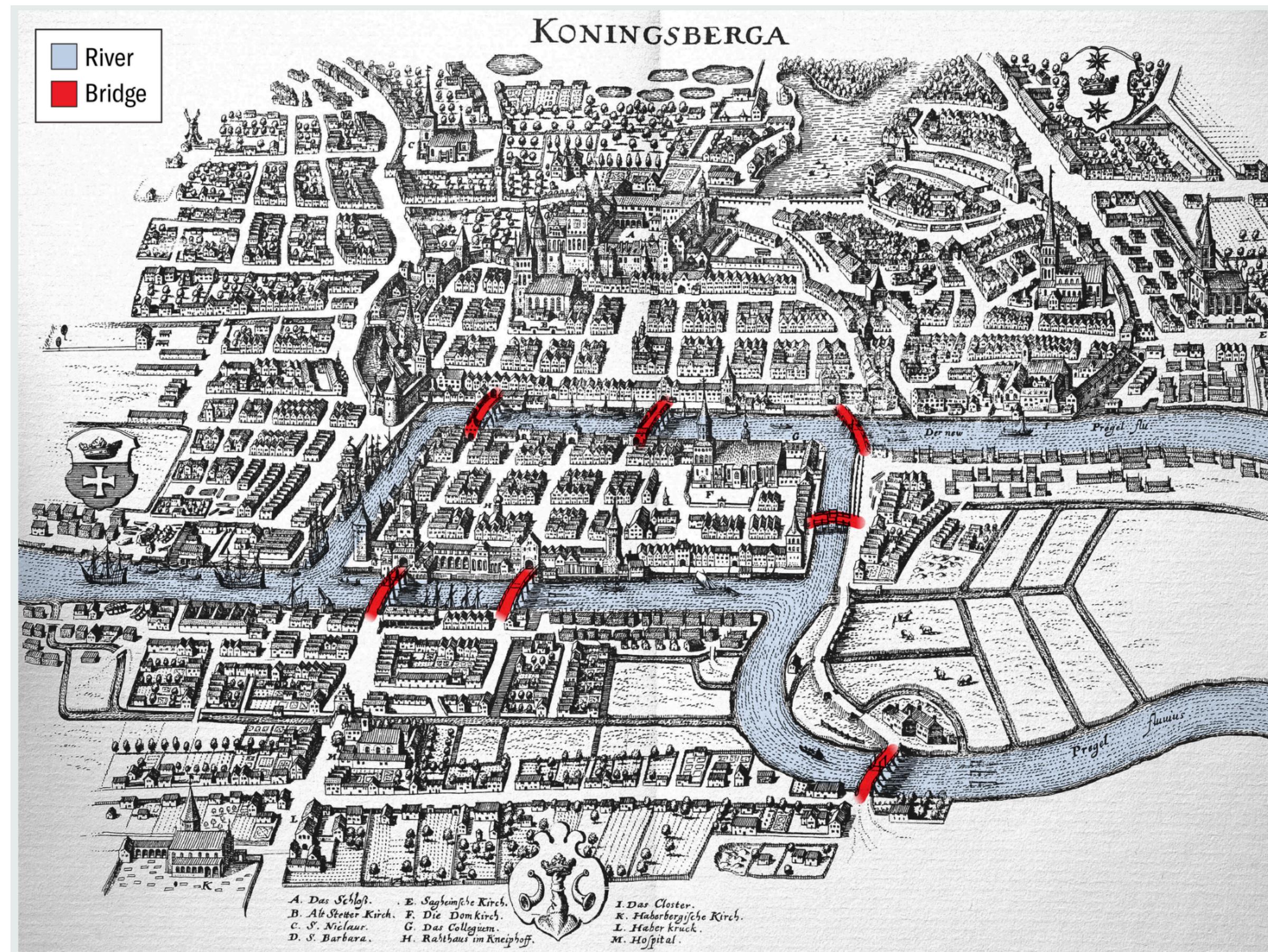
- pop / dequeue
 - move the last element to the root
 - percolate / bubble down
 - current node is not a leaf
 - lower priority than children

```
void* pq_max_dequeue(struct pq* pq) {  
    void* value = pq_max(pq);  
  
    //Ensure pre-conditions are met  
    assert(!pq_isempty(pq));  
  
    //Swap first and last elements  
    _swap(pq, 0, _pq_length(pq) - 1);  
  
    //Remove last element and free memory  
    free(dynarray_get(pq->array, _pq_length(pq) - 1));  
    dynarray_remove(pq->array, _pq_length(pq) - 1);  
  
    //Percolate down from root to maintain max-heap  
    _percolate_down(pq);  
  
    return value;  
}
```



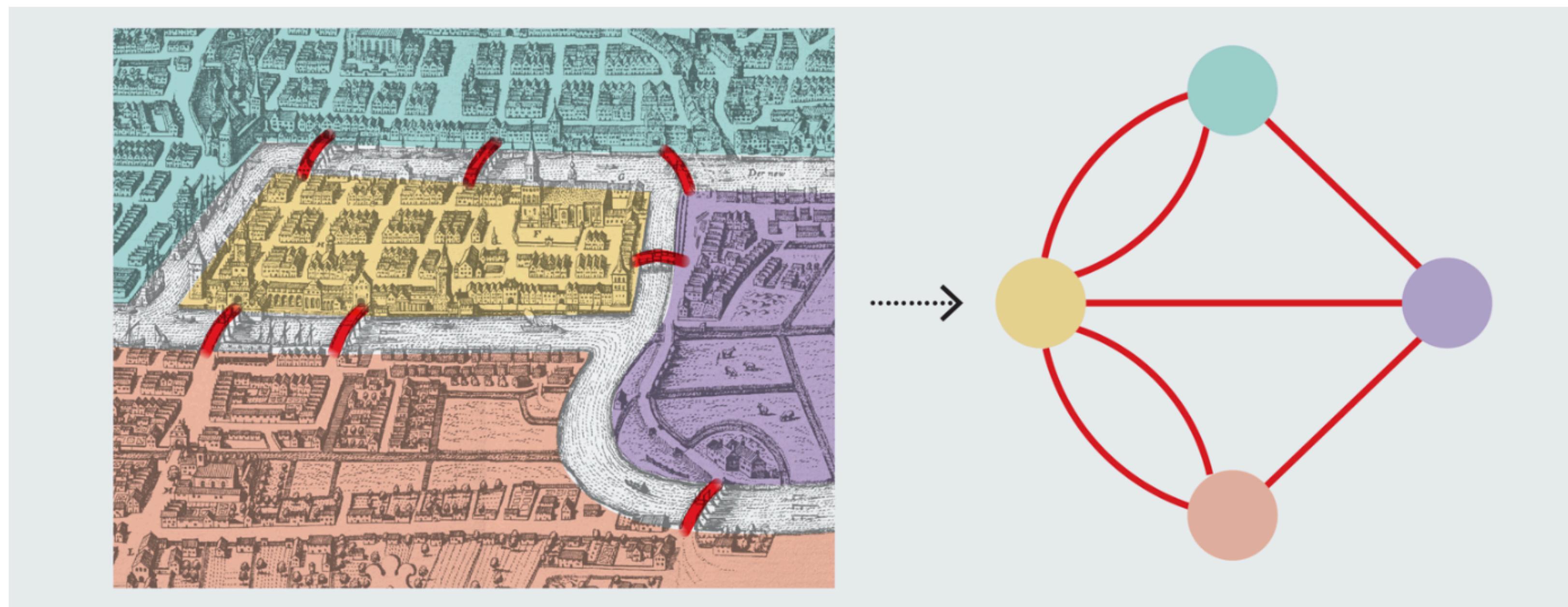
Graph

- Seven Bridges of Königsberg
- find a walking path that can cross each bridge once and only once



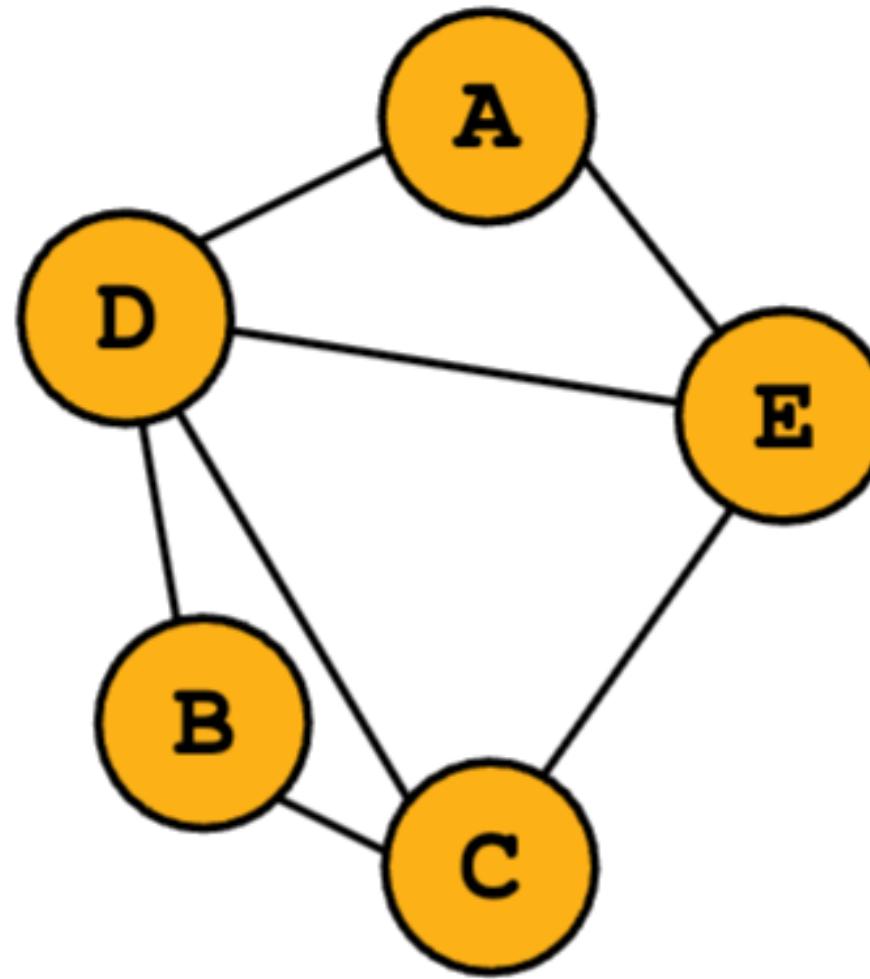
Graph

- Seven Bridges of Königsberg
 - find a walking path that can cross each bridge once and only once
 - Euler's abstraction
 - lands as nodes; bridges as edges



Graph

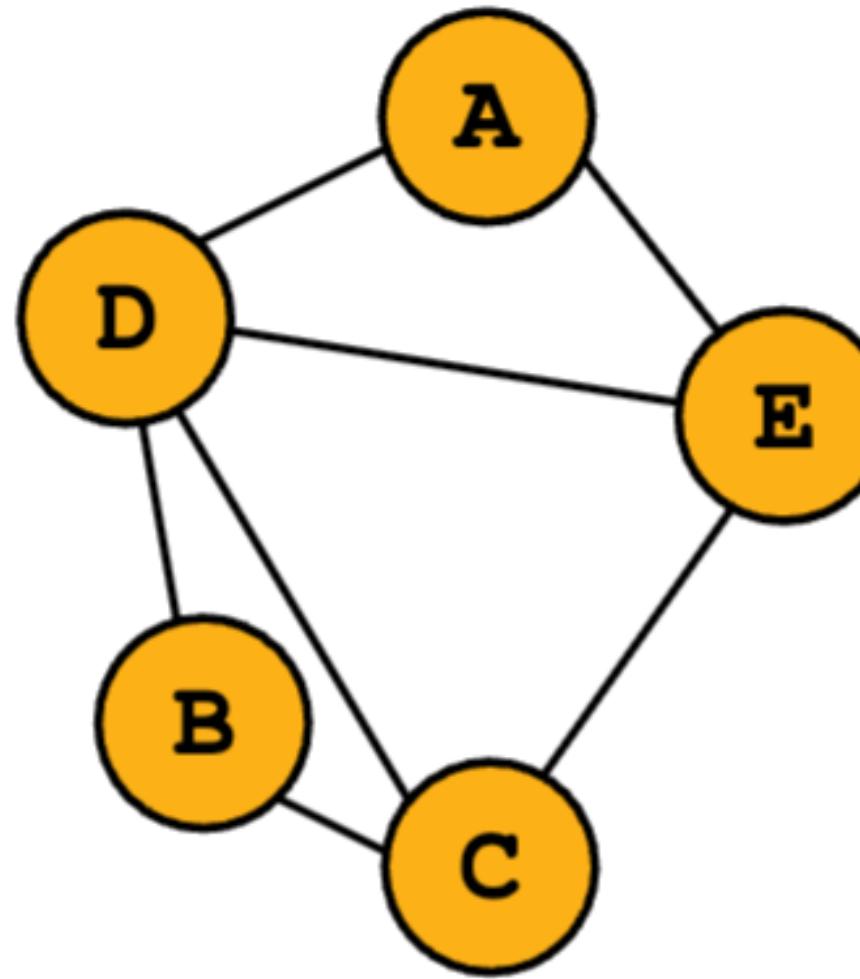
- $G = (V, E)$
- vertices: $V = \{v_1, v_2, \dots, v_n\}$
- edges: $E = \{(v_i, v_j) \mid v_i, v_j \in V\}$



$$V = \{A, B, C, D, E\}, \text{ and}$$
$$E = \{\{A, D\}, \{A, E\}, \{D, B\}, \{D, C\}, \{C, B\}, \{C, E\}\}$$

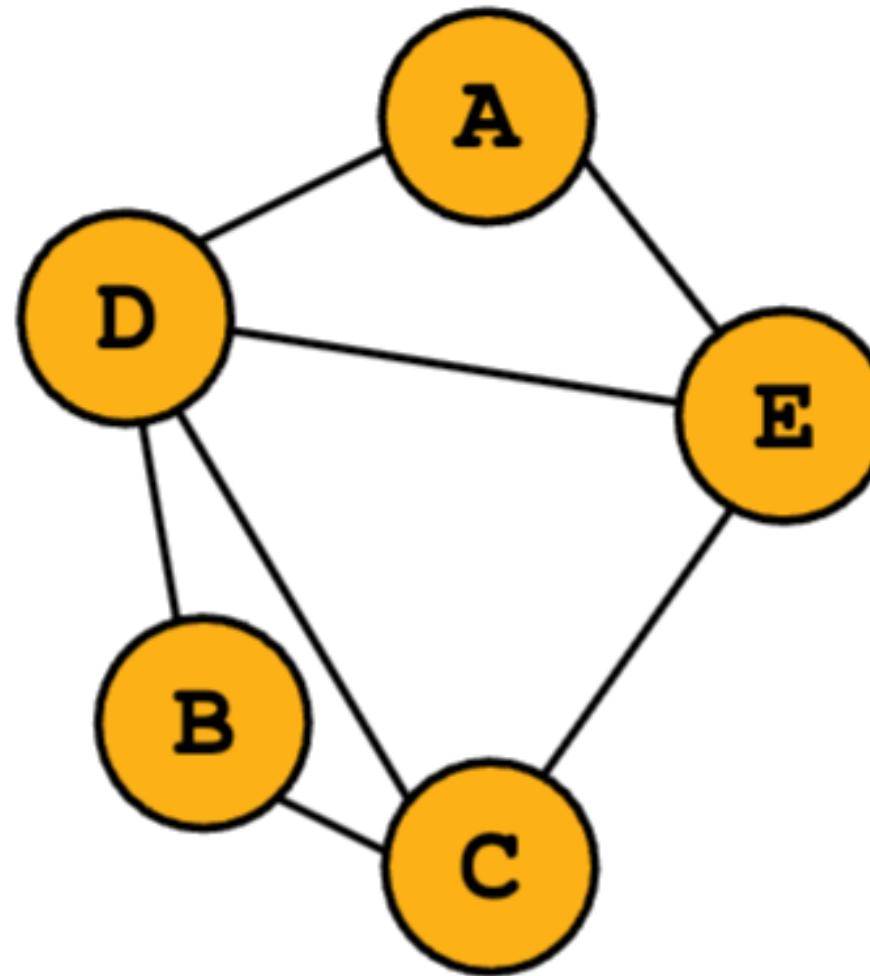
Graph

- examples of graph?



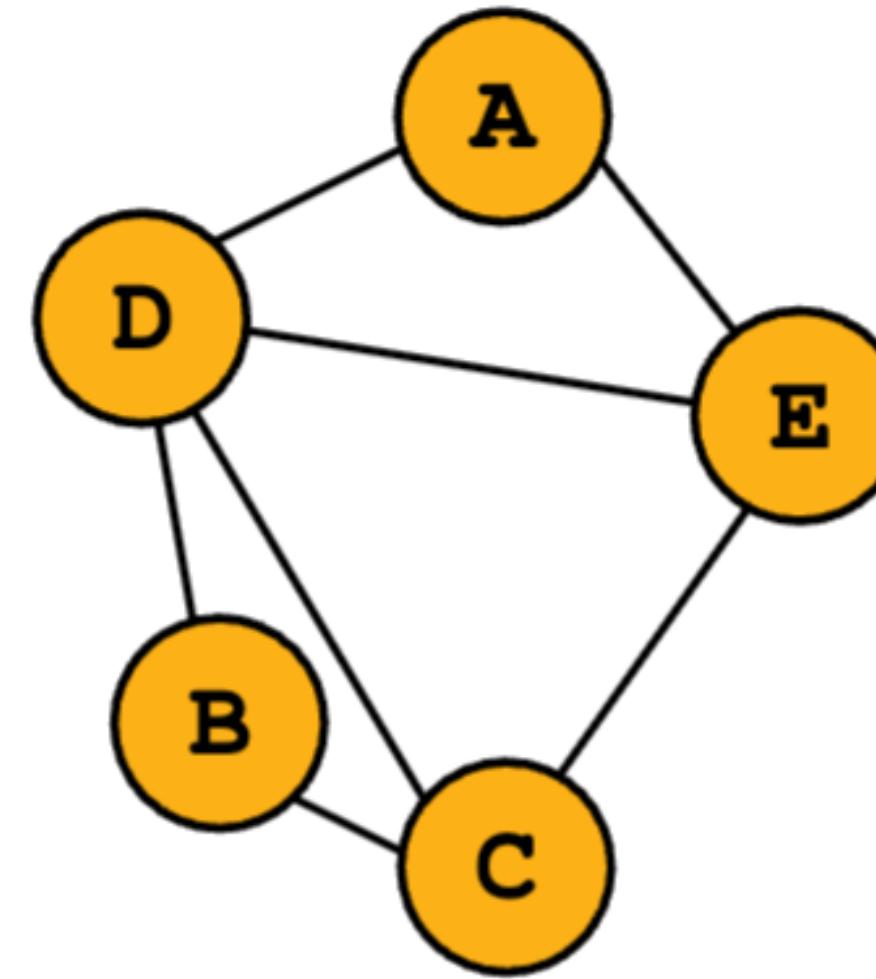
Graph

- examples of graph?
 - google map
 - flight networks
 - social networks
 - course prerequisites graph
 - software dependency graph
 - wikipedia linked documents



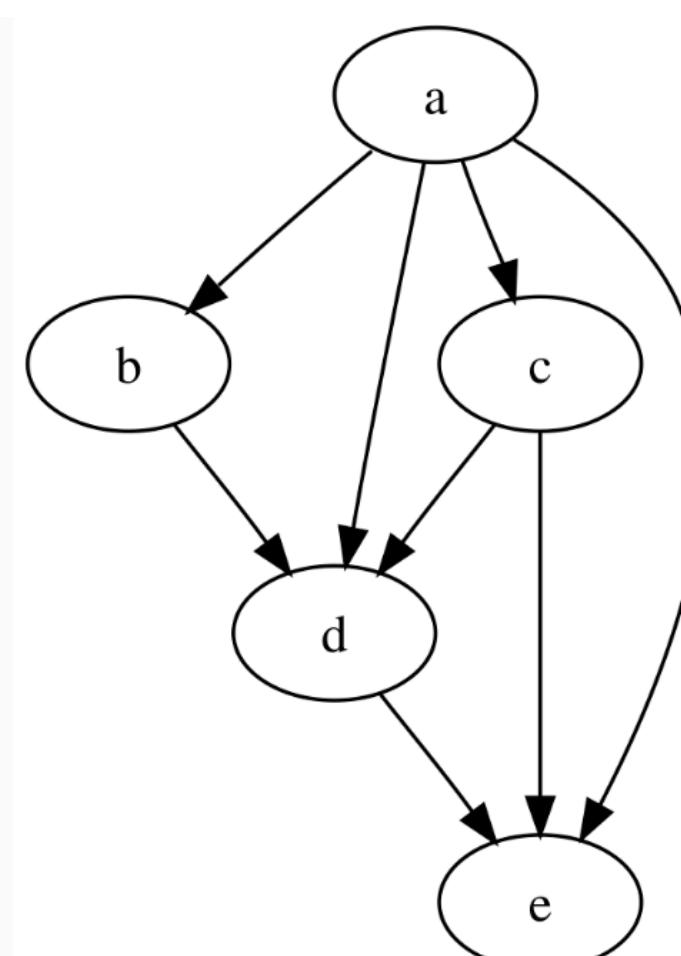
Graph Types

- many dimensions to describe graphs



Graph Types

- many dimensions to describe graphs
 - undirected vs. directed
 - Facebook vs. Twitter
 - sparse vs. dense
 - dense: $|E| \sim |V|^2$ (rarely)
 - sparse: $|E| \ll |V|^2$ (most graphs)
 - six degree of separation (due to hubs)
 - Dunbar's number (~ 150)
 - acyclic vs. cyclic
 - DAG: Directed Acyclic Graph
 - course & software dependence graph



Graph Types

- many dimensions to describe graphs (cont.)
 - connected vs. disconnected
 - undirected:
 - directed: strongly vs. weakly connected
 - strongly connected component (SCC)
 - SCC-DAG

Graph Types

- many dimensions to describe graphs (cont.)

