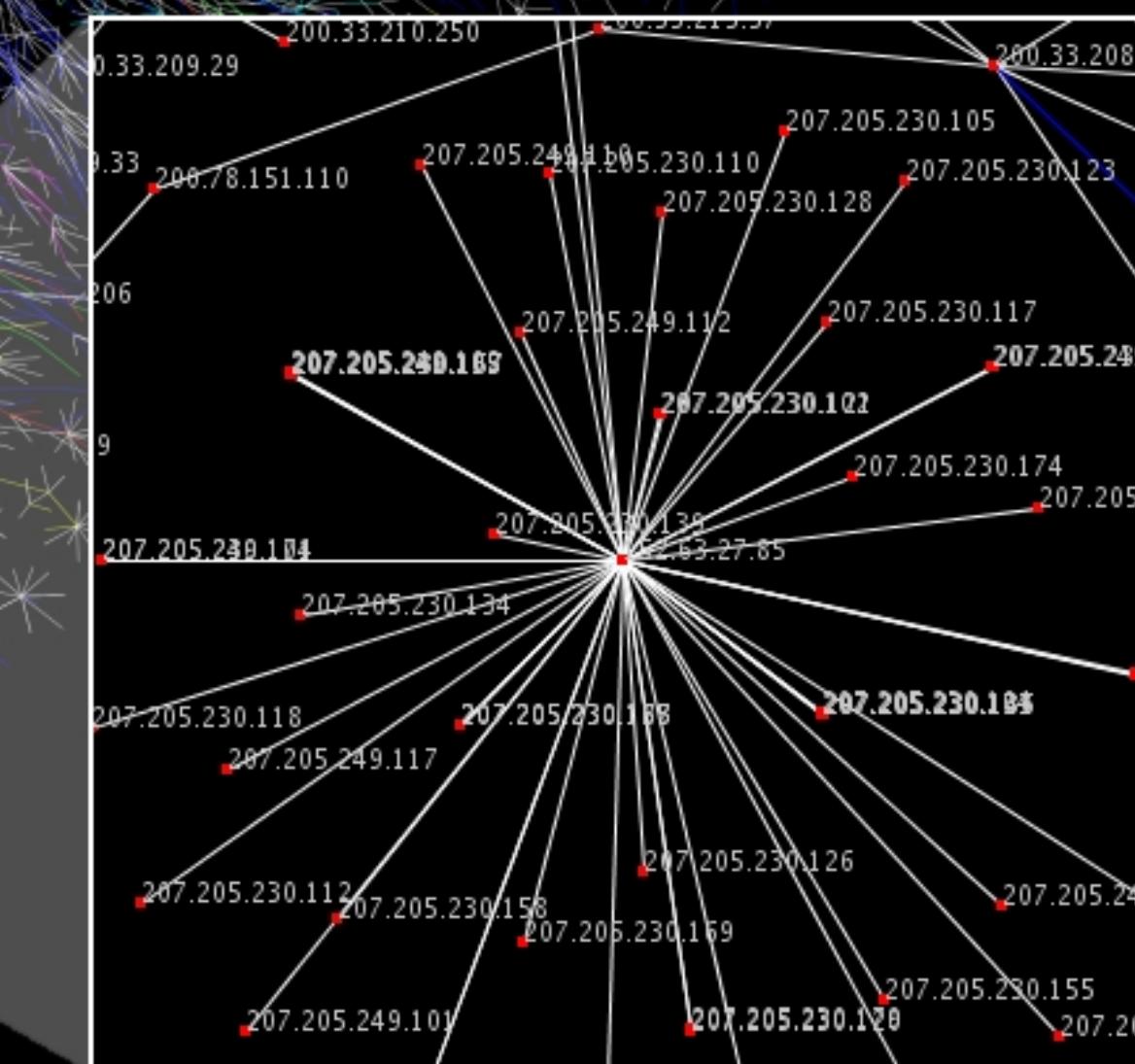
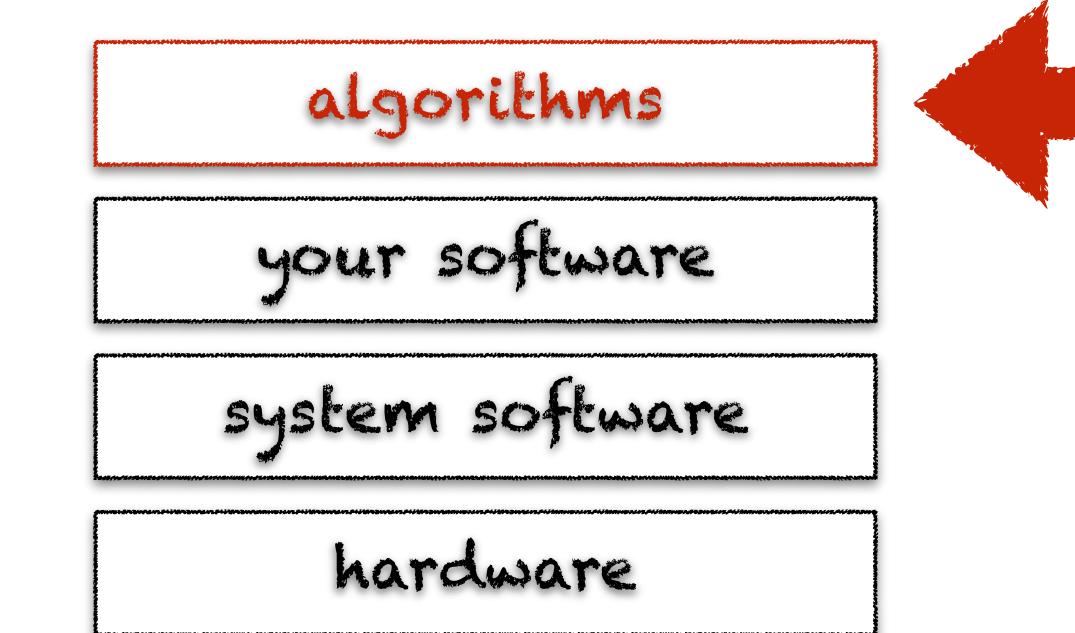


graph algorithm



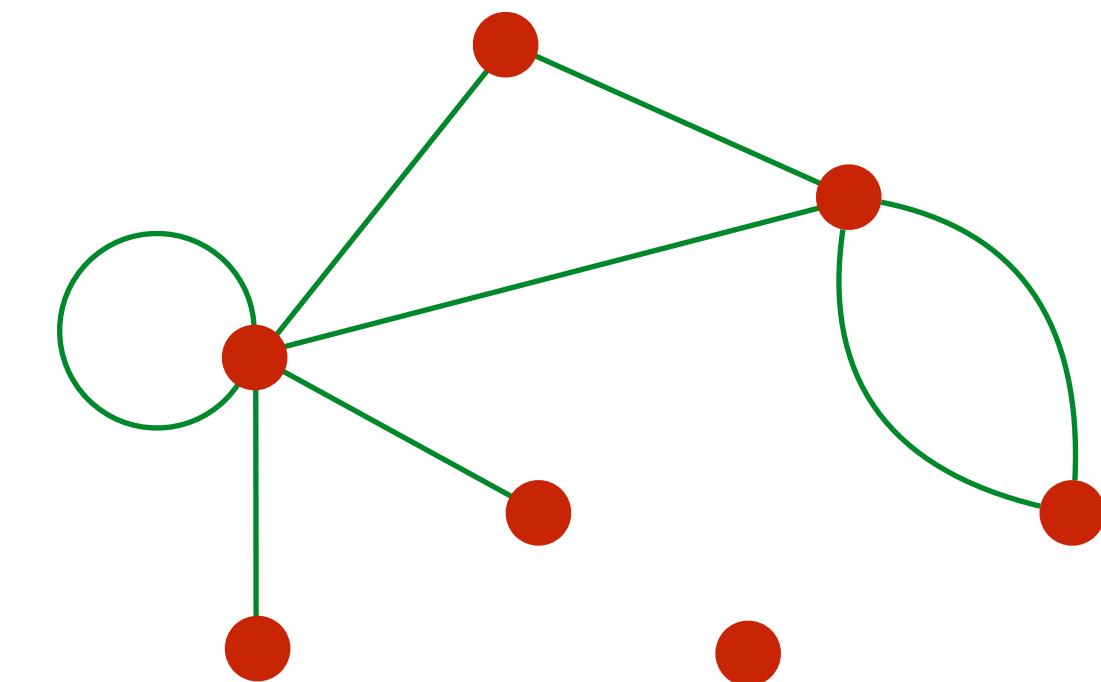
learning objectives



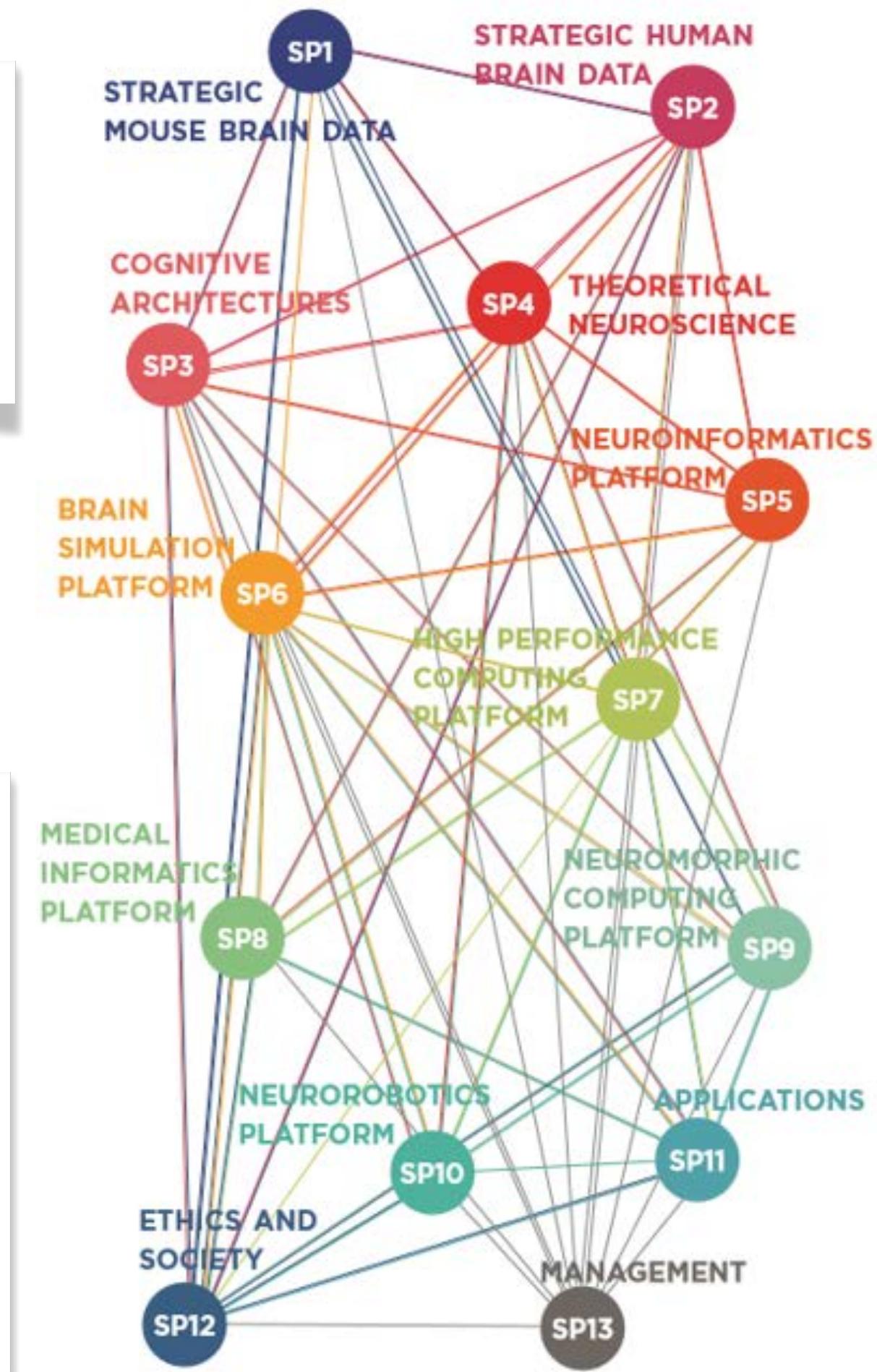
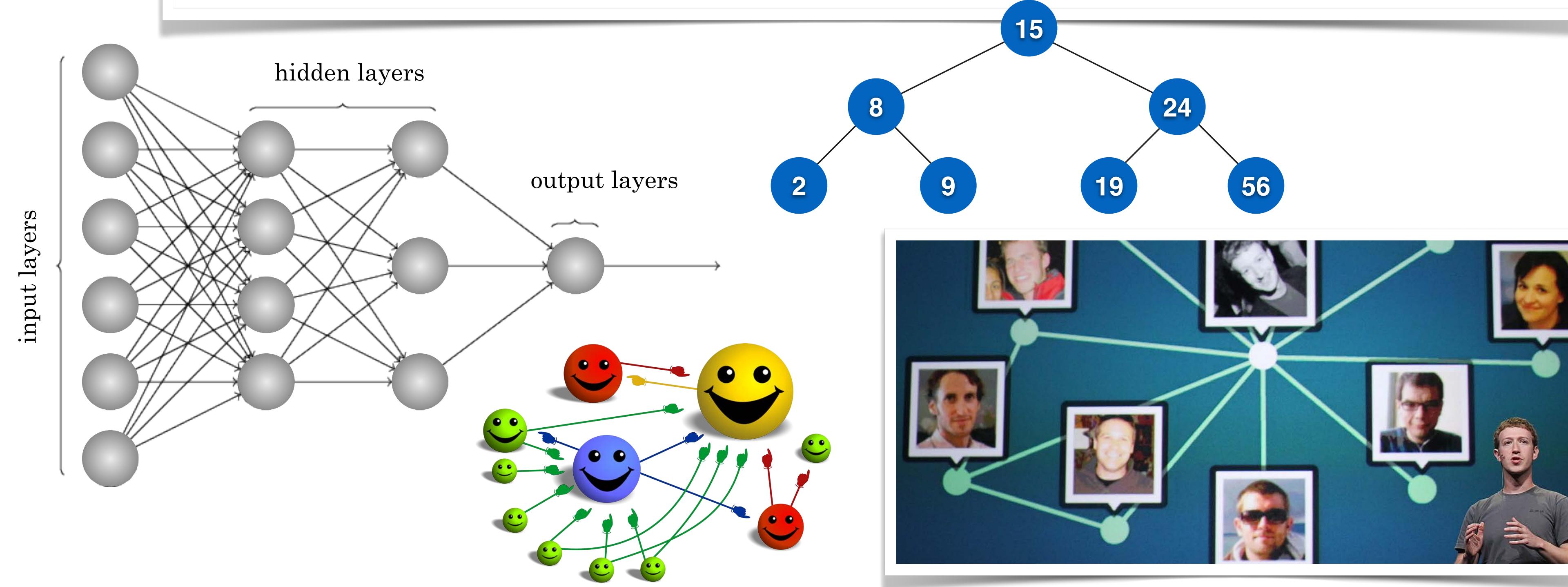
- learn what graphs are in mathematical terms
- learn how to represent graphs in computers
- learn about typical graph algorithms

