Universal Sink Finding Algorithm

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Data Structures for Graphs

Convention

- Vertices: Are always numbered 1, 2, ..., n or 0, 1, ..., n-1.
- |E| = m.
- Size of Input: Usually measured in terms of the number of vertices |V| and the number of edges |E| of the graph.
- That is, there are two relevant parameters describing the size of the input, not just one.

Representation

There are two standard ways to represent a graph G = (V, E):

- as a collection of adjacency lists or
- 2 as an adjacency matrix.

Adjacency-list Representation

- Consists of an array Adj of |V| lists, one for each vertex in V.
- For each $u \in V$, the adjacency list

$$Adj[u] = \{v : (u, v) \in E\}.$$

- That is, Adj[u] consists of all the vertices adjacent to u.
- Alternatively, it may contain pointers to these vertices.
- Note: The vertices in each adjacency list are typically stored in an arbitrary order.

Adjacency-list Representation (Cont.)

- **Directed Graphs:** Sum of the lengths of all adjacency lists is |E|, since an edge (u, v) appears only in Adj[u].
- Undirected Graphs: Sum of the lengths of all adjacency lists is 2|E|, since an edge (u, v) appears both in Adj[u] and Adj[v].
- Memory Requirement: For both directed and undirected graphs, it requires $\Theta(|V| + |E|)$ memory.
- Weighted Graphs: The weight w(u, v) of the edge $(u, v) \in E$ is simply stored with vertex v in u's adjacency list.

Adjacency-matrix Representation

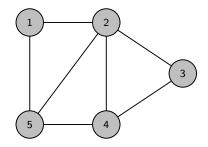
- **Assumption:** Vertices are numbered $1, 2, \dots, |V|$.
- Consists of a $|V| \times |V|$ matrix $A = (a_{ij})$, s.t.,

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

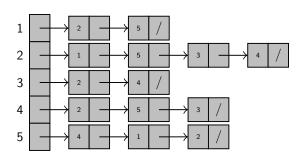
- Memory Requirement: $\Theta(|V|^2)$
- Undirected Graph: Symmetric matrix.
 - Suffices to store only the entries on and above the diagonal.
 - Cutting the memory needed to store the graph almost by half.
- Directed Graph: Not necessarily a symmetric matrix.
- Weighted Graphs: Store

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

Example: Undirected Graph



An undirected graph G



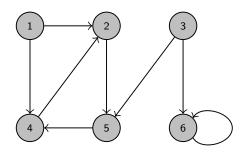
An adjacency-list representation of G

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

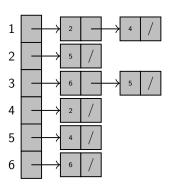
The adjacency-matrix representation of G

Note: The adjacency-matrix is symmetric!

Adjacency-list Representation: Directed Graph Example



An directed graph G



An adjacency-list representation of G

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

The adjacency-matrix representation of G

Note: The adjacency-matrix is **not** necessarily symmetric!

The Adjacency-list Representation: Pros vs. Cons.

Pros:

- Quite Robust: Can be modified to support many other graph variants.
- Space Efficient: Takes $\Theta(|V| + |E|)$ amount of memory.
- Neighbors of a vertex can be computed in optimal time.

Cons:

- No quicker way to determine if a given edge (u, v) is present in the graph than to search for v in the adjacency list Adj[u].
 - Complexity: $\mathcal{O}(|V|)$.
 - Can be remedied by an adjacency-matrix representation of the graph, at the cost of using asymptotically more memory.

The Adjacency-matrix Representation: Pros vs. Cons.

Pros:

• Determining whether there is an edge from u to v takes $\mathcal{O}(1)$.

Cons:

- Computing all neighbors of a given vertex v takes $\mathcal{O}(|V|)$ time.
- Takes $\Theta(|V|^2)$ amount of memory.

The Adjacency-list vs. Adjacency-matrix Representation

- Although the adjacency-list representation is asymptotically at least as efficient as the adjacency-matrix representation, the simplicity of an adjacency matrix may make it preferable when graphs are reasonably small, i.e., |V| is small.
- **Unweighted Graphs:** Rather than using one word of computer memory for each matrix entry, one bit can be used per entry of the matrix, thus making it more space efficient.
- Commonly Used: Adjacency-lists.

Reasons:

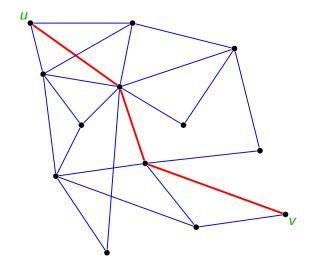
- Graphs in real life are sparse $(|E| \ll |V|^2)$.
- Most algorithms require processing neighbours of each vertex.

But, there are a few exceptions.

Graph Traversal

Graph Traversal

Definition: A vertex v is said to be **reachable** from u if there is a path from u to v.



