

CSOR4231: Analysis of Algorithms

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Last time

- Greedy algorithms
 - Huffman Coding
 - Fractional Knapsack

Today

- Graphs
 - Defs
 - Basic algorithms: BFS, DFS

Graphs

- Graph $G = (N, E)$
- set N of **vertices** (**nodes**) – represents set of objects
- set E of **edges** – represents relation between objects
- **undirected graph**: edge (u, v) same as (v, u)
- **directed graph**: edge (u, v) not same as (v, u)
- Fundamental model!

Modeling by Graphs

- Physical connections

- communication networks: nodes=switches, computers
- electric circuits: nodes=gates, edges=wires
- Chemistry: nodes=molecules, edges=bonds
- Neural networks: nodes=neurons, edges=synapses
- geographical maps: nodes=cities, edges= roads, flights, railroads
- City maps: nodes=intersections, edges=streets
- ...

Modeling by Graphs

- Logical connections & relations
 - social networks: nodes=people, edges=connections
 - data structures: nodes and pointers
 - entity relationship diagrams in data modeling
 - dependency relation
 - control/data flow graphs
 - message flow graphs
 - AI: puzzles, mazes
 - Games: nodes=positions, edges=moves
- Structure of other models in mathematics, CS
 - Graphical models,
 - finite automata, state machines
 - Markov chains,
 - partial orders, lattices, ...

Graphs : Terminology

- $G(N, E)$, Undirected or Directed
- N : nodes (or vertices)
- E : edges (or arcs for directed)
- Main size parameters: $n = |N|, e = |E|$
for simple graphs e is between 0 and n^2
- Degree of nodes
- Paths, simple paths, cycles
- Forest, Tree : undirected acyclic graphs
- DAG : directed acyclic graph
- Sometimes, weighted graph: $w: E \rightarrow R$
- Review Appendix B

Some Basic Problems

- **Reachability:** Which nodes can reach which other nodes
 - given s, t , is there path from s to t ; find such path
 - given s , which nodes can it reach?
- **Graph Structure (Properties):**
 - Given G , is it connected? Biconnected?
 - Connected components, Cycles?
 - Planar? Bipartite? Eulerian? Hamiltonian?
 - Graph isomorphism...
- **Optimization problems:**
 - Shortest Paths between nodes, Longest Path
 - Minimum Spanning Tree
 - Maximum Flow, Minimum Cut, ...

Graph Representation : Nodes

- **Nodes:** Integers $1, \dots, n$, so can index them (build arrays)
- Generally, nodes belong to some domain D (eg. strings).
- Use a dictionary (symbol table) eg. hash map, trie, BST, etc, to map between D and $[n]$, both ways.

- **Example:**

EWR 1

JFK 2

LGA 3

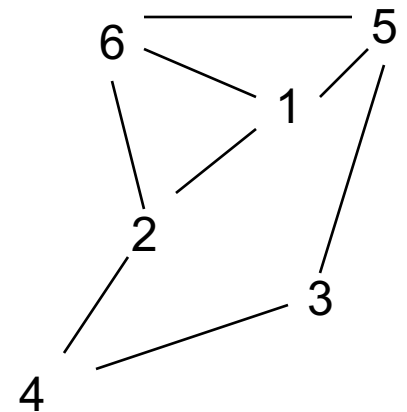
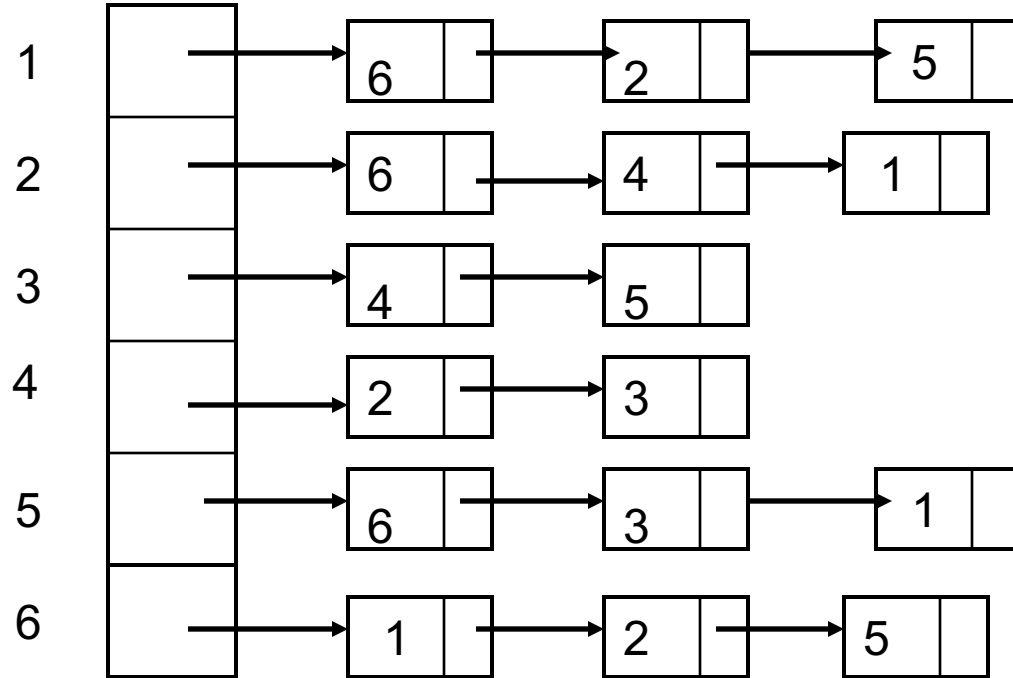
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Graph Representation : Edges