- connected vs. disconnected

- undirected:
- 

- directed: strongly vs. weakly connected

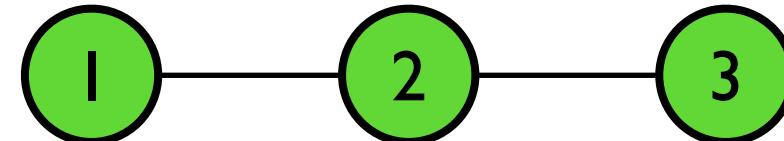
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- SCC-DAG

Graph Types

- many dimensions to describe graphs (cont.)

- connected vs. disconnected

- undirected:

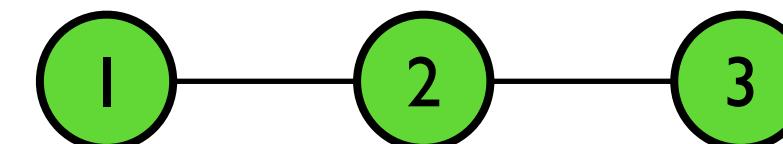
- directed: strongly vs. weakly connected

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- SCC-DAG

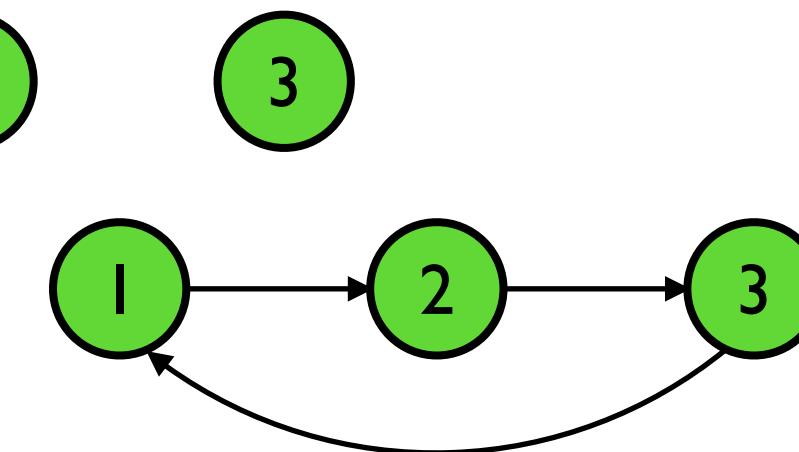
Graph Types

- many dimensions to describe graphs (cont.)

- connected vs. disconnected

- undirected:

- directed: strongly vs. weakly connected



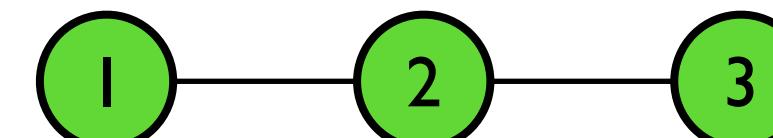
- strongly connected component (SCC)

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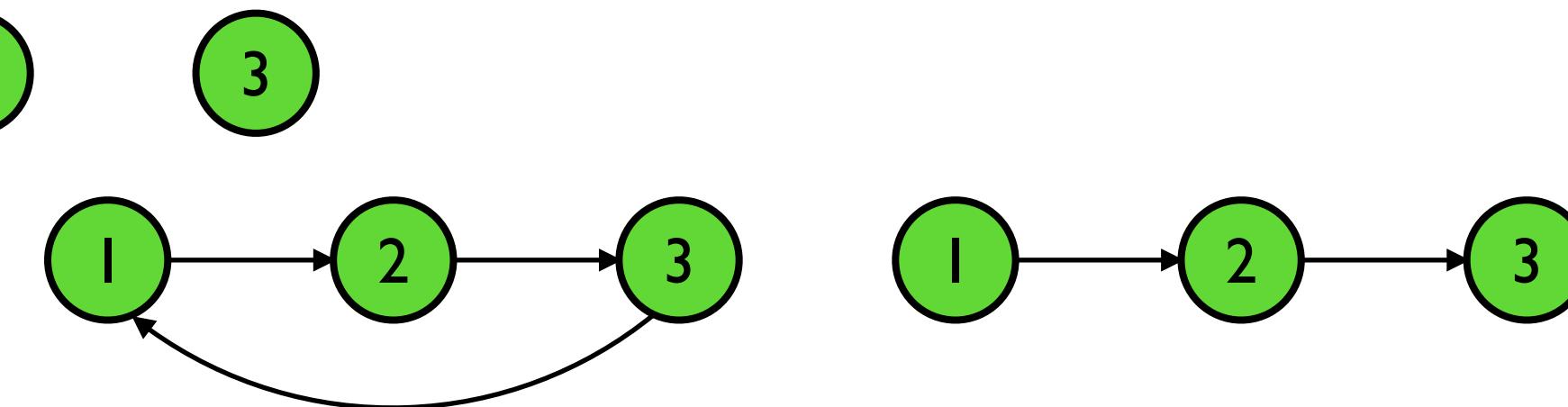
Graph Types

- many dimensions to describe graphs (cont.)

- connected vs. disconnected

- undirected:

- directed: strongly vs. weakly connected



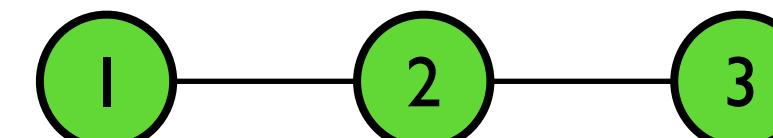
- strongly connected component (SCC)

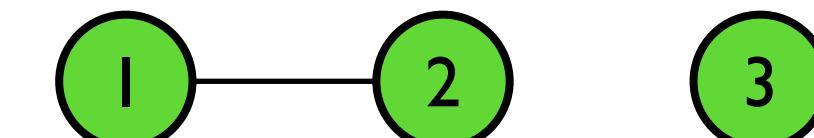
- SCC-DAG

Graph Types

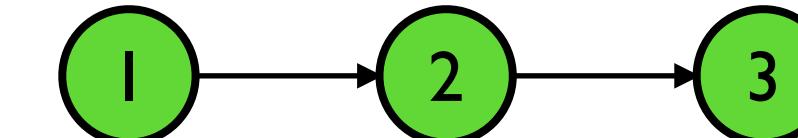
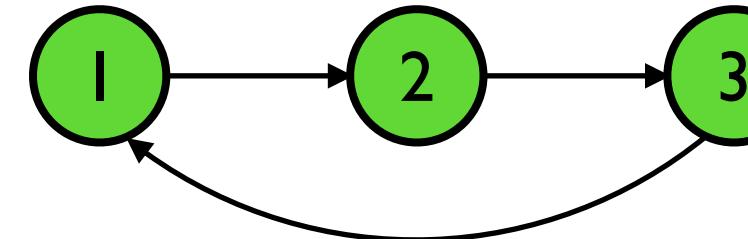
- many dimensions to describe graphs (cont.)

- connected vs. disconnected

- undirected:

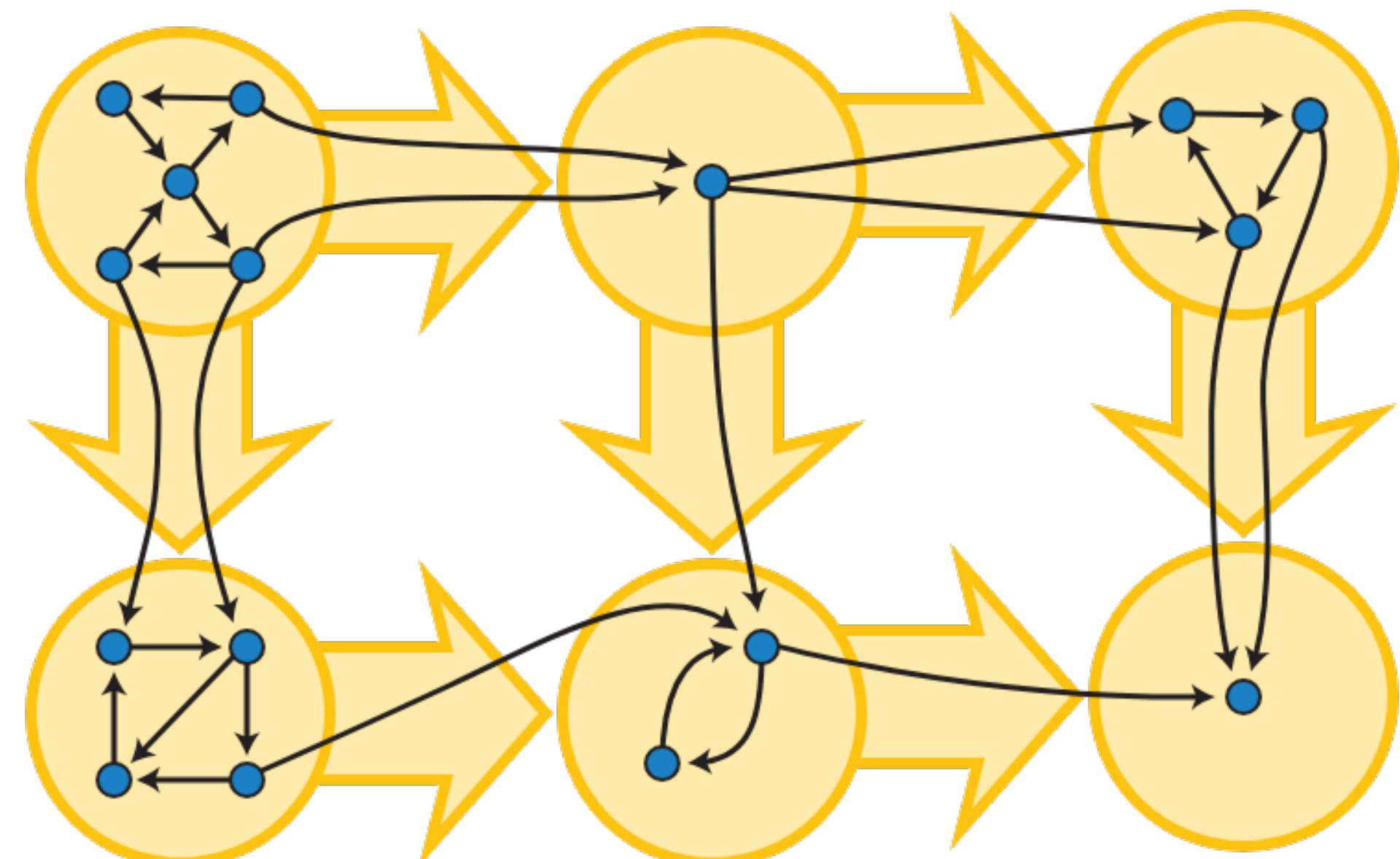


- directed: strongly vs. weakly connected



- strongly connected component (SCC)

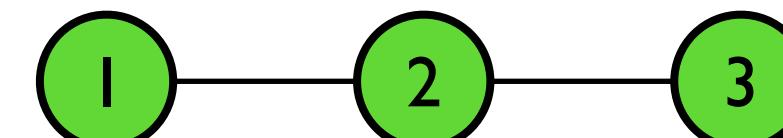
- SCC-DAG



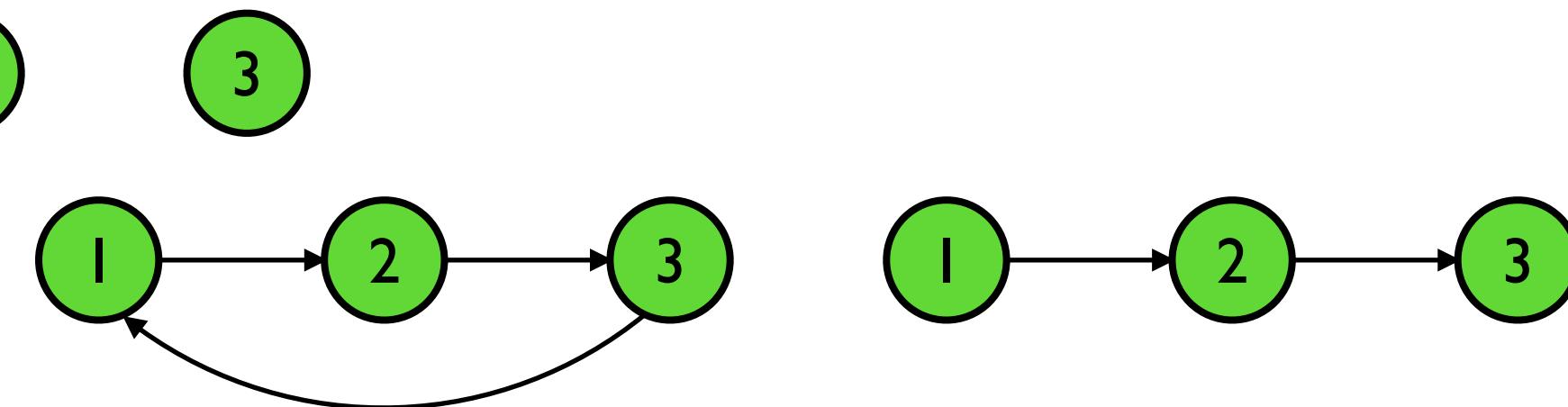
Graph Types

- many dimensions to describe graphs (cont.)

- connected vs. disconnected

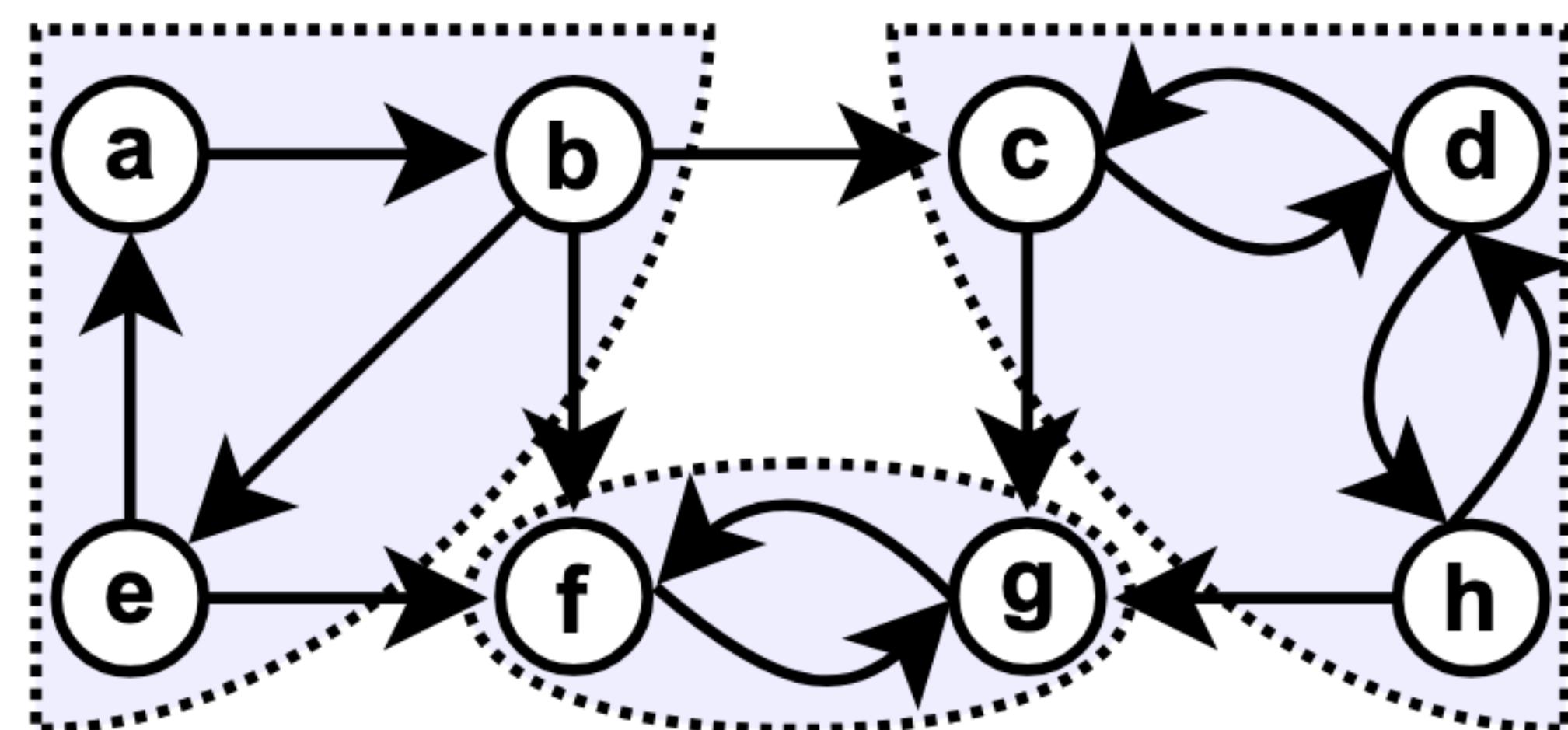
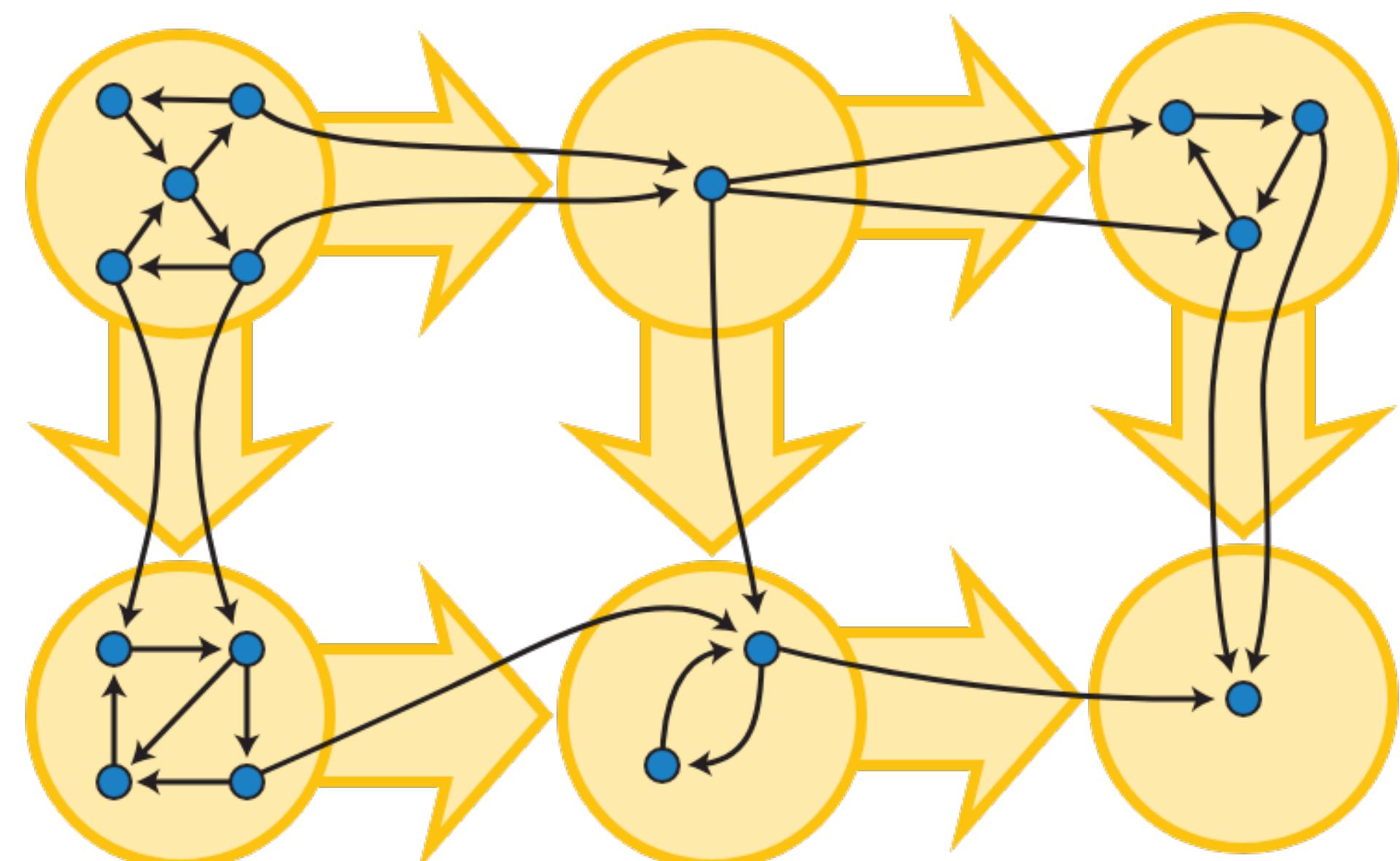
- undirected:

- directed: strongly vs. weakly connected



- strongly connected component (SCC)

- SCC-DAG

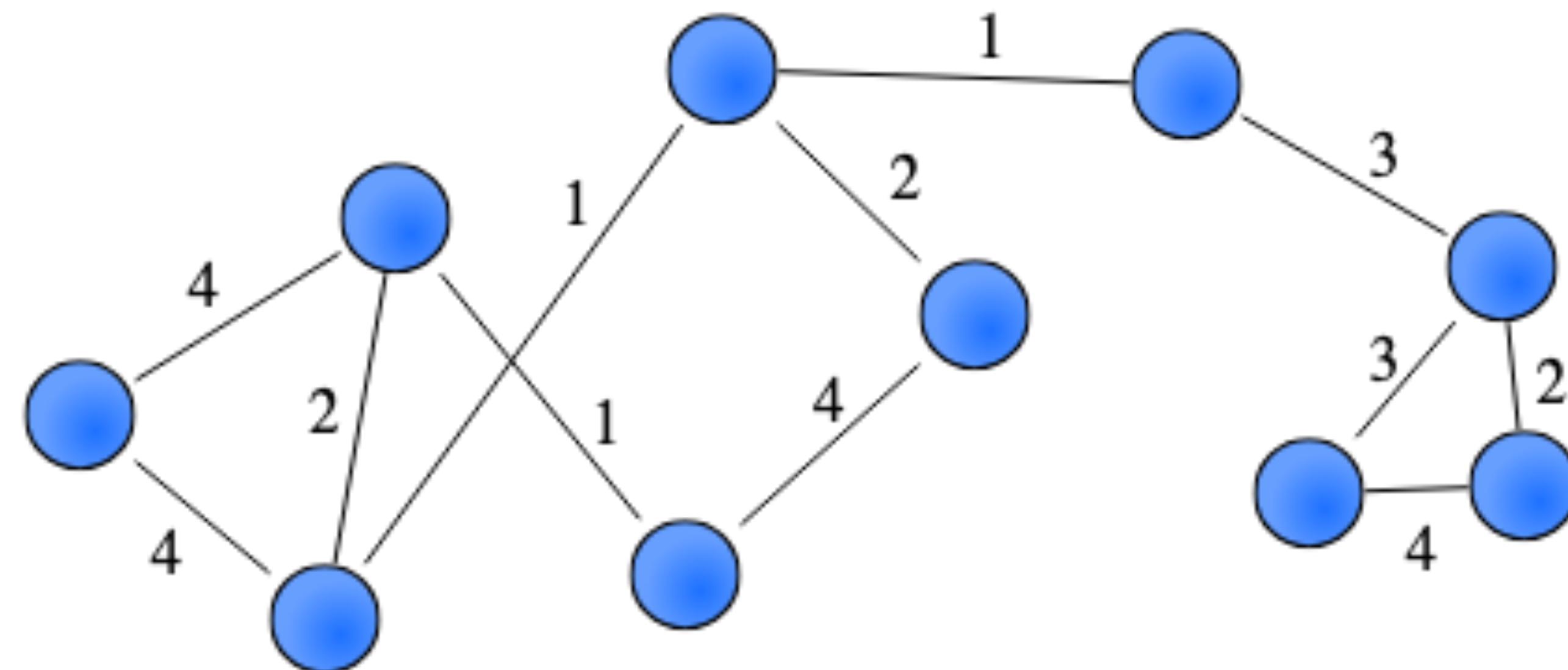


Graph Types

- many dimensions to describe graphs (cont.)

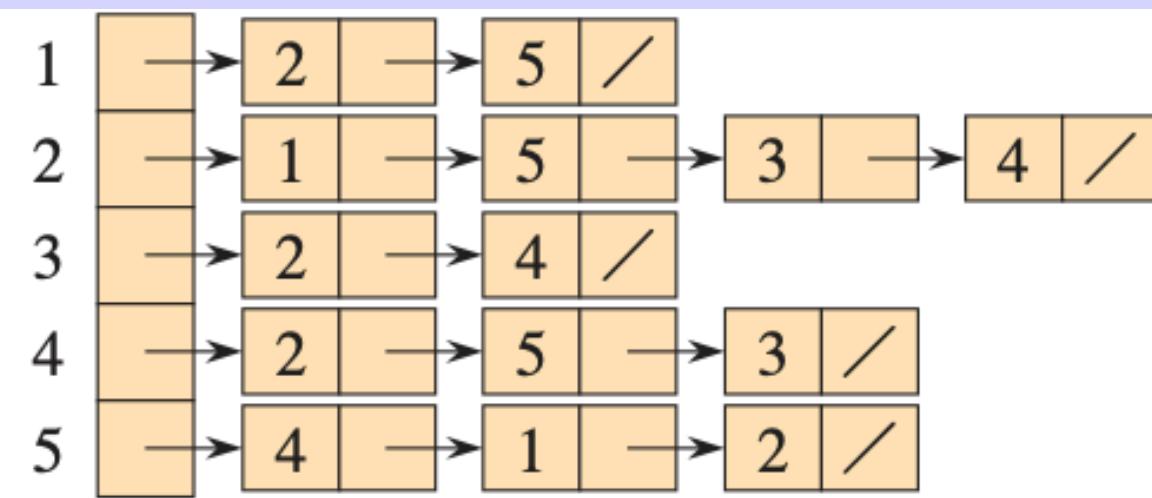
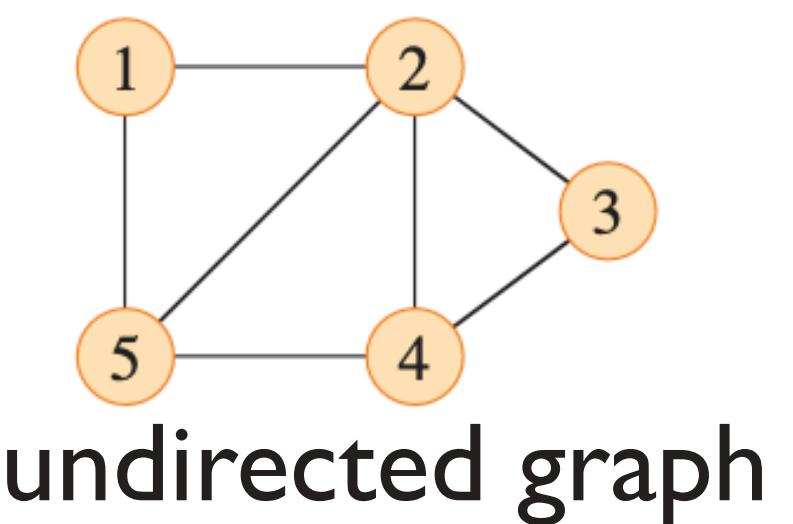
- unweighted vs. weighted

- vertices: $V = \{v_1, v_2, \dots, v_n\}$
- edges: $E = \{(v_i, v_j) \mid v_i, v_j \in V\}$
- weight function: $w : E \rightarrow \mathbb{R}$



Graph Representations

- $G = (V, E)$
- adjacency list
 - $Adj: |V|$ lists
 - $Adj[v]$: v 's neighbors



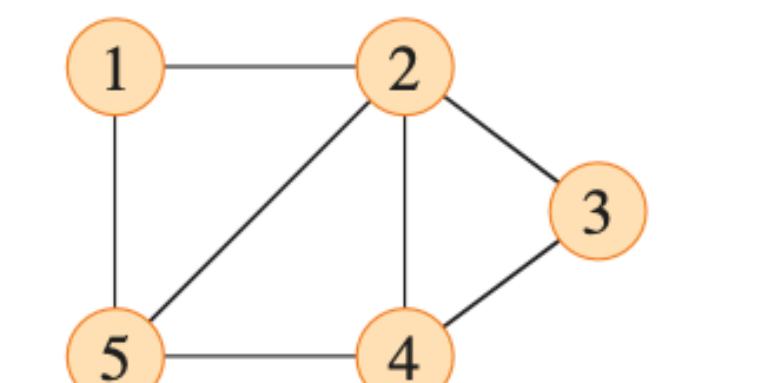
1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

- adjacency matrix
 - $A = (a_{ij})$: $|V| \times |V|$ matrix

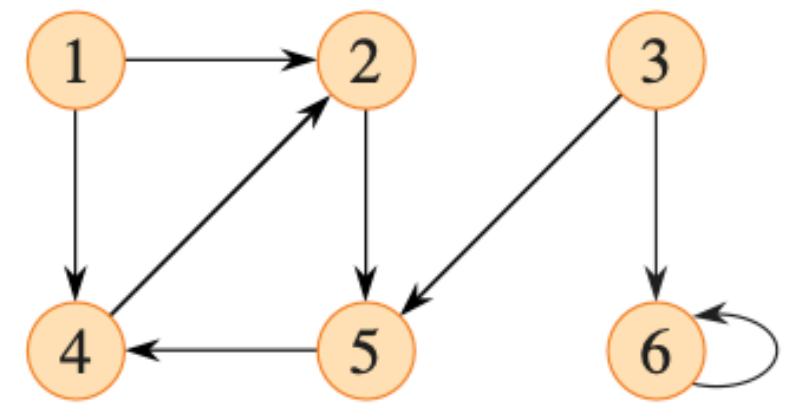
$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

Graph Representations

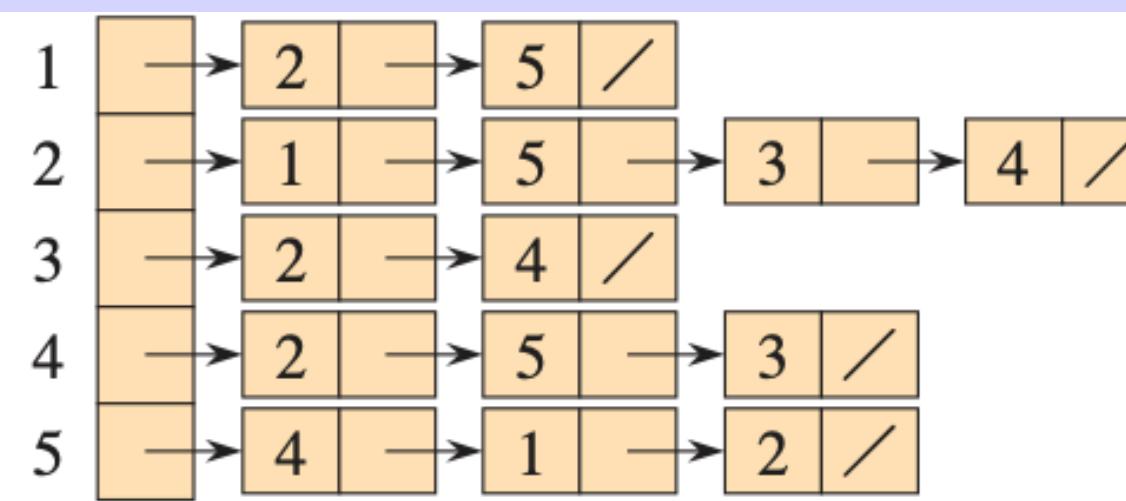
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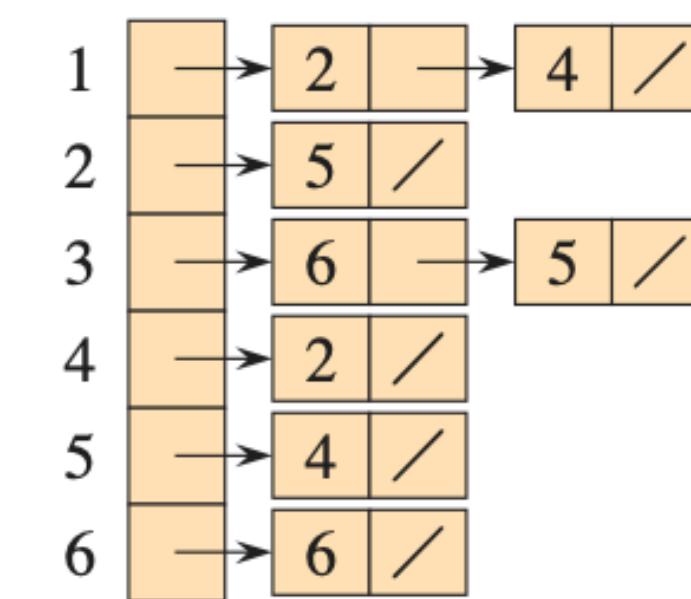
undirected graph



directed graph



1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0



1	2	3	4	5	6
1	0	1	0	1	0
2	0	0	0	0	1
3	0	0	0	0	1
4	0	1	0	0	0
5	0	0	0	1	0
6	0	0	0	0	1

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 - $A = (a_{ij})$: $|V| \times |V|$ matrix

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Graph Representations

- $G = (V, E)$

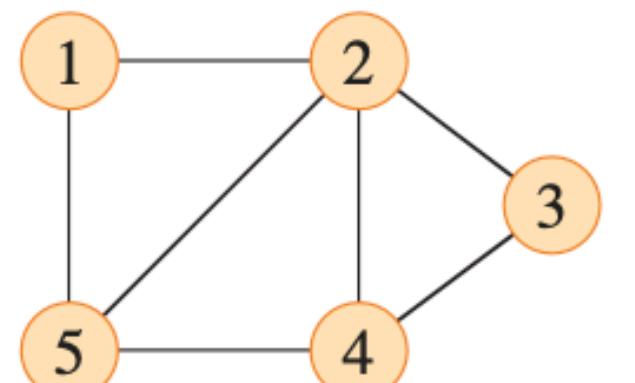
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- $Adj: |V|$ lists

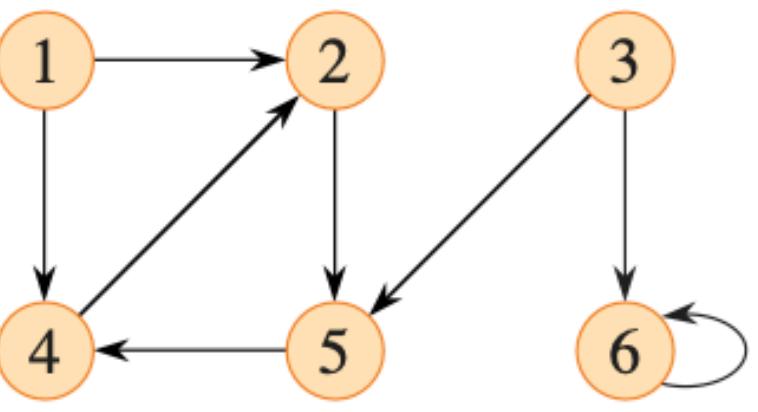
- $Adj[v]$: v 's neighbors

```
typedef struct Node {
    int vertex;
    struct Node* next;
} Node;
```

