

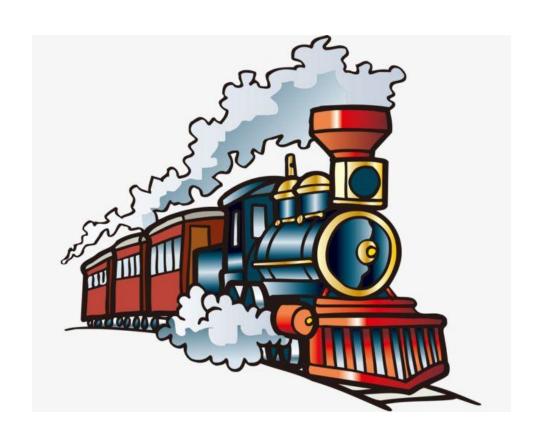
Graph searching



- You want to visit Kolari by train for a nice ski vacation
- Starting from Tampere is it possible at all?
- Which route is most economical in terms of distance, in time, money, ...?
- Routes are represented as a graph of cities
- Estimated times (distances) between cities might be given

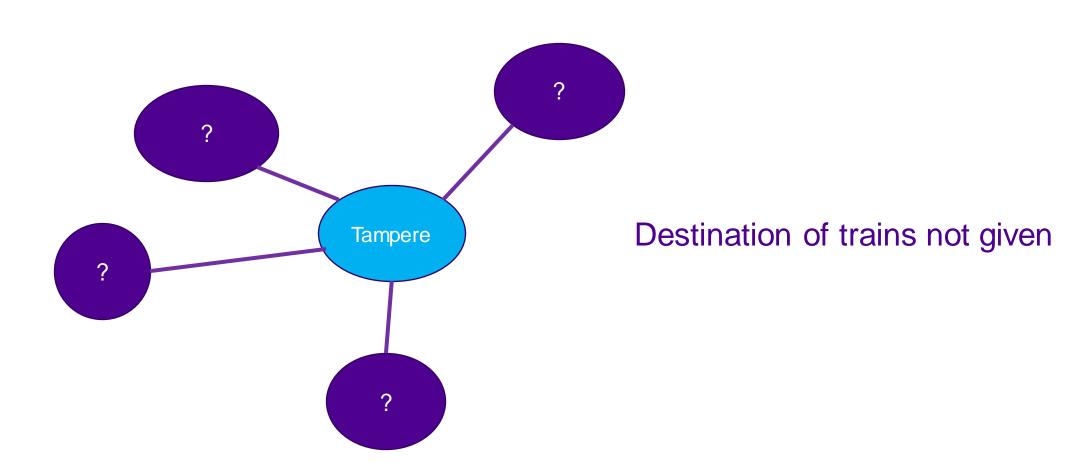


- Start state for our Kolari trip is Tampere train station
- The goal state is Kolari train station
- Primary interest is to find a way to reach Kolari
- Secondary interest is to find the most economical route to Kolari
- The cost of a path is the sum city costs en route
- Each station may have several successors (neighboring cities)
- Depending on our level of knowledge there are different search options





