

MCS 253p

Graphs

Graphs

- $G = (V, E)$
- V : vertices (or nodes). Notation u, v in V
- E : edges. Each connects a pair of vertices
 $E : V \times V; (u, v) \text{ in } E$
- [Link](#)

Example Applications of Graphs

- Airport system
 - Airports are vertices, flights are edges
 - Best path may mean shortest fly time or fewest edges
- Traffic flow
 - Intersection is vertex, roads are edges
 - Best path could be shortest distance, or fastest time
- Computer Network

Variations

- Undirected
 - e.g., electrical wire in a building
- Directed
 - streets (one way or two way), flights
- Weighted edges (notation $w(u, v)$)
 - e.g., time/cost to travel a given flight leg
- Can also have weighted vertices
 - e.g., the time spent waiting at an airport
 - or the cost to stay in a hotel

Undirected Graphs

- degree of a vertex: # of edges connected to the vertex
- complete graph:
 - graph where all pairs of vertices are connected
- bipartite graph:
 - undirected graph whose $V = V_1 \times V_2$ and $E = V_1 \times V_2$
- multigraph: can have multiple edges between the same pair of vertices (including self edges)
- hypergraph: can have edges connect more than two vertices

Directed Graphs

- in-degree: # incoming edges
- out-degree: # outgoing edges
- path: $\langle v_0, v_1, \dots, v_k \rangle$ where $\langle v_i, v_{i+1} \rangle$ in E
- simple path: path where all v_i are distinct
- cycle: a non-trivial simple path plus $\langle v_k, v_0 \rangle$ to close the path
- DAG: directed acyclic graph (contains no cycles)
- strongly connected: digraph whose vertices are all reachable from each other

Representations

- Adjacency list:
 - For each vertex, list its neighbors on outgoing edges
 - Good for sparse graphs
 - Can store weight in linked list node
- Adjacency matrix:
 - Bit matrix to represent presence of edge
 - Good for dense graphs
 - Can store weight in matrix
- Trade-offs
 - Adjacency lists - space efficient, easy to grow
 - Adjacency matrix - fast access, but $O(V^2)$ space

Breadth First Search

- “Distance” refers to number of vertices in path
- Visit vertices in increasing order of distance from starting point
- Use a queue to store vertices “to visit”
- Not necessarily unique
- Explore breadth and then depth
- May find the shortest distance to some node
- Breadth First Search

BFS(G, s)

for each vertex u in $V[G] - [s]$ do

$\text{color}[u] = \text{WHITE}$ (White means undiscovered)

$\text{distance}[u] = \text{Infinity}$ (distance from s)

$\text{previous}[u] = \text{NIL}$ (previous vertex)

$\text{color}[s] = \text{GRAY}$, $\text{distance}[s] = 0$, $\text{previous}[s] = \text{NIL}$, $Q = \{\}$

$Q.\text{ENQUEUE}(s)$

while ! $Q.\text{IS_EMPTY}()$ do

$u = Q.\text{DEQUEUE}()$

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

 if $\text{color}[v] == \text{WHITE}$ then

$\text{color}[v] = \text{GRAY}$ (Gray means discovered, but not expanded)

$\text{distance}[v] = \text{distance}[u] + 1$

$\text{previous}[v] = u$

$Q.\text{ENQUEUE}(v)$

$\text{color}[u] = \text{BLACK}$ (Black means expanded)

Depth first search

- Explores depth then neighbors
- Depth isn't necessarily the same as distance in BFS
- discover vertices before we visit
- Use a stack to store nodes "to visit"
- DFS order may not be unique!
- Variant: Iterative Deepening DFS
- Depth First Search

DFS(G)

for each u in $V[G]$ do

$\text{color}[u] = \text{WHITE}$

$\text{previous}[u] = \text{NIL}$

time = 0

for each u in $V[G]$ do

 if $\text{color}[u] == \text{WHITE}$ then

 DFS_VISIT(u)

Useful Graph Operations

- We will use these in some graph algorithms
- Time stamp each vertex with two time stamps
 - discover time `discovered[u]`
 - finish time `finished[u]`
- Can be used to detect cycles
- Assigned with DFS_VISIT (next slide)

DFS_VISIT(u)

color[u] = GRAY (discovered u)

discovered[u] = ++time

for each v in Adjacent[u] do

 if color[v] == WHITE then

 previous[v] = u

 DFS_VISIT(v)

color[u] = BLACK (done exploring from u)

finished[u] = ++time

Topological Sorting

- Directed Acyclic Graph
- Sorted order in which all of a node's predecessors appear before the node
- Useful if computing some property on graph that requires computations of predecessors
- E.g., useful for scheduling tasks that depend on other tasks

eat food, cook food, mix food, wash food, buy food

Topological Sort Algorithm

- 1 call DFS to compute finishing times for each vertex v
- 2 as each vertex is finished, insert it onto the front of a list
- 3 return the list of vertices

Minimum Spanning Trees (MST)

- Spanning tree
 - Spanning - connects each vertex
 - Tree - acyclic
 - Start with connected, undirected, edge-weighted graph G
 - Sub-graph with a subset of edges that connect all vertices
 - $|E| = |V| - 1$
- Minimum spanning tree
 - Sum total weight of edges is minimized
 - edge weights need not be distinct, MST need not be unique
- Example: wiring a house with minimal cable

