

COMP 476 Advanced Game Development

Session 3
Pathfinding AI, World Representations

Pathfinding AI (Reading: AI for G, Millington § 4.1-4.4)

Lecture Overview

- Pathfinding
- ☐ Dijkstra's Algorithm
- ☐ A* Algorithm
- □ Pathfinding Lists
- **☐** World Representation



Pathfinding

If we need to design an AI...

- ☐ That is able to calculate a <u>suitable route</u> through the game level to get from where it currently is to its goal position ...
- □ ... and that route is to be <u>as short or rapid as</u> <u>possible</u>, or at least, looks smart enough (!) ...
- ☐ ... we need to do ... pathfinding



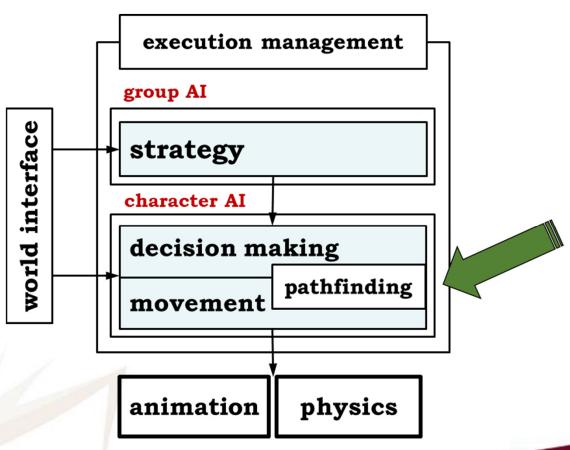
Pathfinding

- □ Sometimes called path planning (but pathfinding is still the usual term used)
- □ Simple task: Work out where to move to reach a goal
- ☐ In Millington and Funge's AI model sits on the border between decision-making and movement.
 - The goal is decided by decision-making AI
 - Pathfinding simply works out how to get there
 - Movement AI gets the character there



Pathfinding AI

Millington and Funge's Model





Using Pathfinding

- □ Vast majority of games use pathfinding solutions that are <u>efficient and easy to implement</u> but pathfinding AI cannot work directly with game level data.
- □ Requires game level <u>to be represented in a</u> <u>perticular data structure</u>:

A directed non-negative weighted graph

☐ In particular, a graph is a <u>simplified version of</u> <u>the level</u>; the better the simplification, the better the path



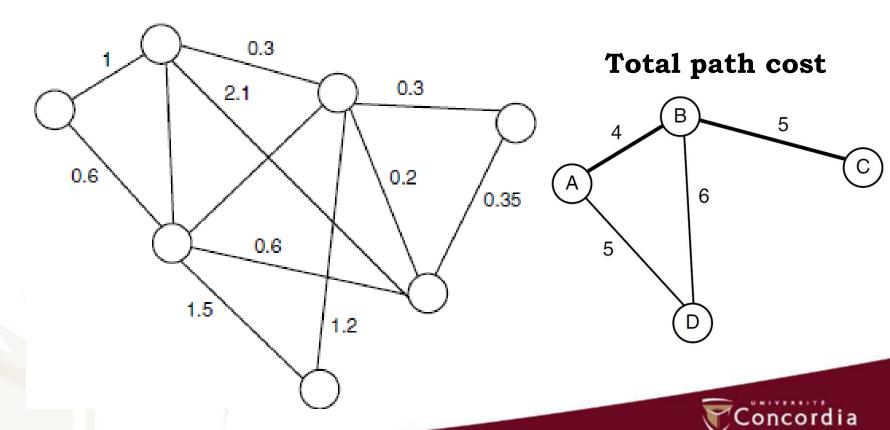
Graphs

- Formal representation:
 - A set of nodes
 - A set of connections/edges

 (an unordered pair of nodes)
- Each <u>node represents</u>
 - A <u>region of the game level</u> (room, section of corridor, platform, etc.)
- □ Connections represent
 - Which locations (nodes) are connected
 - E.g., if a room adjoins a corridor
- A graph splits the whole game level into regions which are connected together

Weighted Graphs

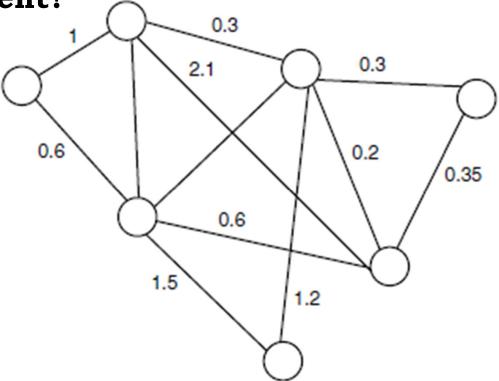
☐ In addition to having nodes and connections, a numerical value or "weight" is used to label each connection with an associated cost value



Graphs in Games

Weights?

