



Graphs (contd.)

References: “Data Structures and Algorithm Analysis”, C. Shaffer, pp. 371–388.

Paths and cycles

Path: A sequence of vertices v_1, v_2, \dots, v_n with an edge from v_i to v_{i+1} for $1 \leq i < n$ is a path of length $n - 1$.

A path is *simple* if all vertices on the path are distinct.

A cycle is a path of length at least 1 that connects v_i to itself.

A cycle is *simple* if the path is simple, except the first and last vertices are the same.

Graph ADT

```
public interface Graph {  
    public class Edge {  
        public int from, to;  
        public Edge(int i, int j) { from = i; to = j; }  
    }  
  
    public void init(int n); // initialize with n vertices  
  
    public int numVertices(); // number of vertices of graph  
    public int numEdges(); // number of edges of graph  
    public void addEdge(int u, int v, int weight); // add edge  
    public Iterator<Edge> getOutgoingEdges(int v); // get out edges from v  
    public void delEdge(int u, int v); // delete edge from u to v  
    public boolean isEdge(int u, int v); // Is there an edge from u to v  
    public int weight(int u, int v); // get weight of edge from u to v  
  
    public void setMark(int u, int val); // mark vertex u with value  
  
    public int getMark(int u); // get value of mark for vertex u  
}
```

Graph traversals (search)

- Many applications require visiting nodes or vertices of the graph in some order exactly once.
- A lot of applications for graph search
 - how connected is a network ?
 - search for an optimal solution in a combinatorial problem
 - AI search (e.g. game playing)

Graph traversals

To insure visiting all vertices:

```
public void graphTraverse(Graph g) {  
    for (int v=0; v < g.numVertices(); v++) {  
        g.setMark(v, UNVISITED);  
    }  
  
    for (int v=0; v < g.numVertices(); v++) {  
        if (g.getMark(v) == UNVISITED) {  
            doTraverse(g, v);  
        }  
    }  
}
```

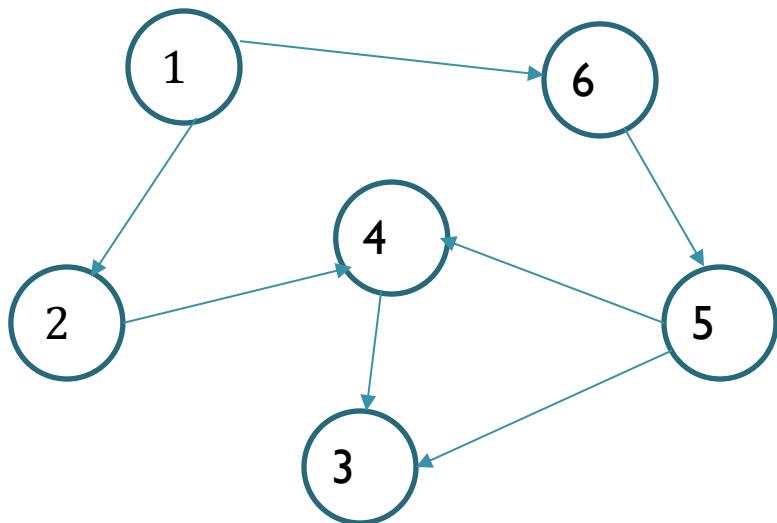
We will next examine a couple of choices for “doTraverse()”.

Depth first search

- similar to backtracking search we did for finding maze path.
Examine a path as far as possible and then back track

```
public void dfs(Graph g, int v) {  
    preVisit(g, v); // add to dfslist()  
    g.setMark(v,VISITED);  
    Iterator<Graph.Edge> outEdgelter = g.getOutgoingEdges(v);  
    while (outEdgelter.hasNext()) {  
        Graph.Edge edge = outEdgelter.next();  
        int w = edge.to;  
        if (g.getMark(w) == UNVISITED) {  
            dfs(g, w);  
        }  
    }  
    postVisit(g, v);  
}
```

Graph search example



Graph search example – DFS tree

- Start vertex : 1; dfslist = {1}

DFS(1) dfslist={1}

DFS(2) dfslist={1, 2}

DFS(6) dfslist={1, 2, 4, 3, 6}

DFS(4) dfslist={1, 2, 4}

DFS(5) dfslist={1, 2, 4, 3, 6, 5}

DFS(3) dfslist={1, 2, 4, 3}

Breadth-first search

Like DFS, but replace stack with a queue.

Visit vertex's neighbors before continuing deeper in the tree.