Graph traversal from vertex u: Visit all vertices which are reachable from u.

Non-triviality of Graph Traversal

• Avoiding Loops: How to avoid visiting a vertex multiple times?

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Ans: Keeping track of vertices already visited.

- Finite number of steps: The traversal must stop in finite number of steps.
- Completeness: We must visit all vertices reachable from the start vertex u.

Some Interesting Graph Algorithmic Problems

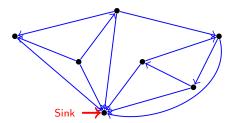
- Are two vertices u and v connected?
- Find all connected components in a graph.
- Is there is a cycle in a graph?
- Compute a path of shortest length between two vertices?
- Is there is a cycle *passing through all vertices*?

Some Interesting Graph Algorithmic Problems

- Are two vertices u and v connected?
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- Is there is a cycle passing through all vertices?
 - Hamiltonian cycle problem.

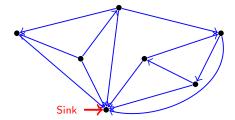
Definition: A vertex s in a given directed graph is said to be a **universal sink** if

- there is no edge emanating from (leaving) s and
- every other vertex has an edge into s.



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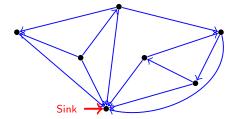
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How many sinks can there be in a graph *G*?

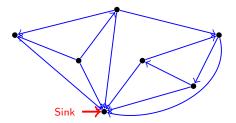
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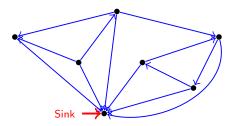
How many sinks can there be in a graph G? At most 1.

Problem: Given a directed graph G = (V, E) in an adjacency matrix representation, design an $\mathcal{O}(|V|)$ time algorithm to determine if there is any sink in G.



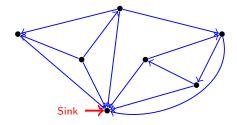
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Hint: We can only look into $\mathcal{O}(|V|)$ entries of the Adjacency matrix M.



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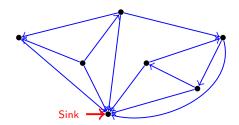
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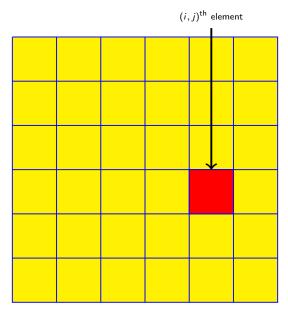
Hint: We can only look into $\mathcal{O}(|V|)$ entries of the Adjacency matrix M.



Question: Can we efficiently verify whether any given vertex *i* is a sink?

Answer: Yes, in $\mathcal{O}(|V|)$ time only. (Look at i^{th} row and j^{th} column of M.)

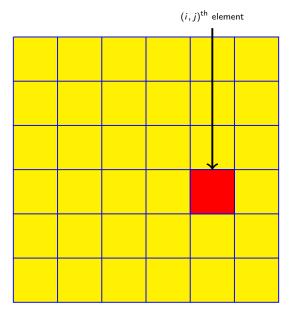
Main Idea



Adjacency Matrix *M*

Question: Can we eliminate |V|-1 non-sink vertices in $\mathcal{O}(|V|)$ look-up into the Adjacency Matrix M?

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Note:

- If M[i,j] = 0, then j cannot be a sink.
- If M[i,j] = 1, then i cannot be a sink.

Algorithm: UniversalSink(M)

```
I/P: A |V| \times |V| adjacency-matrix of a graph G = (V, E).
Begin
  i = 1;
  j = 1;
  while (i \leq |V| \text{ and } j \leq |V|) {
     if (M[i,j] == 1)
       i = i + 1;
     else
       j = j + 1;
  if (i > |V|)
     print "there is no universal sink"
  else
     if (IsSink(M, i) == False)
       print "there is no universal sink"
     else
       print "i is the universal sink"
End
```

UniversalSink(M): Correctness and Complexity

Note:

- Loop terminates when either i > |V| or j > |V|.
- Upon termination, only *i* could possibly be a sink.
 - If i > |V|: There is no sink.
 - If $i \leq |V|$: Then j > |V|. Note that,
 - Vertices k where $1 \le k < i$ cannot be sinks.
 - Vertices k where $i < k \le |V|$ cannot be sinks.
- Check whether i is a sink or not.
 - Out Degree is 0? ith row must be an all zero row.
 - In Degree is |V| 1? i^{th} column must contain all 1's except for the $(i, i)^{th}$ entry, which must be zero by the first condition.

