5SENG003W - Algorithms, Week 9

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RECAP: Graphs

- Graphs consist of vertices connected by edges
- Edges can be directed or undirected
- Representations:
 - Set of vertex pairs
 - Ordered pairs for directed graphs:

```
\{(v_1, v_2), (v_1, v_3), (v_2, v_5), \ldots\}
```

Unordered pairs for undirected graphs:

```
\{\{v_1, v_7\}, \{v_2, v_5\}, \{v_2, v_3\}, \ldots\}
```

- Adjacency matrix
 - **a**[i][j] is 1 if there is an edge (v_i, v_j) and 0 otherwise
- Adjacency lists
 - Each vertex contains a list of outgoing edges
 - Vertices can be implicit: can represent the graph by an array of lists

Recap: Graph properties

- Adjacency between vertices, incidence between vertices and edges
- (In-, out-) degree
- Walks, paths, cycles
- Loops and parallel edges
- Connectedness, acyclicity
- Colours, weights, . . .

More about graph representations

- We have previously seen that graphs can be represented using
 - Adjacency matrices
 - Adjacency lists
- Let us talk about this in more detail. The main questions to ask yourself are
 - How do we represent vertices?
 - How do we represent edges?
 - how do we store additional data (labels, weights, ...)?

Representing vertices

- We can represent vertices using
 - Explicit Vertex objects
 - Containing the adjacency list (makes no sense with an adjacency matrix)
 - Containing all relevant additional data
 - Suitable if number of vertices is initially unknown
 - IDs
 - To index the adjacency matrix row or the adjacency list (in the array of lists)
 - Could be more than a single number (e.g. co-ordinates on a chessboard)

Representing edges

- We can represent edges using
 - Explicit Edge objects
 - Containing the target verted
 - Containing all relevant additional data
 - Suitable if there are parallel edges or complex additional data
 - Just the target vertex
 - ▶ ID or pointer to object

Recap: Searching in graphs

- A common operation on graphs is exploration
 - Searching for some particular vertex
 - Searching for a path between two vertices
 - Searching for a shortest path between two vertices
 - Or just trying to fully explore the graph
- Graph exploration is more complex than in trees or lists
 - Graphs can have cycles
 - Naive search could get stuck in a loop
 - Need to keep track of visited vertices to prevent this
- Previously introduced: depth-first and breadth-first search

Recap: Depth-first search

- In depth-first search, we recursively follow outgoing edges
- To explore a vertex, we
 - Pick a previously unvisited neighbour
 - Recursively explore that neighbour
 - Repeat if (after returning from the recursive call) there are still unvisited neighbours
- Note the difference between visited and explored
 - A node is visited as soon as we encounter it
 - A node is **explored** once we have visited all its neighbours (possibly from a different node)
 - This difference will be represented by the closed and open lists

Recap: Breadth-first search

- In breadth-first search, we explore the graph in "layers"
- Not recursive
- We maintain two data structures:
 - The set of all visited vertices (the closed list)
 - A queue of visited but not yet fully explored vertices (the open list)
- In each iteration, we
 - Check the neighbours of the open list's front element
 - Enqueue those which were not yet visited (and add them to the closed list)
 - Dequeue the front element (it is now fully explored)

Recap: Iterative depth-first search

- We can perform depth-first search in the same iterative style as breadth-first search
- The key change is organising the open list as a stack rather than a queue
- ► In each iteration, we
 - Check if the top element has unvisited neighbours
 - If yes, push one of them onto the open list (and add it to the closed list)
 - If not, pop the top element off (it is now fully explored)
- Alternatively, in each iteration, we
 - Pop the top element of the stack
 - Push all its unvisited neighbours onto the open list (and add them to the closed list)
- The first version means re-examining adjacency lists but will directly produce a path to the target (if any)

Depth-first search: iterative version

The second version of iterative DFS looks like this (assuming for simplicity that Vertex contains List<Vertex> adjacencyList;)

```
public void DFS(Vertex source) {
    // Set up open and closed list
    Stack<Vertex> o = new Stack<Vertex>();
    HashSet<Vertex> c = new HashSet<Vertex>();
    o.push(source);
    c.add(source);
    while(!o.empty()){
        Vertex v = o.pop();
        for(Vertex w : v.adjacencyList)
            if(! c.contains(w)){ // not yet visited
                o.push(w);
                c.add(w);
```

