

Inf 2B: Graphs, BFS, DFS

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Directed and Undirected Graphs

- ▶ A *graph* is a mathematical structure consisting of a set of *vertices* and a set of *edges* connecting the vertices.
- ▶ **Formally:** $G = (V, E)$, where V is a set and $E \subseteq V \times V$.
- ▶ For edge $e = (u, v)$ we say that e is *directed from u to v* .
- ▶ $G = (V, E)$ *undirected* if for all $v, w \in V$:

$$(v, w) \in E \iff (w, v) \in E.$$

Otherwise *directed*.

Directed \sim *arrows* (**one-way**)

Undirected \sim *lines* (**two-way**)

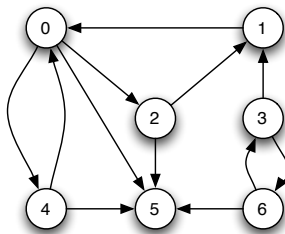
- ▶ We assume V is finite, hence E is also finite.

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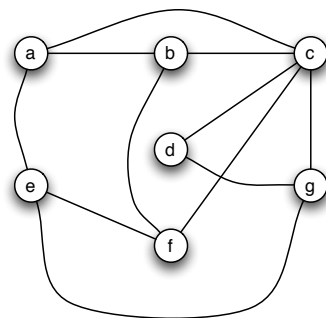
A directed graph

$$\begin{aligned} G &= (V, E), \\ V &= \{0, 1, 2, 3, 4, 5, 6\}, \\ E &= \{(0, 2), (0, 4), (0, 5), (1, 0), (2, 1), (2, 5), \\ &\quad (3, 1), (3, 6), (4, 0), (4, 5), (6, 3), (6, 5)\}. \end{aligned}$$



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An undirected graph



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Examples

- ▶ *Road Maps.*
Edges represent streets and vertices represent crossings (junctions).
- ▶ *Computer Networks.*
Vertices represent computers and edges represent network connections (cables) between them.
- ▶ *The World Wide Web.*
Vertices represent webpages, and edges represent hyperlinks.
- ▶ ...

Adjacency matrices

Let $G = (V, E)$ be a graph with n vertices. Vertices of G are numbered $0, \dots, n - 1$.

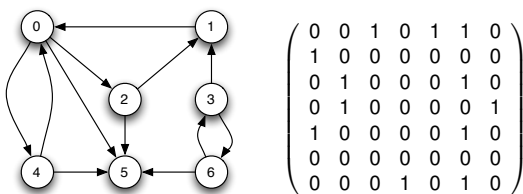
The *adjacency matrix* of G is the $n \times n$ matrix

$$A = (a_{ij})_{0 \leq i, j \leq n-1}$$

with

$$a_{ij} = \begin{cases} 1, & \text{if there is an edge from vertex } i \text{ to vertex } j; \\ 0, & \text{otherwise.} \end{cases}$$

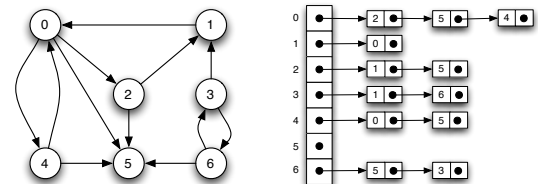
Adjacency matrix (Example)



Adjacency lists

Array with one entry for each vertex v , which is a list of all vertices adjacent to v .

Example



Quick Question

Given: graph $G = (V, E)$, with $n = |V|$, $m = |E|$.
 For $v \in V$, we write $in(v)$ for in-degree, $out(v)$ for out-degree.

Which data structure has faster (asymptotic) worst-case running-time, for *checking if w is adjacent to v* , for a given pair of vertices?

1. Adjacency list is faster.
2. Adjacency matrix is faster.
3. Both have the same asymptotic worst-case running-time.
4. It depends.

Answer: 2. For an Adjacency Matrix we can check in $\Theta(1)$ time. An adjacency list structure takes $\Theta(1 + out(v))$ time.

Quick Question

Given: graph $G = (V, E)$, with $n = |V|$, $m = |E|$.
 For $v \in V$, we write $in(v)$ for in-degree, $out(v)$ for out-degree.

Which data structure has faster (asymptotic) worst-case running-time, for *visiting all vertices w adjacent to v* , for a given vertex v ?

1. Adjacency list is faster.
2. Adjacency matrix is faster.
3. Both have the same asymptotic worst-case running-time.
4. It depends.

Answer: 3. Adjacency matrix requires $\Theta(n)$ time always. Adjacency list requires $\Theta(1 + out(v))$ time. In worst-case $out(v) = \Theta(n)$.

Adjacency Matrices vs Adjacency Lists

	adjacency matrix	adjacency list
Space	$\Theta(n^2)$	$\Theta(n + m)$
Time to check if w adjacent to v	$\Theta(1)$	$\Theta(1 + out(v))$
Time to visit all w adjacent to v .	$\Theta(n)$	$\Theta(1 + out(v))$
Time to visit all edges	$\Theta(n^2)$	$\Theta(n + m)$

Sparse and dense graphs

$G = (V, E)$ graph with n vertices and m edges

Observation: $m \leq n^2$

- ▶ G dense if m close to n^2
- ▶ G sparse if m much smaller than n^2

Graph traversals

A *traversal* is a strategy for visiting all vertices of a graph while respecting edges.