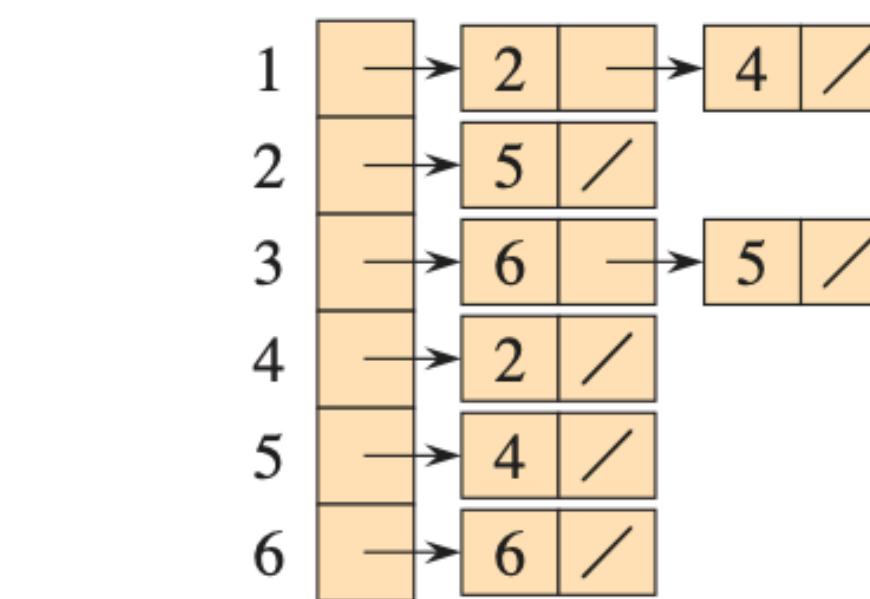
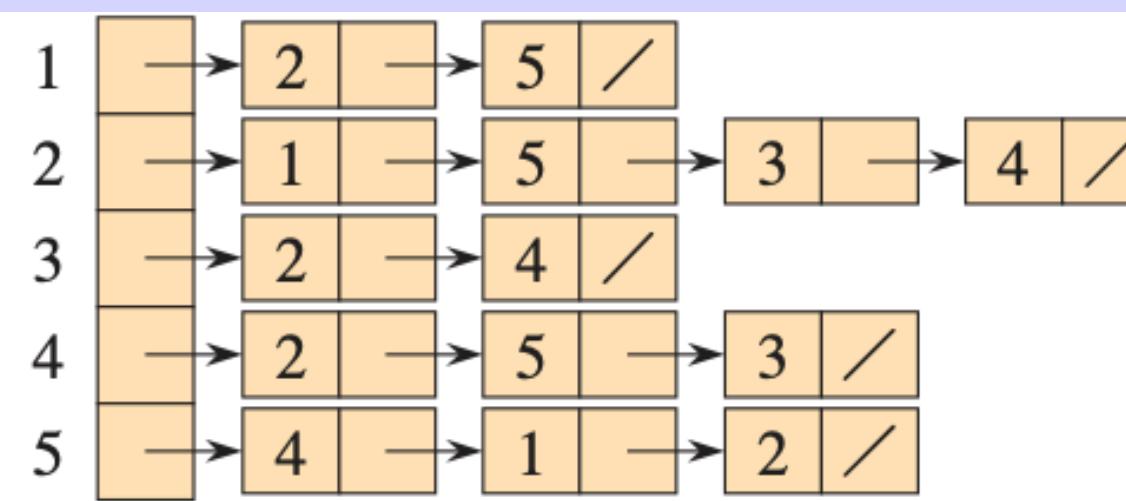
```
Node* Adj [MAX_VERTICES];
```



undirected graph



directed graph



1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

1	2	3	4	5	6
1	0	1	0	1	0
2	0	0	0	0	1
3	0	0	0	0	1
4	0	1	0	0	0
5	0	0	0	1	0
6	0	0	0	0	1

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$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

Graph Representations

- $G = (V, E)$

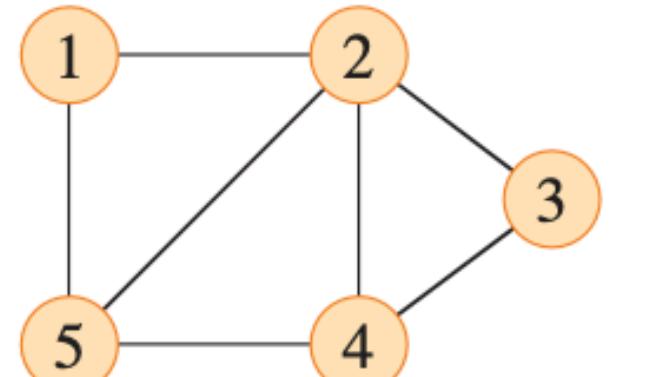
- adjacency list

- $Adj: |V|$ lists

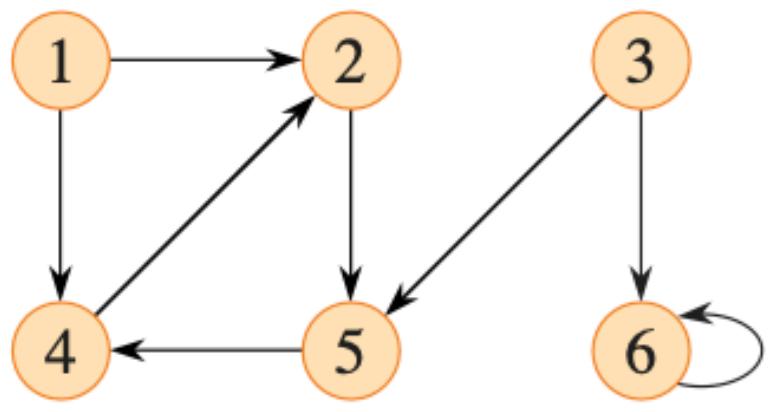
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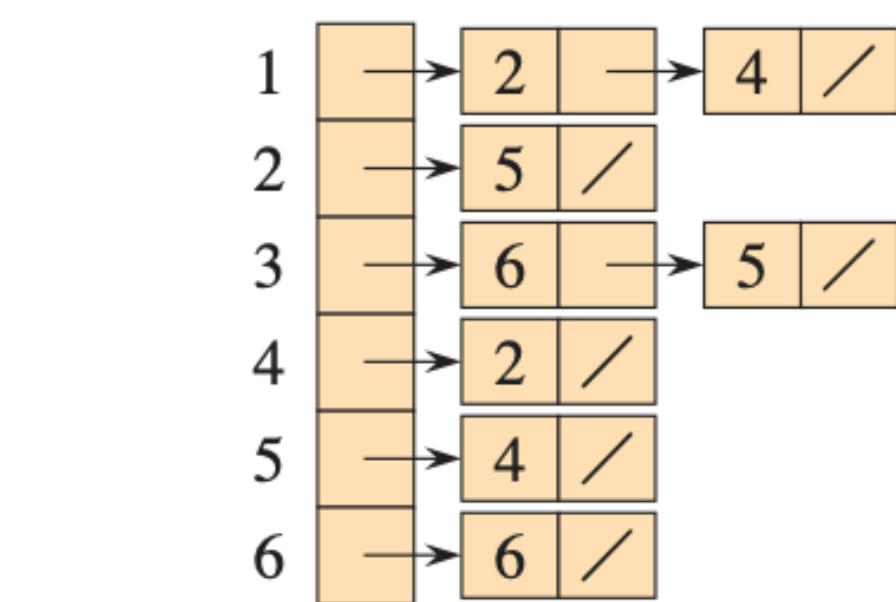
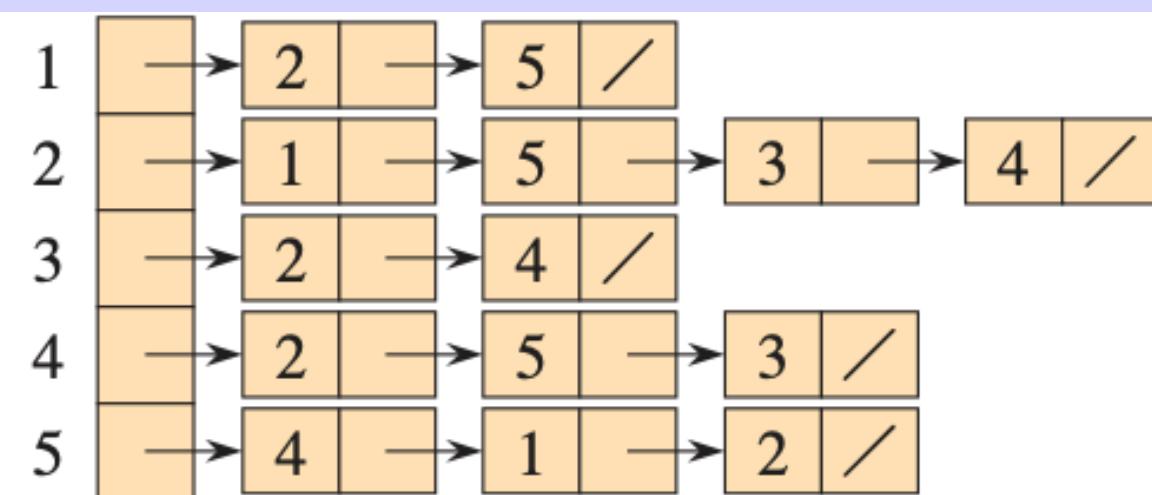
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undirected graph



directed graph



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2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

1	2	3	4	5	6	
1	0	1	0	1	0	
2	0	0	0	0	1	0
3	0	0	0	0	1	1
4	0	1	0	0	0	0
5	0	0	0	1	0	0
6	0	0	0	0	0	1

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- $A = (a_{ij})$: $|V| \times |V|$ matrix

$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

```
int A[MAX_VERTICES][MAX_VERTICES];
```

Graph Representations

- $G = (V, E)$

- adjacency list

- $Adj: |V|$ lists

- $Adj[v]$: v 's neighbors

```
typedef struct Node {
    int vertex;
    struct Node* next;
} Node;
```

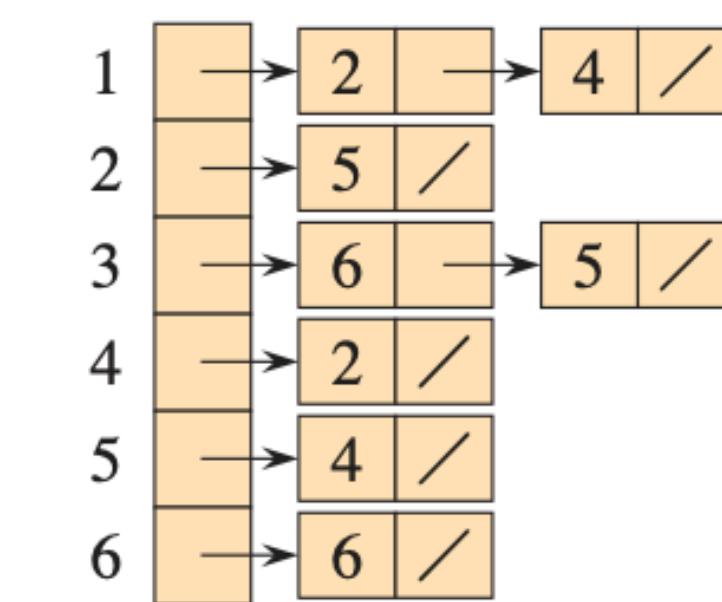
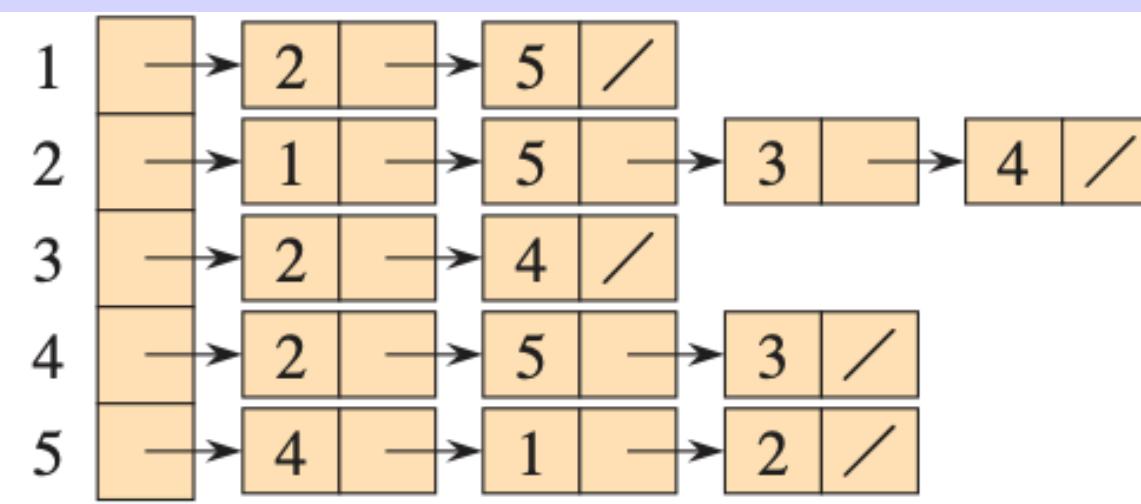
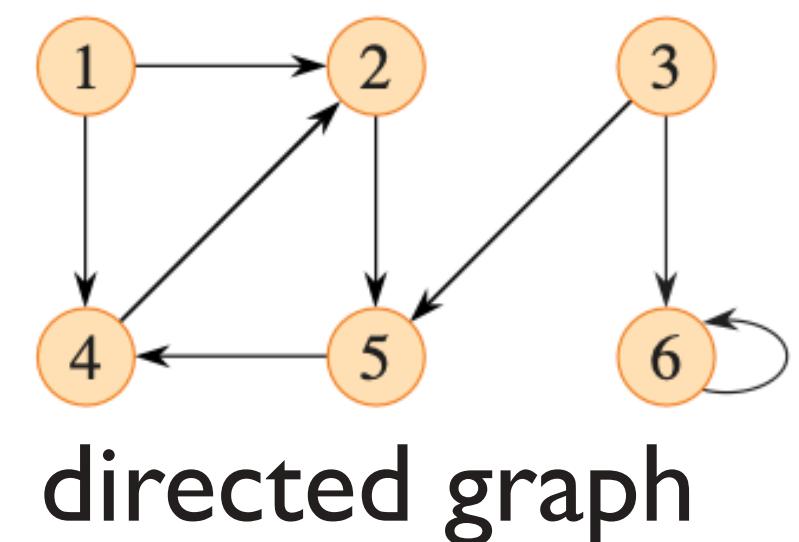
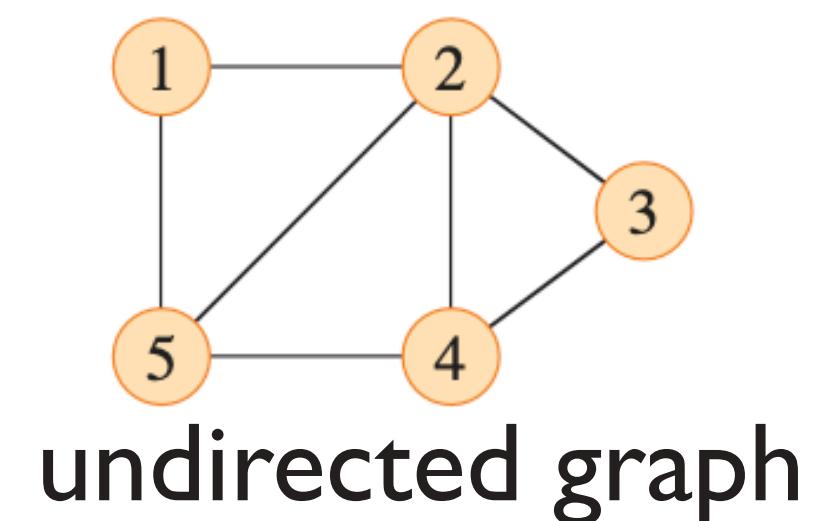
```
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```

- adjacency matrix

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```
int A [MAX_VERTICES] [MAX_VERTICES];
```



	1	2	3	4	5	6
1	0	1	0	0	1	
2	1	0	1	1	1	
3	0	1	0	1	0	
4	0	1	1	0	1	
5	1	1	0	1	0	
6	0	0	0	0	0	1

Feature	Adjacency Matrix	Adjacency List
space complexity	-	-
edge lookup	-	-
iterating neighbors	-	-
best for	-	-
implementation complexity	-	-

Graph Representations

- $G = (V, E)$

- adjacency list

- $Adj: |V|$ lists

- $Adj[v]$: v 's neighbors

```
typedef struct Node {
    int vertex;
    struct Node* next;
} Node;
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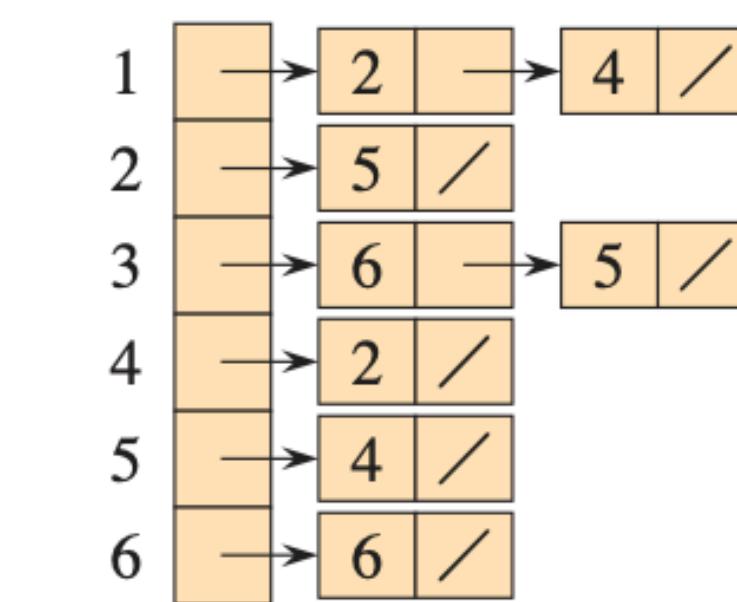
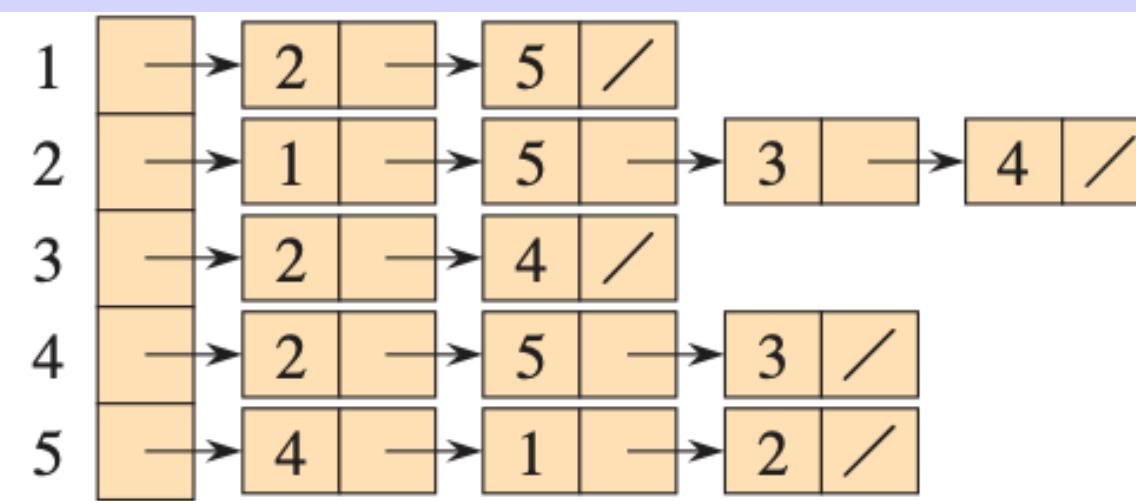
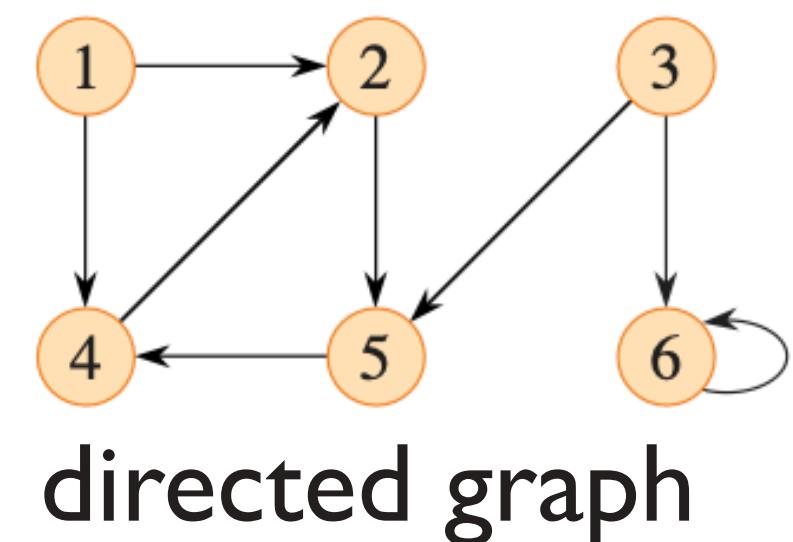
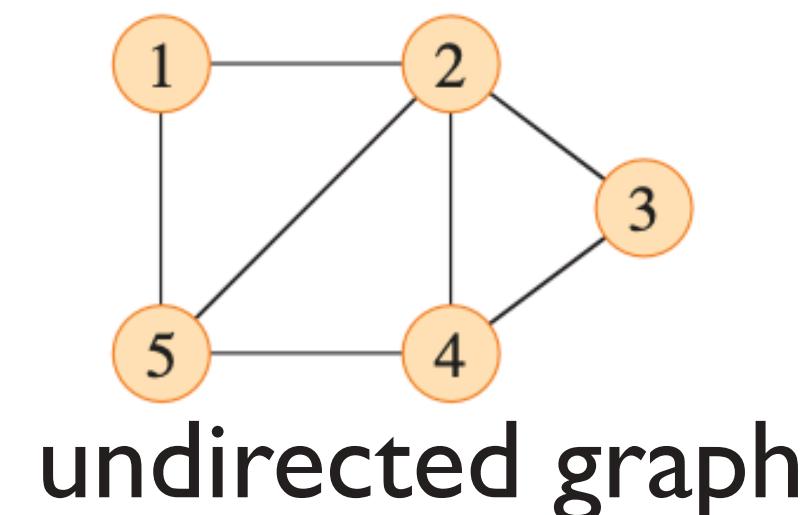
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Node* Adj [MAX_VERTICES];
```