☐ In games, what can we use the weights (costs) to represent?



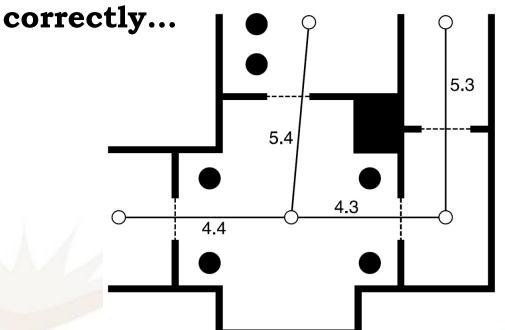


Graphs in Games

Nodes/Connections?

☐ How shall we overlay a weighted graph onto game level geometry?

Suggest some ways to place nodes/connections



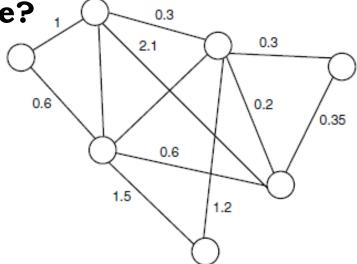
Weighted graph overlaid onto level geometry



Graphs in Games

Now, what's the whole idea?

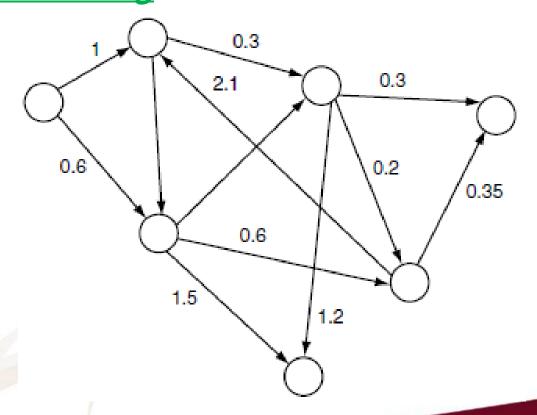
□ So, back to the original idea of pathfinding: <u>How</u> shall we use a weighted graph (like below) to find paths in the game? 1 □ 0.3



- ☐ Can we have negative weights in a graph?
- ☐ Can certain nodes be only connected one way(just like a one-way street)?

Directed Weighted Graphs

□ Connections of graph are <u>directed</u>, or allows movement between 2 nodes to be from <u>one</u> <u>direction only</u>





Directed Weighted Graphs

- ☐ Can be useful in 2 special situations:
 - Reachability between two locations (node A can reach node B, but node B cannot reach node A)
 - Allow both connections in opposite directions to have different weights (the cost to move from node A to node B is different from the cost to move from node B to node A)









Graph Representations

Now, what's the whole idea?

- ☐ Graph class store an array of connection objects for any node
- □ Connection class store cost, 'from' node, 'to' node

```
class Graph:
    # Returns an array of connections (of class
    # Connection) outgoing from the given node
    def getConnections(fromNode)

class Connection:
    # Returns the non-negative cost of the connection
    def getCost()

# Returns the node that this connection came from
    def getFromNode()

# Returns the node that this connection leads to
    def getToNode()
```



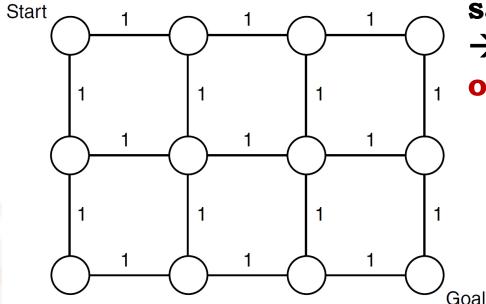
Dijkstra's Algorithm

- □ Named after Edsger Dijkstra, a mathematician
- Originally designed to solve a problem in mathematical graph theory called "<u>shortest</u> <u>path</u>", not for games
- □ Idea: Finding the shortest path from one start point to one goal point
- □ Dijkstra's Algorithm is designed to find the shortest routes to everywhere from an initial point – Wasteful?
- We will examine Dijkstra's because it is a simpler version of the main pathfinding algorithm A*.