Unknown environment

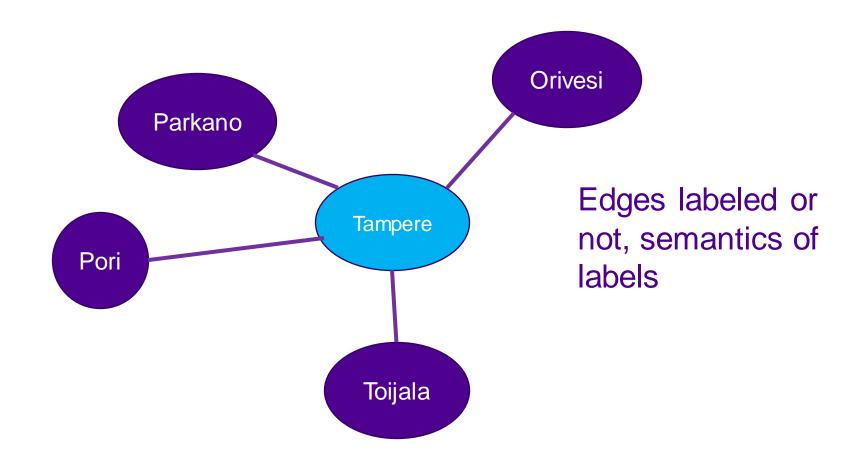






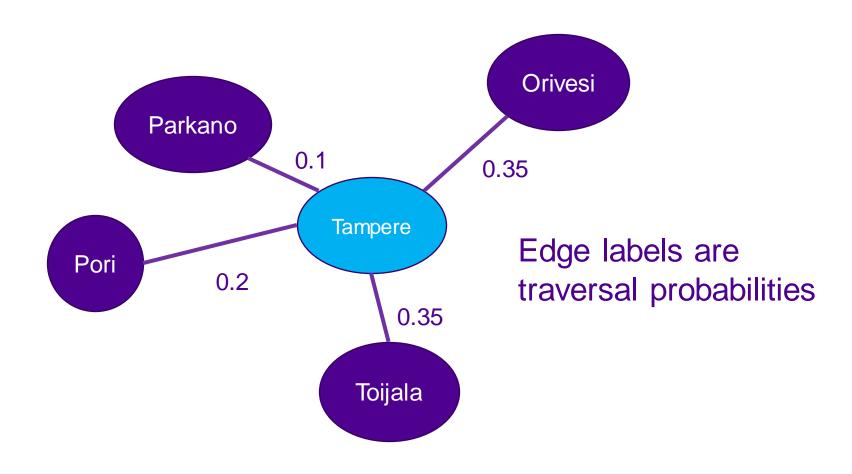


Known deterministic environment

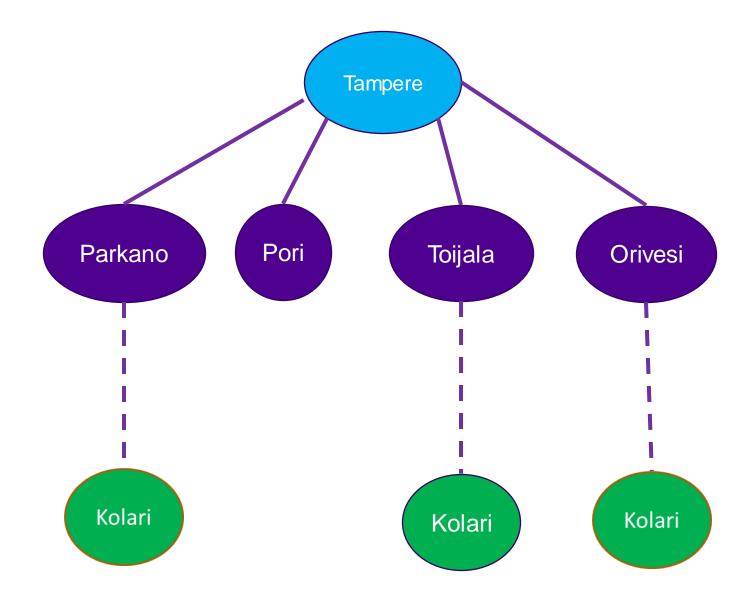




Nondeterministic environment



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Exhaustive search — a.k.a. Brute-force

- Search spaces in general (real world) are huge
- The worst setting is that we have to examine all possible states of the search space (exhaust it)
- Systematically go through the possible states of the world
- Entails keeping track of visited states to avoid looping
- This is particularly true when we aim at optimizing some objective
- Test the satisfiability of a logical formula

$$(x_1 \lor \neg x_2 \lor x_5) \land x_3 \lor \neg x_4$$
$$x_1 = T, x_2 = F, \dots$$

• If the state space is large, exhaustive search is unavoidably slow (in asymptotic sense)



State Space Sizes

World state:

Agent positions: 120

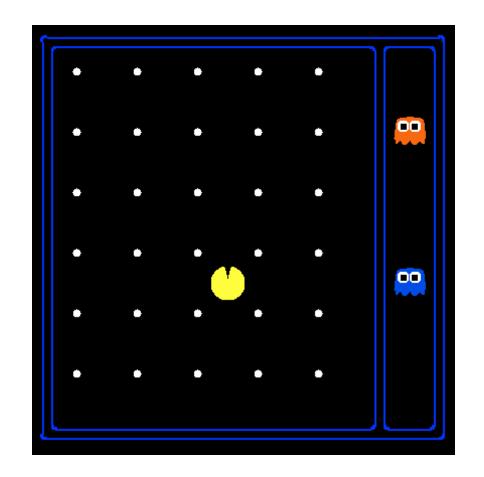
• Food count: 30

• Ghost positions: 12

Agent facing: NSEW

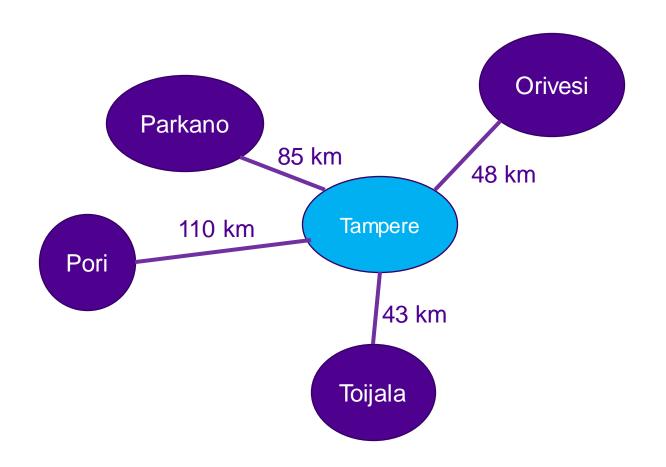
How many

- World states?
 120x(2³⁰)x(12²)x4
- States for pathing?120
- States for eat-all-dots?
 120x(2³⁰)



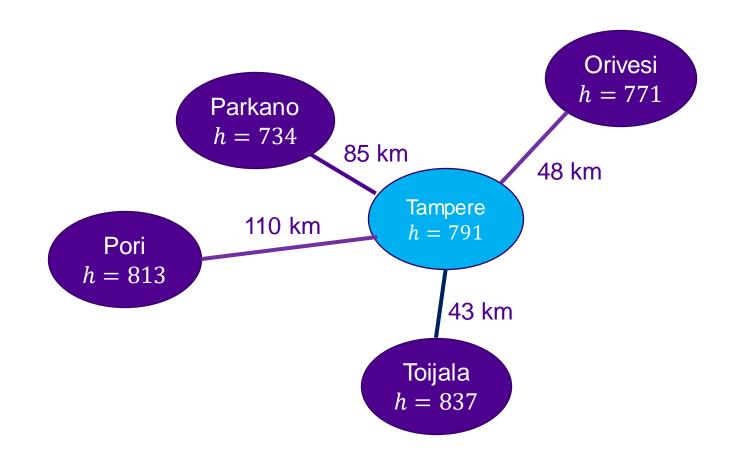


Informed search

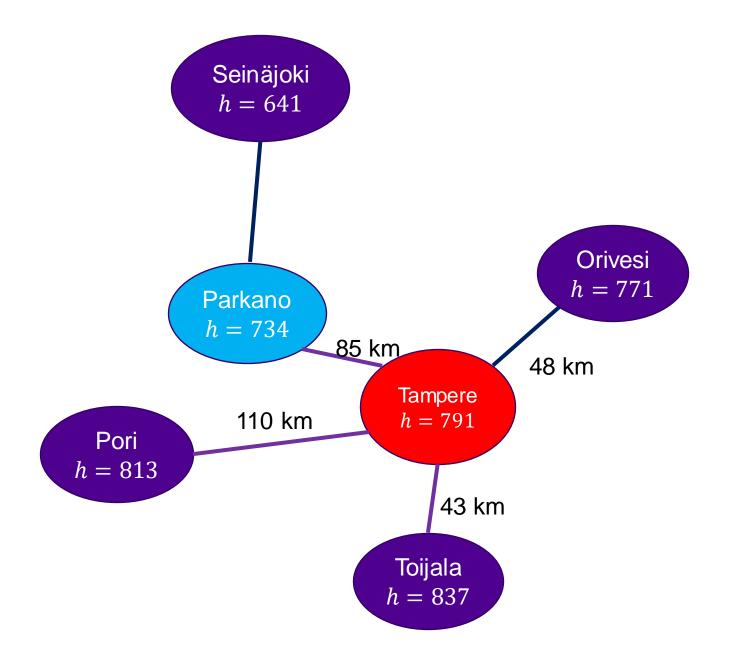




Knowing heuristic info



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Sudoku solved?