- adjacency matrix

- $A = (a_{ij})$: $|V| \times |V|$ matrix

$$a_{ij} = \begin{cases} 1 & \text{if } (i, j) \in E \\ 0 & \text{otherwise} \end{cases}$$

```
int A [MAX_VERTICES] [MAX_VERTICES];
```



1	0	1	0	0	1
2	1	0	1	1	1
3	0	1	0	1	0
4	0	1	1	0	1
5	1	1	0	1	0

1	2	3	4	5	6
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Feature	Adjacency Matrix	Adjacency List
space complexity	$O(V ^2)$	$O(V + E)$
edge lookup	-	-
iterating neighbors	-	-
best for implementation complexity	-	-

Graph Representations

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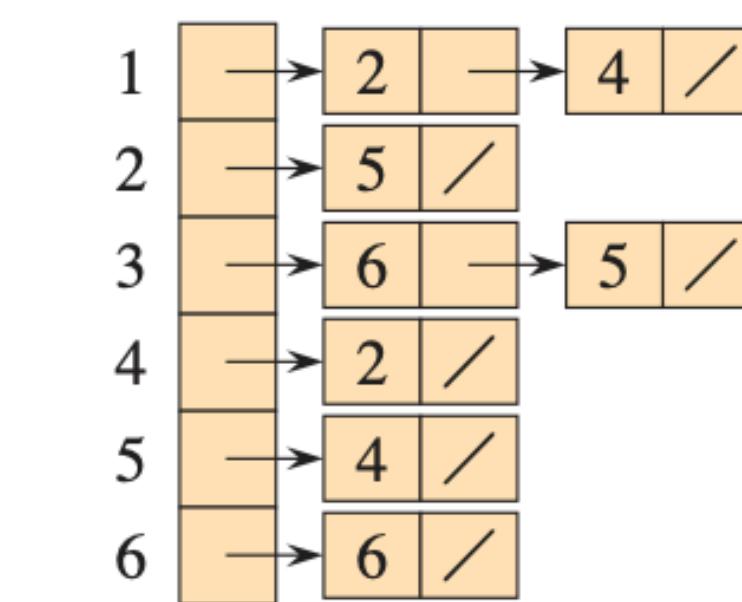
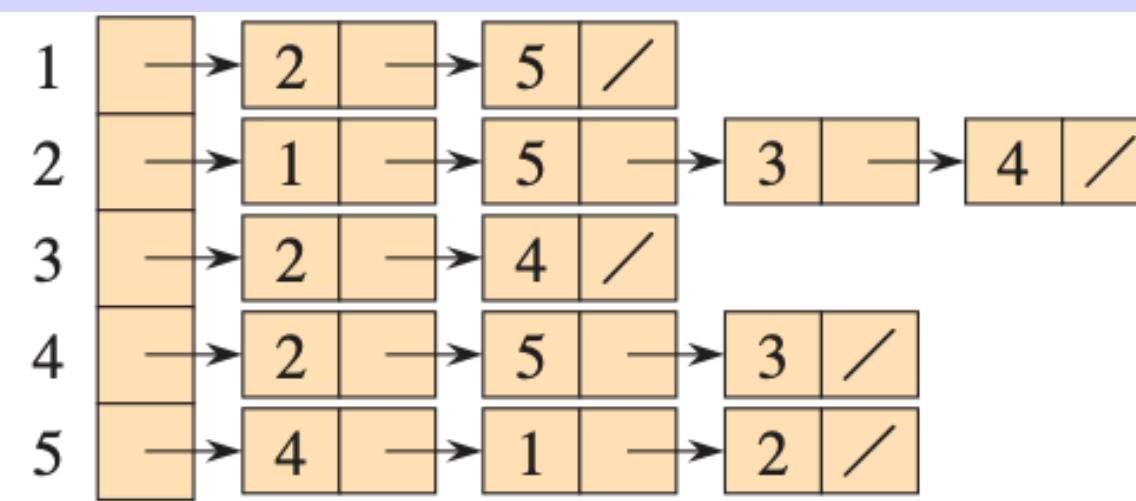
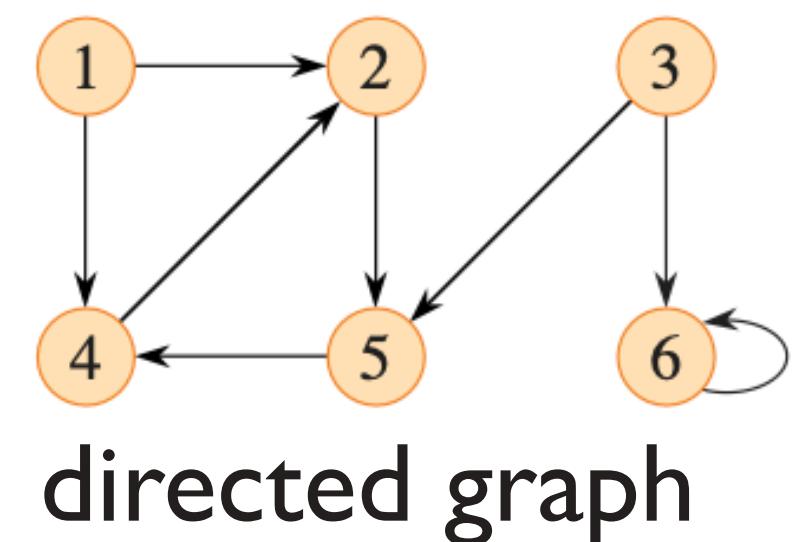
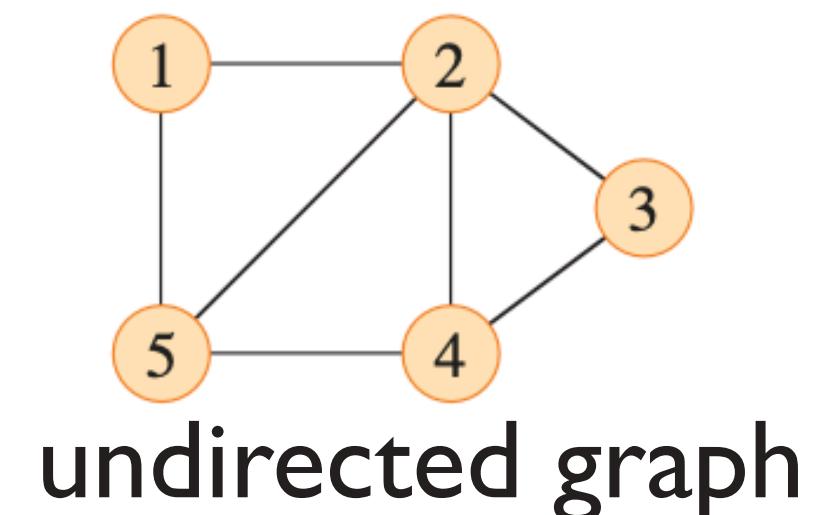
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Feature	Adjacency Matrix	Adjacency List
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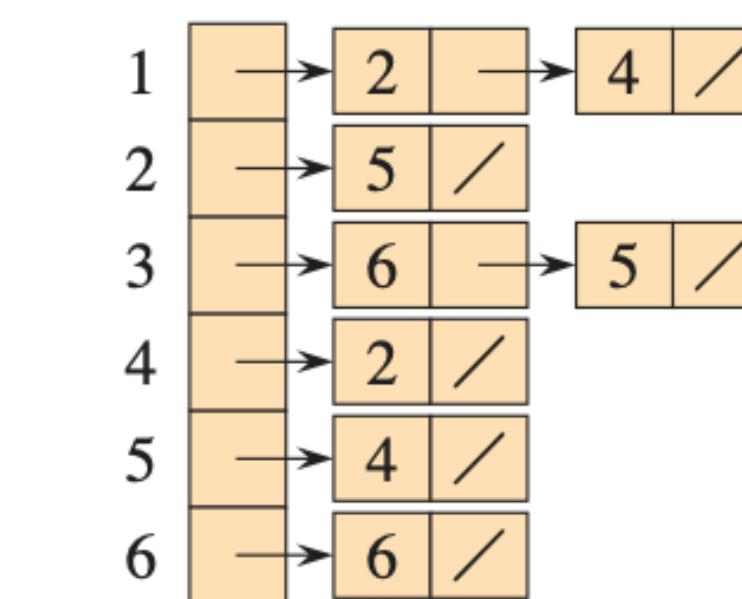
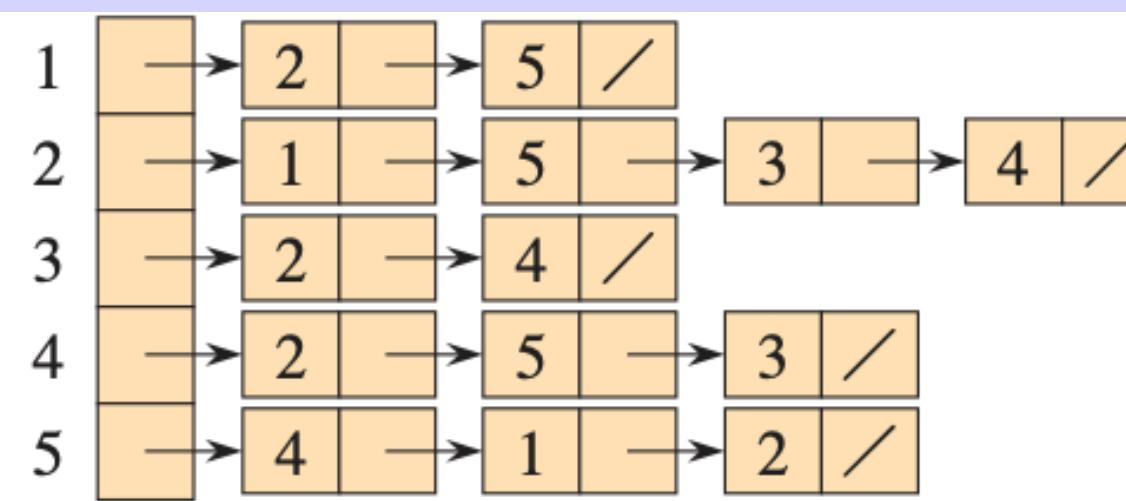
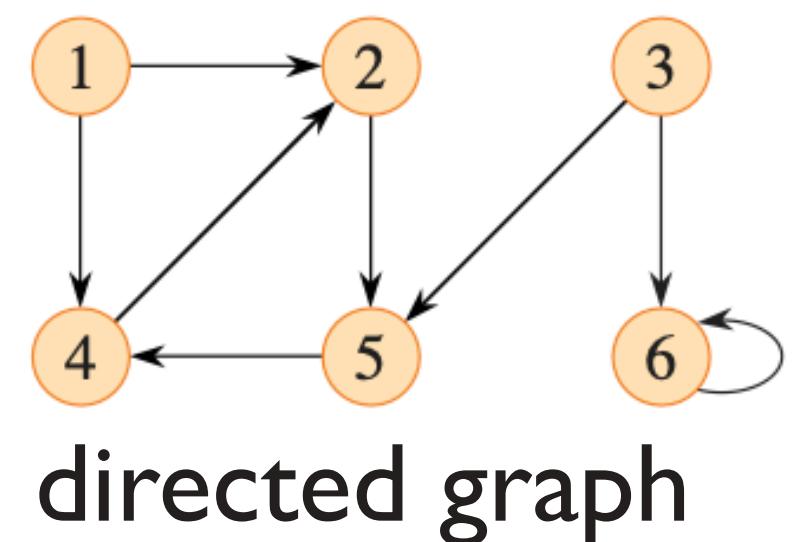
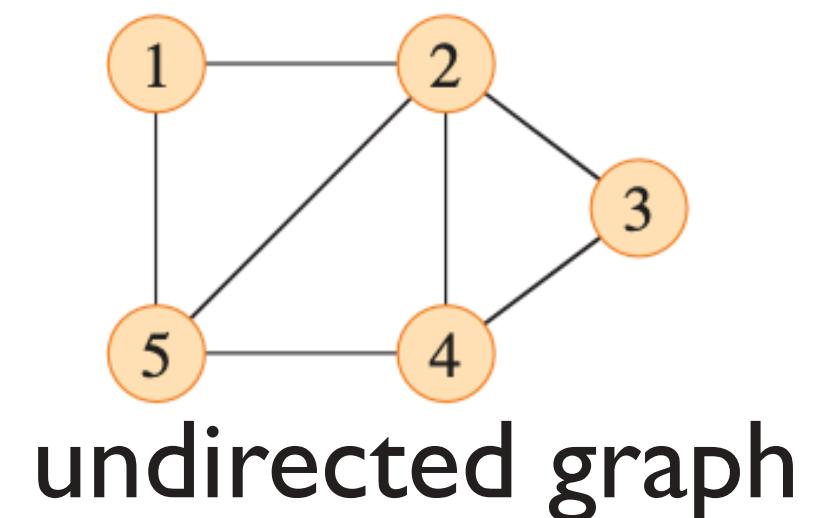
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1	2	3	4	5	6
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6	0	0	0	0	1

Feature	Adjacency Matrix	Adjacency List
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edge lookup	$O(1)$	$O(\text{degree})$
iterating neighbors	$O(V)$	$O(\text{degree})$
best for implementation complexity	-	-

Graph Representations

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- adjacency list

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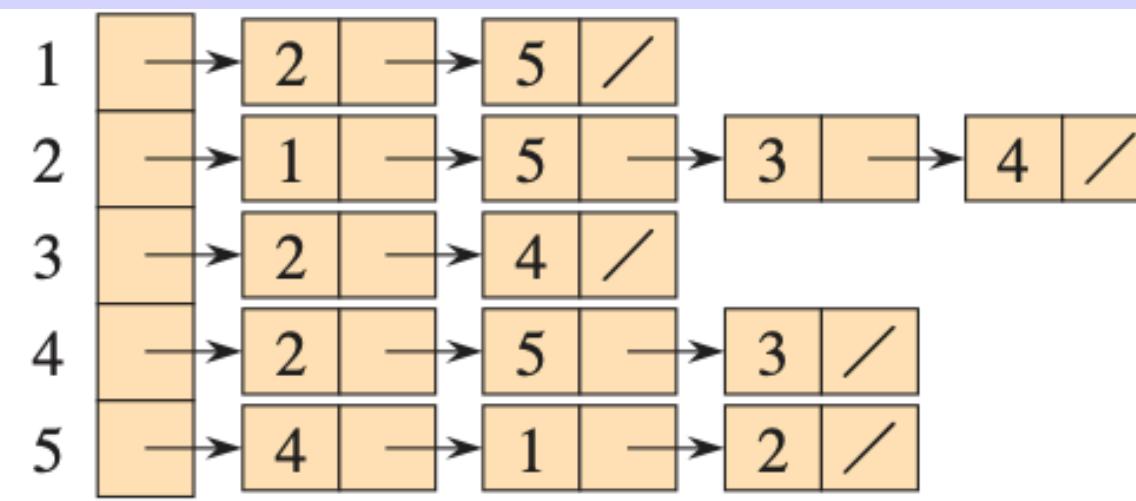
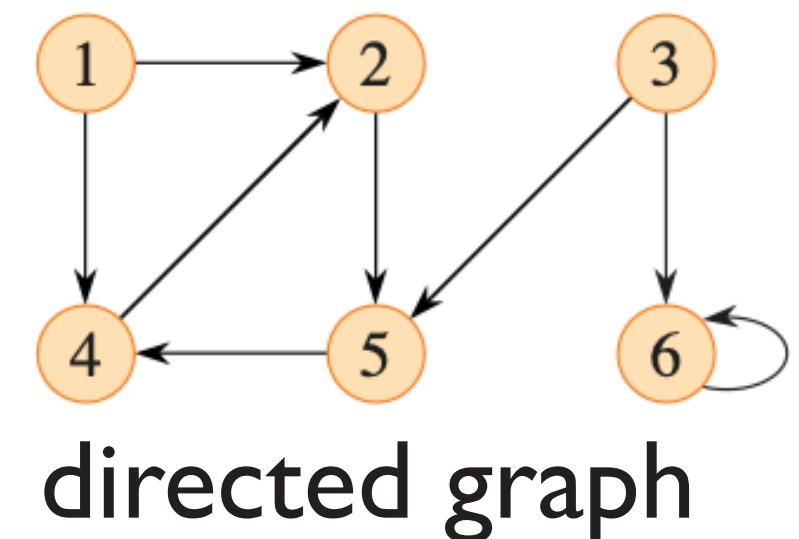
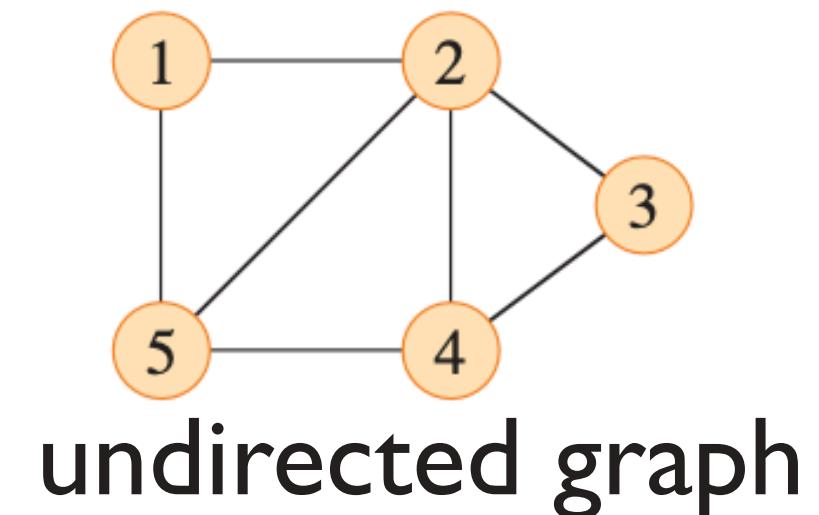
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- adjacency matrix