The Problem

- □ Aim: Given a graph and two nodes (start and goal), generate a path such that the total path cost of that path is minimal among all possible paths from start to goal
- There may be any number of paths with the Start same minimal cost



→ Just returning any one will do



Same Cost Paths

- ☐ The path we expect to be returned consists of a <u>set of connections</u>, <u>not nodes</u>.
- Many games do not consider having more than one connection between any pair of nodes (only one possible path?)
- □ With many connections between any pair of nodes, an optimum path is needed to make the AI look smart! So, keep track of multiple connections, use the connection of these with smallest cost.
- ☐ This is particularly useful if the costs change over the course of the game or between different characters

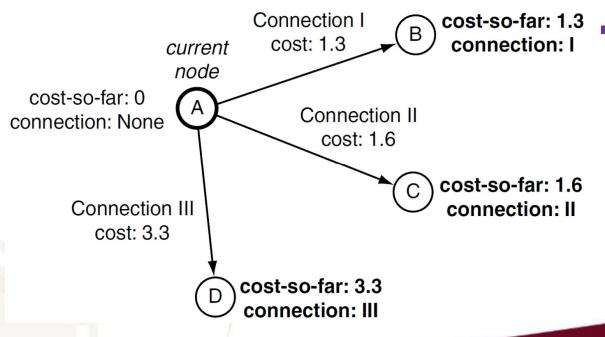
The Algorithm

- □ Spread out from the start node along its connections
- □ Each time, <u>keep a record of the direction</u> it came from with "arrows" (stored as directed edges)
- Eventually, when the goal is reached, <u>follow the</u> <u>"arrows", in reverse</u>, back to the start point to generate the complete minimal route
- Works in iterations
- □ Each iteration: <u>consider one node of a graph</u> and <u>follow its outgoing connections</u> (to other nodes)
- Each iteration's node is called the "current node"



The Algorithm

- ☐ Processing the <u>Current Node</u> at each iteration
 - Consider outgoing connections from current node
 - For each connection, finds the end node and stores the total cost of the path so far as the "cost-so-far" (from the start node)



Each connected node stores a)

<u>cost-so-far</u> and b) which connection to the node was used to get the cost-so-far