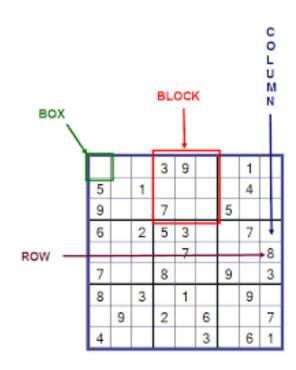
5	3			7				
6			1	9	5			
	9	8					6	
8 4 7				6				3
4			8		3			1 6
7				2				6
	6					2	8	
			4	1	9			5 9
				8			7	9

5	3	4	6	7	8	9	1	2
6	7	2	1	9	5	3	4	8
1	9	8	ო	4	2	5	6	7
8	5	9	7	6	1		2	3
4	2	6	8	5		7	9	1
7	1	3	9	2	4	8	5	6
9	6	1	5	3	7	2	8	4
2	8	7	4	1	9	6	3	5
3	4	5	2	8	6	1	7	9



Rules of Sudoku

- One of each number 1—9 on
 - Each block
 - Each row
 - Each column





Search strategy for Sudoku

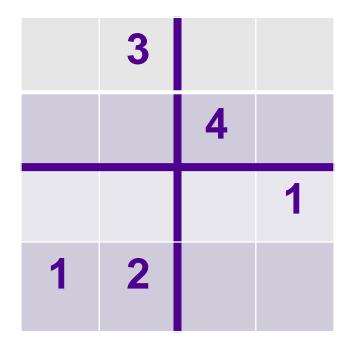
- Start, e.g., from the left bottom corner of the matrix (if empty)
- Attempt numbers 1,2,...,9 in order until there is no illegal configuration according to the rules
- Continue from the next cell
- If at any point you run into an illegal configuration, back-up (recursively) to the last number selection and choose the next candidate.
- Eventually (after having examined enough value combinations) you'll find a solution as long as one exists

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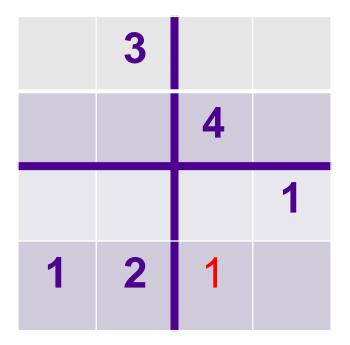
- 1. cell ← bottom left-most cell
- 2. while cell value is predetermined move to next cell <-
- 3. cell value ← 0
- 4. cell value++
- **5.** if cell value \leq 9 and legal configuration
- 6. move to next cell; **jump** to row 2
- 7. if cell value > 9 then
- 8. back-up to the previous not predetermined cell
- **9. jump** to row 4