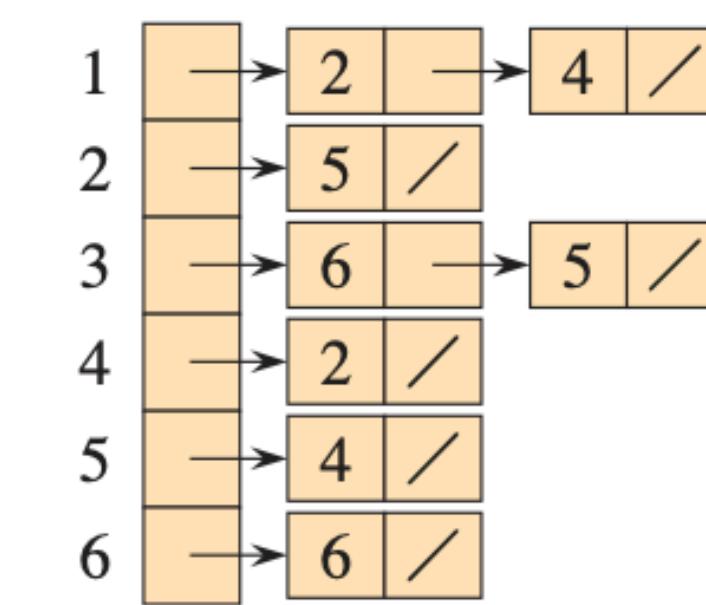
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Feature	Adjacency Matrix	Adjacency List
space complexity	$O(V ^2)$	$O(V + E)$
edge lookup	$O(1)$	$O(\text{degree})$
iterating neighbors	$O(V)$	$O(\text{degree})$
best for implementation complexity	dense graphs	sparse graphs

Graph Representations

- $G = (V, E)$

- adjacency list

- $Adj: |V|$ lists

- $Adj[v]$: v 's neighbors

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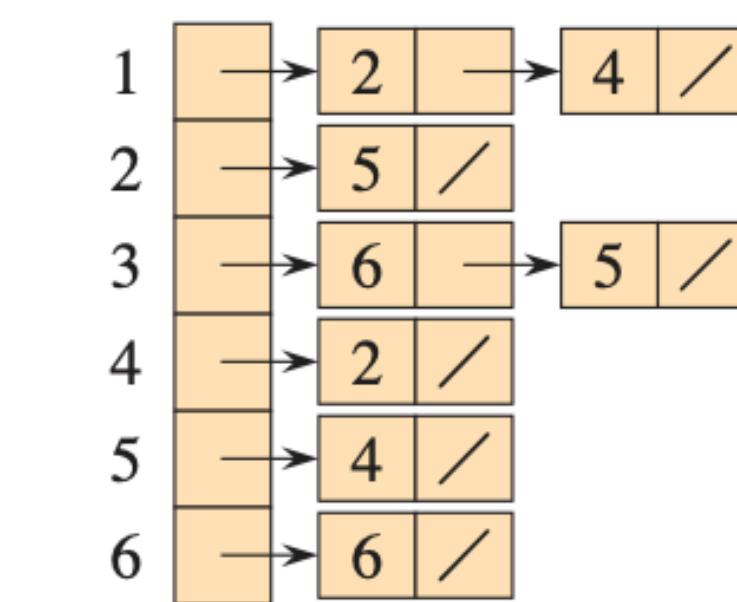
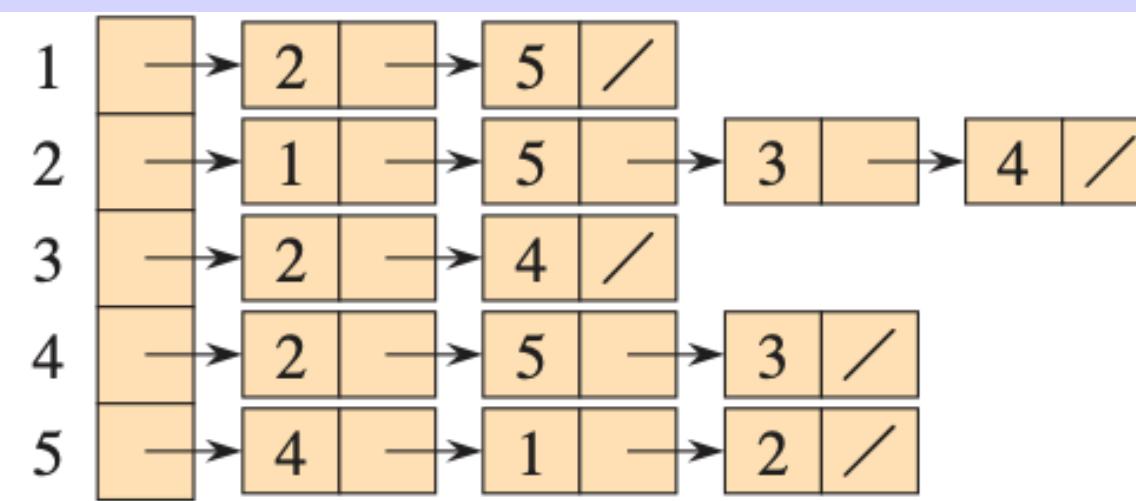
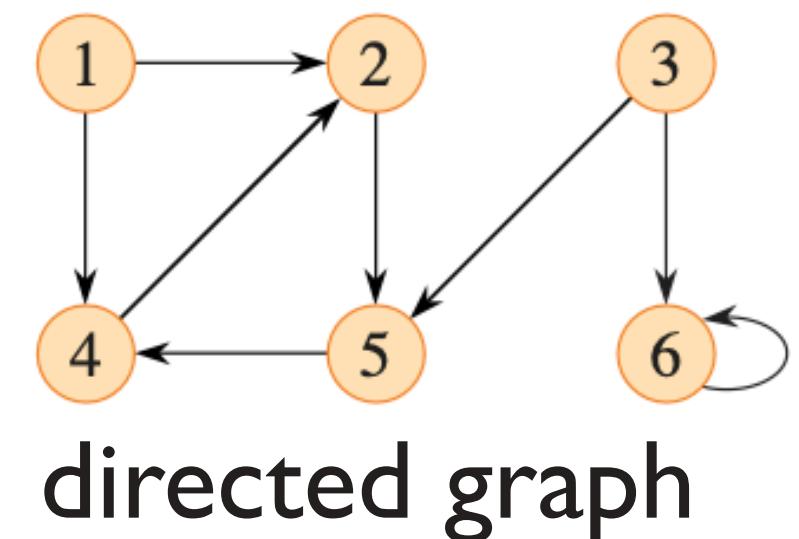
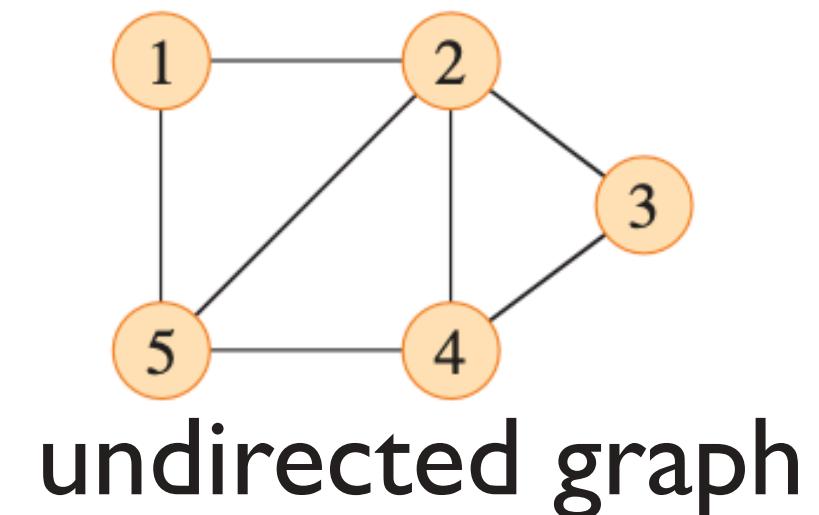
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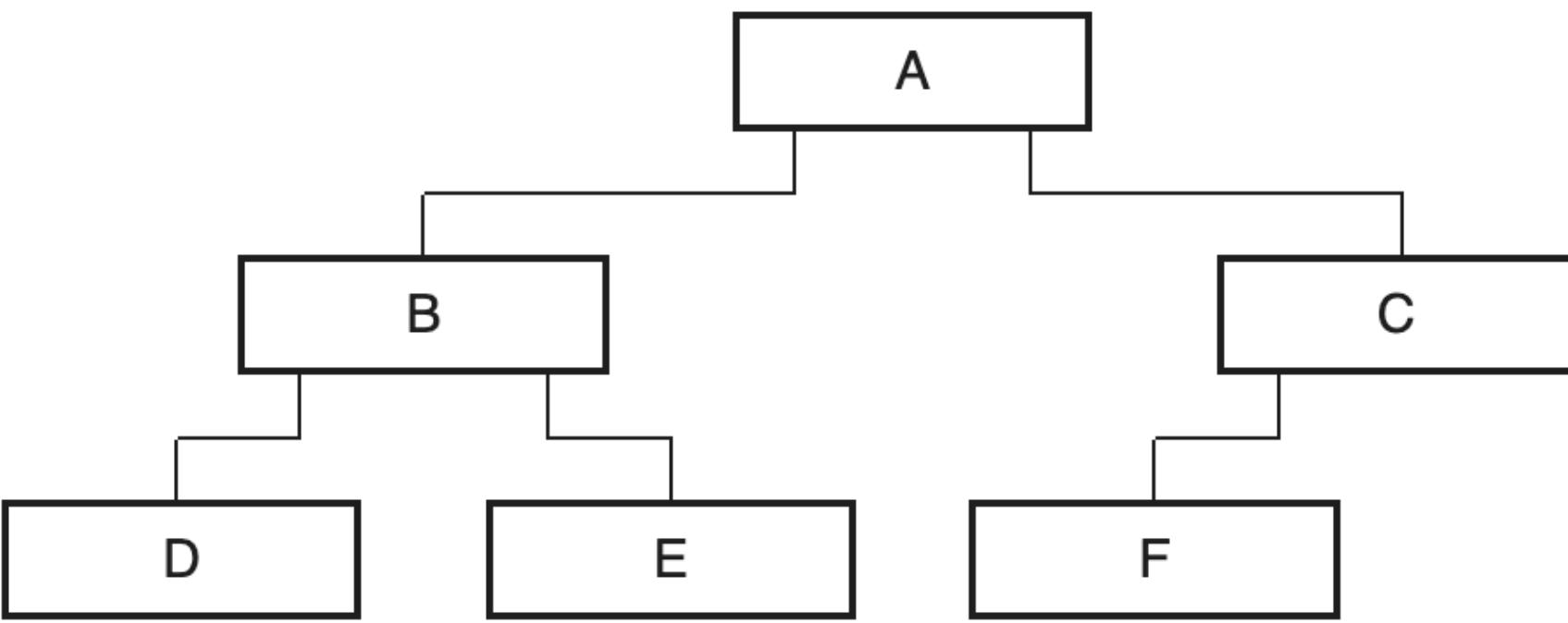


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Feature	Adjacency Matrix	Adjacency List
space complexity	$O(V ^2)$	$O(V + E)$
edge lookup	$O(1)$	$O(\text{degree})$
iterating neighbors	$O(V)$	$O(\text{degree})$
best for implementation complexity	dense graphs simple	sparse graphs slightly complex

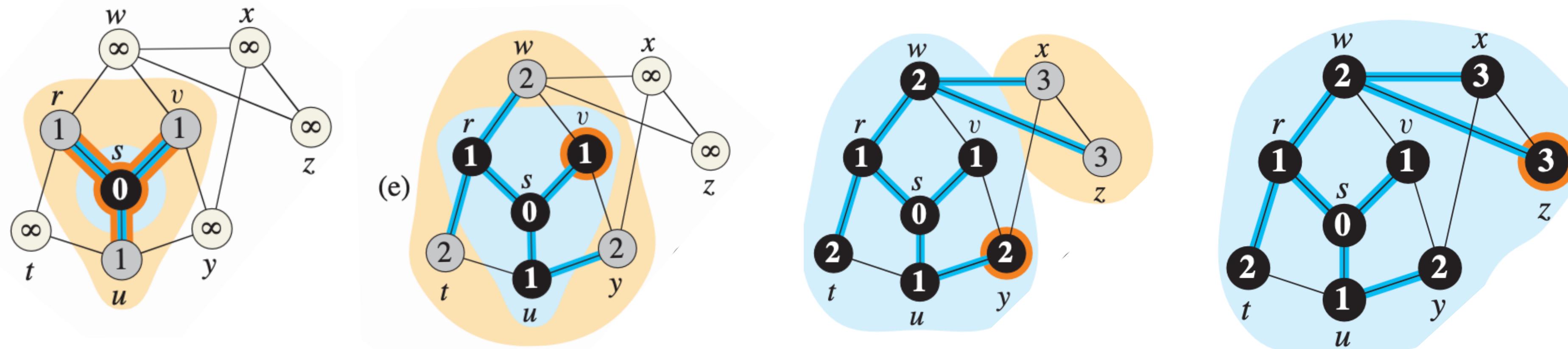
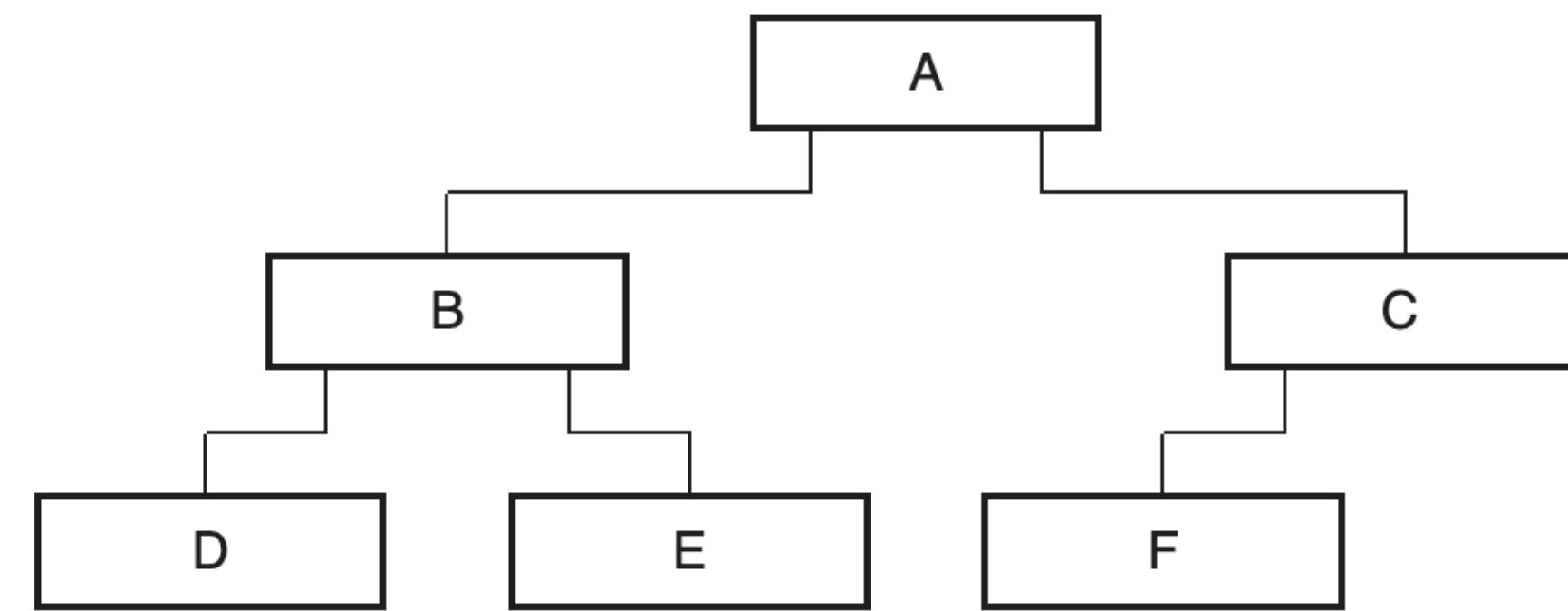
Graph Traversal

- recall binary tree traversals
- Breadth First Search (BFS)
 - level-order: [A | B C | D E F]
- Depth First Search (DFS)
 - pre-order: A (B (D) (E)) (C (F))
 - post-order: ((D) (E) B) ((F) C) A
 - in-order: ((D) B (E)) A ((F) C)
- Tree is actually a special graph!