MST Algorithms

- Two common algorithms:
 - Prim's
 - Kruskal's
- Both are “Greedy algorithms”
 - take the best choice at each opportunity
- Difference is in how next edge is selected

Prim's Algorithm

- Start with single vertex, called the root
- Select the minimum edge connecting u , v such that u is in the MST and v is not
- Grow MST by adding one vertex and one edge at a time
- $O(|V|^2)$ without heaps for priority queue
- $O(|E| \log |V|)$ with heaps
- [Link](#)

Prim's Algorithm

```
def PRIM(G):
```

```
    for each u in V[G]:
```

```
        key[u] = INFINITY;  prev[u] = nil
```

```
    Q.ENQUEUE(G.V);  F = {G.V}
```

```
    while ! Q.ISEMPTY():
```

```
        u = Q.EXTRACT-MIN()
```

```
        foreach v in Adjacent[u]:
```

```
            if v in F  &&  G.w(u, v) < key[v]:
```

```
                F = F - u
```

```
                pr[v] = u
```

```
                key[v] = G.w(u,v)
```

Kruskal's Algorithm

- Start with single-node trees
- Identify cheapest edge
- If edge links different trees && doesn't cause a cycle
 - add the edge
- Adding an edge merges two trees into one
- Minimum Spanning Tree Animation
- $O(|V| \log |E|)$ - much better in practice
- [Link](#)

Kruskal's Algorithm in pseudo code

```
def KRUSKAL(G):
```

```
     $A = \emptyset$ 
```

```
    foreach  $v \in G.V$ :
```

```
        MAKE-SET( $v$ )
```

```
    foreach  $(u, v)$  in  $G.E$  ordered by  $\text{weight}(u, v)$ , increasing:
```

```
        if FIND-SET( $u$ )  $\neq$  FIND-SET( $v$ ):
```

```
             $A = A \cup \{(u, v)\}$ 
```

```
            UNION( $u, v$ )
```

```
    return  $A$ 
```

Disjoint Set

- Sets are required for Kruskal's MST algorithms
- [Link](#)

Kruskal's Algorithm in pseudo C++

// complexity is $O(|E| \lg |E|)$ or $O(|E| \lg |V|)$ if using binary heap for PriorityQueue

```
Vector<Edges> Kruskals(const Graph & g)
{
    for ( auto v : g.vertices() ) {
        vertex.setID = Set::makeSet();
    }
    int N = g.vertices().size();
    PriorityQueue Q;
    for ( auto e : g.edges() )
        Q.insert(e); // Sort Key is edge weight
    Vector<Edges> T;
    while ( T.size() < N-1 ) {
        Edge e = Q.extractMin();
        if ( e.u.setID != e.v.setID ) { // i.e., not in the same set
            T.push_back(e); // add edge e to solution
            Set::union( e.u, e.v ); // join two sub trees into one tree
        }
    }
    return T;
}
```

Single Source Shortest Path

- Given
 - A graph, $G = (V, E)$
 - A single starting vertex, s
- Find lowest cost path to all other vertices
- For un-weighted graph, use BFS
- For weighted graph, we have choices
 - Dijkstra's Algorithm - positive weights
 - Bellman-Ford Algorithm - negative too
 - A* is heuristic, but must know max
 - http://en.wikipedia.org/wiki/A*#Pseudocode

Applications

- Maze games
- Street traffic routing
- Plane trip planning
- Network packet routing
- Robot walking through obstacles
- Dijkstra Animation

Naïve Solution

- "Brute Force" or "Exhaustive Search"
- Enumerate all routes from A to B
- Add up the distances of each route
- Select the shortest
- There could be millions of possibilities
 - cycles make it infinite

Dijkstra's Algorithm

- A greedy algorithm published in 1959
- Keep distance estimates to neighbors of S
- Add neighbor with shortest distance to S and update other neighbors
- Doesn't work with negative edges
- $O((|E|+|V|) \log |V|)$
 - with heap for PriorityQ
- [Dijkstra Algorithm Presentation](#)
- Wikipedia entry
 - http://en.wikipedia.org/wiki/Dijkstra's_algorithm
- [Link STL Link](#)

Dijkstras Algorithm in pseudo C++

// complexity is $O((|E|+|V|) \lg |V|)$ with binary heap for PriorityQueue

```
typedef int Vertex;
```

```
struct Edge {Vertex dest; int weight;};
```

```
void dijkstras(Graph g, Vertex s, int dist[], int prev[]) {
```

```
    PriorityQueue<Vertex> Q;
```

```
    for ( Vertex v : g.getVertices() ) { // initialize
```

```
        dist[v] = INFINITY;           // INT_MAX
```

```
        prev[v] = UNDEFINED;         // -1
```

```
        Q.insert(v);
```

```
    }
```

```
    dist[s] = 0; // distance from start to start is zero, first one extracted below
```

```
    while ( ! Q.isEmpty() ) {
```

```
        Vertex u = Q.extractMin(); // take out vertex with min dist
```

```
        for ( Edge edge : g.outgoingEdges(u) ) { // from adjacency list
```

```
            Vertex v = edge.dest;
```

```
            if ( dist[v] > dist[u] + edge.weight ) { // relax, use edge u to v
```

```
                dist[v] = dist[u] + edge.weight;
```

```
                prev[v] = u;
```

```
                Q.decreaseKey(v); // dist[v] has new value, must move up in Q
```

```
            }
```

```
        }
```

```
    }
```

Sample Graph Problems

1.