

CSE 326: Data Structures

Graph Algorithms Graph Search

Lecture 23

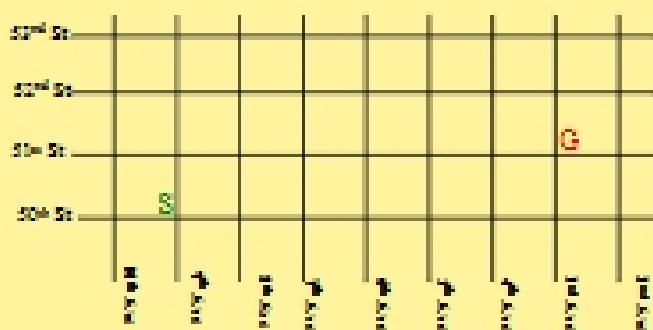
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Problem: Large Graphs

- ❑ It is expensive to find optimal paths in large graphs, using BFS or Dijkstra's algorithm (for weighted graphs)
- ❑ How can we search large graphs efficiently by using "commonsense" about which direction looks most promising?

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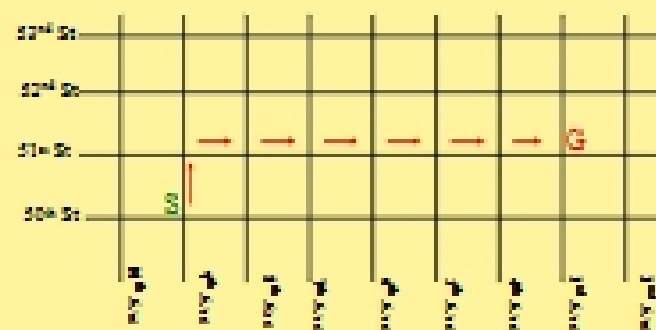
Example



Plan a route from 9th & 50th to 17th & 51st

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Example



Plan a route from 9th & 50th to 17th & 51st

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Best-First Search

- The *Manhattan distance* ($\Delta x + \Delta y$) is an estimate of the distance to the goal
 - It is a *search heuristic*
- ❑ Best-First Search
 - Order nodes in priority to minimize estimated distance to the goal
- ❑ Compare: BFS / Dijkstra
 - Order nodes in priority to minimize distance from the start

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Best-First Search

Open – Heap (priority queue)

Criteria – Smallest key (highest priority)

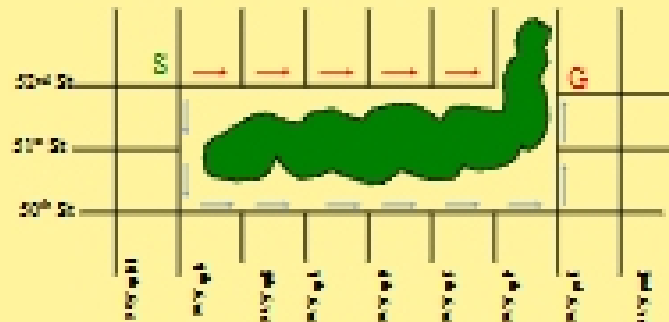
$h(n)$ – heuristic estimate of distance from n to closest goal

```
• Best_First_Search( Start, Goal_Test)
• insert(Start, h(Start), heap);
• repeat
•   if (empty(heap)) then return fail;
•   Node := deleteMin(heap);
•   if (Goal_Test(Node)) then return Node;
•   for each child of node do
•     if (Child not already visited) then
•       insert(Child, h(Child), heap);
•   end
•   Mark Node as visited;
• end
```

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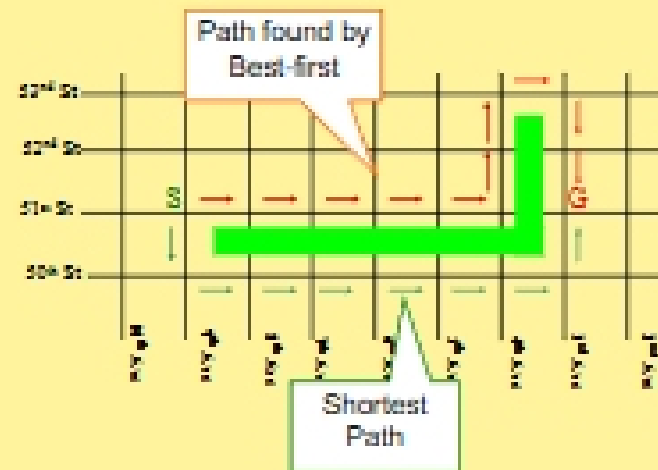
Obstacles

- Best-FS eventually will expand vertex to get back on the right track



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Non-Optimality of Best-First



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Improving Best-First

- ❑ Best-first is often tremendously faster than BFS/Dijkstra, but might stop with a non-optimal solution
- ❑ How can it be modified to be (almost) as fast, but guaranteed to find optimal solutions?
- ❑ A* - Hart, Nilsson, Raphael 1968
 - One of the first significant algorithms developed in AI
 - Widely used in many applications