why graphs?

intuitively, a graph is formed by vertices and edges between vertices



graphs are used in numerous fields to model relationships (edges) between elements (vertices)



what's a graph?

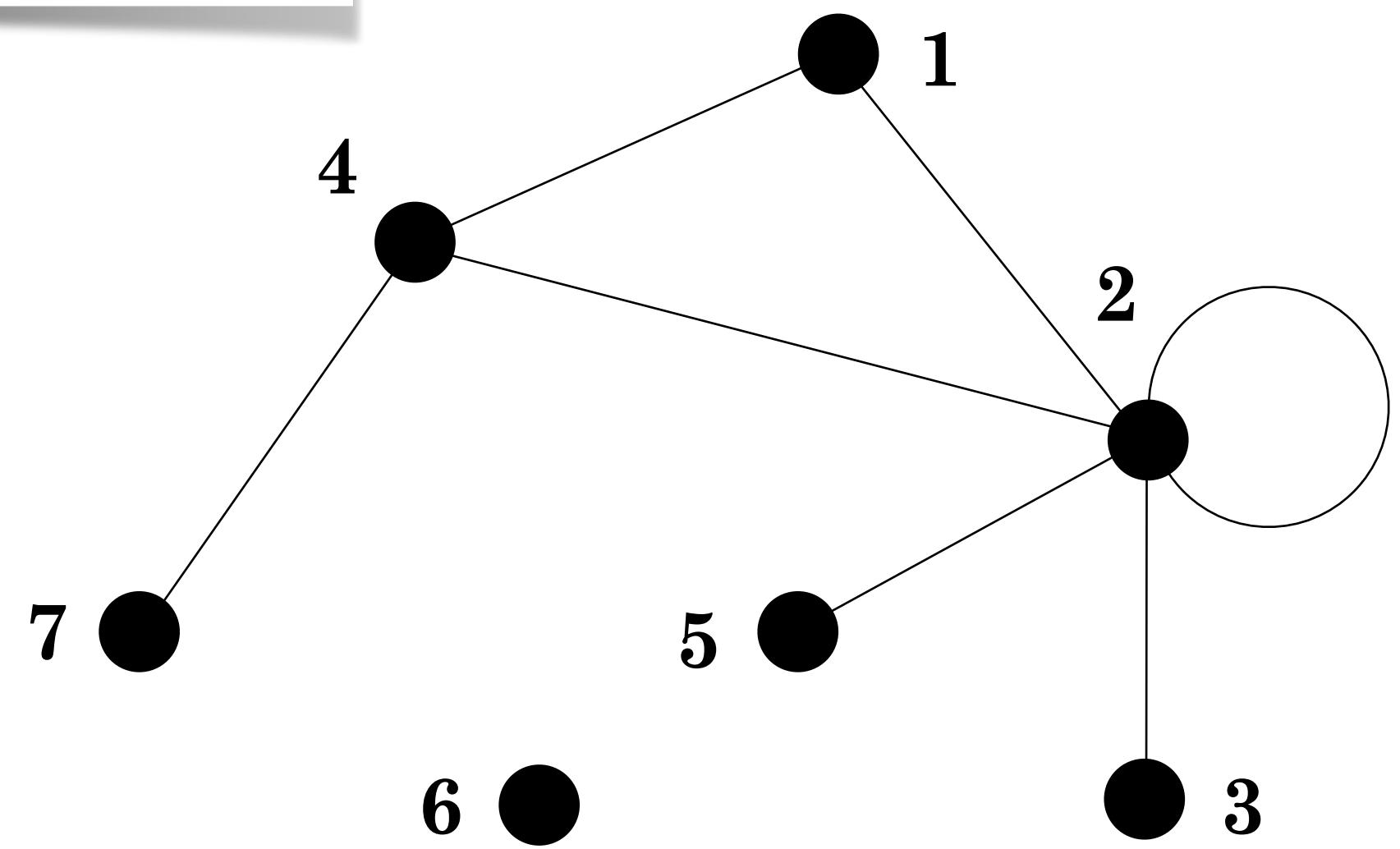
formally, a graph is a tuple $G = (V, E)$ of sets,
where V is a set of vertices (or nodes or points)
and E is a set of edges such that:

$$E \subseteq V \times V$$

example:

$$V = \{1, 2, 3, 4, 5, 6, 7\}$$

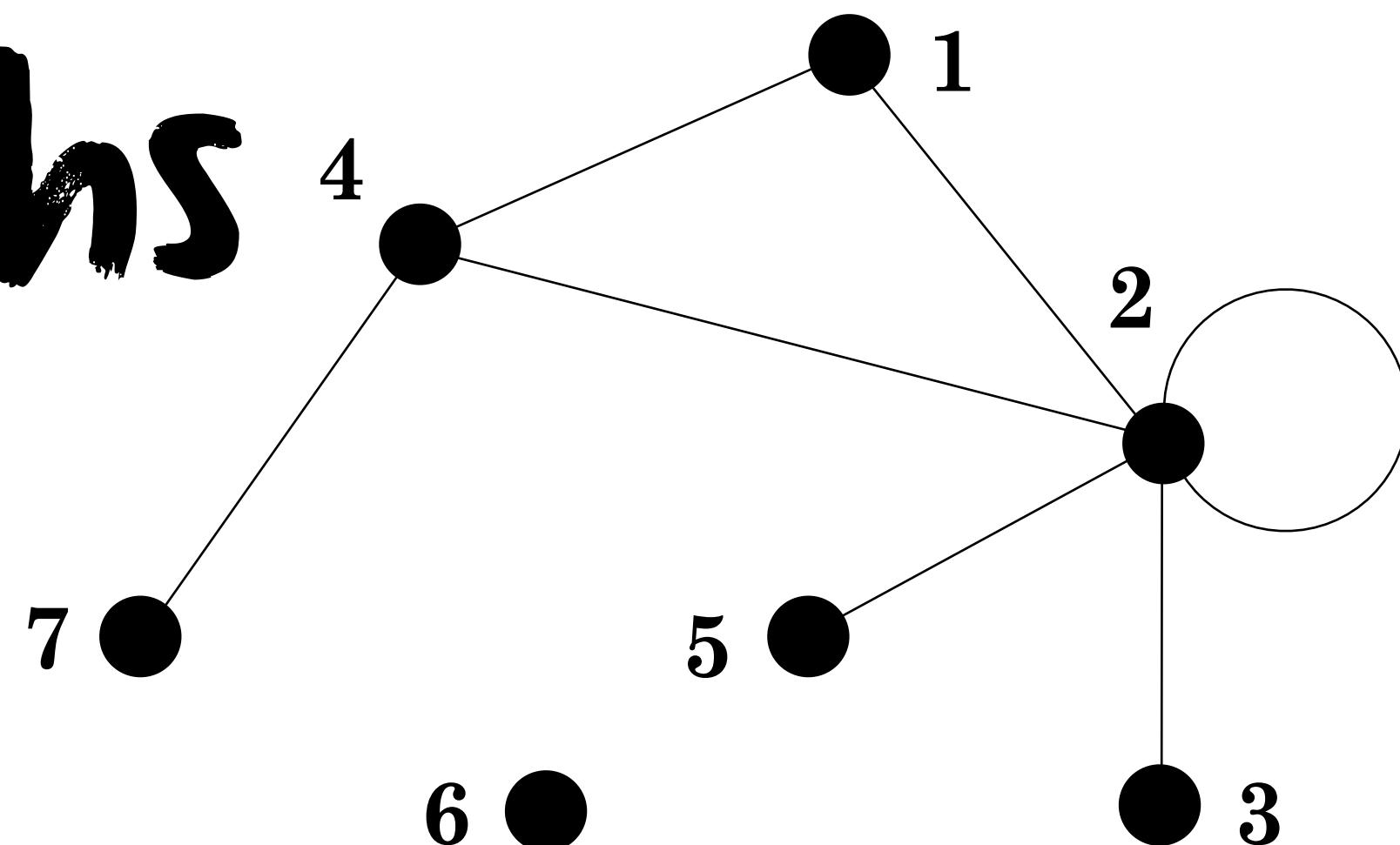
$$E = \{\{1, 2\}, \{1, 4\}, \{2\}, \{2, 3\}, \{2, 4\}, \{2, 5\}, \{4, 7\}\}$$



undirected:

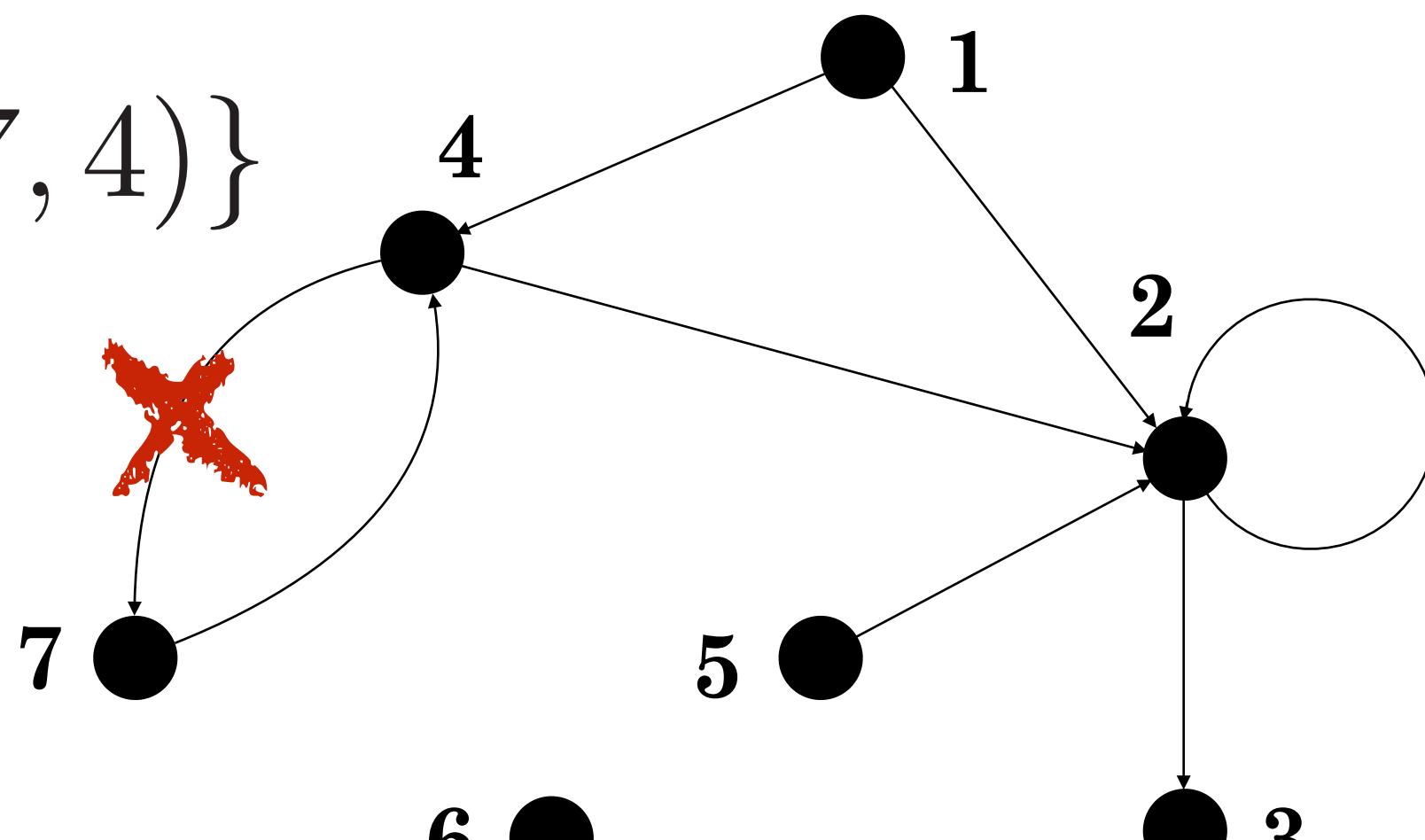
$$V = \{1, 2, 3, 4, 5, 6, 7\}$$

$$E = \{\{1, 2\}, \{1, 4\}, \{2\}, \{2, 3\}, \{2, 4\}, \{2, 5\}, \{4, 7\}\}$$



directed:

$$E = \{(1, 2), (1, 4), (2, 2), (2, 3), (4, 2), (4, 7), (5, 2), (7, 4)\}$$



oriented:

$$E = \{(1, 2), (1, 4), (2, 2), (2, 3), (4, 2), (4, 7), (5, 2), (7, 4)\}$$

notations & metrics

let G be graph, $G.V$ denotes its set of vertices and $G.E$ its set of edges

the edge between vertices x and y is noted $\{x,y\}$, (x,y) or simply xy

the order of G , written $|G|$, is the number of its vertices, whereas $\|G\|$ denotes its number of edges

graph G is sparse if $\|G\| \ll |G|^2$ and it is dense if $\|G\| \approx |G|^2$

two vertices x and y are adjacent or neighbors if $xy \in G$

if all the vertices of G are pairwise adjacent, then G is complete

notations & metrics

a path from vertex x to vertex y is a sequence $\langle v_0, v_1, \dots, v_k \rangle$ of vertices $v_i \in V$ where $x = v_0$ and $y = v_k$, such that $\forall i \in \{1, \dots, k\} : (v_{i-1}, v_i) \in E$

a graph is connected if every pair of vertices is connected via a path

a path $\langle v_0, v_1, \dots, v_k \rangle$ is a cycle if vertices $v_0 = v_k$

we can store attributes in vertices and edges using the dotted notation,
e.g., $v.\text{color}$ stores a color attribute in vertex v , while $e.\text{weight}$ and
 $(x,y).\text{weight}$ store a weight attribute in edge e and edge (x,y) respectively

notations & metrics

let $G = (V, E)$ and $G' = (V', E')$ be two graphs, if $V' \subseteq V$ and $E' \subseteq E$,
then G' is a **subgraph** of G , which we write $G' \subseteq G$

let $G = (V, E)$ and $G' = (V', E')$ be two graphs and $G' \subseteq G$,
if $V' = V$, G' is a **spanning subgraph of G**

the **degree** (or **valency**) of a vertex v is the
number of neighbors of v and is noted $d(v)$

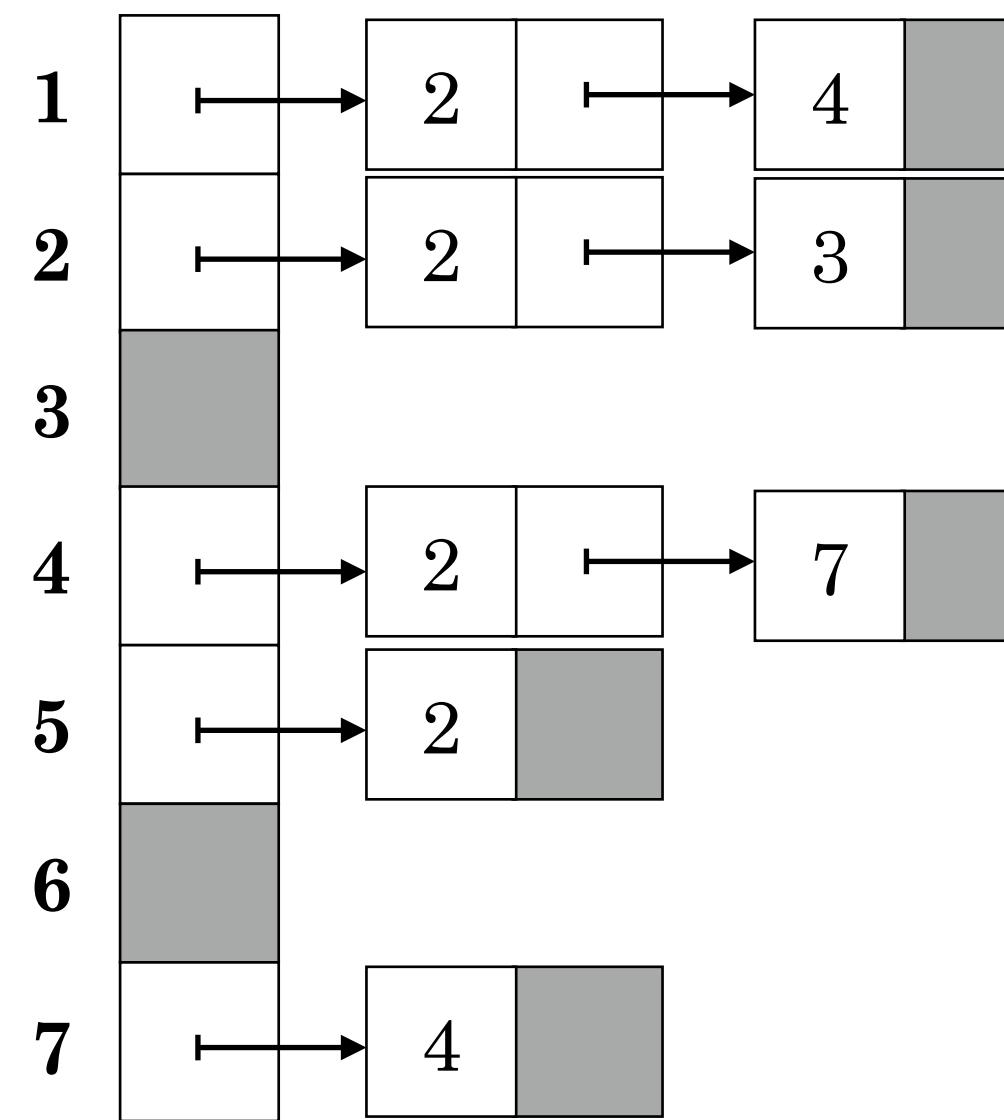
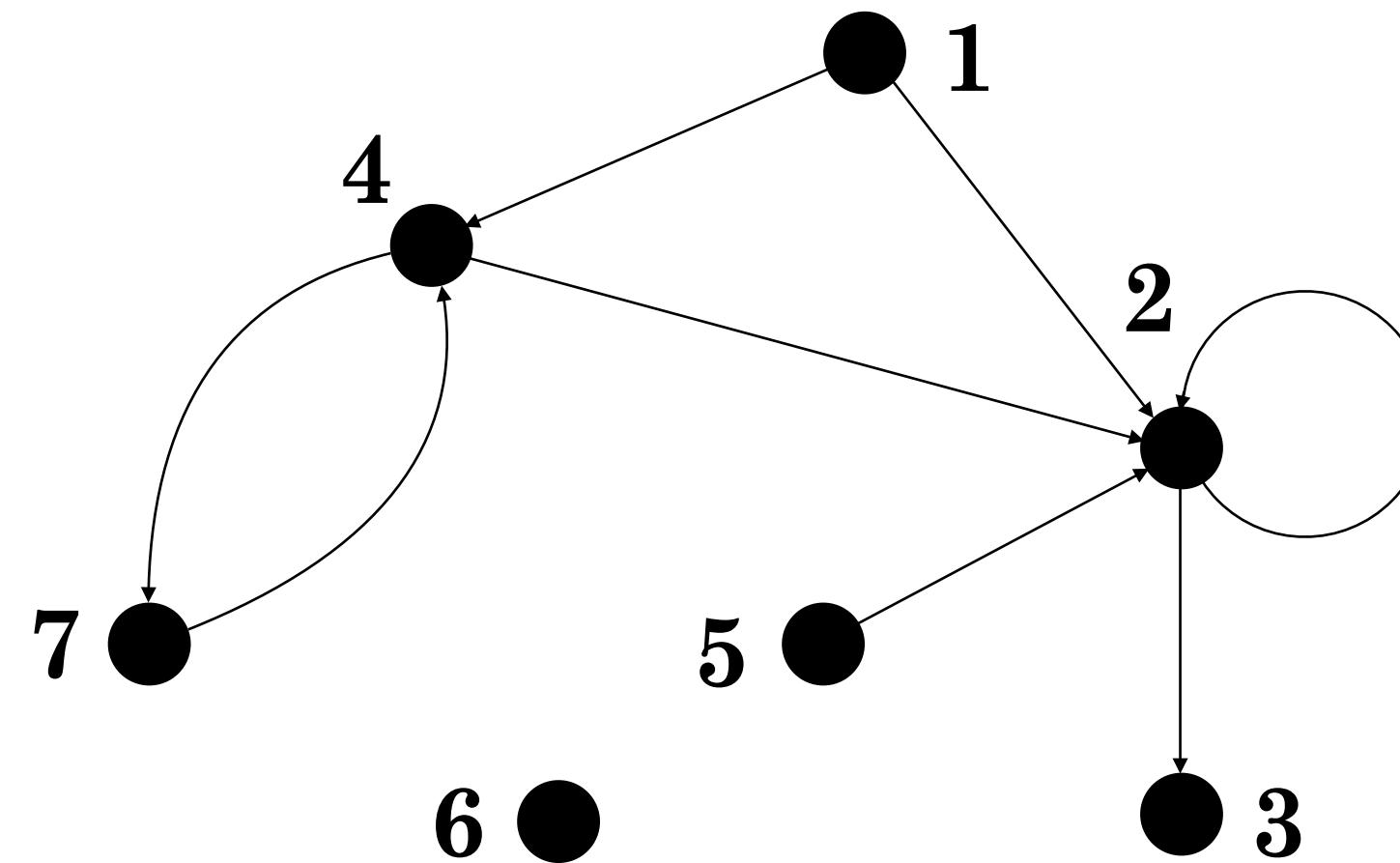
we defined $\delta(G) = \min \{ d(v) \mid v \in V \}$
as the **minimum degree** of G

we defined $\Delta(G) = \max \{ d(v) \mid v \in V \}$
as the **maximum degree** of G

we defined $d(G) = \frac{1}{|V|} \sum_{v \in V} d(v)$ as the **average degree** of G

representing graphs

directed



adjacency list

	1	2	3	4	5	6	7
1	0	1	0	1	0	0	0
2	0	1	1	0	0	0	0
3	0	0	0	0	0	0	0
4	0	1	0	0	0	0	1
5	0	1	0	0	0	0	0
6	0	0	0	0	0	0	0
7	0	0	0	1	0	0	0