Aside: eliminating recursion

- Iterative DFS is an example of a more general idea:
 We can get rid of recursion by using a stack (to store pending branches)
- Example: pre-order traversal in a tree looks like this:

```
public class TreeNode{
   public int data;
   public TreeNode leftChild, rightChild;
}

public void preOrder(TreeNode n) {
   if(n != null) {
       System.out.println(n.data);
       preOrder(n.leftChild);
       preOrder(n.rightChild);
   }
}
```

Aside: eliminating recursion

 Example: an iterative version of pre-order traversal (using java.util.Stack)

- This code could still be improved (some vertices get pushed and then immediately popped)
- Think about how to do this with in-order or post-order

Exploration for pathfinding

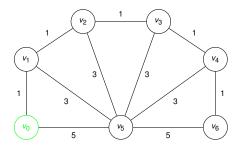
- The traversal algorithms explore the entire graph
- When searching for a path to a target this is not needed. Instead we can
 - Abort the search as soon as the target is found
 - Keep track of the path taken, e.g. using a predecessor map
 - This map represents a spanning tree with predecessors being parents
- Either algorithm will find a path if one exists
- BFS also guarantees that the path found has minimal length

Why breadth-first search works

- BFS advances in "layers":
 - For a vertex v, let f(v) be the distance from the start to v.
 - ▶ BFS first visits first all v with f(v) = 0, then f(v) = 1 etc
- Why is this the case?
 - At first, the open list only contains the start vertex, which is the only vertex s with f(s) = 0
 - All vertices with f(v) = k + 1 are reached in one step from vertices with f(v) = k
 - So they are all enqueued while the vertices with f(v) = k are getting visited

Breadth-first search breaks for weighted graphs

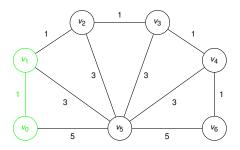
- In a weighted graph, vertices are no longer found ordered by distance
- For example, vertices with f(v) = 5 may have edges to vertices with f(v) = k for **any** k > 5.
- Idea: could use a sorted data structure like a binary search tree for the open list to fix this
 - ► Then insertion and extraction in the open list both are in O(log n)
- But we only ever need the minimal-distance vertex
 - So we can use a **priority queue** based on a **min-heap**.
- This gives us Dijkstra's Algorithm.
 - Finds shortest paths from the start to all other targets
 - Easy to turn into a single-target version



- Open list:
 - v_0
- Closed list:

$$\{v_0\mapsto 0\}$$

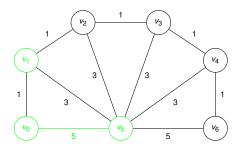
- ▶ Step 5: Vertex v_2 inserted in front of v_5
- Done: found shortest distances to all vertices



- Open list:
 - $v_0 v_1$
- Closed list:

$$\{v_0\mapsto 0,v_1\mapsto 1\}$$

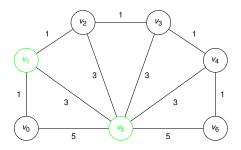
- ► Step 5: Vertex v_2 inserted in front of v_5
- Done: found shortest distances to all vertices



- Open list:
 - $v_0 \ v_1 \ v_5$
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_5 \mapsto 5\}$$

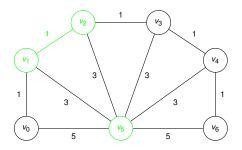
- ► Step 5: Vertex v_2 inserted in front of v_5
- Done: found shortest distances to all vertices



- Open list:
 - V₁ V₅
- Closed list:

$$\{v_0\mapsto 0,v_1\mapsto 1,v_5\mapsto 5\}$$

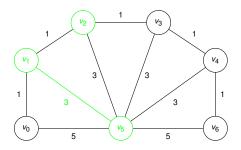
- ► Step 5: Vertex v_2 inserted in front of v_5
- Done: found shortest distances to all vertices



- Open list:
 - $V_1 \ V_2 \ V_5$
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_5 \mapsto 5\}$$