- #1: List of edges, i.e. pairs (i, j)
- Space = $O(e)$
- #2: Adjacency matrix indexed by nodes: 2D boolean array $A[i, j]$ is 1 if edge (i, j) is in E , 0 otherwise
- Undirected graph: matrix is symmetric $A[i, j] = A[j, i]$
- Directed graph: A maybe asymmetric
- Space = $O(n^2)$
- Good for dense graphs: #edges close to $(\# \text{ nodes})^2$

#3: Adjacency list representation

- Array of lists = adjacency lists of nodes



Space proportional to $n + e$

Adjacency list representation (2)

- Undirected Graphs:
- Every edge (i, j) leads to two entries:
 j in $\text{adj}[i]$ and i in $\text{adj}[j]$
- Sometimes it is helpful to connect the two entries so that we can access easily one from the other. That is, node in adj list contains in addition a link to the other node
- Also, sometimes may use doubly linked lists
- Directed Graphs: every edge appears once

Nonuniqueness of representation

- A graph on node set $N=\{1,\dots,n\}$ has one adjacency matrix representation but many different adjacency list representations (lists with different orderings)
 - Can determine if two adj list representations represent same graph in $O(n+e)$ time
 - just sort each list
- A graph with vertex names from a general domain has many representations depending how vertex names are mapped to $[n]=\{1,\dots,n\}$
- **Graph isomorphism problem:** Determine whether two representations are isomorphic, i.e. same graph except for the vertex mapping to $[n]$
 - Is there vertex permutation that preserves the edges?
 - Much harder problem. Complexity is open!

Comparison between representations

- | Rep.\Task | Space | Edge (i,j) ? | List edges(i) |
|---------------|-------|--------------|---------------|
| List of edges | e | e | e |
| Adj matrix | n^2 | 1 | n |
| Adj lists | $n+e$ | $\deg(i)$ | $\deg(i)$ |
- Often in practice, sparse graphs \rightarrow use Adj lists
 - Can augment Adj lists with a **dictionary of edges** to answer quickly edge queries with space $O(e)$
 - eg. hash table: $(i,j) \rightarrow$ hash value

Simple Problems

- Assume **Adj lists** representation
- Print all edges

For **Directed** Graphs:

for $i = 1$ **to** n { **foreach** j in $Adj[i]$ print (i, j) }

For **Undirected** Graphs:

for $i = 1$ **to** n { **foreach** j in $Adj[i]$: **if** $(i < j)$ print (i, j) }

// so that every edge is printed only once

- Construct Adjacency matrix
 - Initialize matrix to 0
 - Traverse every Adj list and change entry to 1
- Construct reverse of a directed graph
- Compute degree of nodes
-

Graph Searching: Single Source Reachability

Input: directed (undirected) graph $G = (N, E)$, “source” node s
Which nodes are reachable from node s ?