adjacency matrix

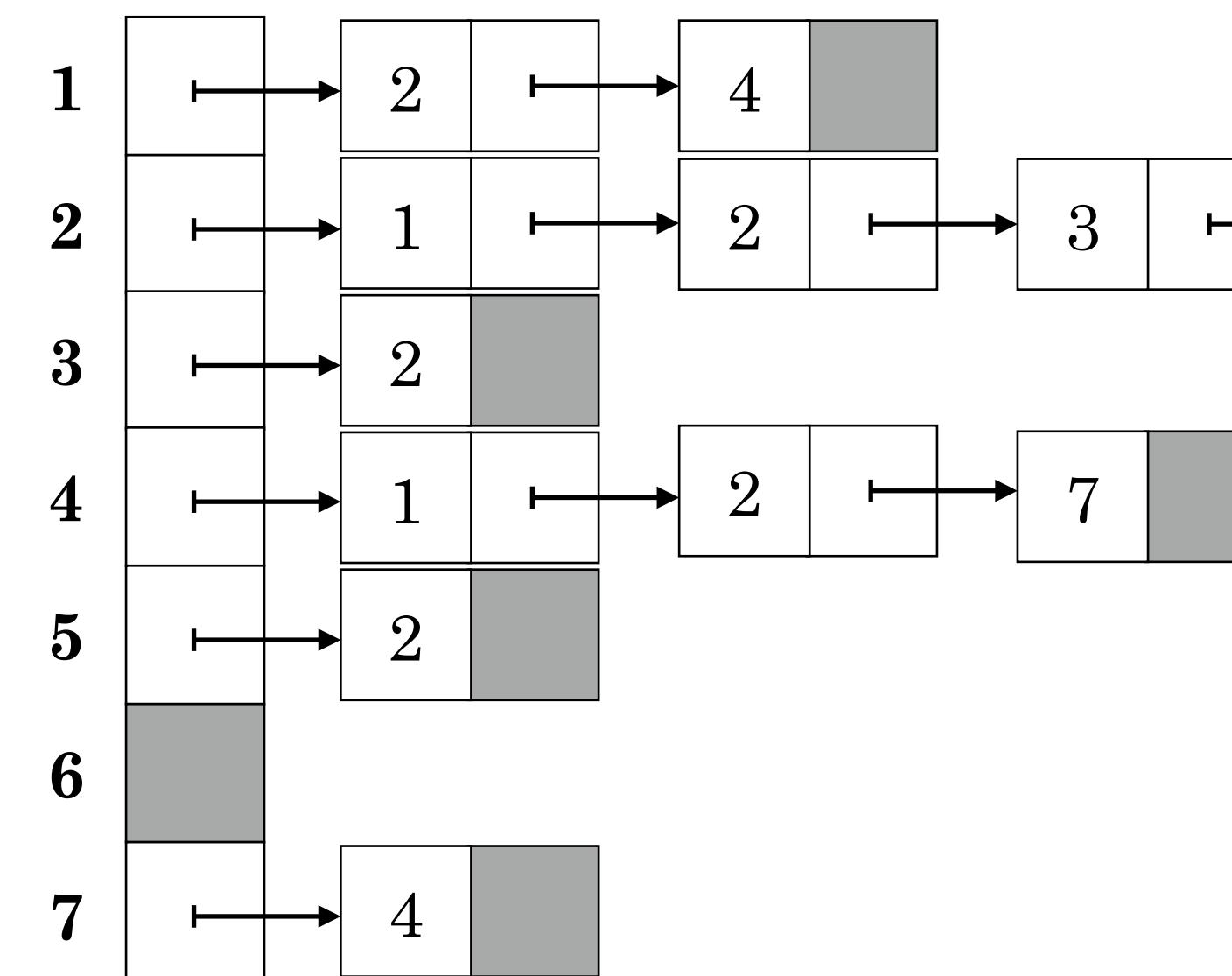
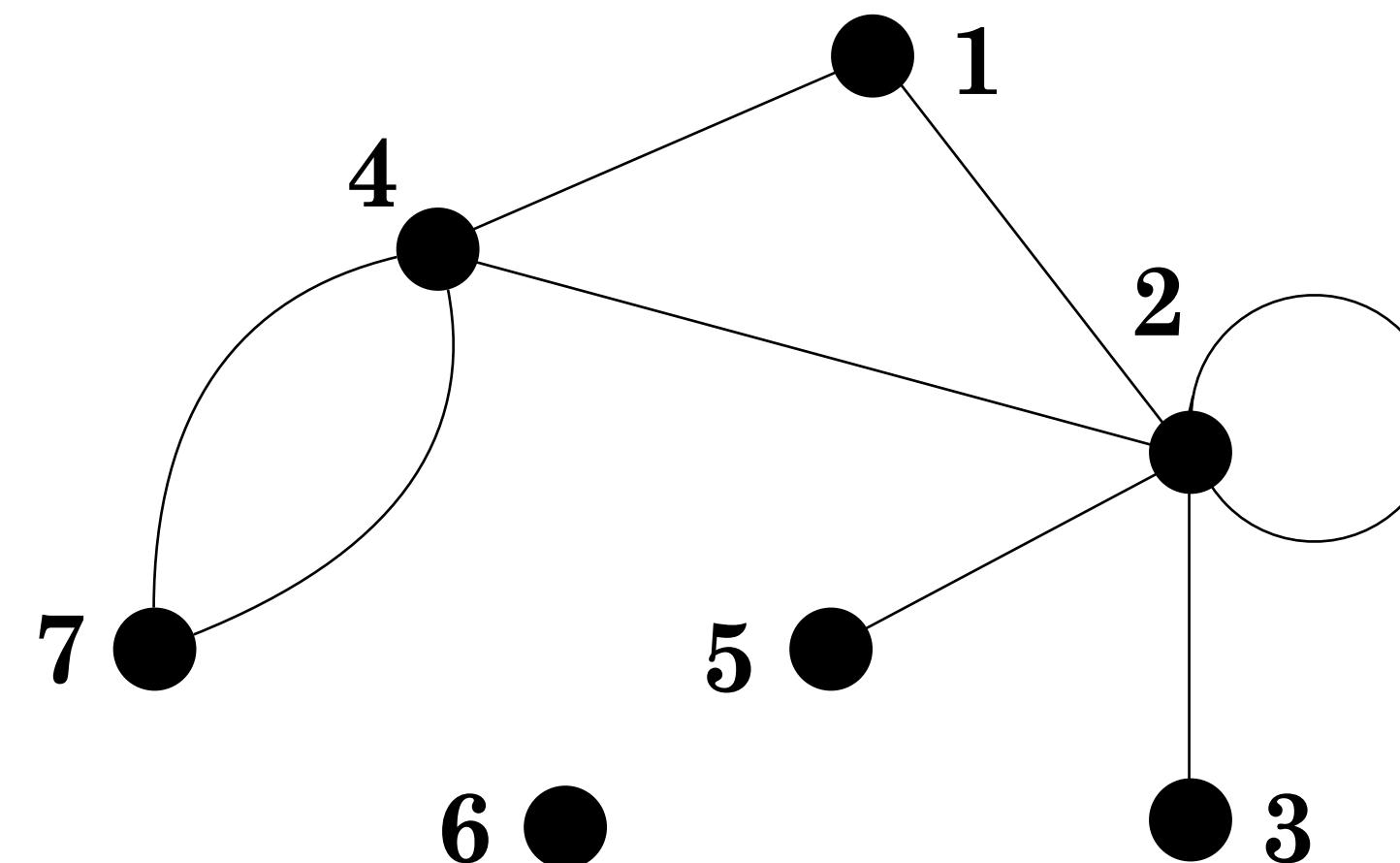
an **adjacency list** is best suited for representing a **sparse graph**

most graph algorithms rely on adjacency lists

an **adjacency matrix** is best suited for representing a **dense graph** or when the algorithm needs to know quickly if there exists an edge connecting two vertices

representing graphs

undirected



adjacency list

	1	2	3	4	5	6	7
1	0	1	0	1	0	0	0
2	1	1	1	1	1	0	0
3	0	1	0	0	0	0	0
4	1	1	0	0	0	0	1
5	0	1	0	0	0	0	0
6	0	0	0	0	0	0	0
7	0	0	0	1	0	0	0

adjacency matrix

typical problems

breadth-first search

minimum spanning tree

single-source shortest paths

breadth-first search

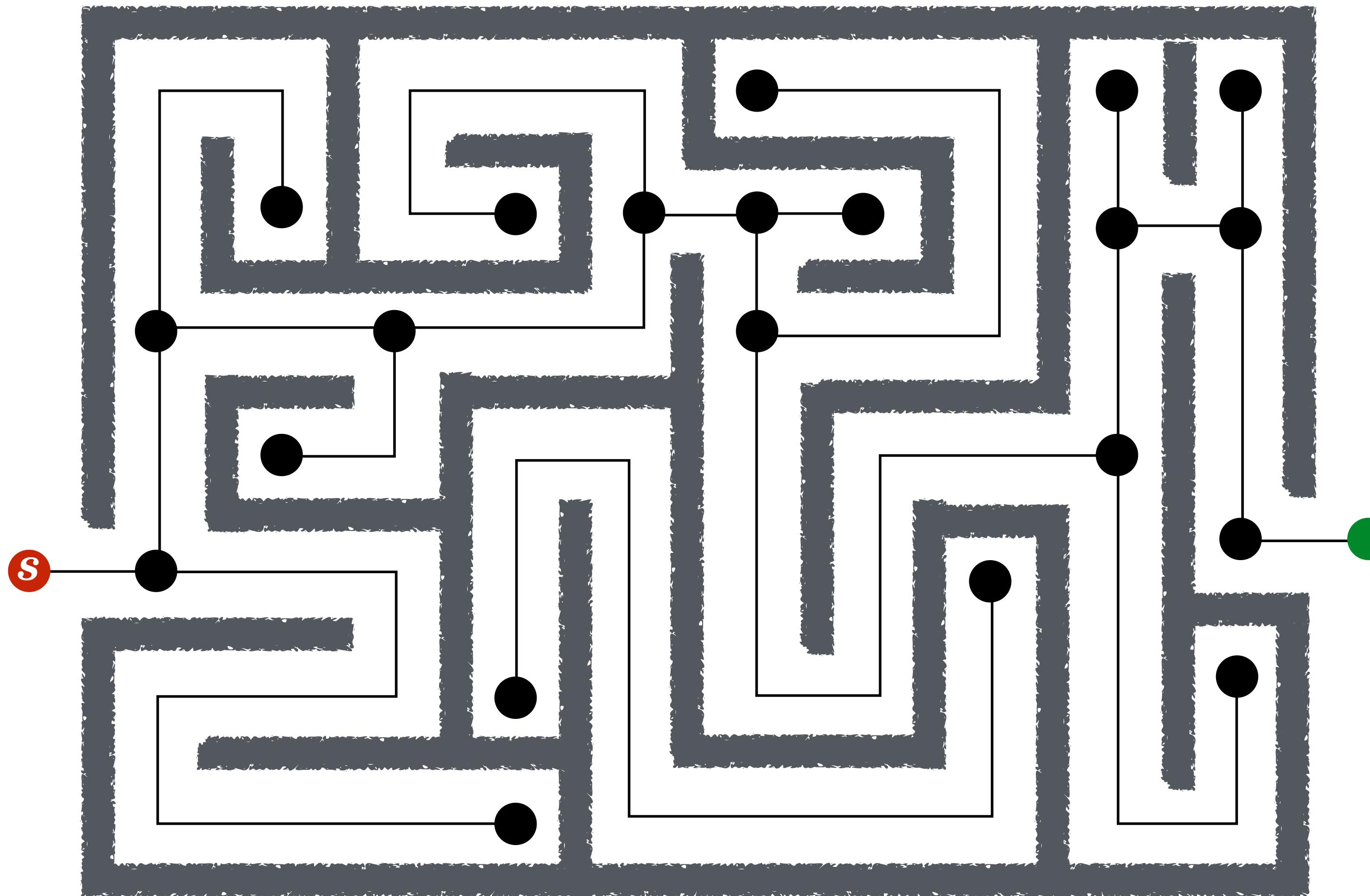
given graph G and a source vertex $s \in G$,
it discover every vertex reachable from s

it computes the distance from s to every vertex $v \in G$

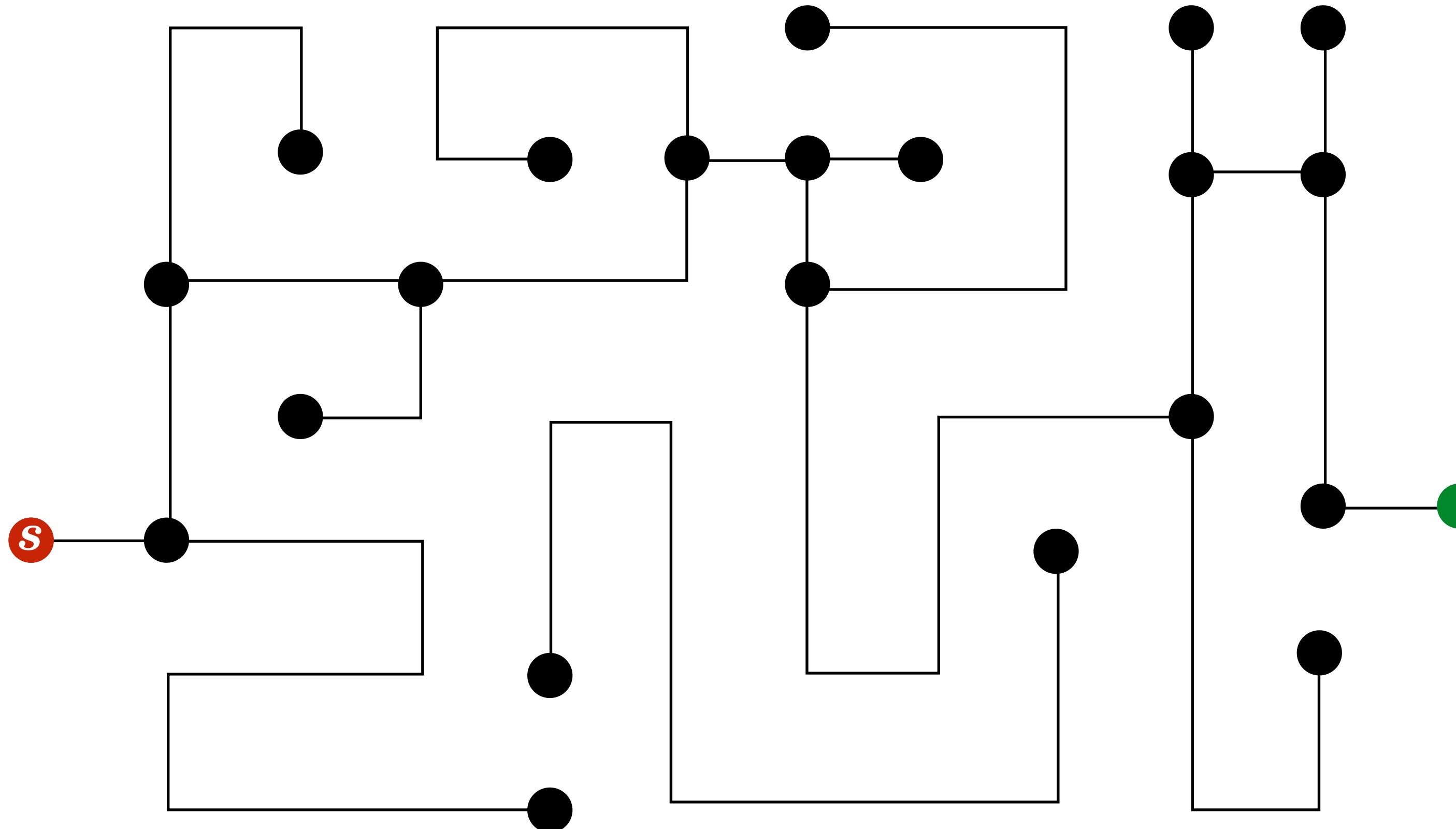
it produces a breadth-first tree rooted at s
that contains all reachable vertices from s

the search is said to be breadth-first because it
discovers all vertices at distance k from s before
discovering any vertices at distance $k + 1$