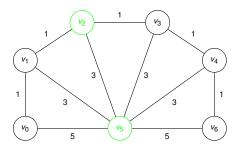
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - $V_1 \ V_2 \ V_5$
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_5 \mapsto 4\}$$

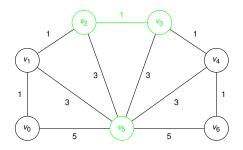
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - *V*₂ *V*₅
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_5 \mapsto 4\}$$

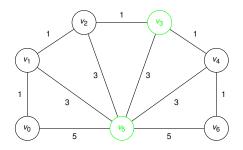
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



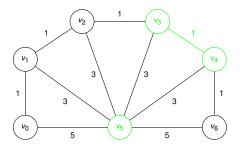
- Open list:
 - *V*₂ *V*₃ *V*₅
- Closed list:

$$\{\textit{v}_0 \mapsto \textit{0}, \textit{v}_1 \mapsto \textit{1}, \textit{v}_2 \mapsto \textit{2}, \textit{v}_3 \mapsto \textit{3}, \textit{v}_5 \mapsto \textit{4}\}$$

- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



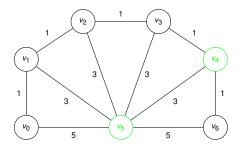
- Open list:
 V₃ V₅
- Closed list:
 - $\{v_0 \mapsto 0, v_1 \mapsto 1, v_2 \mapsto 2, v_3 \mapsto 3, v_5 \mapsto 4\}$
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - *V*₃ *V*₅ *V*₄
- Closed list:

$$\{v_0\mapsto 0, v_1\mapsto 1, v_2\mapsto 2, v_3\mapsto 3, v_4\mapsto 4, v_5\mapsto 4\}$$

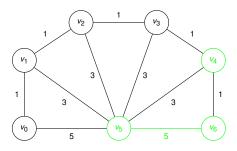
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - *V*₅ *V*₄
- Closed list:

$$\{v_0 \mapsto 0, v_1 \mapsto 1, v_2 \mapsto 2, v_3 \mapsto 3, v_4 \mapsto 4, v_5 \mapsto 4\}$$

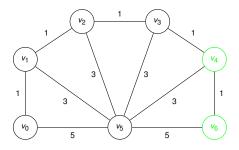
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - v_5 v_4 v_6
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_3 \mapsto 3, \textit{v}_4 \mapsto 4, \textit{v}_5 \mapsto 4, \textit{v}_6 \mapsto 9\}$$

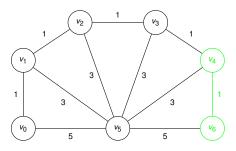
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - *V*₄ *V*₆
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_3 \mapsto 3, \textit{v}_4 \mapsto 4, \textit{v}_5 \mapsto 4, \textit{v}_6 \mapsto 9\}$$

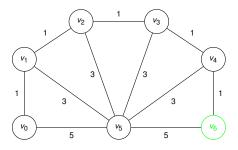
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - *V*₄ *V*₆
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_3 \mapsto 3, \textit{v}_4 \mapsto 4, \textit{v}_5 \mapsto 4, \textit{v}_6 \mapsto 5\}$$

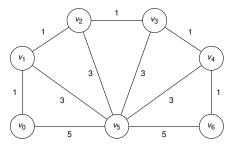
- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
 - v_6
- Closed list:

$$\{\textit{v}_0 \mapsto 0, \textit{v}_1 \mapsto 1, \textit{v}_2 \mapsto 2, \textit{v}_3 \mapsto 3, \textit{v}_4 \mapsto 4, \textit{v}_5 \mapsto 4, \textit{v}_6 \mapsto 5\}$$

- Step 5: Vertex v₂ inserted in front of v₅
- Done; found shortest distances to all vertices



- Open list:
- ► Closed list: $\{v_0 \mapsto 0, v_1 \mapsto 1, v_2 \mapsto 2, v_3 \mapsto 3, v_4 \mapsto 4, v_5 \mapsto 4, v_6 \mapsto 5\}$
- ▶ Step 5: Vertex v_2 inserted in front of v_5
- Done: found shortest distances to all vertices