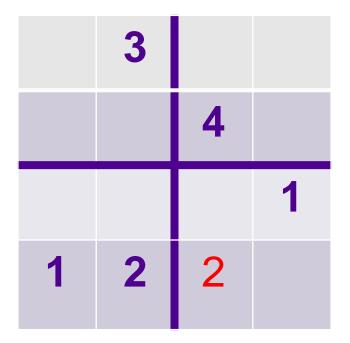
Mini Sudoku



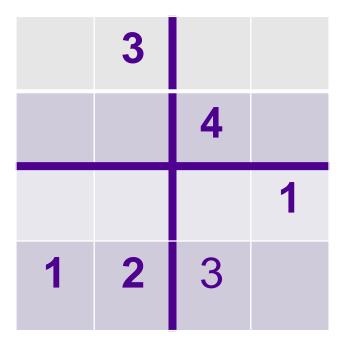




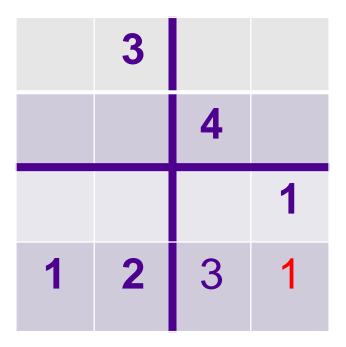




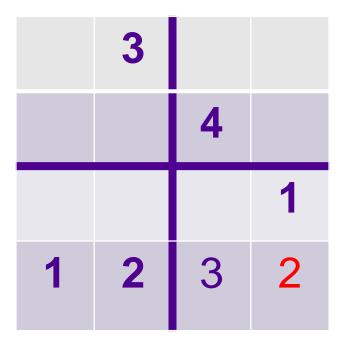




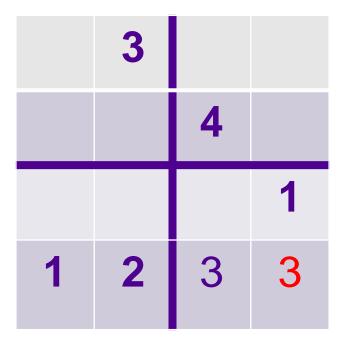




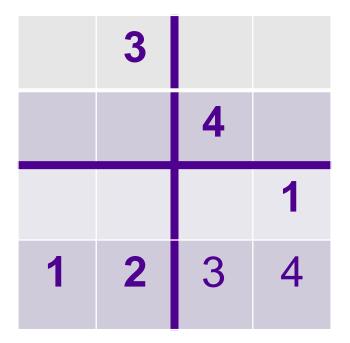




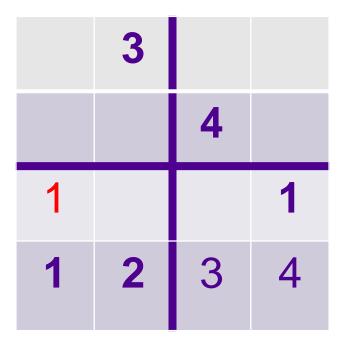




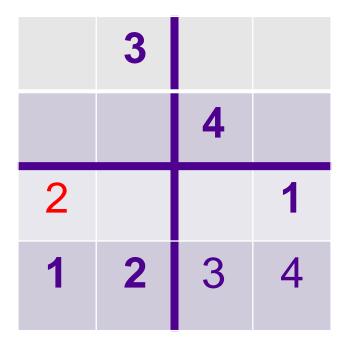




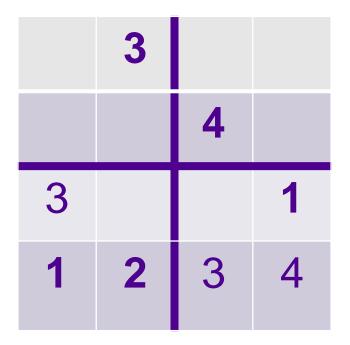




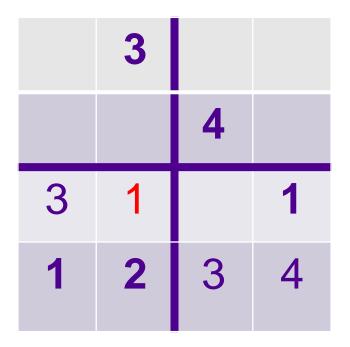




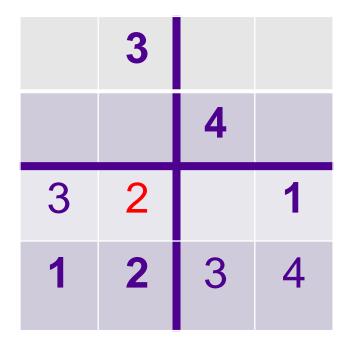




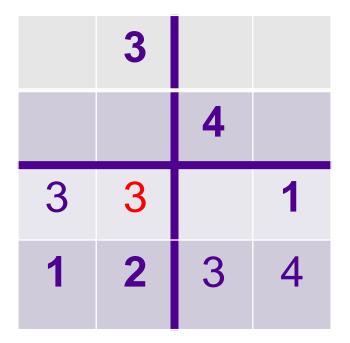




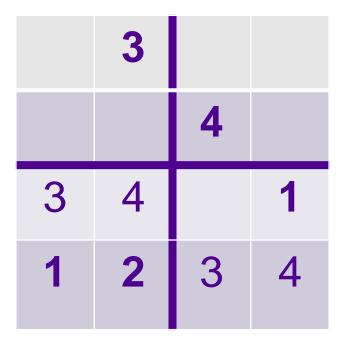




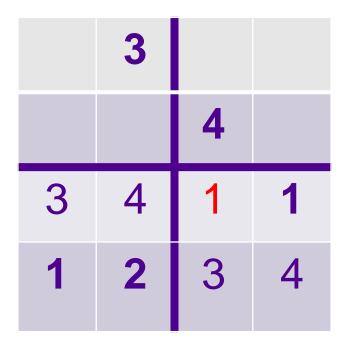




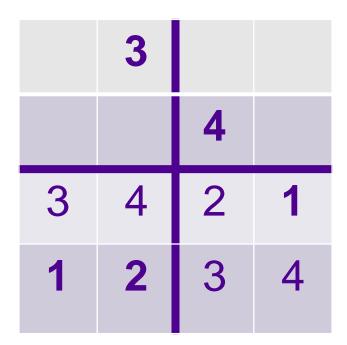




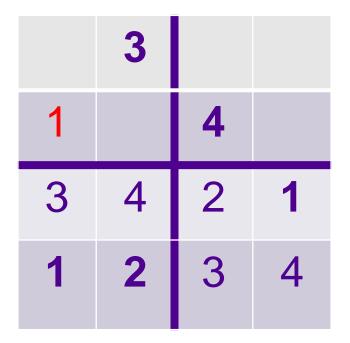




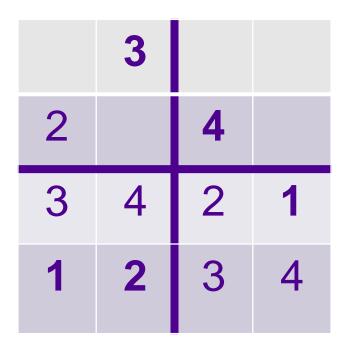




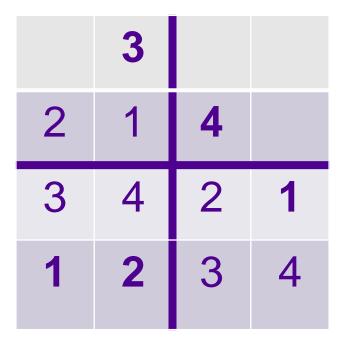




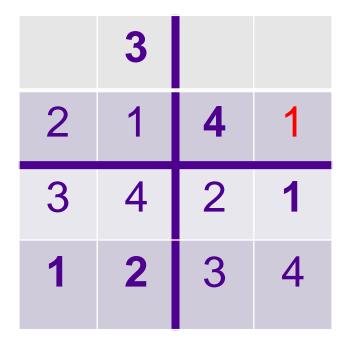




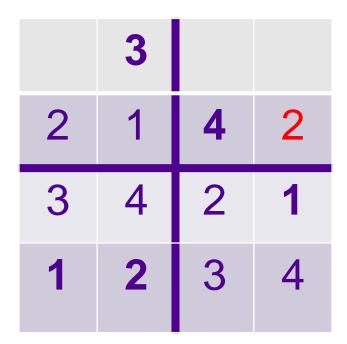














1	3		
2	1	4	3
3	4	2	1
1	2	3	4



2	3		
2	1	4	3
3	4	2	1
1	2	3	4



3	3		
2	1	4	3
3	4	2	1
1	2	3	4



4	3		
2	1	4	3
3	4	2	1
1	2	3	4



4	3	1	