Each vertex has initial mark 0 (unvisited). As explore graph, set mark of a vertex to one greater than mark of vertex that precedes it in BFS.

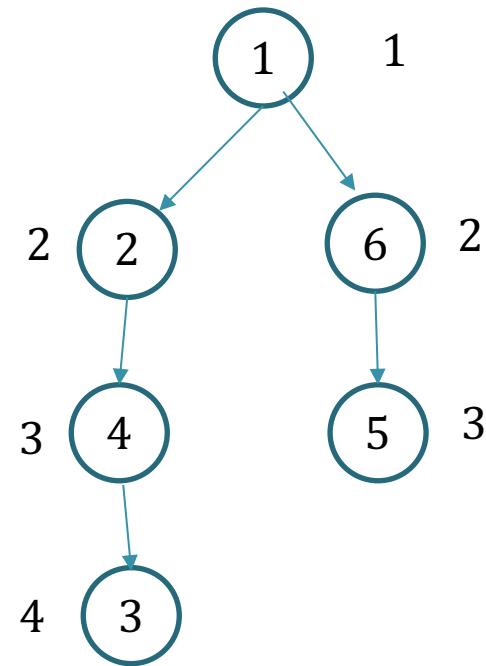
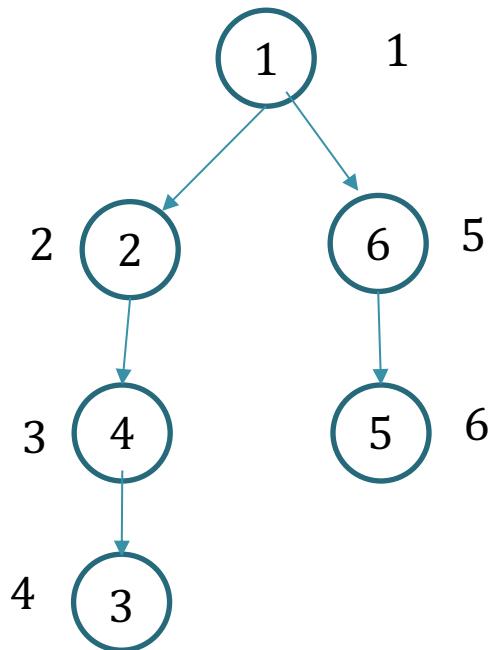
Mark indicates the level of a vertex in the BFS tree

```
public void bfs(Graph g, int start) {
    Queue<Integer> vertexQueue = new LinkedList<Integer>();
    vertexQueue.add(start);
    g.setMark(start, 1);
    while (!vertexQueue.isEmpty()) {
        int v = vertexQueue.remove(); // remove from head of the queue
        preVisit(g, v);
        Iterator<Graph.Edge> outEdgelter = g.getOutgoingEdges(v);
        while (outEdgelter.hasNext()) {
            Graph.Edge edge = outEdgelter.next();
            int w = edge.to;
            if (g.getMark(w) == 0) {
                g.setMark(w, g.getMark(v)+1);
                vertexQueue.add(w);
            }
        }
        postVisit(g, v);
    }
}
```

Graph search example - BFS

- Start vertex : 1; $Q = \{1\}$; $\text{mark}(1) = 1$
- Remove 1 from Q , add 2,6 to Q ; $Q = \{2,6\}$ $\text{mark}(2) = \text{mark}(6) = \text{mark}(1) + 1 = 2$
- Remove 2 , add 4 to Q ; $Q = \{6, 4\}$ $\text{mark}(4) = \text{mark}(2) + 1 = 3$
- Remove 6 , add 5 to Q ; $Q = \{4, 5\}$ $\text{mark}(5) = \text{mark}(6) + 1 = 3$
- Remove 4 , add 3 to Q ; $Q = \{5,3\}$ $\text{mark}(3) = \text{mark}(4) + 1 = 4$
- Remove 5 , no vertex to add; $Q = \{3\}$
- Remove 3, no vertex to add; $Q = \{\}$