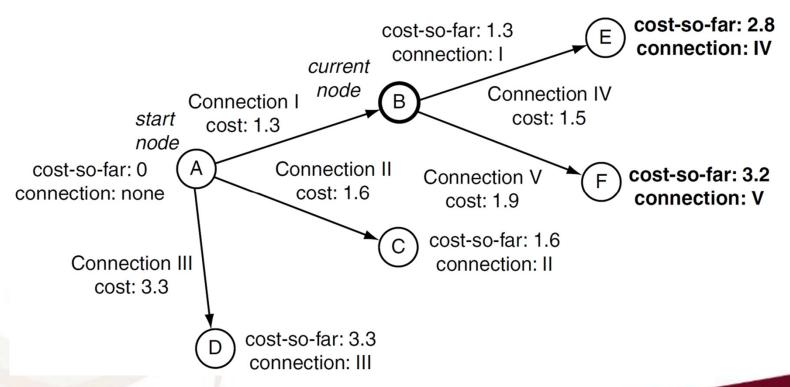


The Algorithm

☐ After 1st iteration, for neighboring unvisited nodes

Cost-so-far = Sum of connection cost +

cost-so-far of current node





Node Lists

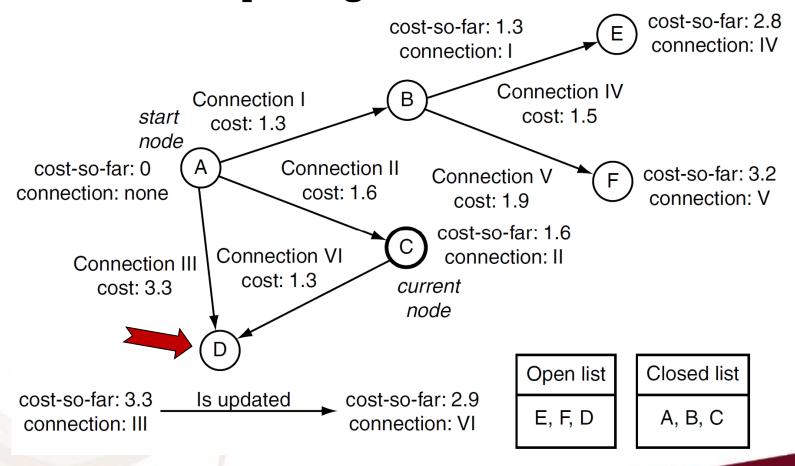
- ☐ Keeps track of all the nodes seen so far in 2 lists:
 - Open List
 - Records all nodes seen, but haven't been processed in its own iteration yet
 - Closed List
 - Records all nodes that have been processed in its own iteration (all "current node" s)
- Nodes in neither list are "unvisited"
- □ At each iteration, a node from the open list with smallest cost-so-far is chosen (and processed), then removed from the open list and placed in the closed list

Cost-so-far for Open and Closed Nodes

- □ What happens when we arrive at an open or closed node (again) during an iteration?
- ☐ If there are <u>new values</u> of the cost-so-far that are lower than the existing value of node, an <u>update</u> of that node is needed, regardless of which list it's in:
 - The open nodes that have their values updated simply stay on the open list
 - For closed nodes simply force the algorithm to recalculate and re-propagate new values:
 - ✓ Remove node from closed list, place back into open list with its values revised
 - ✓ It will be processed once again and have its connections reconsidered

A few iterations later...

■ Notice the updating done on node D





Terminating the Algorithm

- □ Basic Dijkstra's Algorithm <u>terminates when the</u> <u>open list is empty</u> (no more nodes to process), and all nodes are in the closed list
- ☐ In practical pathfinding, we can terminate earlier once we have
 - placed the goal node on the open list, and
 - the goal node is the node with the smallest costso-far on the open list (an added heuristic)
- Why do we add this heuristic rule?

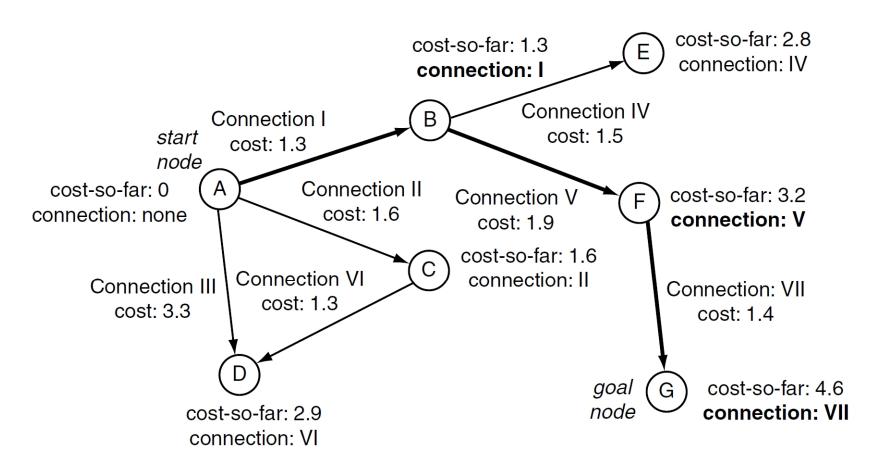


Retrieving the Path

- ☐ Final step
- □ Start from goal node, <u>look back at the connection</u> that was used to arrive to this node; then for the node at the other end of the connection, <u>look at the connection used to arrive at that node</u>; etc. Continue until reaching the start goal
- ☐ The list of connections <u>need to be reversed to obtain the right path order</u>



Final Result



Connections working back from goal: VII, V, I Final path: I, V, VII



Data Structures & Interfaces

- ☐ Graph Rarely a performance bottleneck (can be created offline)
- □ Simple List (for reporting connections in Graph)
 - Not performance critical, use a basic linked list (list) or resizable array (vector)
- Pathfinding Lists
 - Large Open and Closed lists can affect performance. <u>Need optimization to perform</u> <u>these operations</u>:
 - A. Adding/removing entry,
 - B. Finding smallest element,
 - C. Finding entry in list corresponding to a particular node