2	1	4	3
3	4	2	1
1	2	3	4



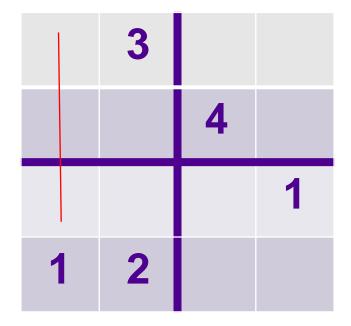
4	3	1	1
2	1	4	3
3	4	2	1
1	2	3	4



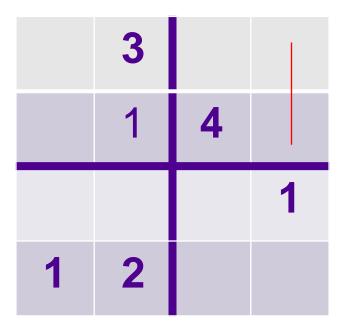
4	3	1	2
2	1	4	3
3	4	2	1
1	2	3	4



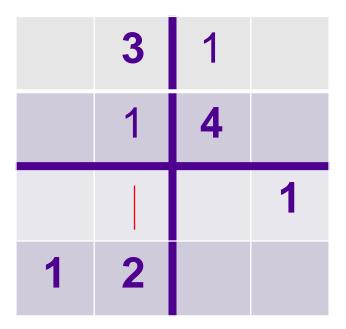
How'd we do it



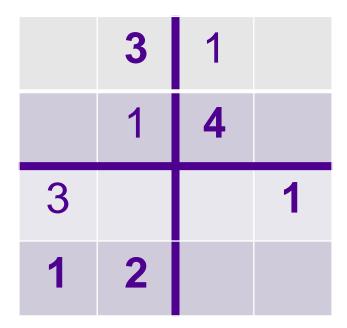




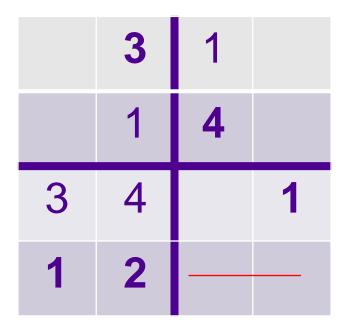




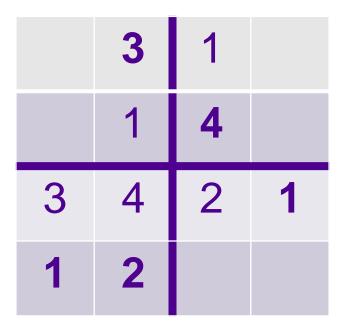




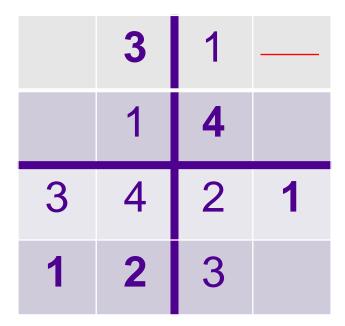




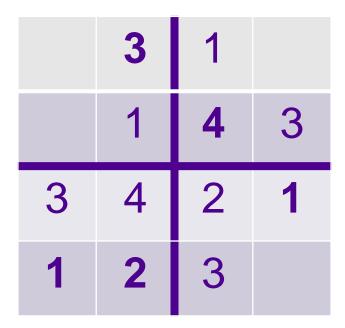














	3	1	
	1	4	3
3	4	2	1
1	2	3	4



	3	1	2
	1	4	3
3	1	2	4
3	4	2	1



	3	1	2
2	1	4	3
3	4	2	1
1	2	3	4



4	3	1	2
2	1	4	3
3	4	2	1
1	2	3	4



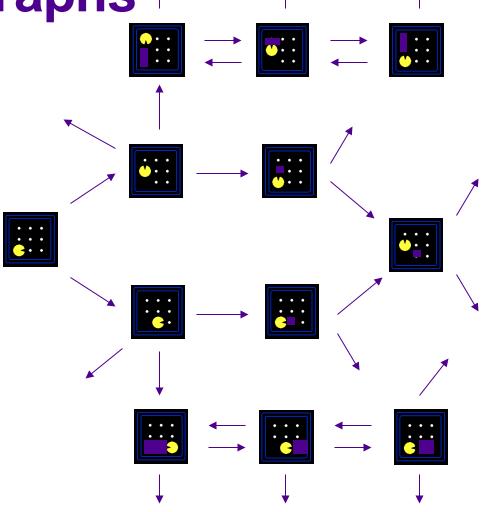
Search space

- In the worst case scenario, we have to look at all possible Sudoku configurations before finding the one and only solution
- Such a enumerative search is not usually feasible due to its inefficiency
- In two-player turn-taking board games the solution search happens in a game tree: White takes the first move in chess from among those available to it according to the rules of the game