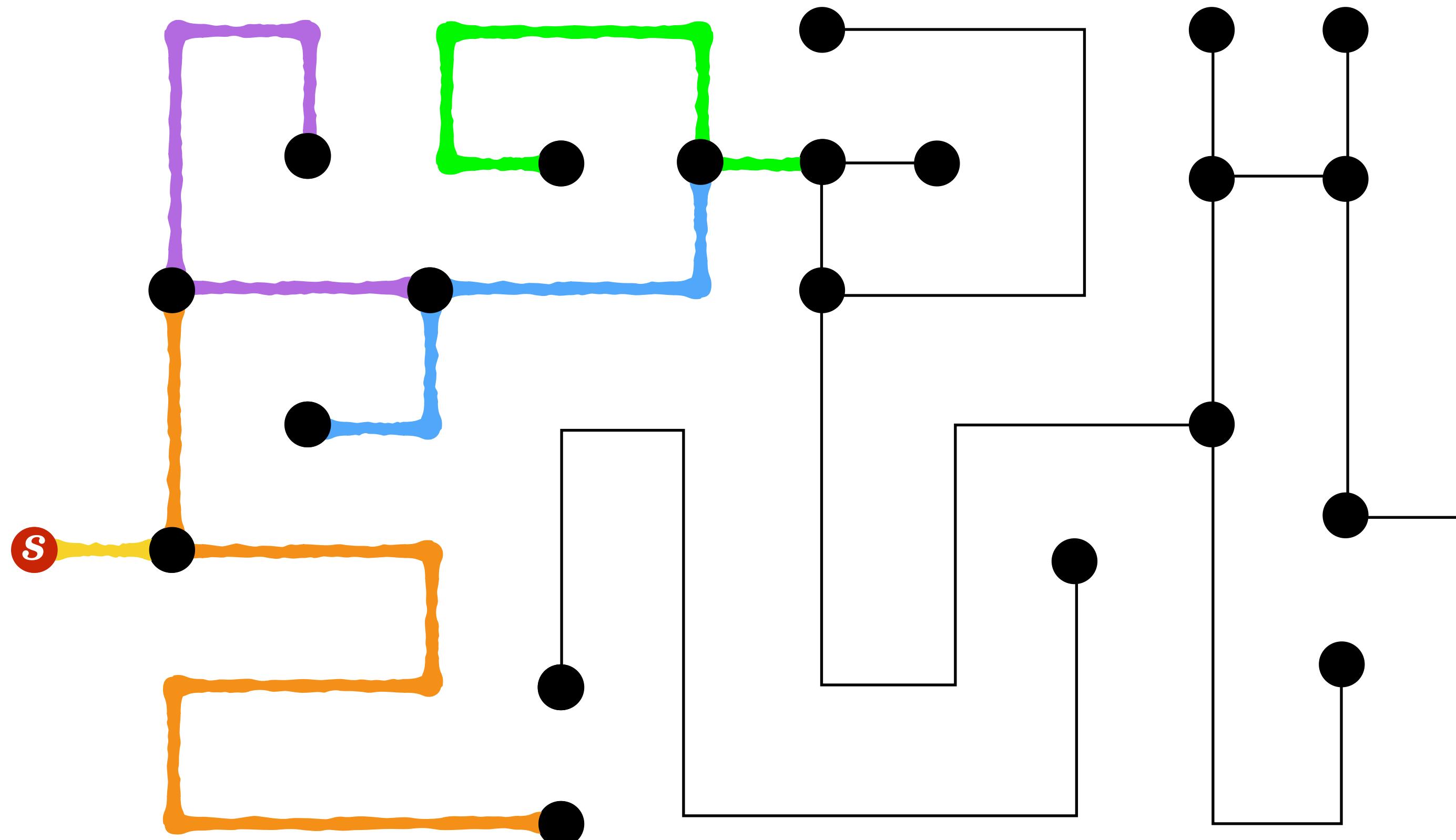
breadth-first search



breadth-first search



breadth-first search



1 —

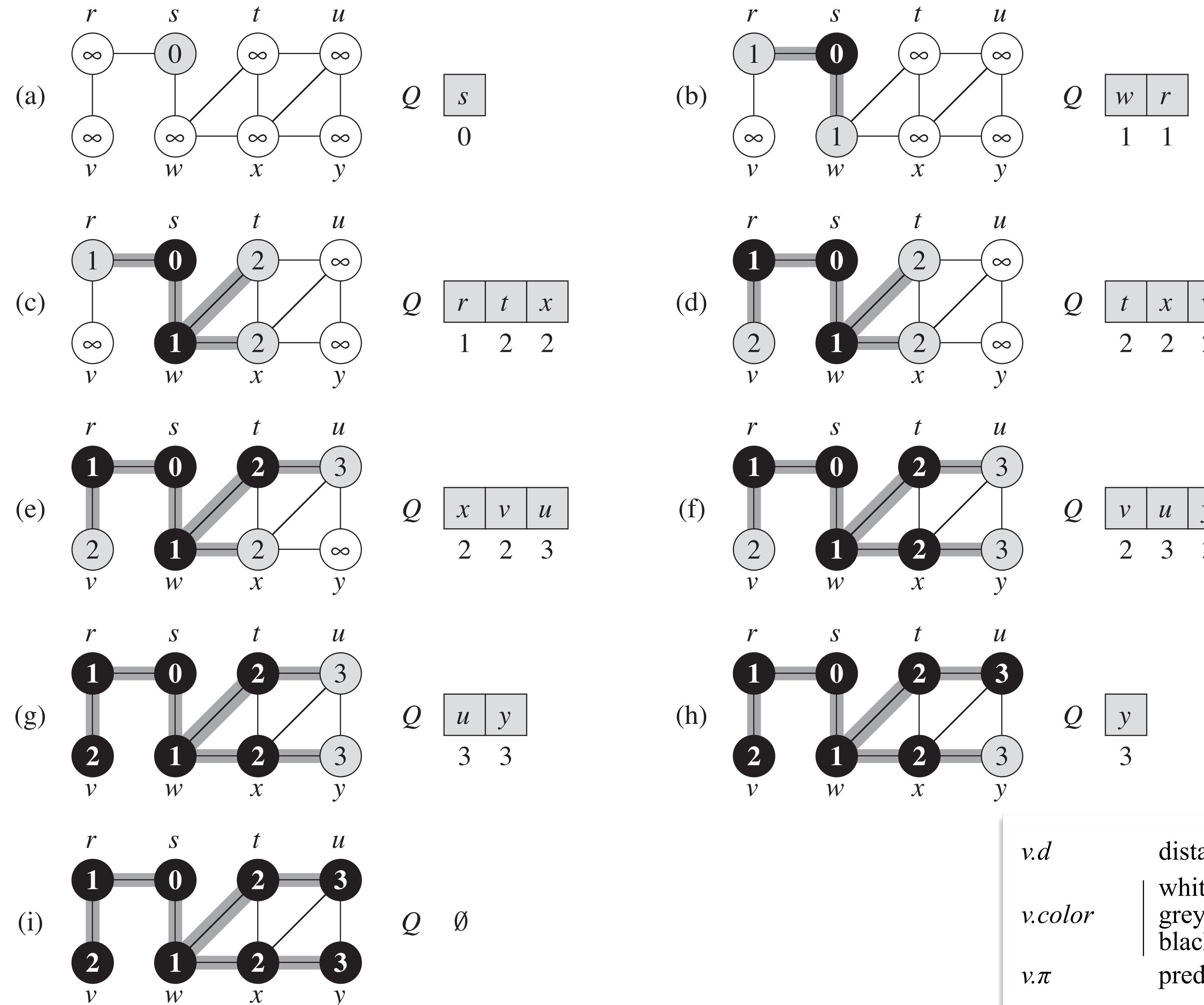
2 —

3 —

4 —

5 —

breadth-first search



BFS(G, s)

```

1  for each vertex  $u \in G.V - \{s\}$ 
2     $u.color = \text{WHITE}$ 
3     $u.d = \infty$ 
4     $u.\pi = \text{NIL}$ 
5     $s.color = \text{GRAY}$ 
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  $v \in G.Adj[u]$ 
13       if  $v.color == \text{WHITE}$ 
14          $v.color = \text{GRAY}$ 
15          $v.d = u.d + 1$ 
16          $v.\pi = u$ 
17         ENQUEUE( $Q, v$ )
18        $u.color = \text{BLACK}$ 
```

$v.d$

distance from source s

$v.color$

white : undiscovered

grey : discovered with some neighbors discovered

black : discovered with all neighbors discovered

$v.\pi$

predecessor in breadth-first tree

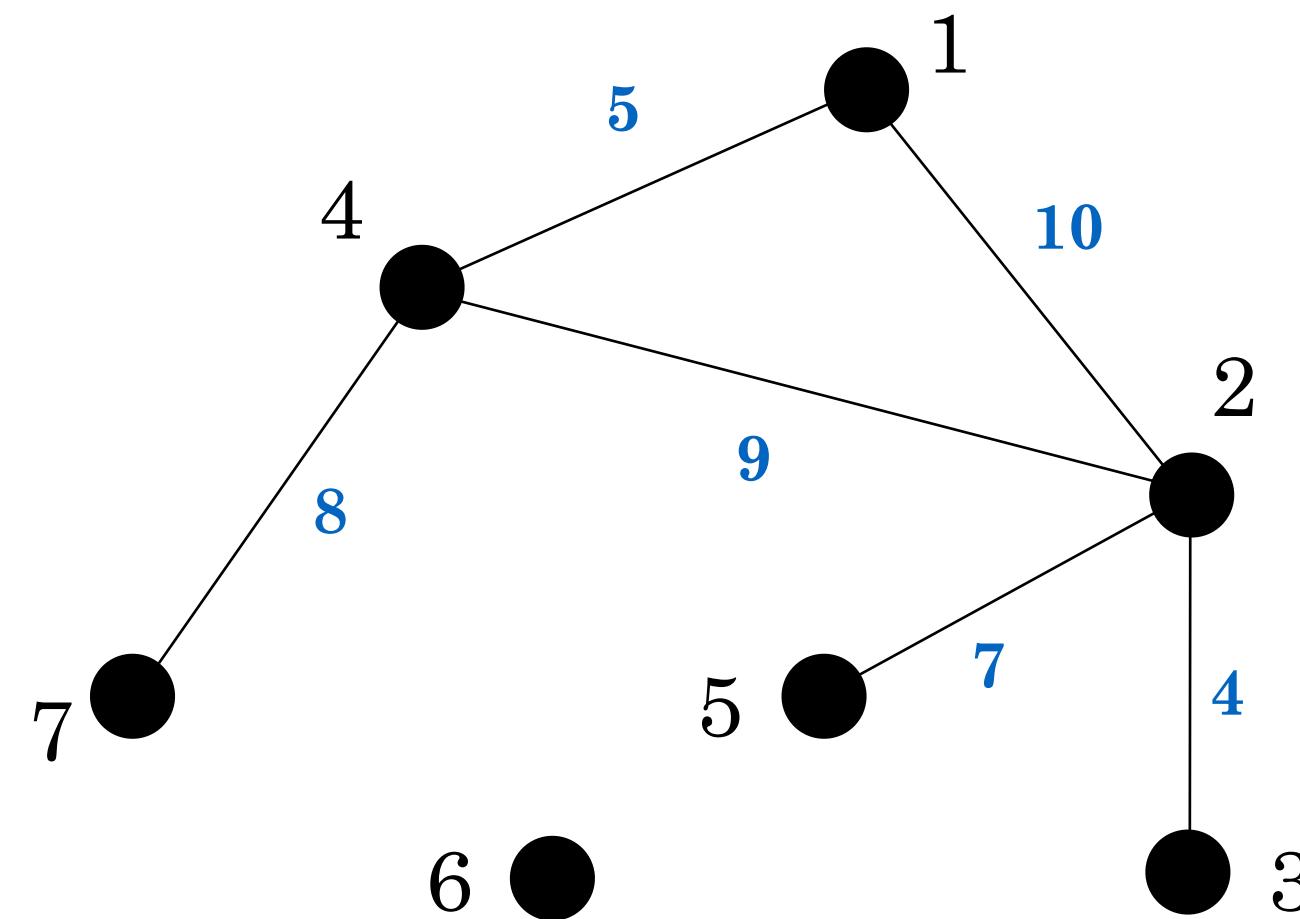
typical problems

breadth-first search

minimum spanning tree

single-source shortest paths

minimum spanning tree



a **weighted graph** $G_w = (G, w)$ is a tuple composed of a graph $G = (V, E)$ and of a **function** $w : E \rightarrow \mathbb{R}$ associating a **weight** w_e to each edge $e \in E$

a **minimum (weight) spanning tree** of graph $G_w = (G, w)$ is a connected subgraph (V', E') such that:

1

$$V' = V$$

2

(V', E') does not contain any cycles

3

$\sum_{e \in E'} w_e$ is minimal across all subgraphs fulfilling 1 and 2

minimum spanning tree

a disjoint-set data structure maintains a collection $S = \{S_1, S_2, \dots, S_k\}$ of disjoint dynamic sets where each set is identified by a member of the set known as its representative

a disjoint-set data structure supports the following operations:

MAKE-SET(x) creates a new set whose only member and its representative is x

UNION(x, y) merges the dynamic sets that contain x and y , say S_x and S_y , into a new set that is the union of these two sets

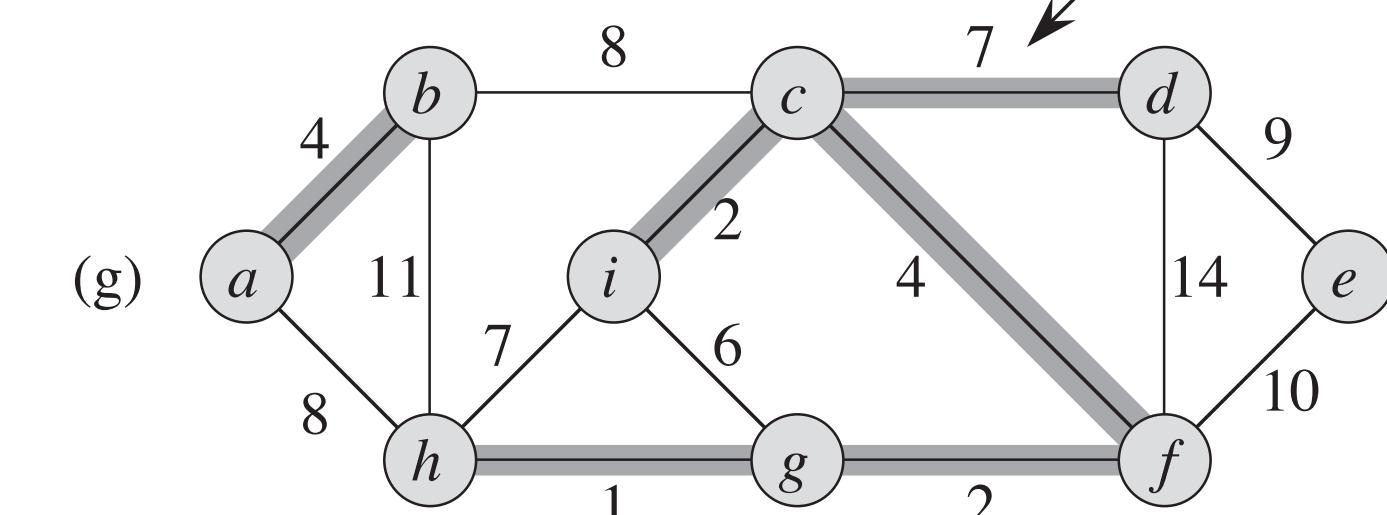
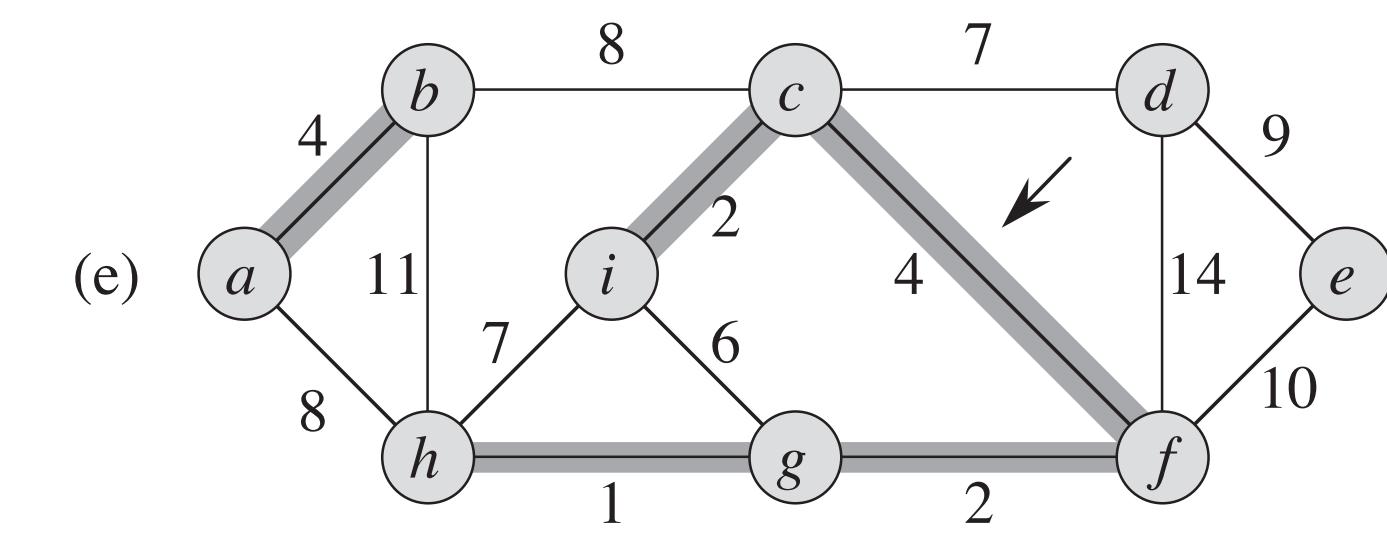
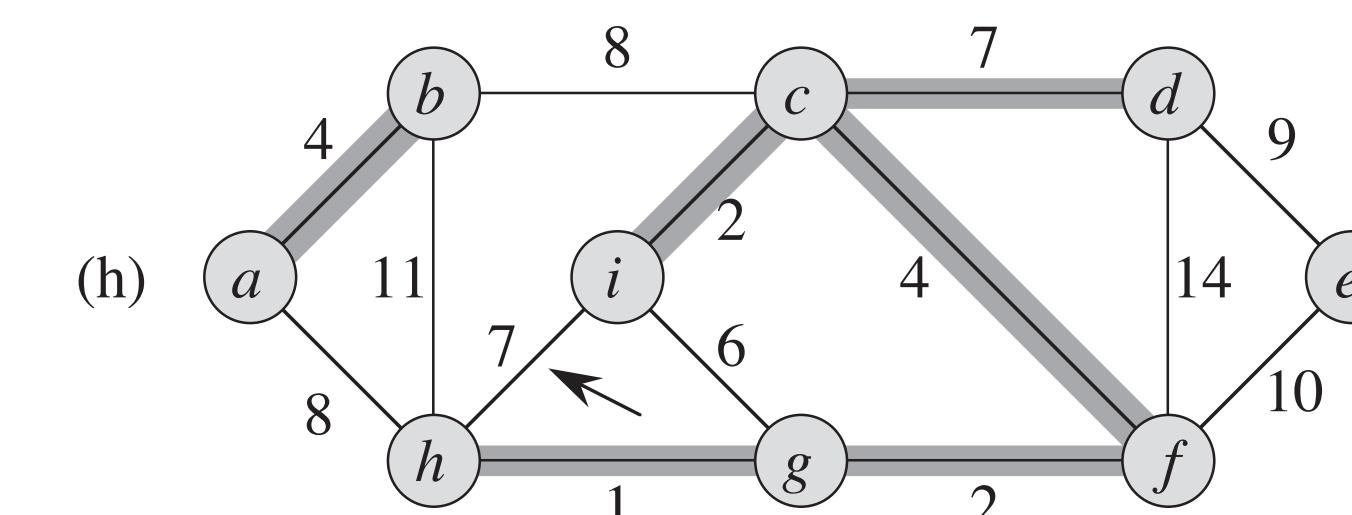
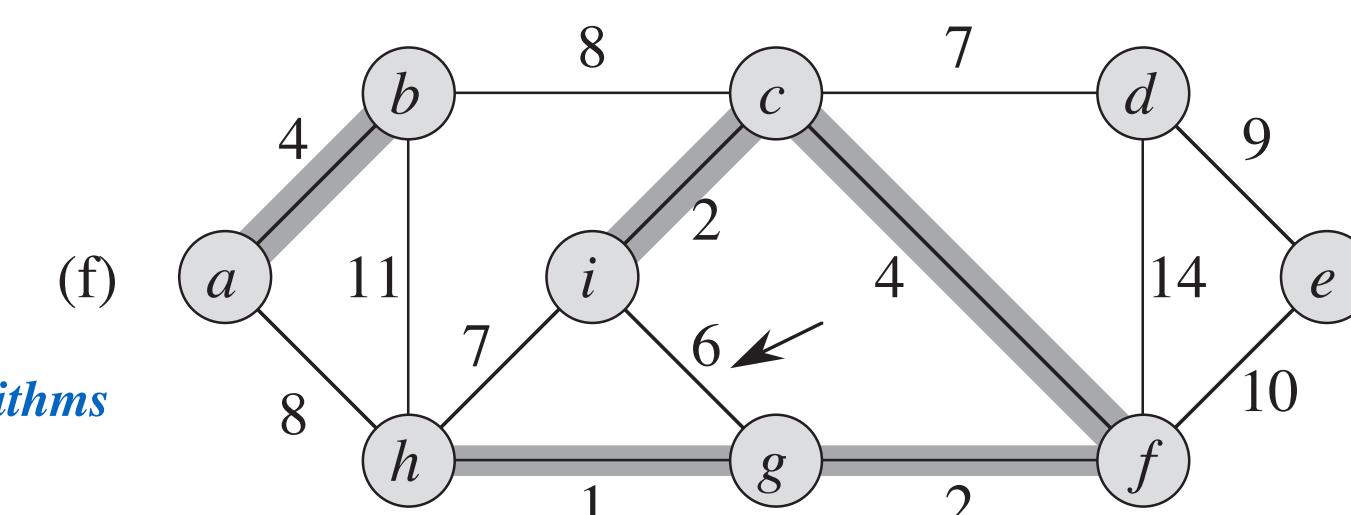
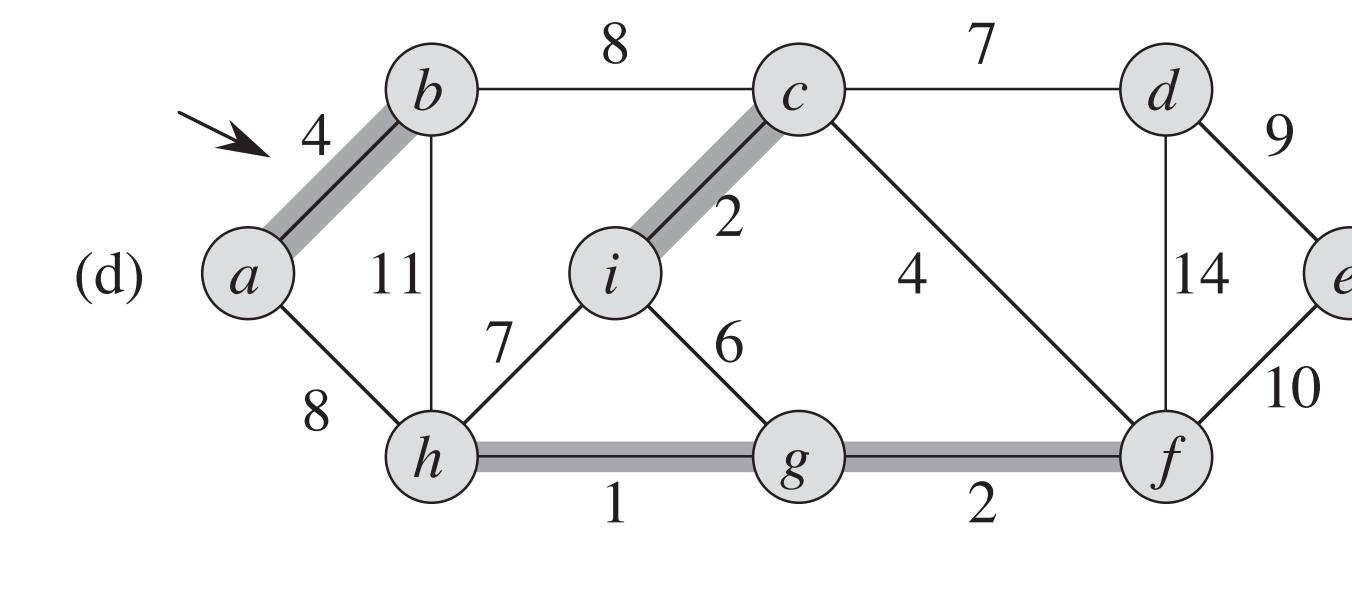
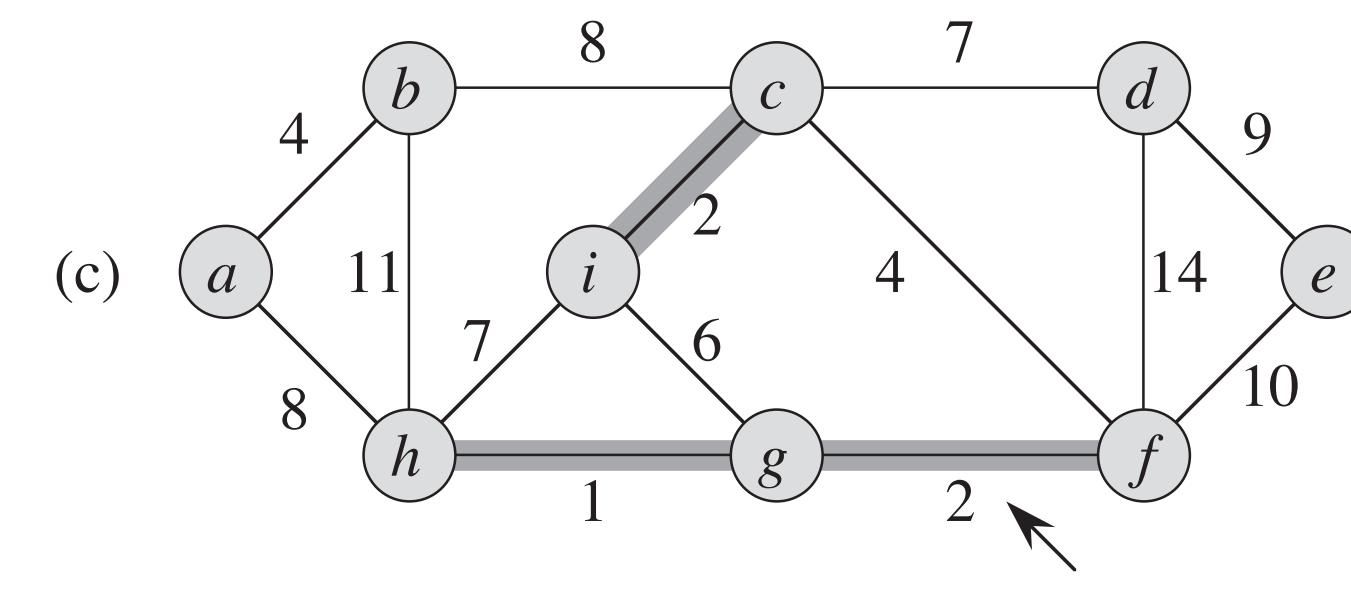
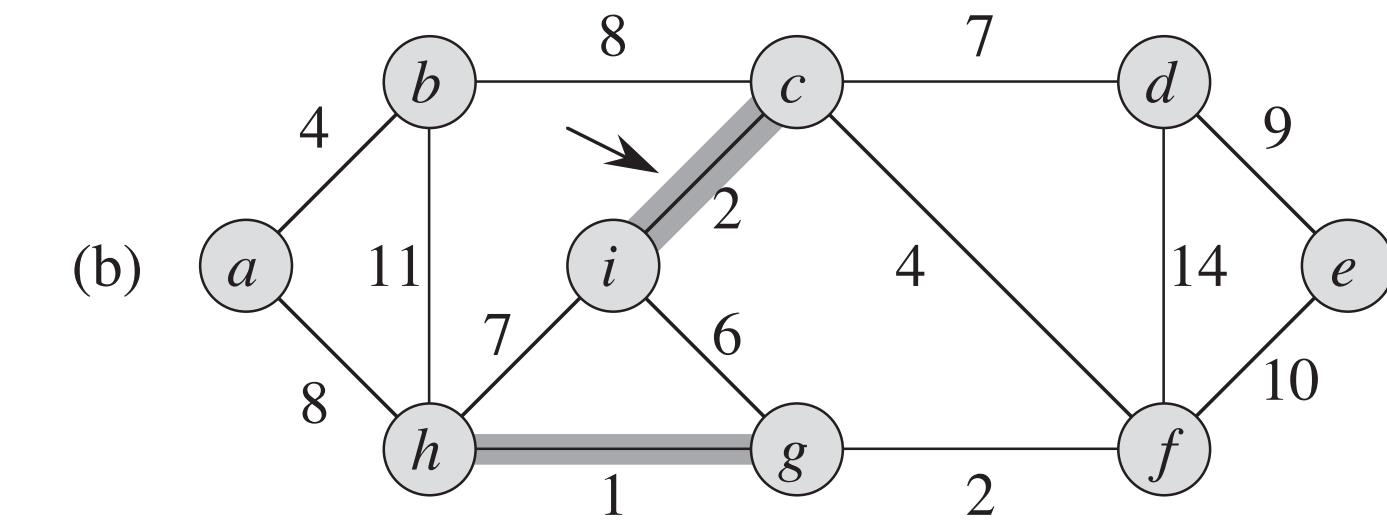
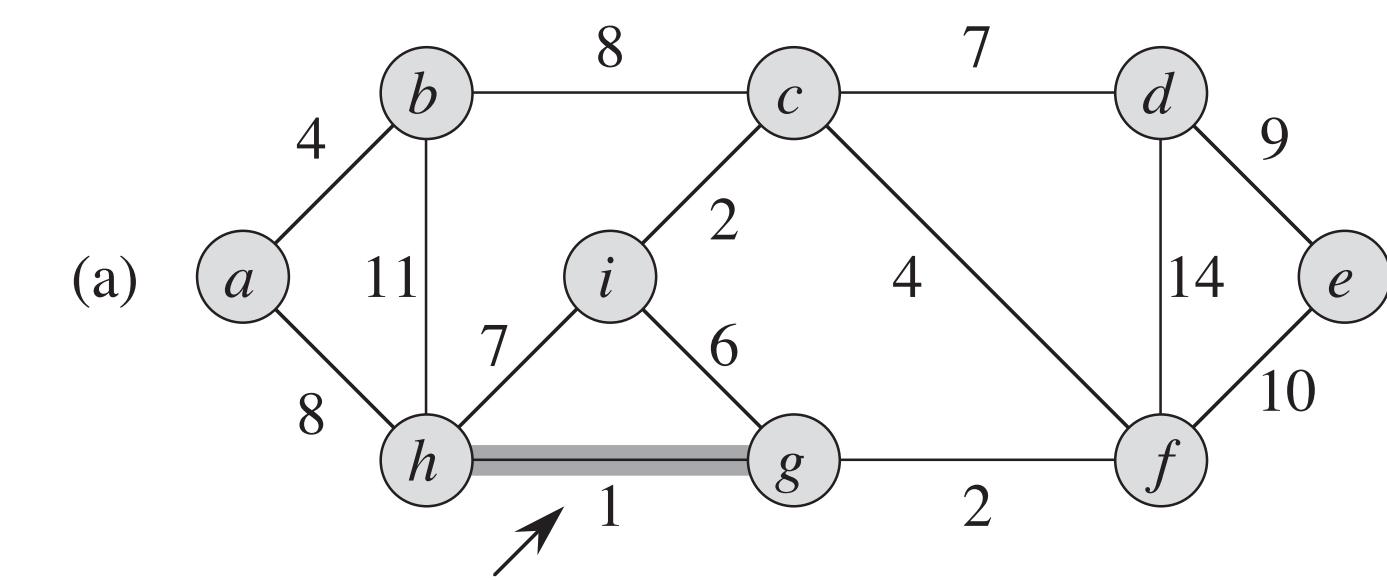
FIND-SET(x) returns the representative of the set containing x