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A*

- Exactly like Best-first search, but using a different criteria for the priority queue:
 - minimize (distance from start) + (estimated distance to goal)
- priority $f(n) = g(n) + h(n)$
 - $f(n)$ = priority of a node
 - $g(n)$ = true distance from start
 - $h(n)$ = heuristic distance to goal

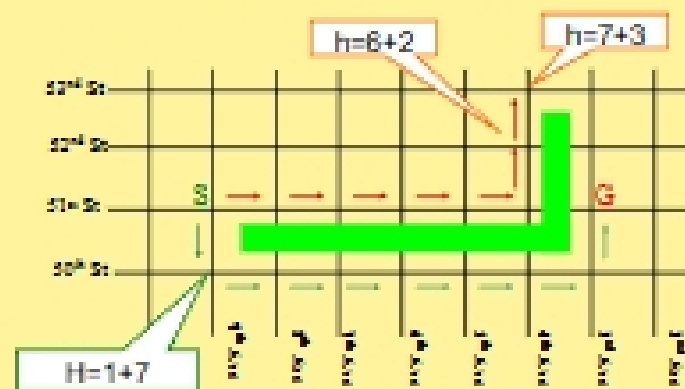
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Optimality of A*

- Suppose the estimated distance is *always* less than or equal to the true distance to the goal
 - heuristic is a lower bound
- Then: when the goal is removed from the priority queue, we are guaranteed to have found a shortest path!

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A* in Action



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Application of A*: Speech Recognition

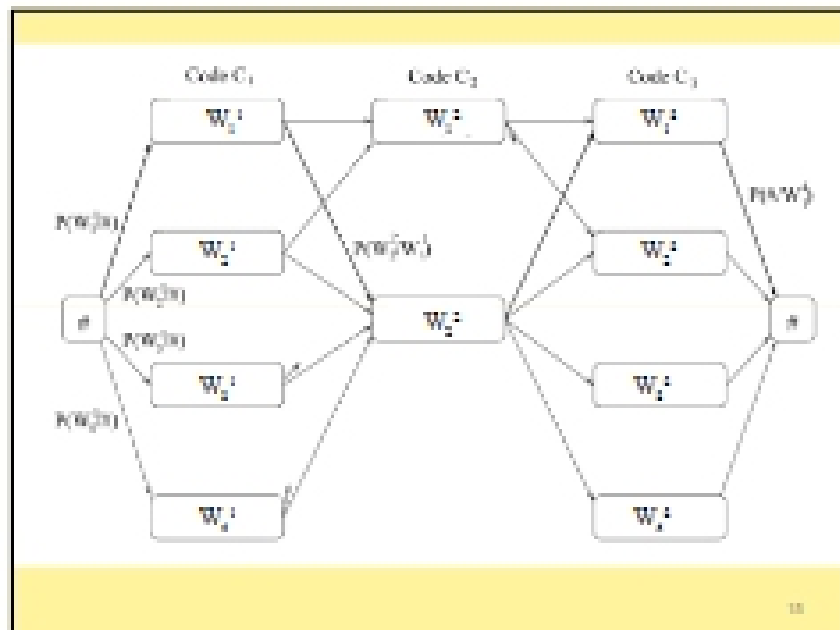
- (Simplified) Problem:
 - System hears a sequence of 3 words
 - It is unsure about what it heard
 - For each word, it has a set of possible “guesses”
 - E.g.: Word 1 is one of { “hi”, “high”, “I” }
 - What is the most likely sentence it heard?

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Speech Recognition as Shortest Path

- Convert to a shortest-path problem:
 - Utterance is a “layered” DAG
 - Begins with a special dummy “start” node
 - Next: A layer of nodes for each word position, one node for each word choice
 - Edges between every node in layer i to every node in layer $i+1$
 - Cost of an edge is smaller if the pair of words frequently occur together in real speech
 - Technically: $-\log$ probability of co-occurrence
 - Finally: a dummy “end” node
 - Find shortest path from start to end node

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Summary: Graph Search

- Depth First
 - Little memory required
 - Might find non-optimal path
- Breadth First
 - Much memory required
 - Always finds optimal path
- Iterative Depth-First Search
 - Repeated depth-first searches, little memory required
- Dijkstra's Short Path Algorithm
 - Use BFS for weighted graphs
- Best First
 - Can visit fewer nodes
 - Might find non-optimal path
- A*
 - Can visit fewer nodes than BFS or Dijkstra
 - Optimal if heuristic estimate is a lower-bound

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Dynamic Programming

- Algorithmic technique that systematically records the answers to sub-problems in a table and re-uses those recorded results (rather than re-computing them).
- Simple Example: Calculating the Nth Fibonacci number.

$$\text{Fib}(N) = \text{Fib}(N-1) + \text{Fib}(N-2)$$

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Floyd-Warshall

```

• for (int k = 1; k <= V; k++)
•   for (int i = 1; i <= V; i++)
•     for (int j = 1; j <= V; j++)
•       if ( ( M[i][k] + M[k][j] ) < M[i][j] )
           M[i][j] = M[i][k] + M[k][j]
    
```

Invariant: After the k th iteration, the matrix includes the shortest paths for all pairs of vertices (i,j) containing only vertices $1..k$ as intermediate vertices

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