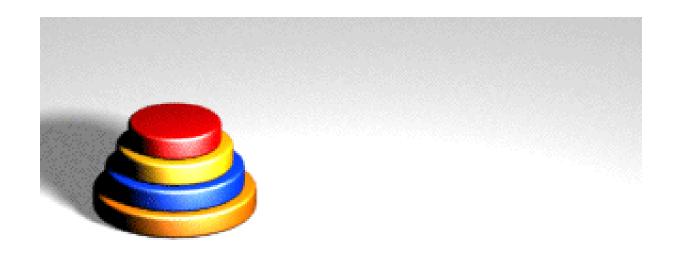
State space graphs †





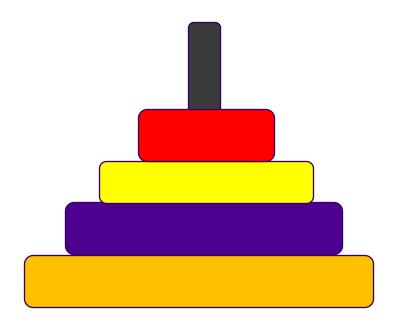
Tower(s) of Hanoi

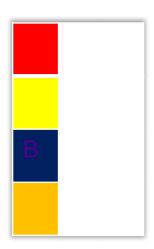
- The objective of the puzzle is to move the entire stack to another rod, obeying the following simple rules:
- 1. Only one disk can be moved at a time.
- 2. Each move consists of taking the upper disk from one of the stacks and placing it on top of another stack or on an empty rod.
- 3. No larger disk may be placed on top of a smaller disk.
- With 3 disks, the puzzle can be solved in 7 moves. The minimal number of moves required to solve a Tower of Hanoi puzzle is $2^n 1$, where n is the number of disks.