minimum spanning tree - Kruskal's algorithm

MST-KRUSKAL(G, w)

```

1   $A = \emptyset$ 
2  for each vertex  $v \in G.V$ 
3      MAKE-SET( $v$ )
4  sort the edges of  $G.E$  into nondecreasing order by weight  $w$ 
5  for each edge  $(u, v) \in G.E$ , taken in nondecreasing order by weight
6      if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ )
7           $A = A \cup \{(u, v)\}$ 
8          UNION( $u, v$ )
9  return  $A$ 
```

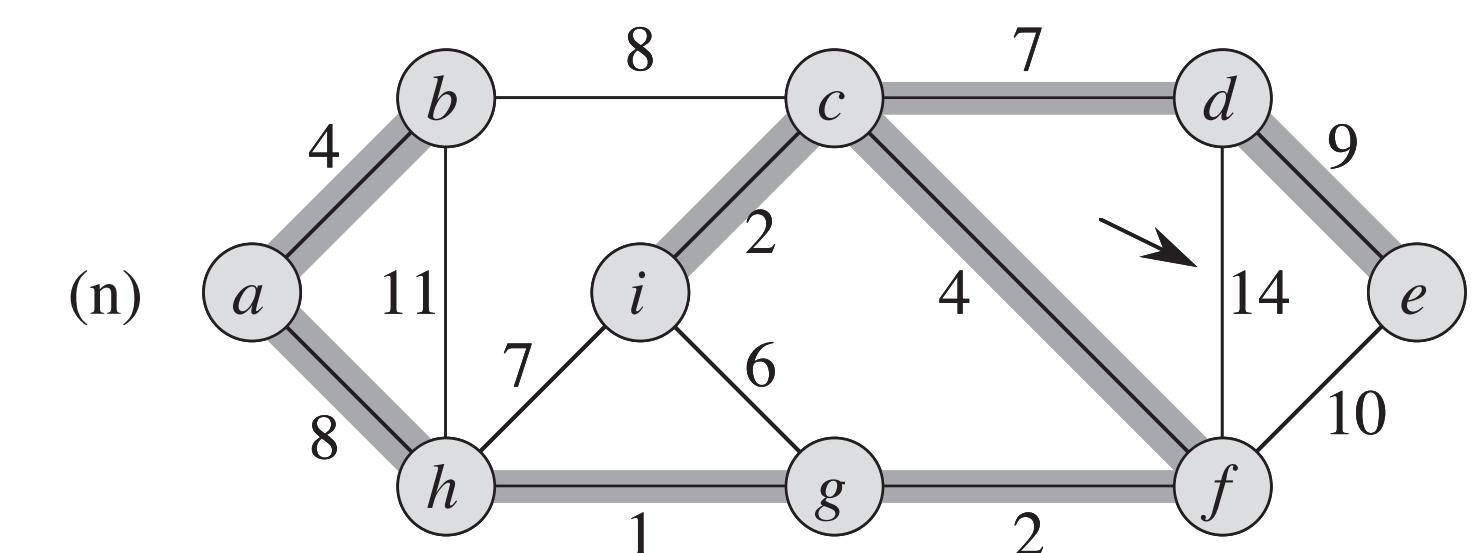
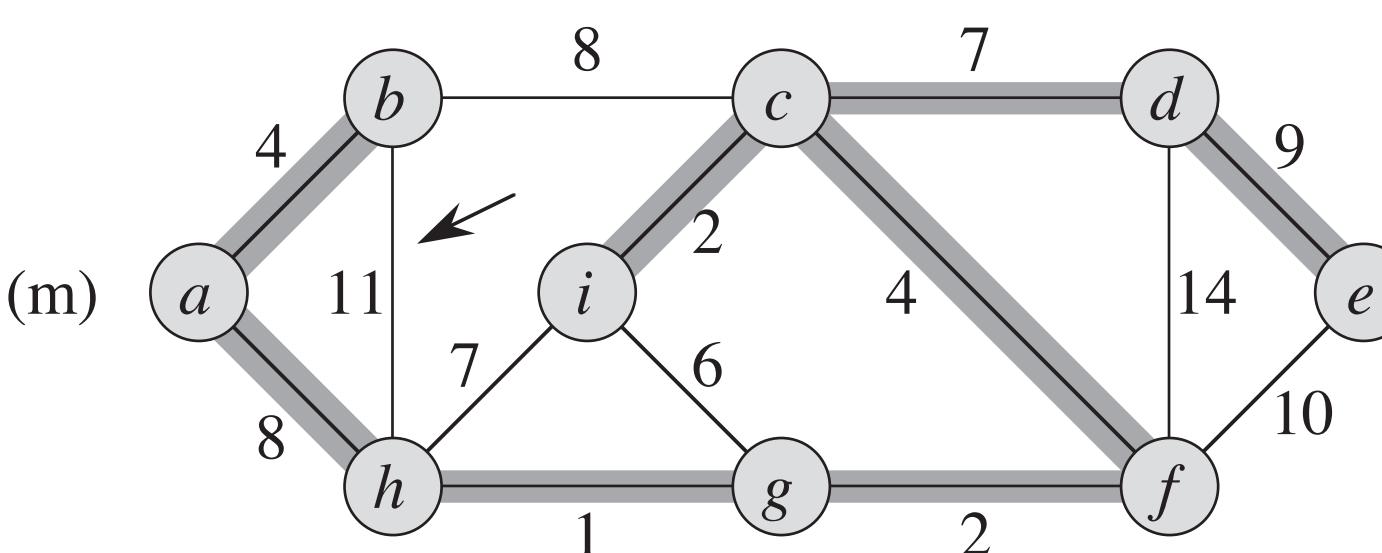
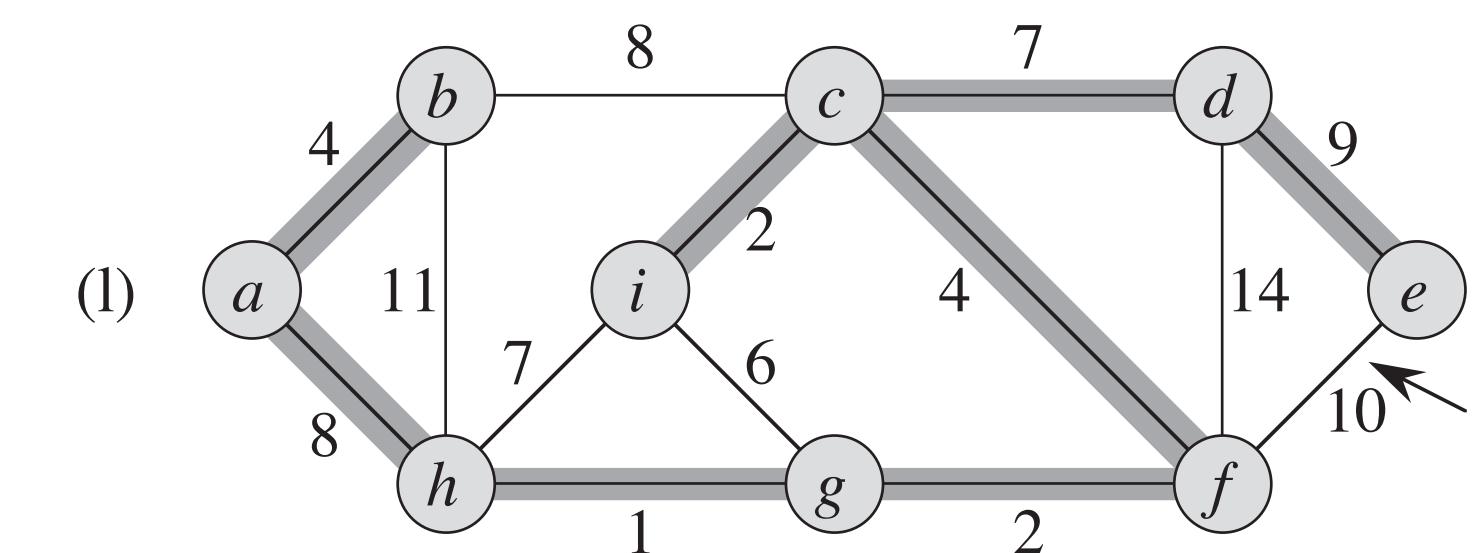
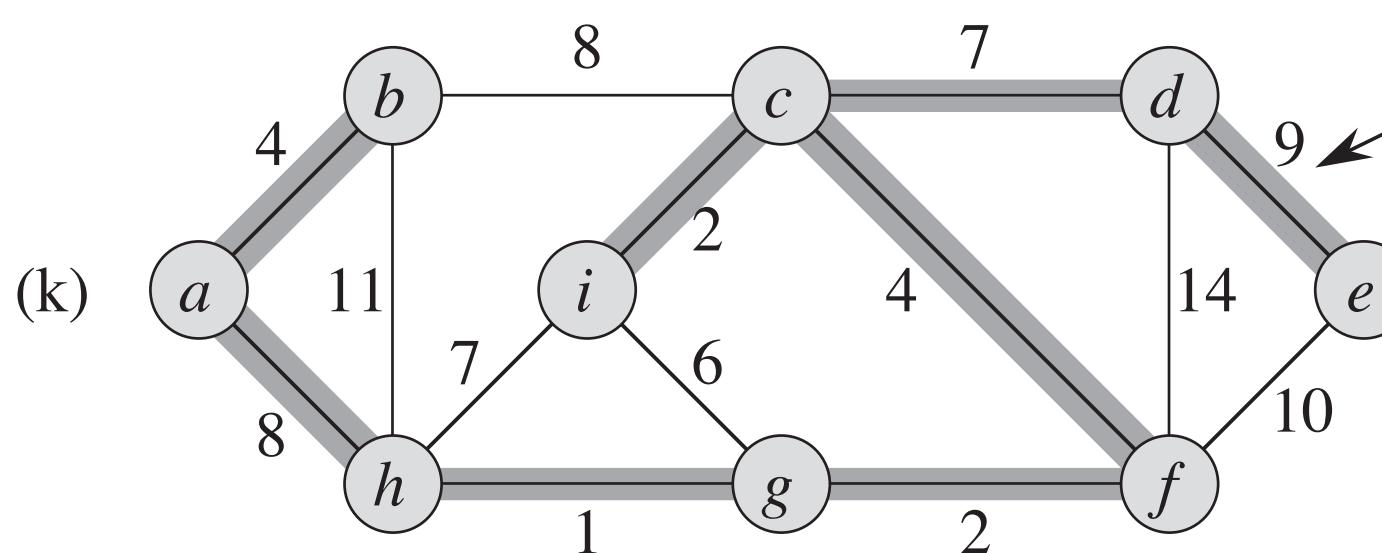
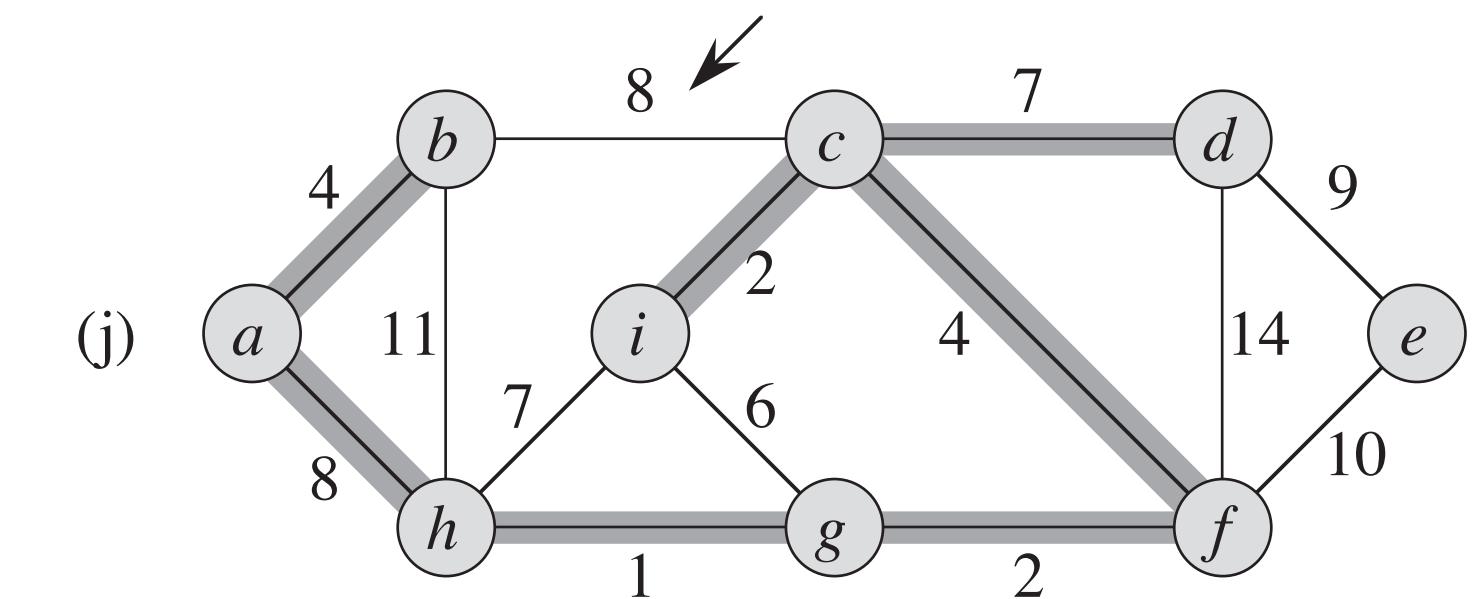
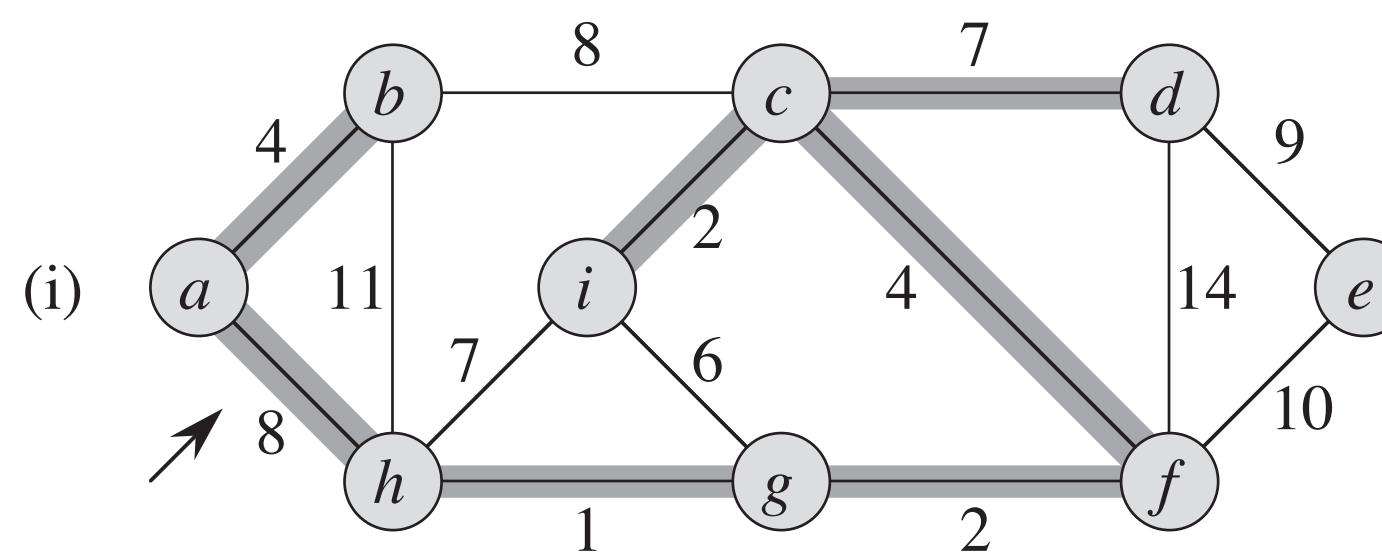


minimum spanning tree - Kruskal's algorithm

MST-KRUSKAL(G, w)

```

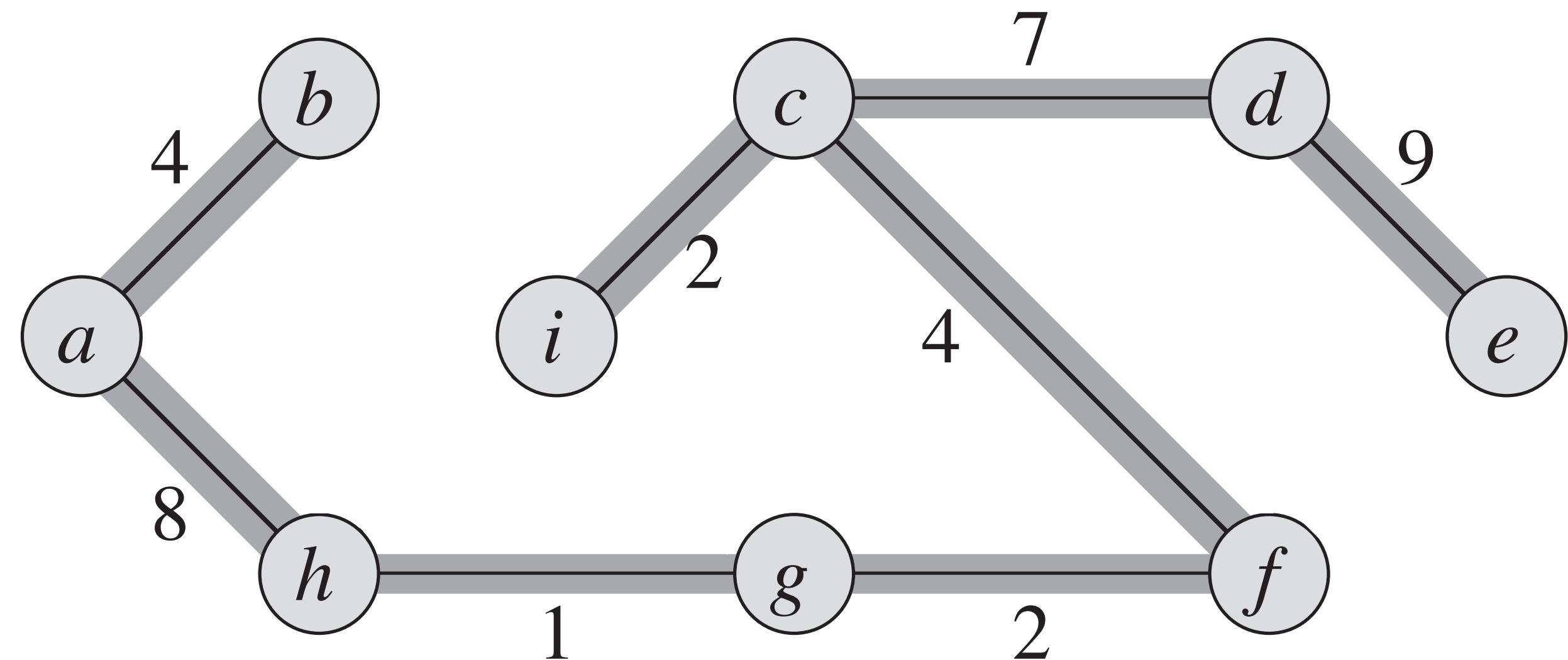
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minimum spanning tree - Kruskal's algorithm

MST-KRUSKAL(G, w)

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7      $A = A \cup \{(u, v)\}$ 
8     UNION( $u, v$ )
9 return  $A$ 
```



the Kruskal's algorithm is **greedy**, i.e., it makes
locally optimal choice at each step