DFS and BFS trees



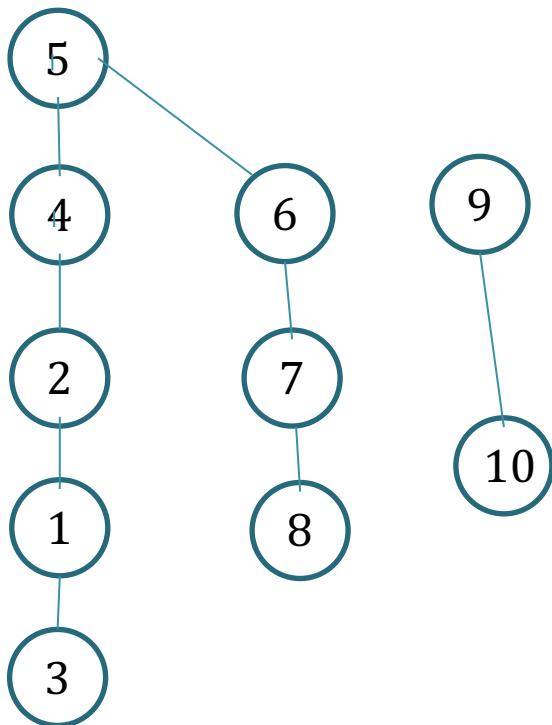
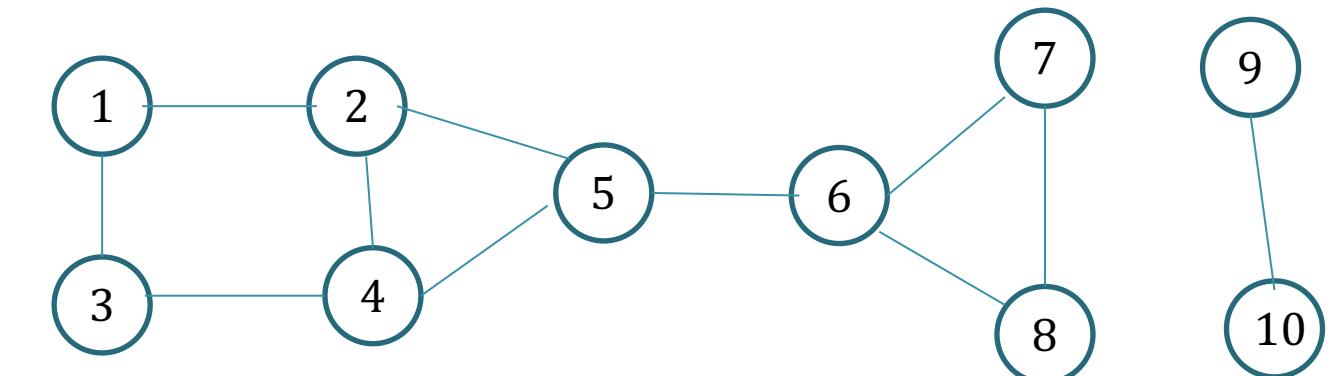
Time complexities

- Both DFS and BFS take $O(n+m)$ time provided the operations `hasNext()` and `next()` of edge iterator and `preVisit()` and `postVisit()` operations take constant time.

Here n is number of nodes and m is number of edges.

Undirected graph connectivity

- A graph is connected if there is a path between every pair of vertices in G .
- A connected component is a maximal subgraph of G which is connected.
- Can use BFS and DFS to find connected components of a G in $O(n+m)$ time.
- A spanning tree of a connected graph G is a subgraph of G which is connected and has minimum number of edges. It has $n - 1$ edges where $n = |V|$
- BFS and DFS provide spanning trees.
- A spanning forest is a collection of spanning trees of a graph that contains all the vertices.