Graph BFS

- tree BFS can be generalized to graph BFS
- level from the root \rightarrow distance from the source

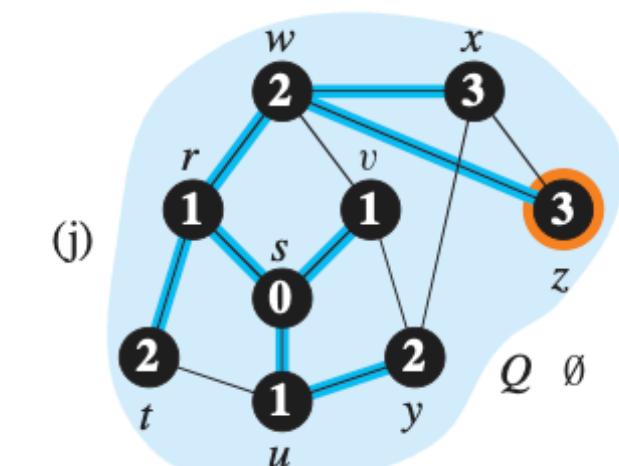
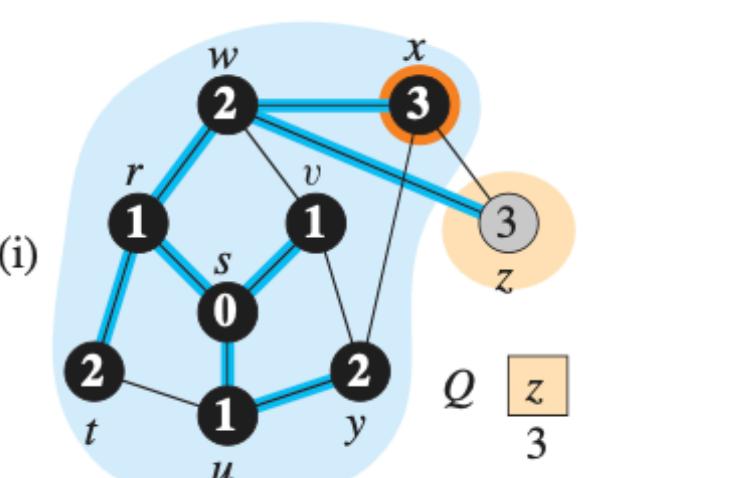
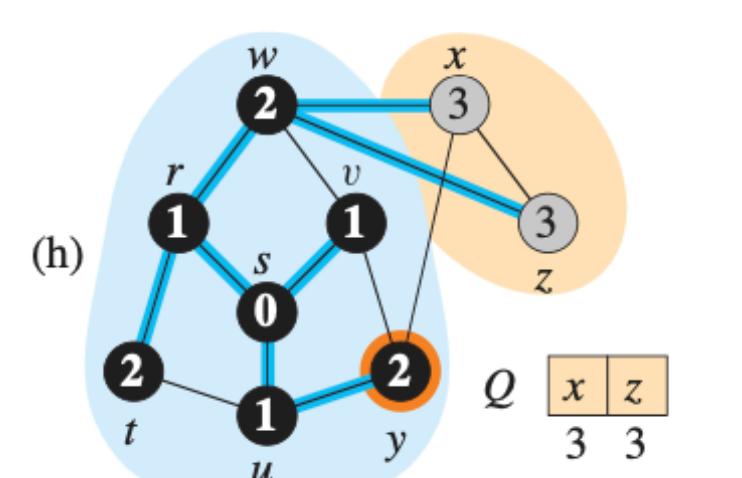
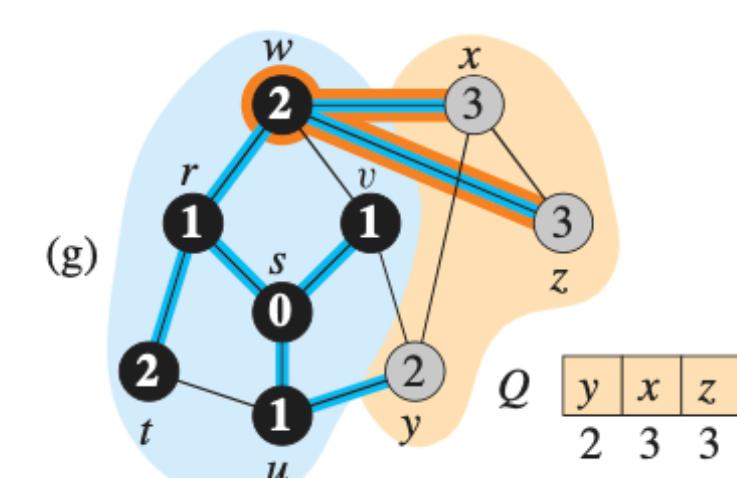
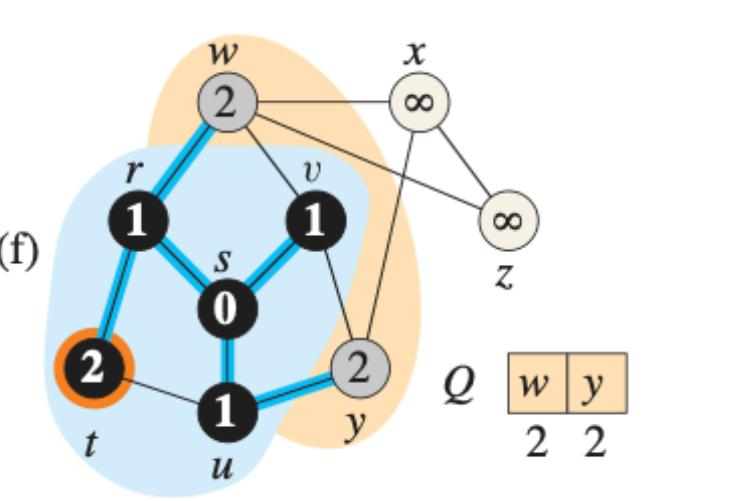
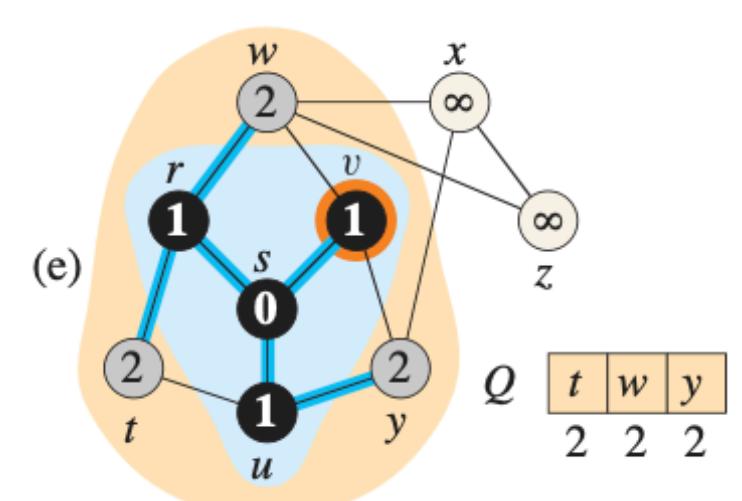
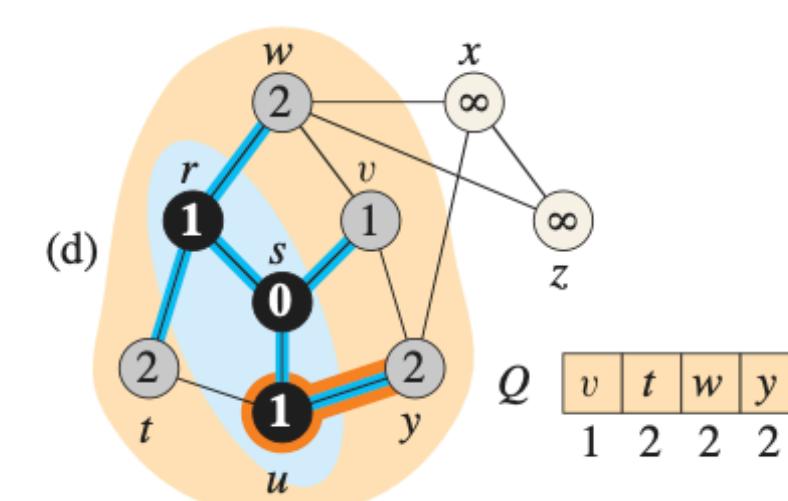
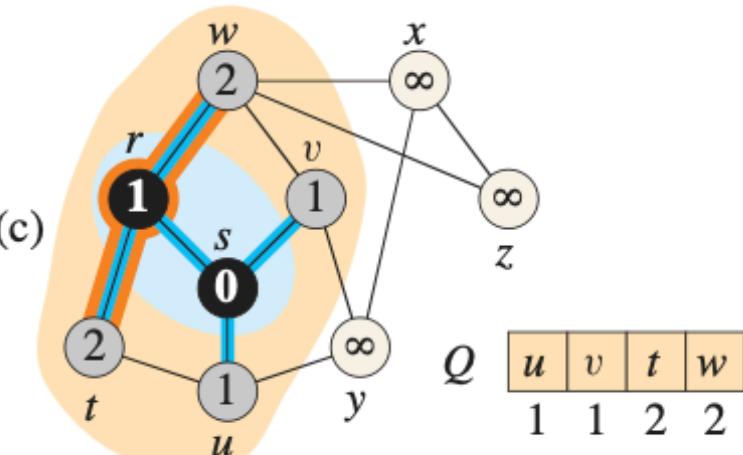
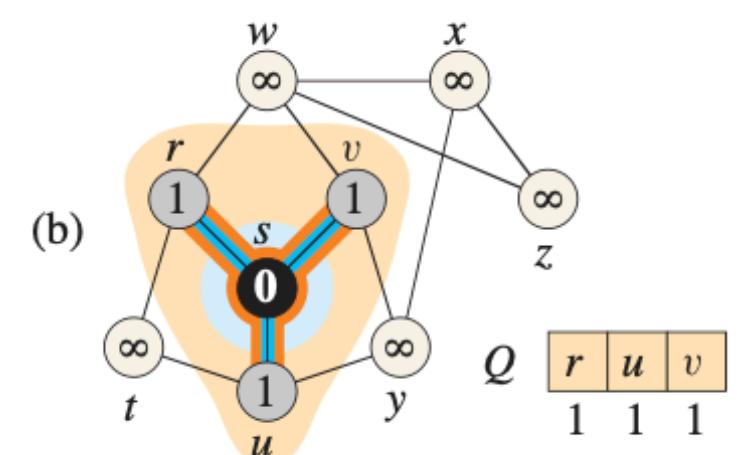
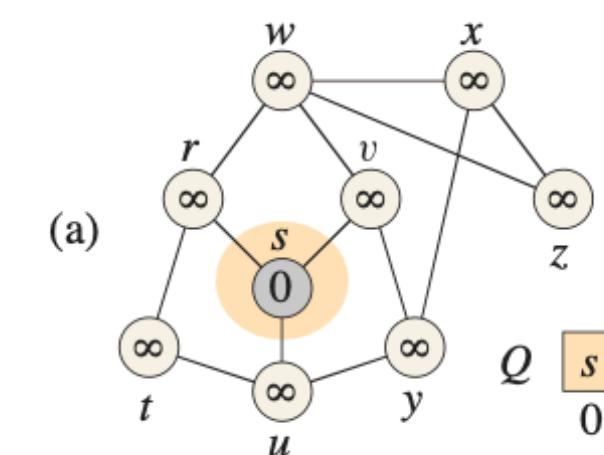


Graph BFS via Queue

white: undiscovered

gray: being visited

black: visited



$\text{BFS}(G, s)$

```

1  for each vertex  $u \in G.V - \{s\}$ 
2     $u.\text{color} = \text{WHITE}$ 
3
4
5   $s.\text{color} = \text{GRAY}$ 
6
7
8   $Q = \emptyset$ 
9  ENQUEUE( $Q, s$ )
10 while  $Q \neq \emptyset$ 
11    $u = \text{DEQUEUE}(Q)$ 
12   for each vertex  $v$  in  $G.\text{Adj}[u]$ 
13     if  $v.\text{color} == \text{WHITE}$ 
14        $v.\text{color} = \text{GRAY}$ 
15
16
17     ENQUEUE( $Q, v$ )
18    $u.\text{color} = \text{BLACK}$ 

```

// search the neighbors of u
// is v being discovered now?

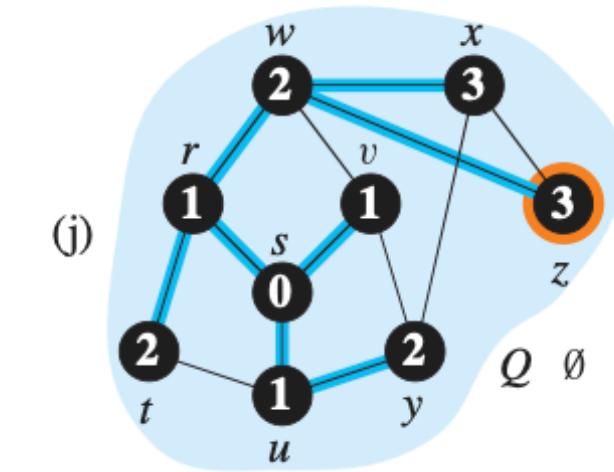
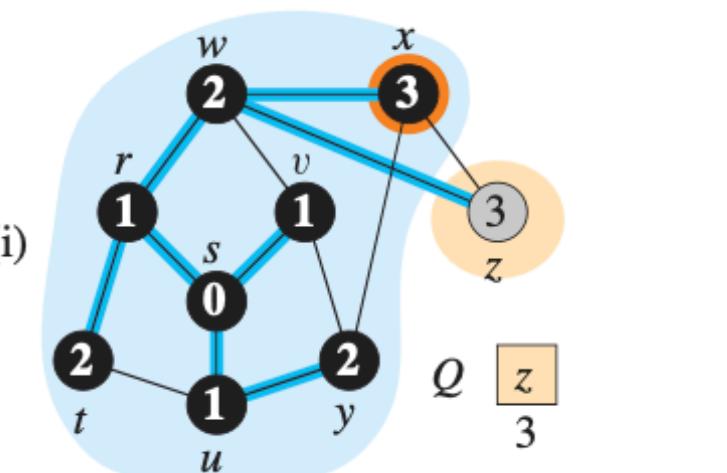
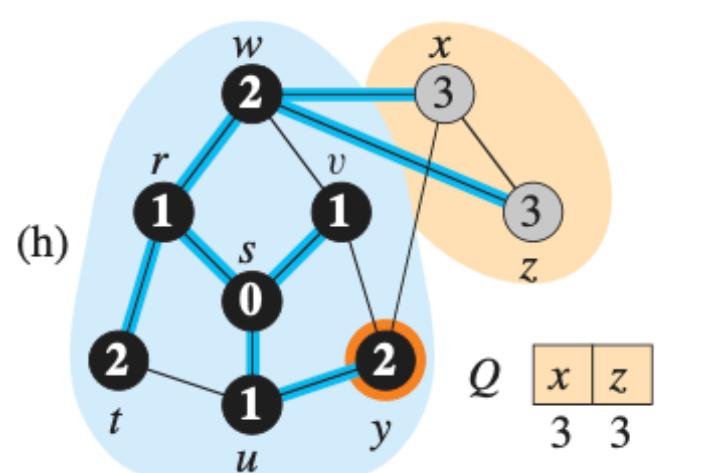
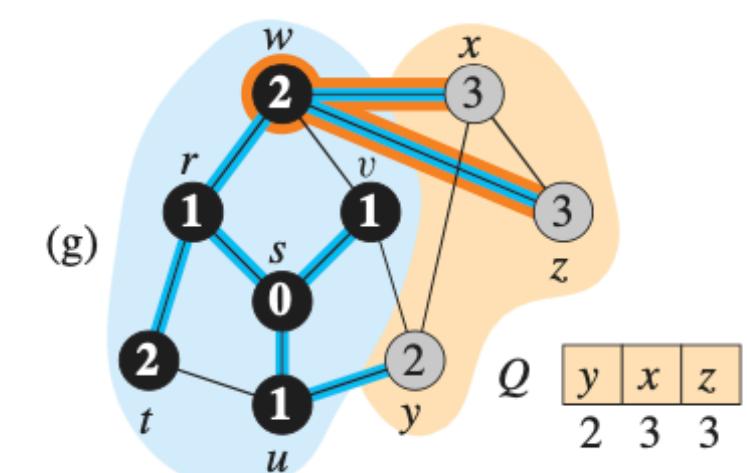
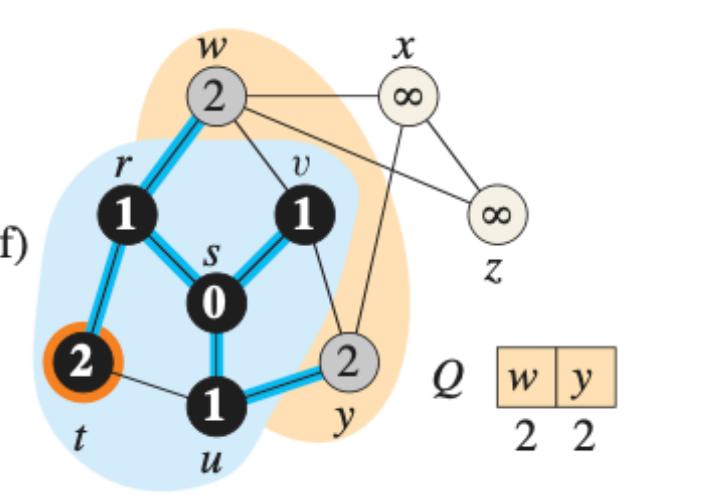
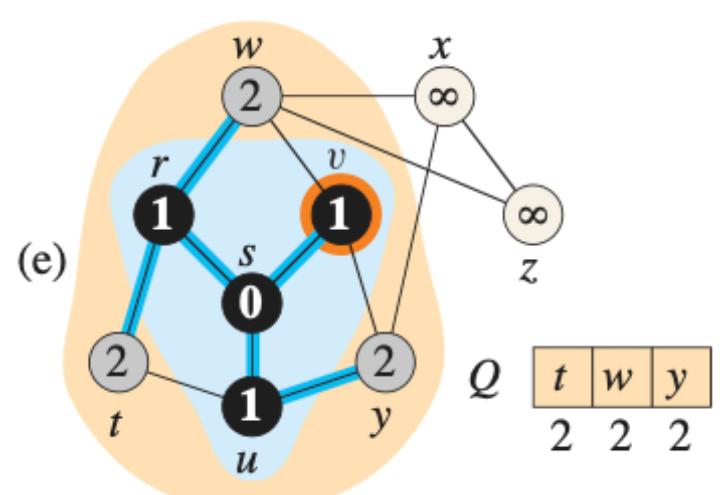
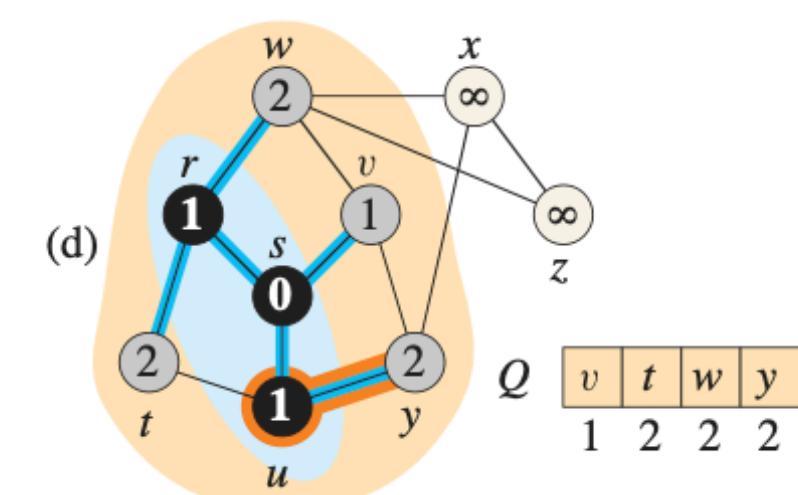
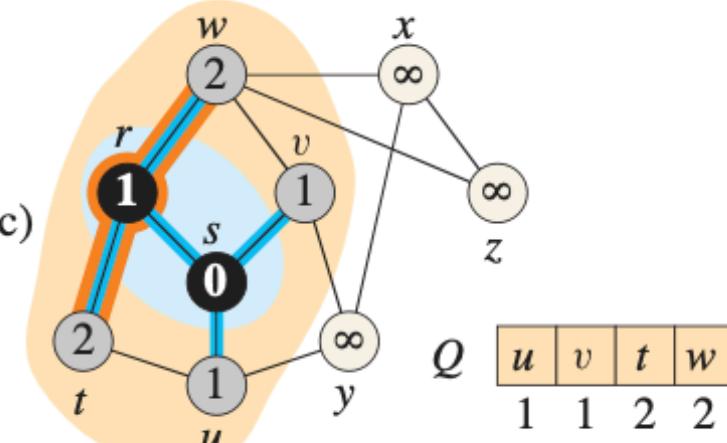
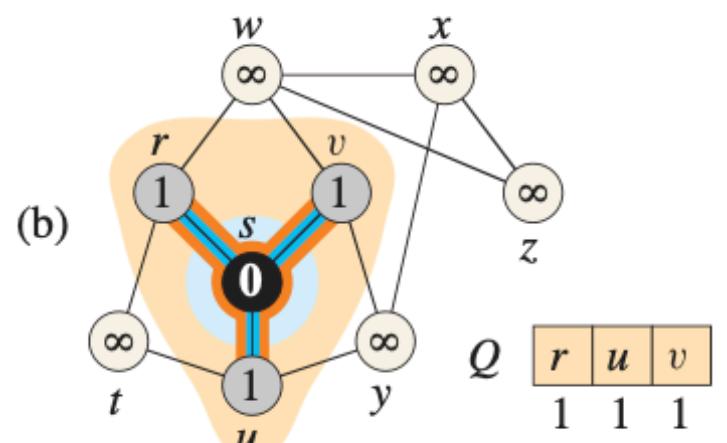
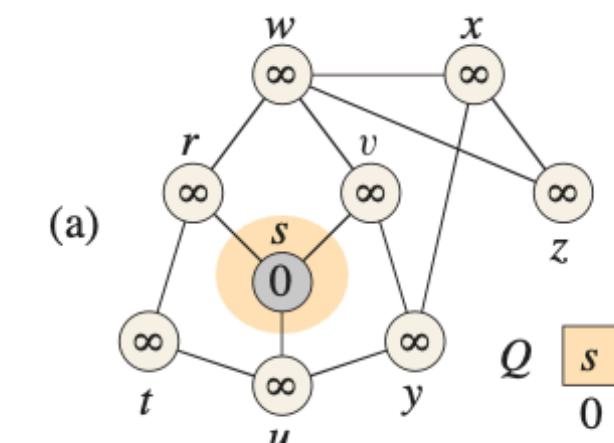
// v is now on the frontier
// u is now behind the frontier