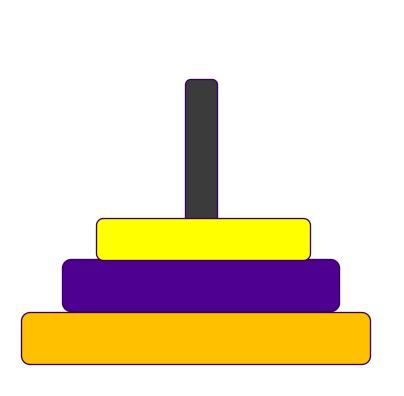


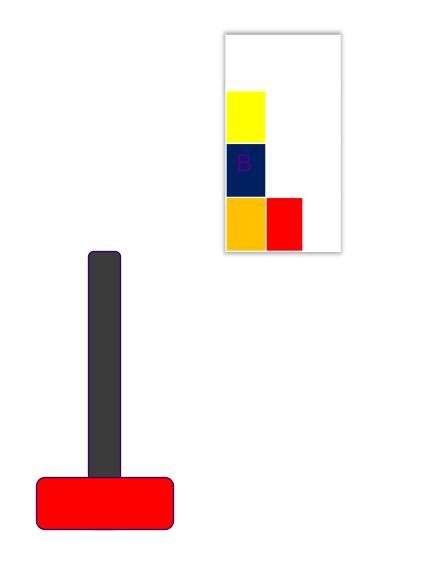
States & Legal Actions



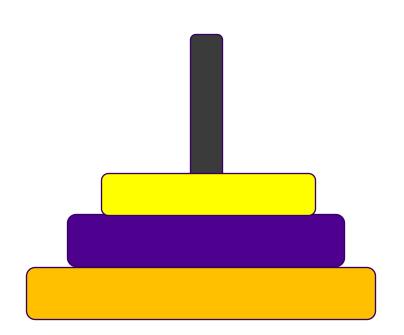


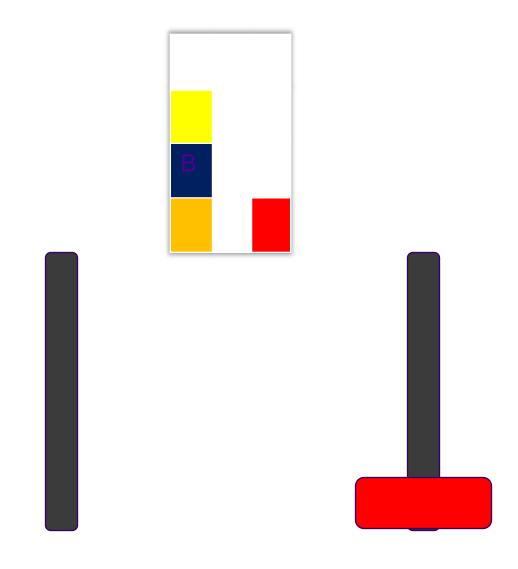




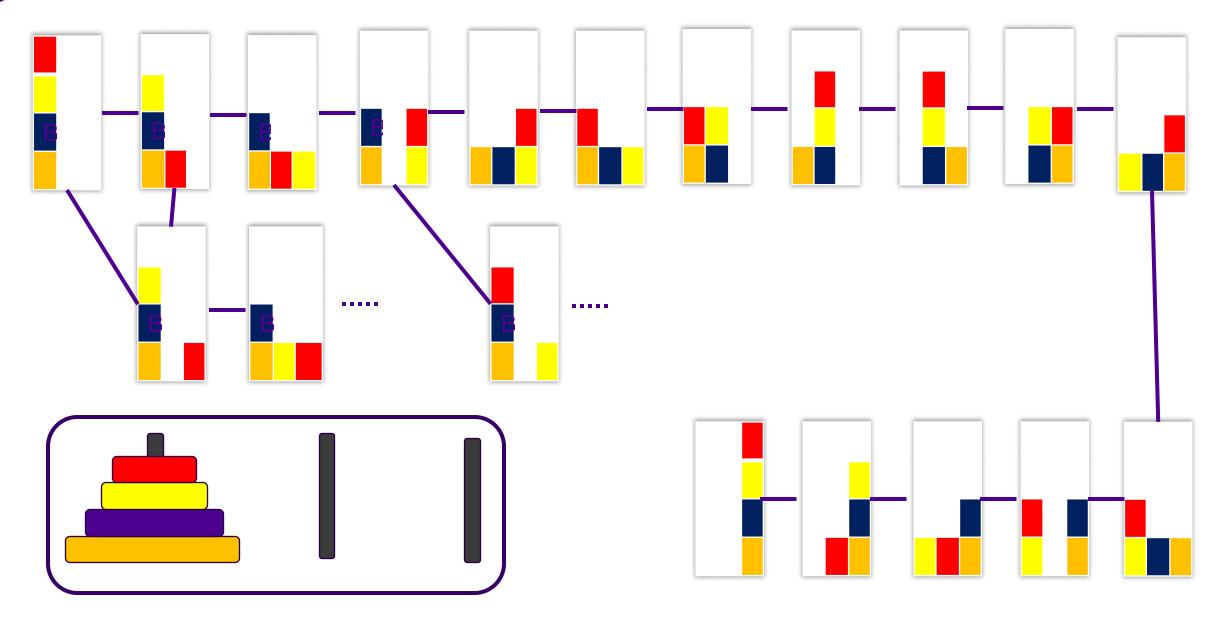








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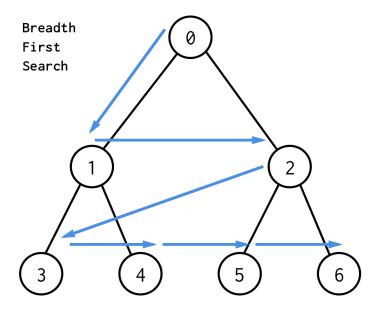


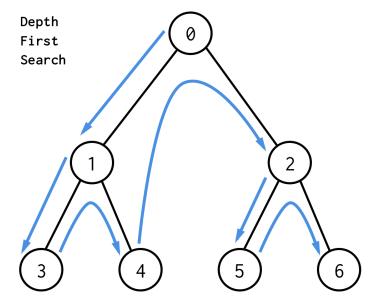


- Search may end up looping or oscillating in the search space never proceed to goal
- Maintain visited state information to avoid repeating states
- Turn the search to a tree rather than a graph
- Need a systematic way of traversing the search space
- A complete search algorithm is guaranteed to find a solution, when one exists
- An optimal search strategy finds the optimal solution



Blind search strategies







BFS & DFS

- Blind since there is no indication where the goal state might reside
- Hence, in the worst case all nodes need to be examined
- The strategies differ in the number of nodes that they must retain

BFS

- Has to hold on to all nodes it has seen until the goal state is found
- The positive side is that it guarantees finding a goal if one exists: complete
- Also guarantees the optimality of the goal (if step costs are equal)

DFS

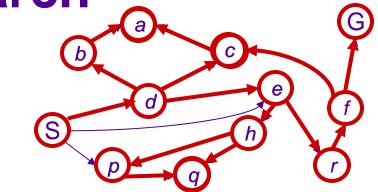
- May discard paths that have already been examined
- May lose itself to an infinite search path
- Might miss the shallowest goal state

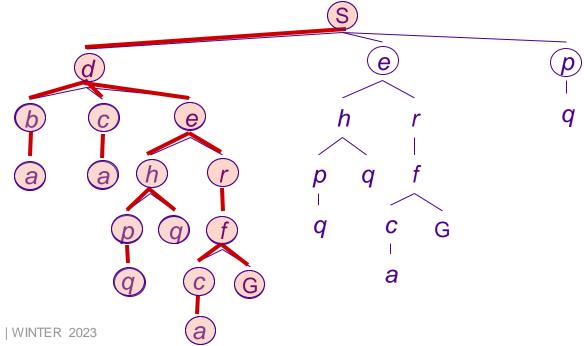


Depth-First Search

Strategy: expand a deepest node first

Implementation: Fringe is a LIFO stack





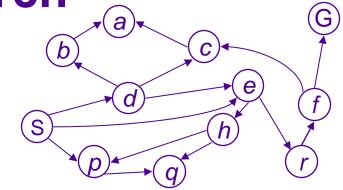


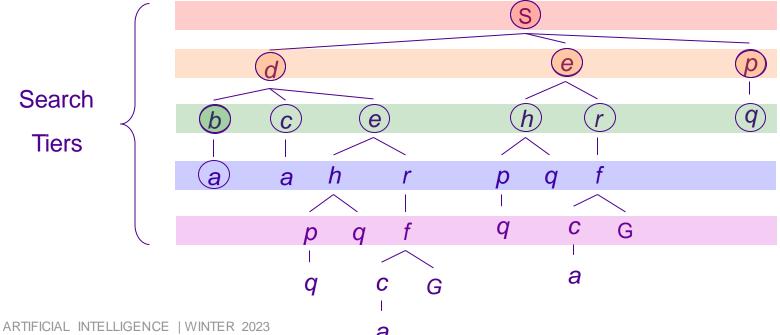
Breadth-First Search

Strategy: expand a shallowest node first

Implementation: Fringe

is a FIFO queue







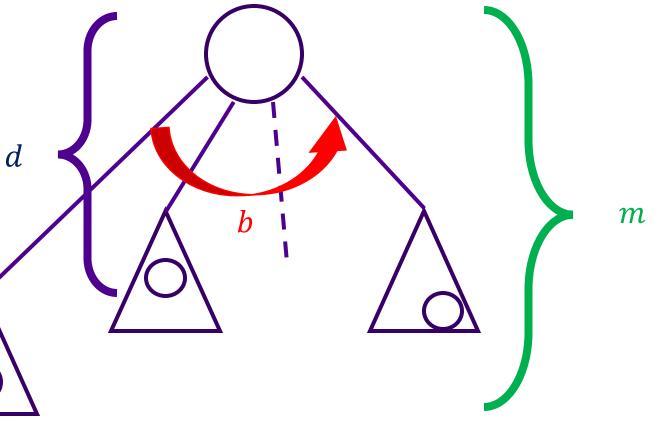
Search tree parameters

• The parameters:

• b branching factor (max or avg.)