2 connected components –
DFS spanning forest

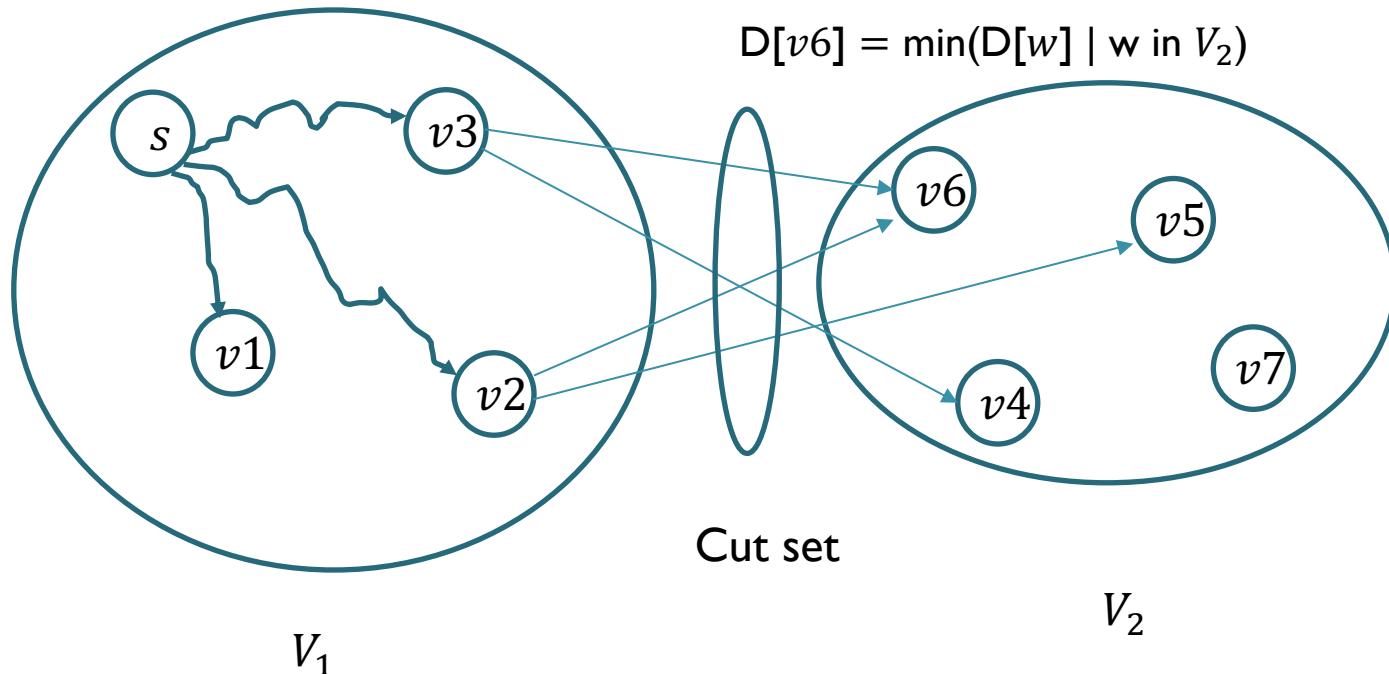
Similar BFS spanning forest exists. Note BFS gives path from u to v in G with smallest number of edges.

Single source shortest path problem

- **Given :**A directed graph $G = (V,E)$ with weights on edges ($w(e)$ for $e \in E$) and a source vertex $s \in V$
- **Required:** Find the shortest path length $D[v]$ from s to v for all $v \in V$
- Example : shortest route between cities
Length of path = sum of weights of edges in the path
- Negative weights with cycles can cause shortest path between vertices to have infinite number of edges (not converge)
- Focus on non-negative weights on edges

Dijkstra's shortest path alg.

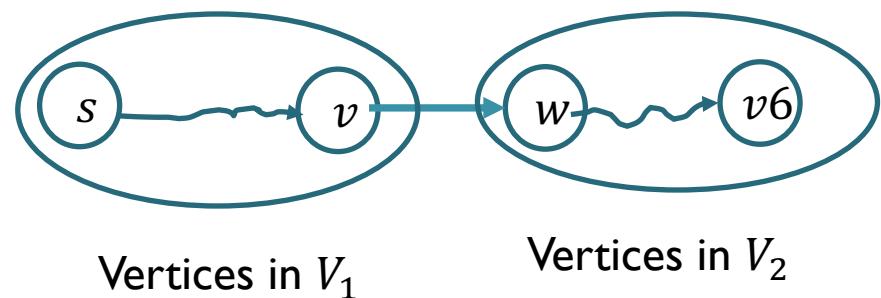
- **Cut** – Partition of V into disjoint subsets V_1 and V_2 ($V_2 = V - V_1$)
Cut-set – $\{(u,v) \in E \mid u \in V_1 \text{ and } v \in V_2\}$
- **Greedy approach idea :**
 - (a) Have a cut where V_1 is set of vertices for which we know shortest path length from s ; initially $V_1 = \{s\}$
 - (b) For all $v \in V_2$, we know shortest path length from s to v with vertices in path (except v) restricted to V_1
(at most one edge in cut-set). Let it be $D[v]$
 - (c) At each step we can move one vertex from V_2 to V_1 if V_2 is not empty. Why ?
Let $D[u] = \min_{v \in V_2} D[v]$.
Cannot have a better path from s to u that includes a vertex from V_2
 $\rightarrow D[u]$ is shortest path from s to u



We know shortest path from s to every vertex in V_1

For every vertex in V_2 , we know shortest of all paths from s with intermediate vertices only in V_1

Can you find a shorter path from s to $v6$ that goes through a vertex say w in V_2 ?



Length of this path $\geq D[w] \geq D[v6]$ hence not TRUE !