Dijkstra's algorithm in a nutshell

- In a nutshell, Dijkstra's algorithm does the following:
 - For each vertex v, keep track of the shortest known distance d(s, v) from s to v
 - Many versions of the algorithm initially set all distances to ∞ except s, where it is 0
 - Our closed list represents those vertices where we have found a non-∞ distance
 - In each iteration, expand a vertex v
 - This means exploring its outgoing edges (v, x)
 - v must be a previously unexpanded vertex with minimal d(s, v)
 - Our open list contains the candidates for this
 - ▶ During the expansion, update d(s, x) if d(s, v) + w(v, x) is smaller

Dijkstra's algorithm: details

- We want to be able to output actual paths
- So we keep track of the predecessor of each vertex (i.e. from which other vertex we reached it, getting the shortest distance)
- Final output in the example:

Vertex	Distance	Shortest path
v_0	0	<i>v</i> ₀
<i>v</i> ₁	1	v_0, v_1
<i>V</i> ₂	2	v_0, v_1, v_2
<i>V</i> ₃	3	v_0, v_1, v_2, v_3
V_4	4	v_0, v_1, v_2, v_3, v_4
<i>V</i> ₅	4	v_0, v_1, v_5
<i>v</i> ₆	5	$v_0, v_1, v_2, v_3, v_4, v_6$

- We make a few changes to our existing code:
- We use a dedicated Vertex class
 - This allows us to store additional information in them, which will be helpful soon
- ► We use maps (java.util.HashMap or std::map)
 - In adjacency lists to keep track of weights

```
public class Vertex{
   public int id;

   public HashMap<Vertex, Integer> adjacencyList;

   /* ... */
}
```

In the closed list to keep track of shortest path lengths

- We make a few changes to our existing code:
- We use a wrapper class containing
 - A vertex
 - The length of the shortest known path from s to v

```
public class WrappedVertex{
   public Vertex vertex;

  // shortest currently known distance from the start
   public int distance;

   public WrappedVertex(Vertex v, int d) {
        vertex = v;
        distance = d;
   }
}
```

in order to store vertices in a heap

- We adapt the MinHeap class
 - It now stores wrapped vertices
 - It uses the current distance to figure out how far to sift up a new entry

```
public class MinHeap {
    ArrayList<WrappedVertex> items;
    public MinHeap() {
        items = new ArrayList<WrappedVertex>();
    public void insert (Vertex v. int d) {
        WrappedVertex newItem = new WrappedVertex(v, d);
        int index = items.size();
        items.add(newItem);
        // Determine position for insertion
        while (index > 0) {
            int parent = (index-1)/2: // parent index
            if(items.get(parent).distance > newItem.distance) { // sift up
                items.set(index, items.get(parent));
                index = parent:
            else
                break:
        items.set(index, newItem);
   /* ... */
```

- Finally, we adapt the search method we have seen before to use a heap as its open list
- Note that if we find a new, shorter path to a vertex v, we
 - Create a new WrappedVertex with the new distance
 - Insert it in the open list
 - We may have several wrapped versions of the same vertex
 - ► They will have different distances
 - ▶ The one with the smallest distance will be handled first
 - The others will just be discarded eventually

The structure of the main loop looks like this:

```
public void Dijkstra(Vertex source) {
  // Set up open, closed, and predecessor lists
  MinHeap o = new MinHeap():
  HashMap<Vertex, Integer> c = new HashMap<>();
  HashMap<Vertex, Vertex> p = new HashMap<>();
  o.insert(source, 0):
  c.put(source, 0);
  p.put(source, null);
  while(!o.empty()){
      WrappedVertex w = o.extractMinimum();
      if(w.distance == c.get(w.vertex)){ // w is up to date
         /* TO DO: complete this
             Output the distance and shortest path (follow the chain of predecessors)
             For every edge e out of w.vertex,
               - check if e gives a new shortest path to its target
               - update lists if ves
         */
```

Dijkstra's algorithm: analysis

- Dijkstra's algorithm is an example of a greedy algorithm
 - It always picks the option that currently looks best
- It expands each vertex (i.e. examines its outgoing edges) once
- Its complexity depends on details such as graph representation, sparsity, etc.
 A general lower bound is Θ(|E| + |V|²).