Practical Performance of Dijkstra's

- □ Depends mostly on performance of operations in pathfinding list(s) data structure
- ☐ Theoretically (n nodes, m average number of connections per node):
 - Time complexity (execution speed): O(nm)
 - Space complexity (memory): O(nm) worst-case
- **□** Select implementations:
 - Min Priority Queue: $O(n^2 + nm)$
 - Fibonacci Heap: O(n logn + nm)
- □ Note: $1 < m \le n-1$ for a connected graph with n > 2



Weaknesses

Searches whole graph indiscriminately for shortest path Start Start Wasteful for point-to-point pathfinding Suffers from terrible amount of "fill" Start Goal Goal Start (nodes that were considered but Key Open nodes never made Closed nodes Unvisited nodes part of the NB: connections are hidden for simplicity final route)



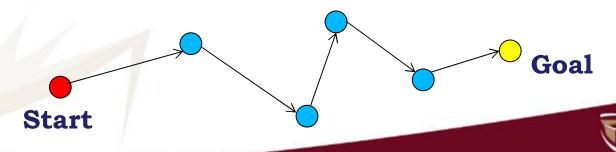
A* Algorithm

- ☐ Most widely used for pathfinding in games
 - Simple to implement
 - Efficient
 - Lots of <u>scope for optimization</u>
- □ Unlike Dijkstra algorithm, A* is designed for point-to-point pathfinding, not for solving the shortest path problem in graph theory.
 - Always returns a single path from source to goal
 - The "best" possible path



The Problem

- ☐ Given a graph (a directed non-negative weighted graph) and two nodes in that graph (a start and a goal node)
- □ Generate a path → the total path cost of the path is minimal among all paths from start to goal
- ☐ If there are more than one possible solution, <u>any</u> <u>minimal path will do</u>
- □ Path should consist of a list of connections from the start node to goal node



Differences from Dijkstra

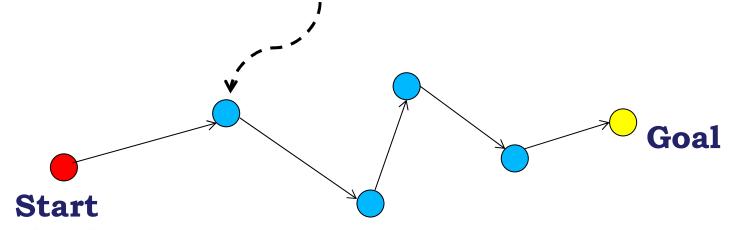
- □ Dijkstra → Always considered the open node with lowest cost-so-far for processing
- \square A* \rightarrow Consider node that most likely to lead to shortest overall path
- "most likely" → controlled by a "heuristic", an estimated rule of thumb
- ☐ If accurate heuristic used: Efficient
- ☐ If bad heuristic used: <u>May perform worse than</u> <u>Dijkstra!</u>



"heuristic" for A*

☐ Think about how an "estimated total cost" should be calculated?...

How to <u>estimate total cost from</u> <u>here to goal</u> for this node?

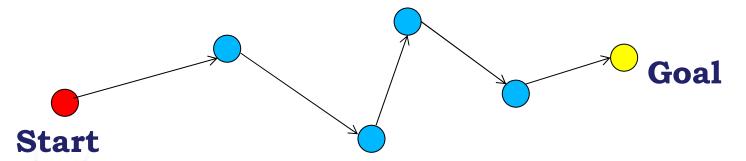


☐ Think of a good heuristic to use?