BFS = breadth-first search

DFS = depth-first search

General strategy:

1. Let v be an arbitrary vertex
2. Visit all vertices reachable from v
3. If there are vertices that have not been visited, let v be such a vertex and go back to (2)

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Graph Searching (general Strategy)

Algorithm searchFromVertex(G, v)

1. **mark** v
2. put v onto schedule S
3. **while** schedule S is not empty **do**
4. remove a vertex v from S
5. **for all** w adjacent to v **do**
6. **if** w is not marked **then**
7. **mark** w
8. put w onto schedule S

Algorithm search(G)

1. ensure that each vertex of G is not marked
2. initialise schedule S
3. **for all** $v \in V$ **do**
4. **if** v is not marked **then**
5. searchFromVertex(G, v)

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Three colour view of vertices

- ▶ Previous algorithm has vertices in one of two states: *unmarked* and *marked*. Progression is $unmarked \rightarrow marked$
- ▶ Can also think of them as being in one of three states (represented by colours):
 - ▶ *White*: not yet seen (not yet investigated).
 - ▶ *Grey*: put on schedule (under investigation).
 - ▶ *Black*: taken off schedule (completed).

Progression is

$white \rightarrow grey \rightarrow black$

We will use the three colour scheme when studying an algorithm for topological sorting of graphs.

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BFS

Visit all vertices reachable from v in the following order:

- ▶ v
- ▶ all neighbours of v
- ▶ all neighbours of neighbours of v that have not been visited yet
- ▶ all neighbours of neighbours of neighbours of v that have not been visited yet
- ▶ etc.

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BFS (using a Queue)

Algorithm bfs(G)

1. Initialise Boolean array *visited*, setting all entries to FALSE.
2. Initialise *Queue* *Q*
3. **for all** $v \in V$ **do**
4. **if** *visited*[*v*] = FALSE **then**
5. bfsFromVertex(*G*, *v*)

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BFS (using a Queue)

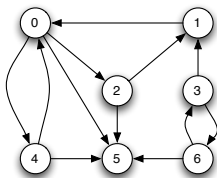
Algorithm bfsFromVertex(*G*, *v*)

1. *visited*[*v*] = TRUE
2. *Q.enqueue*(*v*)
3. **while not** *Q.isEmpty*() **do**
4. $v \leftarrow Q.dequeue()$
5. **for all** *w* adjacent to *v* **do**
6. **if** *visited*[*w*] = FALSE **then**
7. *visited*[*w*] = TRUE
8. *Q.enqueue*(*w*)

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Algorithm bfsFromVertex(*G*, *v*)

1. *visited*[*v*] = TRUE
2. *Q.enqueue*(*v*)
3. **while not** *Q.isEmpty*() **do**
4. $v \leftarrow Q.dequeue()$
5. **for all** *w* adjacent to *v* **do**
6. **if** *visited*[*w*] = FALSE **then**
7. *visited*[*w*] = TRUE
8. *Q.enqueue*(*w*)



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Quick Question

Given a graph $G = (V, E)$ with $n = |V|$, $m = |E|$, what is the worst-case running time of BFS, in terms of m, n ?

1. $\Theta(m + n)$
2. $\Theta(n^2)$
3. $\Theta(mn)$
4. Depends on the number of components.

Answer: 1. To see this need to be careful about bounding running time for the loop at lines 5–8. *Must* use the Adjacency List structure.

Answer: 2. if we use adjacency matrix representation.

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DFS

Visit all vertices reachable from v in the following order:

- ▶ v
- ▶ some neighbour w of v that has not been visited yet
- ▶ some neighbour x of w that has not been visited yet
- ▶ etc., until the current vertex has no neighbour that has not been visited yet
- ▶ Backtrack to the first vertex that has a yet unvisited neighbour v' .
- ▶ Continue with v' , a neighbour, a neighbour of the neighbour, etc., backtrack, etc.

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DFS (using a stack)

Algorithm dfs(G)

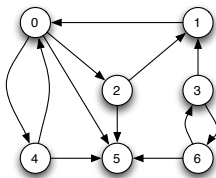
1. Initialise Boolean array *visited*, setting all to FALSE
2. Initialise *Stack S*
3. **for all** $v \in V$ **do**
4. **if** *visited*[v] = FALSE **then**
5. dfsFromVertex(G, v)

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DFS (using a stack)

Algorithm dfsFromVertex(G, v)

1. $S.push(v)$
2. **while not** $S.isEmpty()$ **do**
3. $v \leftarrow S.pop()$
4. **if** *visited*[v] = FALSE **then**
5. *visited*[v] = TRUE
6. **for all** w adjacent to v **do**
7. $S.push(w)$



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Recursive DFS

Algorithm dfs(G)

1. Initialise Boolean array *visited* by setting all entries to FALSE
2. **for all** $v \in V$ **do**
3. **if** *visited*[v] = FALSE **then**
4. dfsFromVertex(G, v)

Algorithm dfsFromVertex(G, v)

1. *visited*[v] \leftarrow TRUE
2. **for all** w adjacent to v **do**
3. **if** *visited*[w] = FALSE **then**
4. dfsFromVertex(G, w)

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Analysis of DFS

$G = (V, E)$ graph with n vertices and m edges

Without recursive calls:

- ▶ $\text{dfs}(G)$: time $\Theta(n)$
- ▶ $\text{dfsFromVertex}(G, v)$: time $\Theta(1 + \text{out-degree}(v))$

Overall time:

$$\begin{aligned} T(n, m) &= \Theta(n) + \sum_{v \in V} \Theta(1 + \text{out-degree}(v)) \\ &= \Theta\left(n + \sum_{v \in V} (1 + \text{out-degree}(v))\right) \\ &= \Theta\left(n + n + \sum_{v \in V} \text{out-degree}(v)\right) \\ &= \Theta\left(n + \sum_{v \in V} \text{out-degree}(v)\right) \\ &= \Theta(n + m) \end{aligned}$$