SEARCH(G, s)

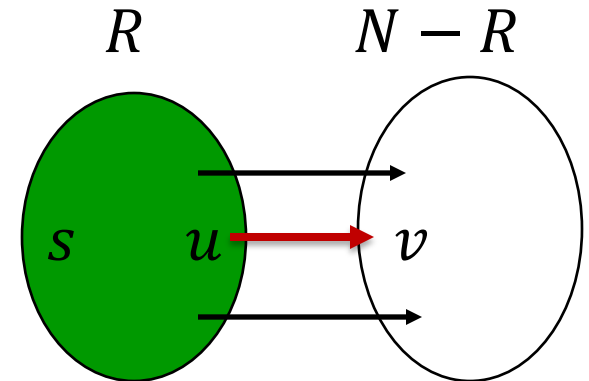
$R = \{s\}$

while \exists edge from R to $N - R$

$\{ (u, v) = \text{such an edge}$

$R = R \cup \{v\}$

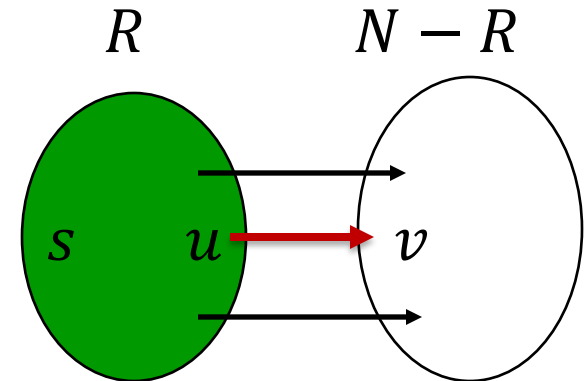
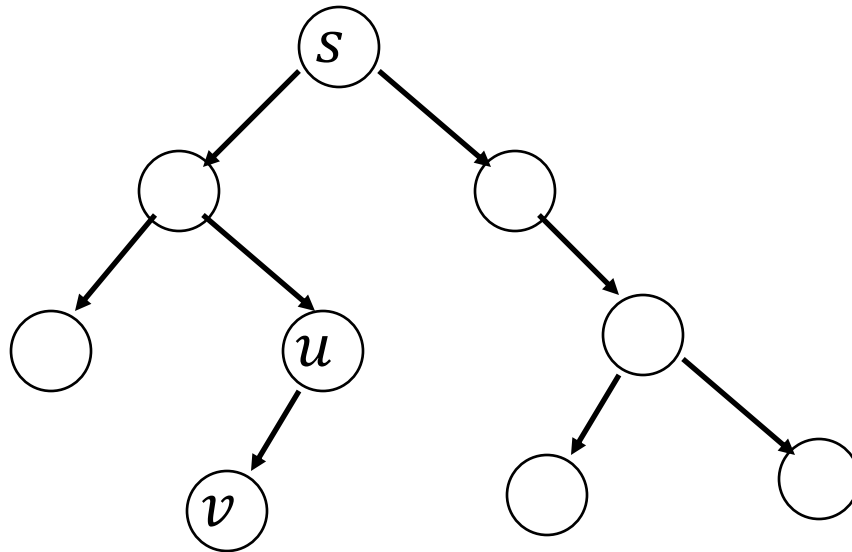
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At the end, R = set of reachable nodes

SEARCH => Reachability/Search Tree

- Rooted tree with root s
- Includes all nodes of R
- parent $p[v] = \text{node } u \text{ that added } v \text{ to } R$



Correctness proof

Theorem: At the end, R = set of nodes that are reachable from s

Proof:

- $v \in R \Rightarrow v$ reachable

in particular, exists $s \rightarrow v$ path in Search tree

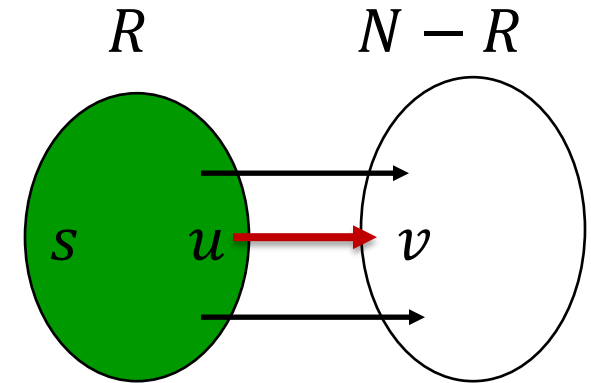
- v reachable $\Rightarrow v \in R$

by induction on length of $s \rightarrow v$ path

IH(i): all nodes at distance $\leq i$ are in R

Search Strategies

- In general, many (u, v) edges from R to $N - R$
- Which to choose?
- Different policies...
- Different algorithms useful in different contexts:
 - Breadth-First Search (BFS) → BFS tree
 - Depth-First Search (DFS) → DFS tree
 - Dijkstra's algorithm → shortest path tree
 - Prim's algorithm → minimum spanning tree
 - ...



Breadth First Search

Policy: (u, v) where u is **earliest** node added to R (F-I-F-O)

Queue Q keeps reached, unprocessed nodes

white: $\in N - R$: not reached

gray: $R \cap Q$: reached, active (did not explore its neighbors)

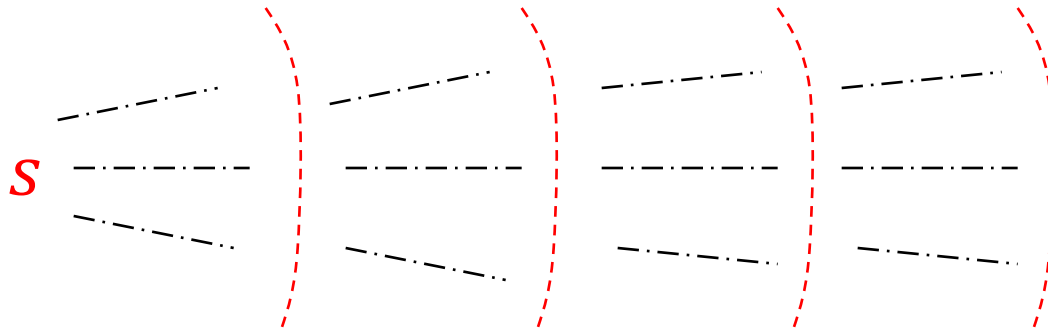
black: $R \setminus Q$: done

Search = Wave, out of s traveling along the graph's edges

$d[v]$: time the wave reaches node v

= distance from the source s

= length of shortest path from s to v



Breadth First Search: Algorithm

BFS(G, s)