"heuristic" for A*

- □ A node with smallest estimated-total-cost should have:
 - A relatively small cost-so-far (Dijkstra only has this)
 - A relatively small estimated distance to go to reach goal (the heuristic)



estimated-total-cost = cost-so-far + heuristic cost



Pathfinding Lists

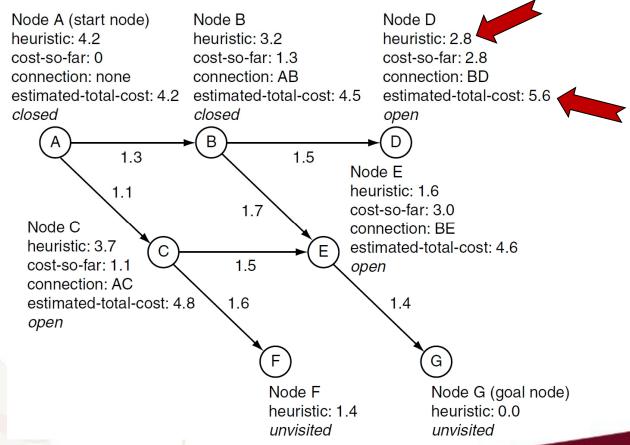
- ☐ Just like Dijkstra, A* also keeps an open list of nodes and a closed list of nodes
 - Nodes are moved to open list as they are found at end of connections (previously "unvisited")
 - Nodes are moved to closed list as they are processed in their own iteration
- ☐ Unlike Dijkstra, node with smallest estimatedtotal-cost is selected from the open list, not the node with smallest cost-so-far



Using the "heuristic"

□ A* stores 2 additional values – heuristic value

and estimated total cost





Calculating Costs

- estimated-total-cost = cost-so-far + Heuristic cost
 - As with Dijkstra's algorithm, the cost-so-far is calculated using the actual total cost of the path travelled so far from the start to the current node
 - The heuristic cost cannot be negative (obviously!)
- ☐ To update a <u>previously visited node</u>
 - Compare cost-so-far, not the estimated total cost (with heuristic added)
 - This is the only reliable "real" value. Heuristics involve estimation (and are usually fixed values)

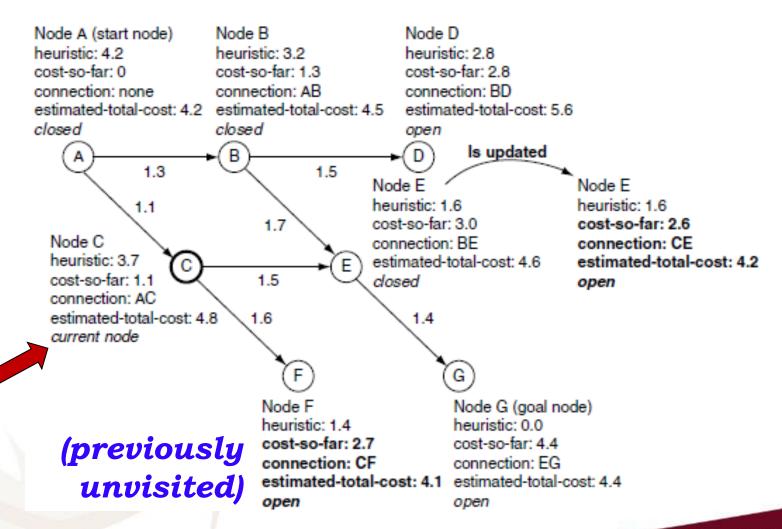


Updating Costs

- ☐ If there are new values of the cost-so-far that are lower than the existing value of node, an update of that node is needed, as with Dijkstra, regardless of which list it is in:
 - The open nodes that have their values revised simply stay on the open list
 - For closed nodes, remove the node from closed list, place it back into open list with its values revised - to be processed once again and have its connections reconsidered



Updating Costs





Terminating the Algorithm

- □ A* terminates when the goal node is the smallest node on the open list (with smallest estimatedtotal-cost)
- ☐ This does not guarantee that shortest path is found. Why?
- ☐ It is natural for A* to run a bit longer to generate a guaranteed optimal result
- □ One way: Require that A* terminates when node in open list with smallest cost-so-far has a costso-far greater than cost of path found to goal (almost like Dijkstra)
- ☐ Drawback: Same fill problem as Dijkstra!!



Algorithm Performance

- Depends on data structure used
 - Time complexity, O(lm)
 - Space complexity, O(lm)

- 1 = amount of "fill"
- 1: number of nodes whose total estimated-pathcost is less than that of goal, < n from Dijkstra
- m: average number of outgoing connections from each node
- □ Heuristic calculation is usually O(1) execution time and memory, so no effect on overall performance. Can be even calculated offline in some cases.
- □ Refer to [M&F] for more detail on pseudocode and data structures involved