Graph BFS via Queue

white: undiscovered

gray: being visited

black: visited



$\text{BFS}(G, s)$

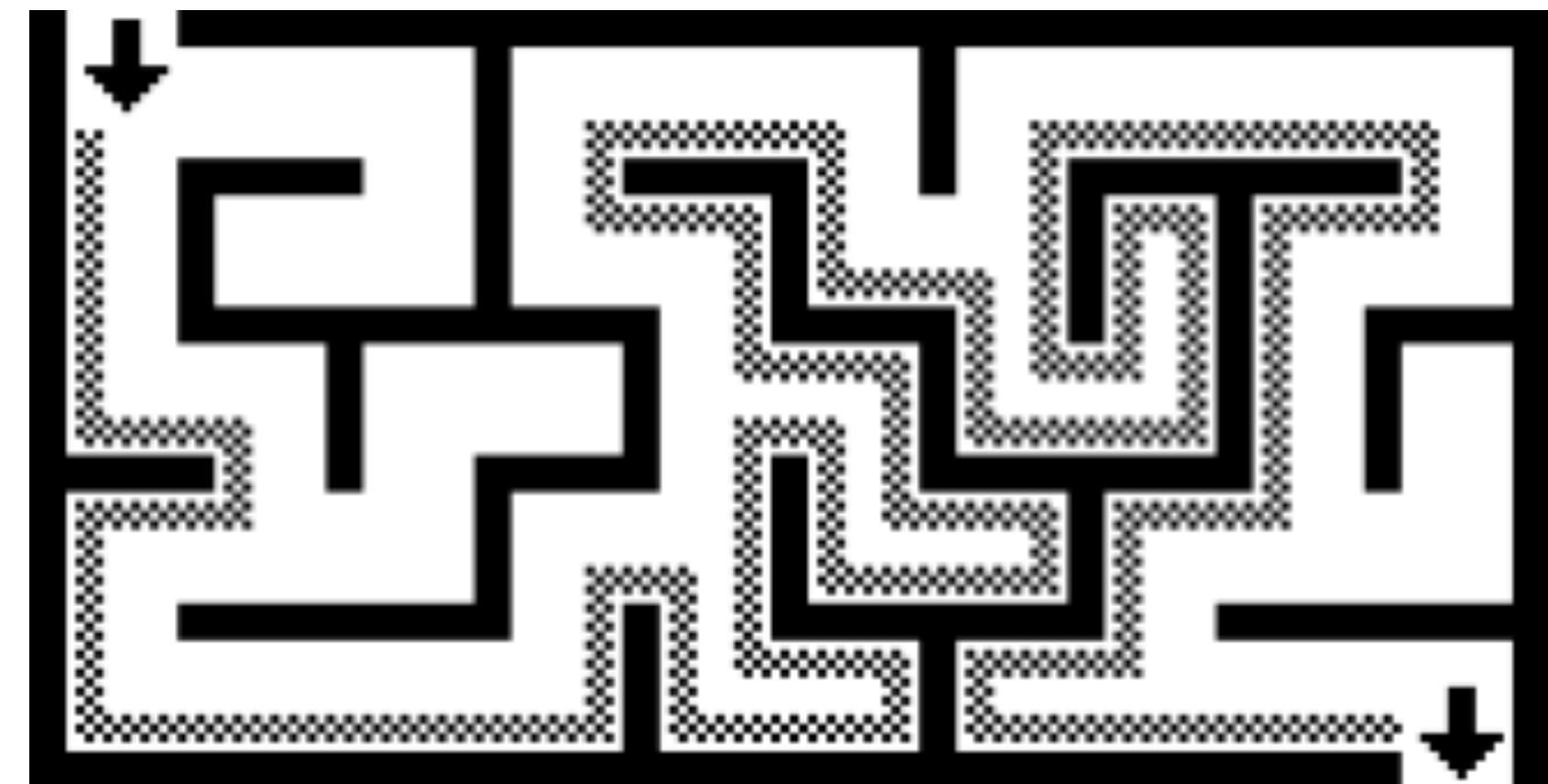
```

1 for each vertex  $u \in G.V - \{s\}$ 
2    $u.\text{color} = \text{WHITE}$ 
3    $u.d = \infty$  //distance to source
4    $u.\pi = \text{NIL}$  //parent in BFS
5    $s.\text{color} = \text{GRAY}$ 
6    $s.d = 0$ 
7    $s.\pi = \text{NIL}$ 
8    $Q = \emptyset$ 
9   ENQUEUE( $Q, s$ )
10  while  $Q \neq \emptyset$ 
11     $u = \text{DEQUEUE}(Q)$ 
12    for each vertex  $v$  in  $G.\text{Adj}[u]$  // search the neighbors of  $u$ 
13      if  $v.\text{color} == \text{WHITE}$  // is  $v$  being discovered now?
14         $v.\text{color} = \text{GRAY}$ 
15         $v.d = u.d + 1$ 
16         $v.\pi = u$ 
17        ENQUEUE( $Q, v$ ) //  $v$  is now on the frontier
18     $u.\text{color} = \text{BLACK}$  //  $u$  is now behind the frontier

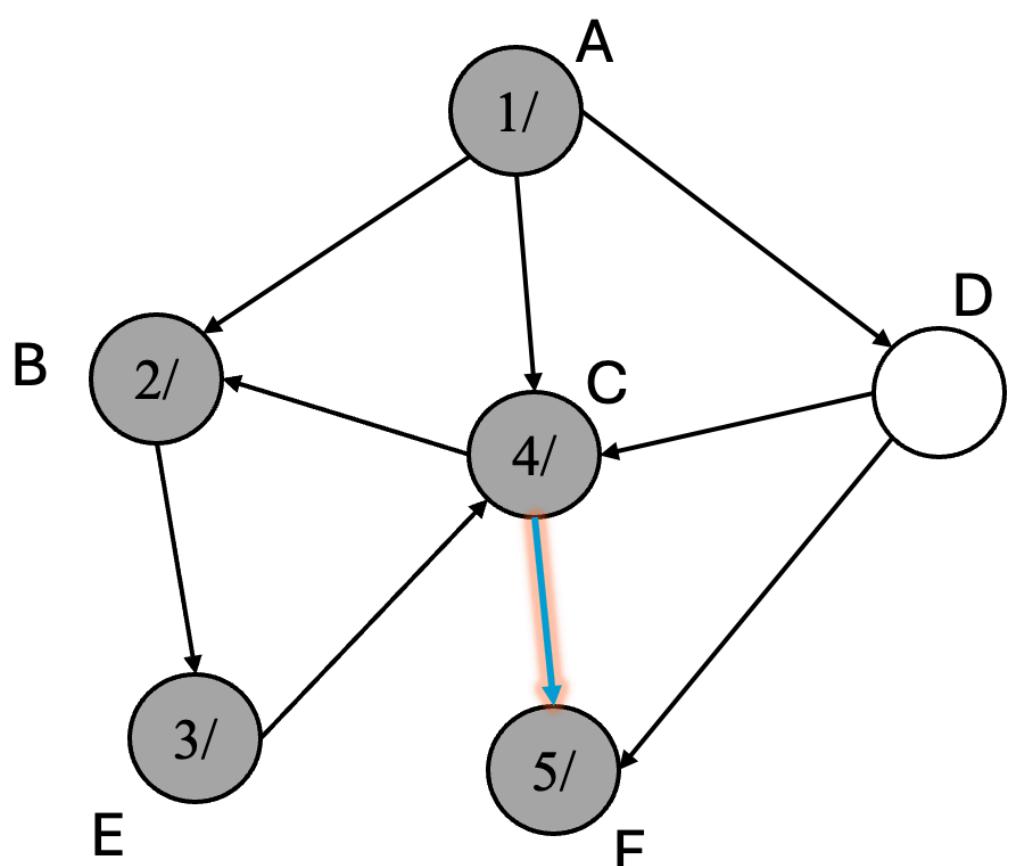
```

Graph DFS

- how to solve a maze
 - go as far as possible until hitting a wall
 - backtrack



- Depth First Search (DFS)
 - go as deep as possible while avoiding revisit
 - backtrack

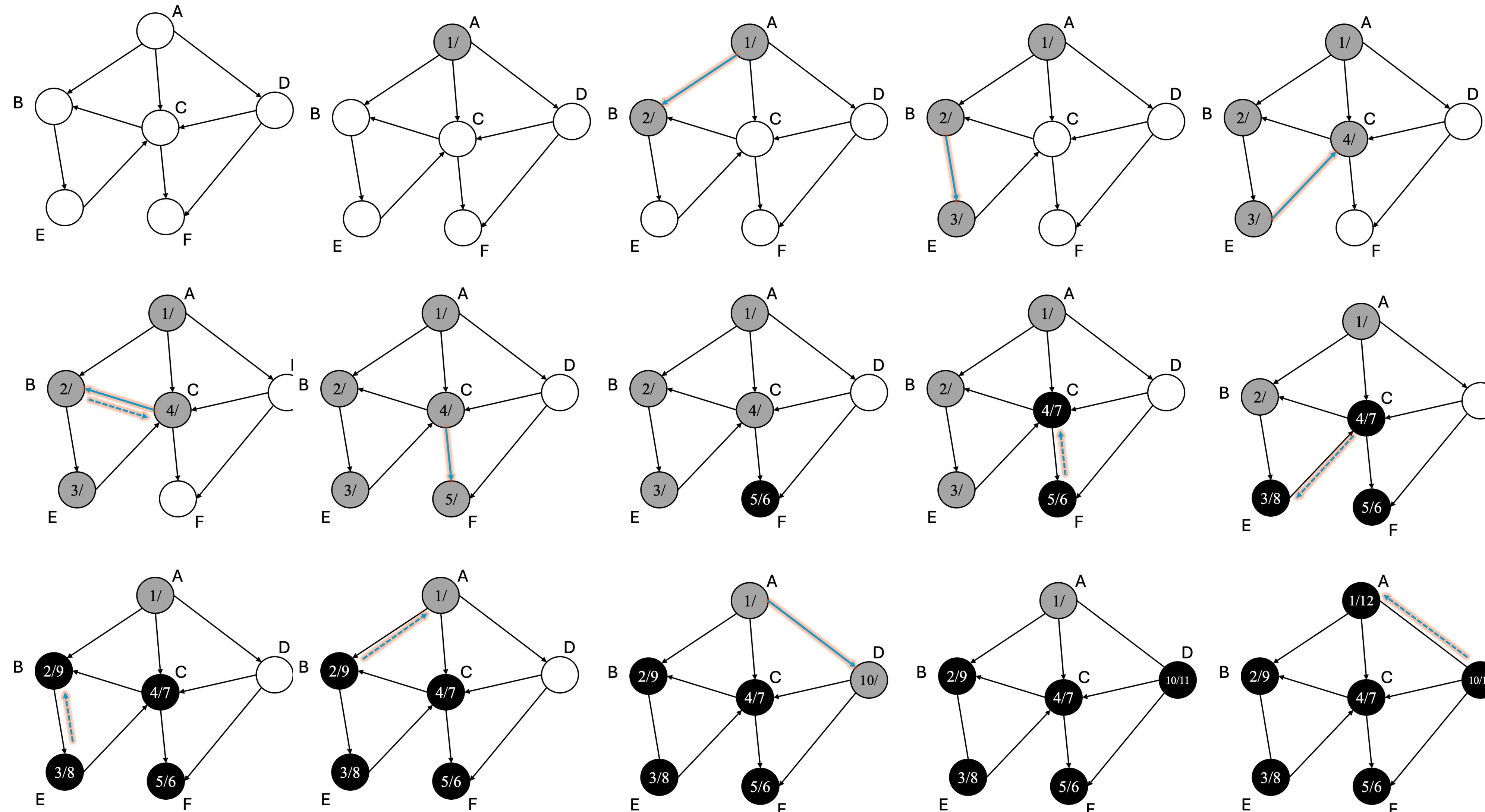


Graph DFS via Recursion

white: undiscovered

gray: being visited

black: visited



$\text{DFS}(G)$

```

1 for each vertex  $u \in G.V$ 
2    $u.\text{color} = \text{WHITE}$ 
3
4    $time = 0$ 
5   for each vertex  $u \in G.V$ 
6     if  $u.\text{color} == \text{WHITE}$ 
7        $\text{DFS-VISIT}(G, u)$ 

```

$\text{DFS-VISIT}(G, u)$

```

1    $time = time + 1$ 
2    $u.d = time$  //discover time
3    $u.\text{color} = \text{GRAY}$ 
4   for each vertex  $v$  in  $G.\text{Adj}[u]$ 
5     if  $v.\text{color} == \text{WHITE}$ 

```

$\text{DFS-VISIT}(G, v)$

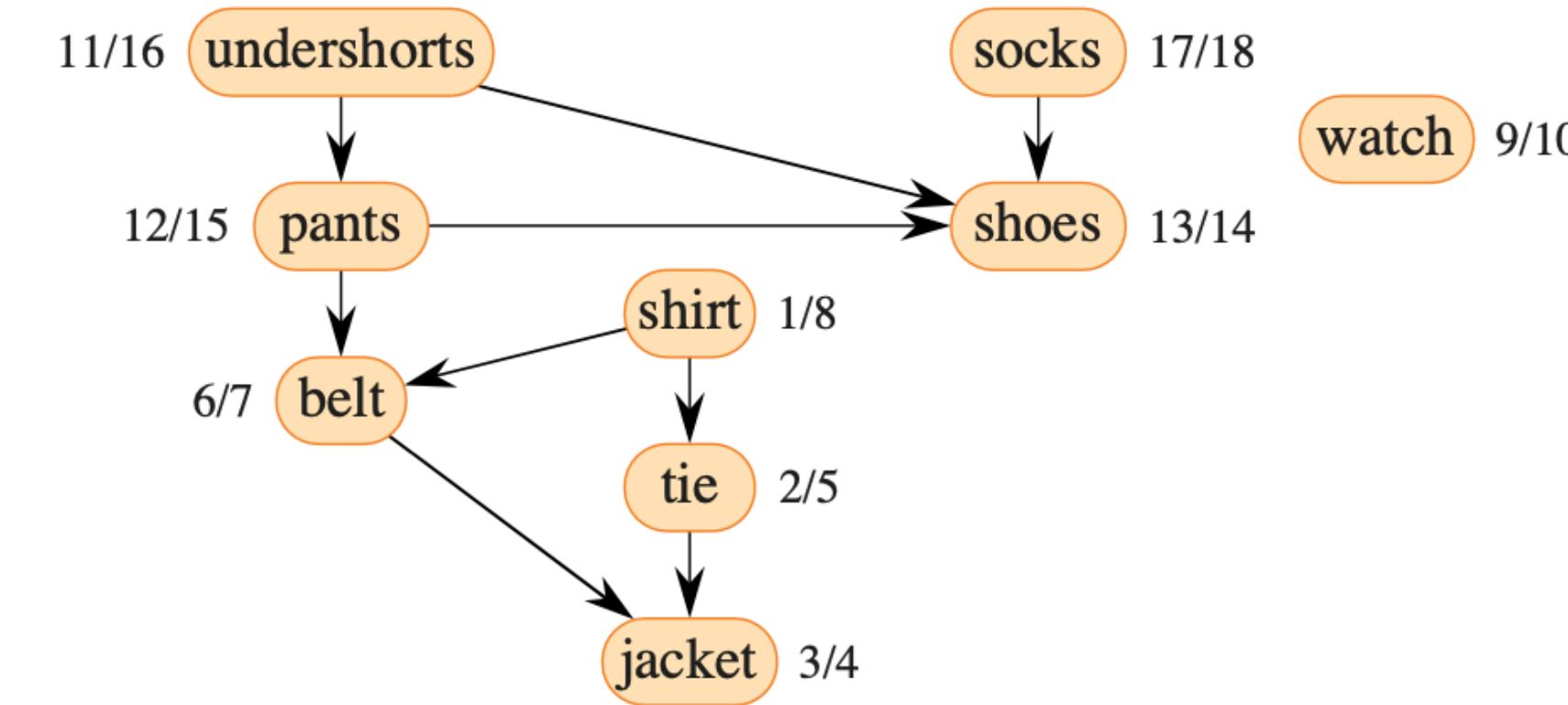
```

7
8    $time = time + 1$ 
9    $u.f = time$  //finish time
10   $u.\text{color} = \text{BLACK}$ 

```

Topological Sort

- DAG (Directed Acyclic Graph)
 - course dependence graph
 - clothes wearing graph
- topological order
 - u comes before v if $(u, v) \in E$
- Topological sort
 - DFS-based
 - Kahn's Algorithm (BFS-Based)
 - keep removing nodes with an in-degree of 0



Topological Sort

- DAG (Directed Acyclic Graph)

- course dependence graph

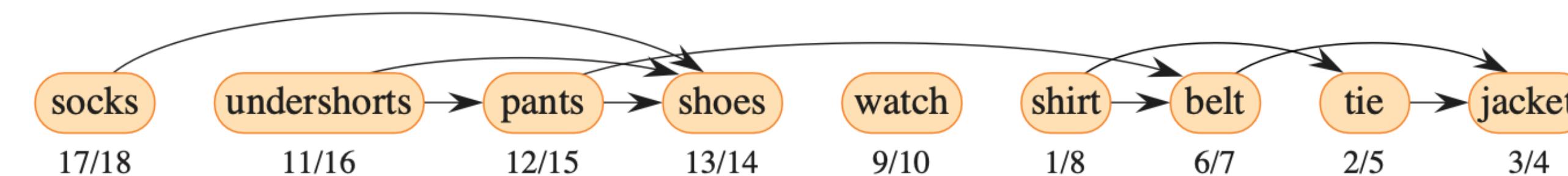
- clothes wearing graph

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- u comes before v if $(u, v) \in E$

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- Kahn's Algorithm (BFS-Based)

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Topological Sort

- DAG (Directed Acyclic Graph)

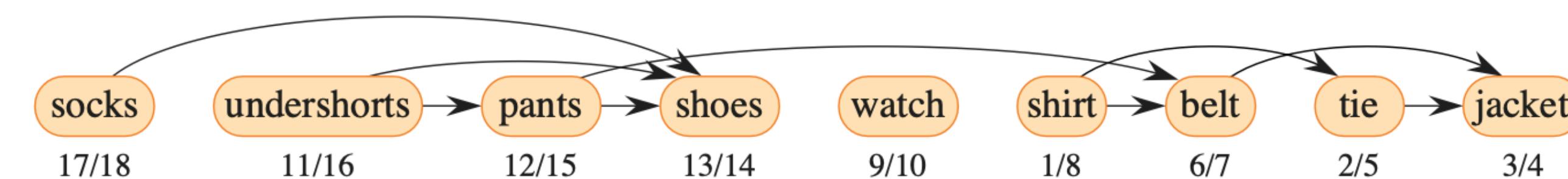
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- Kahn's Algorithm (BFS-Based)

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