So $D[v6]$ is shortest path from s to $v6$. We can add $v6$ to V_1

Updating $D[v]$'s

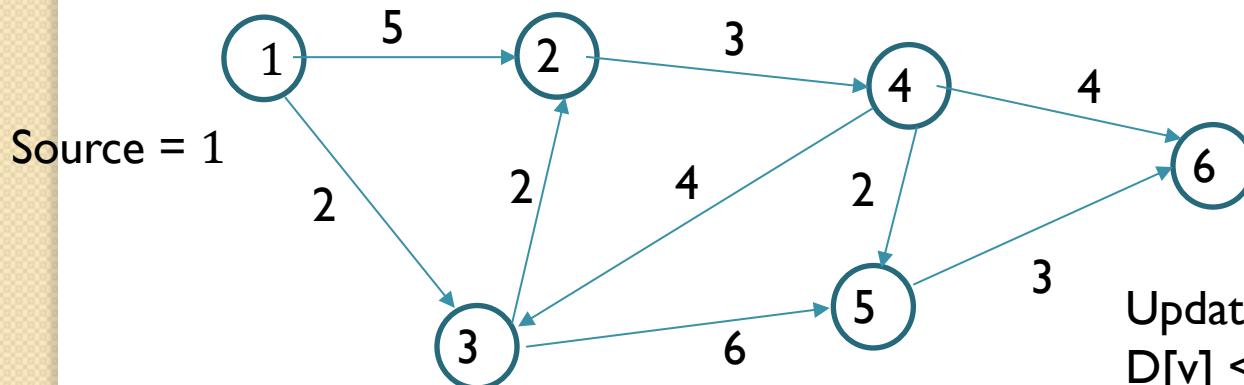
- Once we add a new vertex, set V_1 has changed. This means $D[v]$ must be updated for remaining vertices in V_2

Do we need to consider all vertices in V_2 ?

No. Only vertices adjacent to newly added vertex (e.g. $v6$)

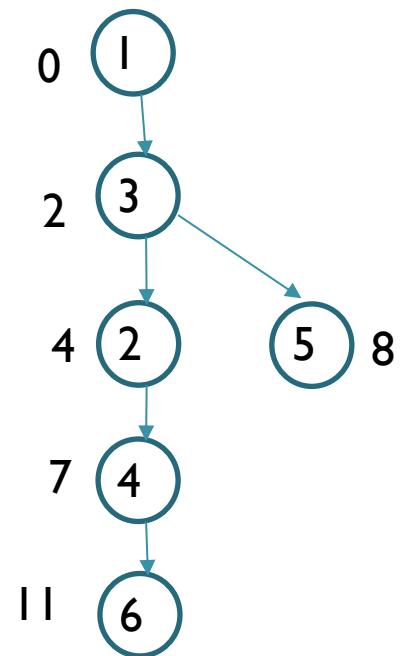
- For each adjacent vertex w of $v6$, update $D[w] = \min(D[w], D[v6] + w((v6, w)))$

Dijkstra's algorithm example



$D[v]$	1	2	3	4	5	6
Add 1; update 2,5,6	0(*)	5	2	∞	∞	∞
Add 3 update 2,5	0(*)	4	2(*)	∞	8	∞
Add 2 update 4	0(*)	4 (*)	2(*)	7	8	∞
Add 4 update 6	0(*)	4(*)	2 (*)	7(*)	8	11
Add 5; no updates	0(*)	4(*)	2 (*)	7 (*)	8(*)	11

Add v6



Dijkstra(G, w, s)

```
1 for each  $v \in V$  do  $d[v] \leftarrow +\infty$ 
2  $d[s] \leftarrow 0$ 
3  $S \leftarrow \emptyset$ 
4  $Q \leftarrow V$ 
5 while  $Q \neq \emptyset$ 
6    $u \leftarrow \text{ExtractMin}(Q)$ 
7    $S \leftarrow S \cup \{u\}$ 
8   for each  $v \in Adj[u]$ 
9     if  $d[v] > d[u] + w(u, v)$  then  $d[v] \leftarrow d[u] + w(u, v)$ 
```

The total time required by Dijkstra's algorithm is

$T = O(|V|) + |V| \cdot T_{\text{ExtractMin}} + |E| \cdot T_{\text{Update}}$. For the unordered array priority queue, this becomes

$O(|V|) + |V| \cdot O(|V|) + |E| \cdot O(1) = O(|V|^2)$. For the binary heap, we get

$T = O(|V|) + |V| \cdot O(\lg |V|) + |E| \cdot O(\lg |V|) = O(|E| \lg(|V|))$