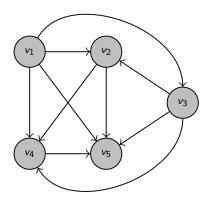
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Worst-case Complexity: $\mathcal{O}(|V|)$

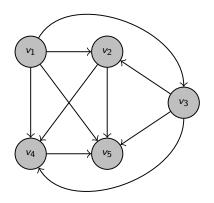
- At most 2|V| for the while loop.
- 2|V| for IsSINK(M, i).

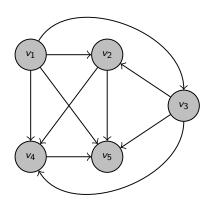


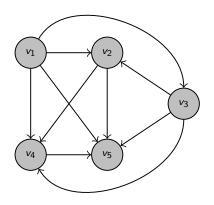
$$\begin{pmatrix}
0 & 1 & 1 & 1 & 1 \\
0 & 0 & 0 & 1 & 1 \\
0 & 1 & 0 & 1 & 1 \\
0 & 0 & 0 & 0 & 1
\end{pmatrix}$$

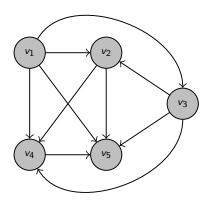
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0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

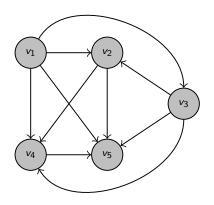
$$\begin{pmatrix}
v_1 \\
v_2 \\
v_3 \\
v_4 \\
v_5 \\
v_6 \\
v_7 \\
v_8 \\
v_8 \\
v_9 \\$$





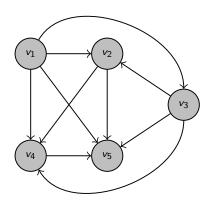


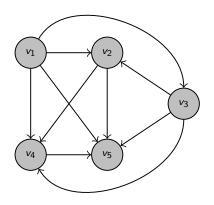




$$\begin{pmatrix}
0 \to 1 & 1 & 1 & 1 \\
0 \to 0 \to 0 \to 1 & 1 \\
0 & 1 & 0 & 1 & 1 \\
0 & 1 & 0 & 1 & 1 \\
0 & 0 & 0 & 0 & 1 \\
0 & 0 & 0 & 0 & 0
\end{pmatrix}$$

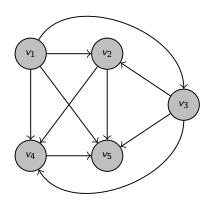
$$\begin{array}{c}
\mathbf{v}_{1} \\
\mathbf{v}_{2} \\
\mathbf{v}_{3} \\
\mathbf{v}_{4} \\
\mathbf{v}_{5} \\
\mathbf{v}_{7} \\
\mathbf{v}_{8} \\
\mathbf{v}_{8} \\
\mathbf{v}_{9} \\
\mathbf{v}_{9}$$





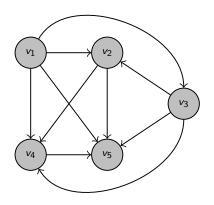
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0 \to 1 & 1 & 1 & 1 \\
0 \to 0 \to 0 \to 1 & 1 \\
0 & 0 \to 0 \to 1 & 1 \\
0 & 1 & 0 & 1 & 1 \\
0 & 0 & 0 & 0 \to 1
\end{pmatrix}$$

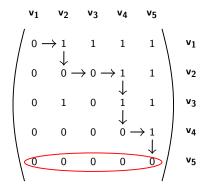
$$\begin{vmatrix}
v_1 \\
v_2 \\
\vdots \\
v_4 \\
\vdots \\
v_5
\end{vmatrix}$$

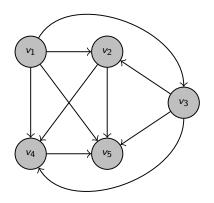


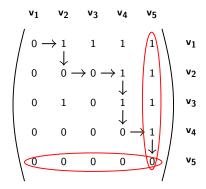
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0 \to 1 & 1 & 1 & 1 \\
0 \to 0 \to 0 \to 1 & 1 \\
0 & 1 & 0 & 1 & 1 \\
0 & 1 & 0 & 1 & 1 \\
0 & 0 & 0 & 0 \to 1 \\
0 & 0 & 0 & 0 \to 1
\end{pmatrix}$$

$$\begin{vmatrix}
v_1 \\
v_2 \\
v_3 \\
v_4 \\
v_5
\end{vmatrix}$$









Thank You for your kind attention!

Books and Other Materials Consulted

- Introduction to Algorithms by Thomas H Cormen, Charles E Leiserson, Ronald L Rivest, Clifford Stein.
- 2 Graph Theory part taken from Discrete Mathematics Lecture Notes (M. Tech (CS), Monsoon Semester, 2007) taught by Prof. Palash Sarkar (ASU, ISI Kolkata).
- Taken from Prof. Surendar Baswana (CSE, IIT Kanpur) lecture slides.

Questions!!