

## §5 Path Finding

- common problem in computer games
  - routing characters, troops etc.
- computationally intensive problem
  - complex game worlds
  - high number of entities
  - dynamically changing environments
  - real-time response



## Problem statement

- given a start point  $s$  and a goal point  $r$ , find a path from  $s$  to  $r$  minimizing a given criterion
- search problem formulation
  - find a path that minimizes the cost
- optimization problem formulation
  - minimize cost subject to the constraint of the path

## The three phases of path finding

1. discretize the game world
  - select the waypoints and connections
2. solve the path finding problem in a graph
  - let waypoints = vertices, connections = edges, costs = weights
  - find a minimum path in the graph
3. realize the movement in the game world
  - aesthetic concerns
  - user-interface concerns

## Discretization

- waypoints (vertices)
  - doorways, corners, obstacles, tunnels, passages, ...
- connections (edges)
  - based on the game world geometry, are two waypoints connected
- costs (weights)
  - distance, environment type, difference in altitude, ...
- manual or automatic process?
  - grids, navigation meshes

## Grid

- regular tiling of polygons
  - square grid
  - triangular grid
  - hexagonal grid
- tile = waypoint
- tile's neighbourhood = connections



## Navigation mesh

- convex partitioning of the game world geometry
  - convex polygons covering the game world
  - adjacent polygons share only two points and one edge
  - no overlapping
- polygon = waypoint
  - middlepoints, centre of edges
- adjacent polygons = connections



### Solving the convex partitioning problem

- minimize the number of polygons
  - points:  $n$
  - points with concave interior angle (notches):  $r \leq n - 3$
- optimal solution
  - dynamic programming:  $O(r^2 n \log n)$
- Hertel–Mehlhorn heuristic
  - number of polygons  $\leq 4 \times$  optimum
  - running time:  $O(n + r \log r)$
  - requires triangulation
    - running time:  $O(n)$  (at least in theory)
    - Seidel's algorithm:  $O(n \lg^* n)$  (also in practice)

### Path finding in a graph

- after discretization form a graph  $G = (V, E)$ 
  - waypoints = vertices ( $V$ )
  - connections = edges ( $E$ )
  - costs = weights of edges ( $weight: E \rightarrow \mathbf{R}_+$ )
- next, find a path in the graph



### Graph algorithms

- breadth-first search
  - running time:  $O(|V| + |E|)$
- depth-first search
  - running time:  $\Theta(|V| + |E|)$
- Dijkstra's algorithm
  - running time:  $O(|V|^2)$
  - can be improved to  $O(|V| \log |V| + |E|)$

### Heuristical improvements

- best-first search
  - order the vertices in the neighbourhood according to a heuristic estimate of their closeness to the goal
  - returns optimal solution
- beam search
  - order the vertices but expand only the most promising candidates
  - can return suboptimal solution

### Evaluation function

- expand vertex minimizing
 
$$f(v) = g(s \rightsquigarrow v) + h(v \rightsquigarrow t)$$
- $g(s \rightsquigarrow v)$  estimates the minimum cost from the start vertex to  $v$
- $h(v \rightsquigarrow t)$  estimates (heuristically) the cost from  $v$  to the goal vertex
- if we had exact evaluation function  $f^*$ , we could solve the problem without expanding any unnecessary vertices