for each $v \in N \setminus \{s\}$ **do** $\{d[v] = \infty; p[v] = \perp\}$

$d[s] = 0$

$p[s] = \perp$

$Q = \{s\}$

while $Q \neq \emptyset$ **do**

$\{ u = \text{Dequeue}(Q)$

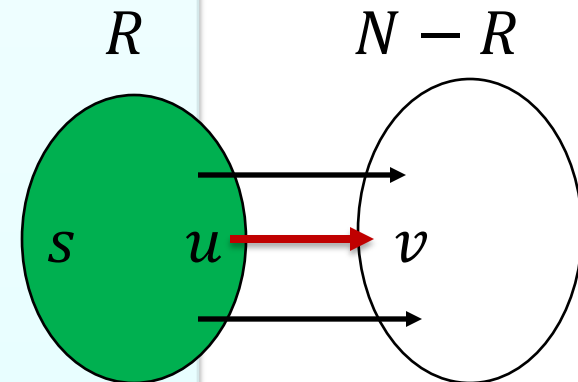
for each $v \in \text{Adj}[u]$ **do**

if $d[v] = \infty$ **then**

$\{ d[v] = d[u] + 1; p[v] = u;$

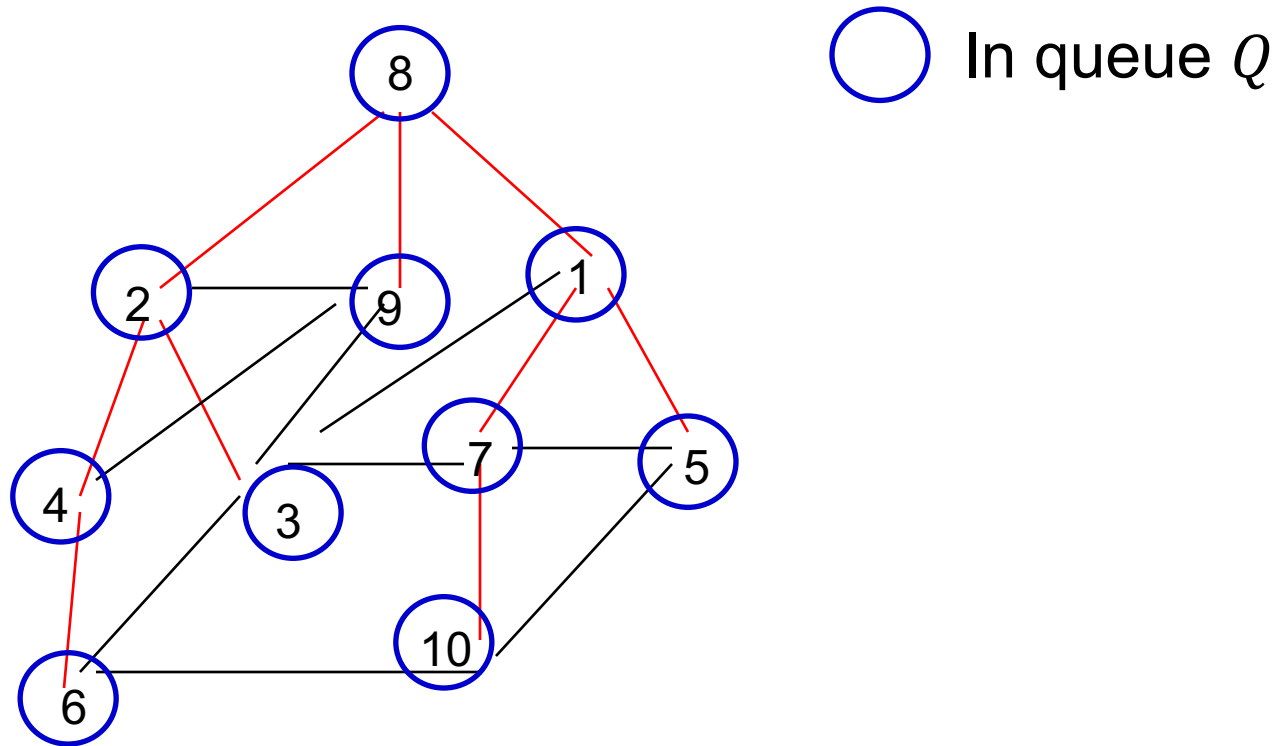
$\text{Enqueue}(Q, v) \}$

$\}$



Time Complexity: $O(n + e)$

Example

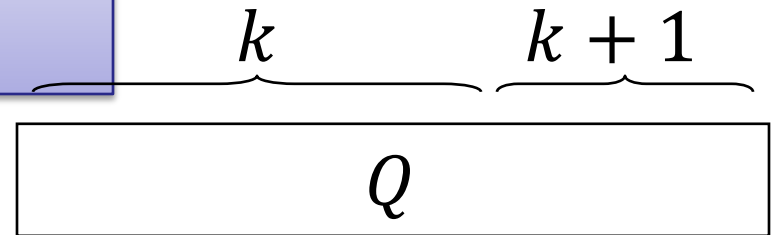


BFS Invariants

Reached nodes $R = \{v \mid d[v] < \infty\}$

$d[u]: \leq k$ if $u \in \text{Done}$

$= k$ or $k + 1$ if $u \in Q$:



Theorem: $\forall v: d[v] = \text{length of shortest } s \rightsquigarrow v \text{ path}$

Proof:

- \geq : length of $s \rightsquigarrow v$ path in BFS tree = $d[v]$

By induction on the time that v was reached (on $d[v]$)

- \leq : By induction on length of shortest $s \rightsquigarrow v$ path

Consider shortest path $s \rightsquigarrow u - v$

by IH, $d[u] \leq \text{length of } s \rightsquigarrow u \text{ path}$

after u processed, $d[v] \leq d[u] + 1 \leq \text{length of } s \rightsquigarrow v \text{ path}$

BFS Tree => Partition into Layers

- Layer 0: $L_0 = \{s\}$
- Layer i : $L_i = \{v \mid d[v] = i\}$
- Undirected graphs: edges connect nodes in same layer or adjacent layers
- Directed graphs: edges can go only to next layer, to same layer or to previous layers (any number back)