• d depth of the (shallowest) goal

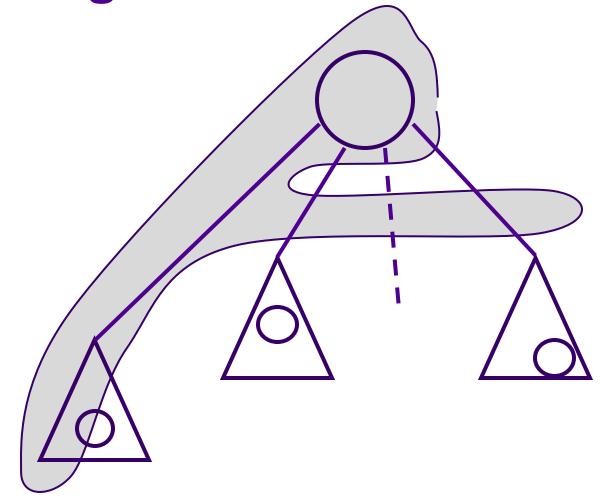
• m max length of any path in the state space, could be ∞



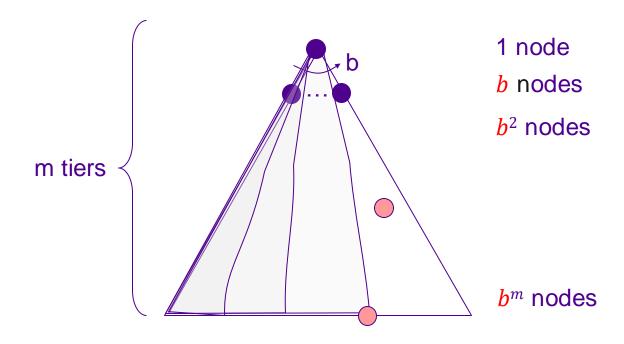


Brief analysis of the strategies: DFS

- Very modest memory requirements
 - Stores only one root-leaf path, along with the remaining unexpanded sibling nodes for each node on the path
 - DFS requires storage of only **bm** nodes
- The only goal node may be in the branch of the tree that is examined the last
 - In the worst case also DFS takes an exponential time: $O(b^m)$
- At its worst $m \gg d$,
 - the time taken by DFS may be much more than that of BFS





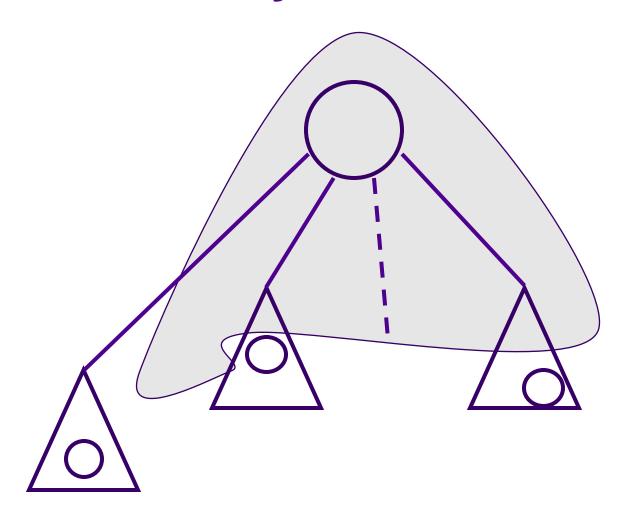








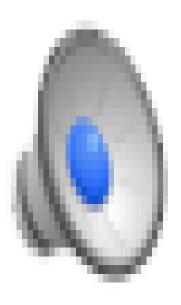
BFS analysis



- In the worst case BFS expand all but the last node at level *d*
- Every node that is generated must remain in memory, because it may belong to the solution path
- Thus the worst-case time and space complexities of BFS are

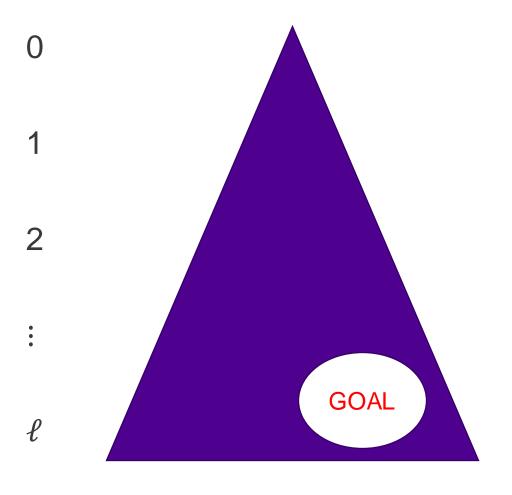
$$b + b^2 + b^3 + \dots + b^d = O(b^d)$$







Depth-limited search



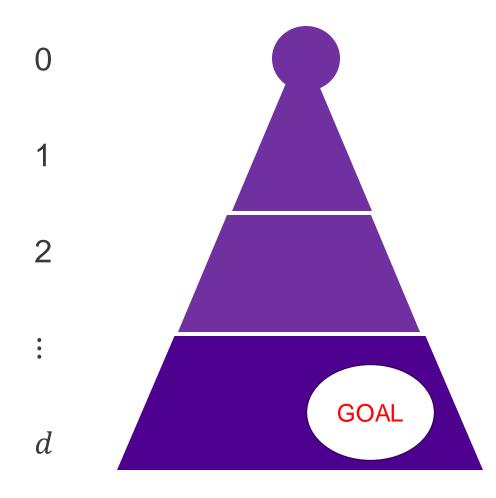
- To avoid infinite branches, limit the search to depth ℓ and use DFS
- Guaranteed to find the a goal, if the shallowest goal is at

 $\mathsf{depth} \leq \ell$



Iterative Deepening

- To combine the good sides of BFS
 & DFS use Iterative Deepening:
- Let the parameter value grow gradually $\ell=0,1,2,...$ until the goal is found
- If a goal node exist, it will eventually be found
- Furthermore, we will find the optimal (shallowest) goal

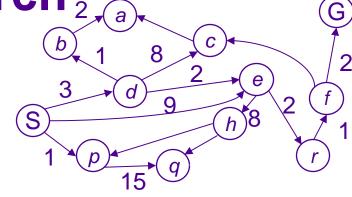


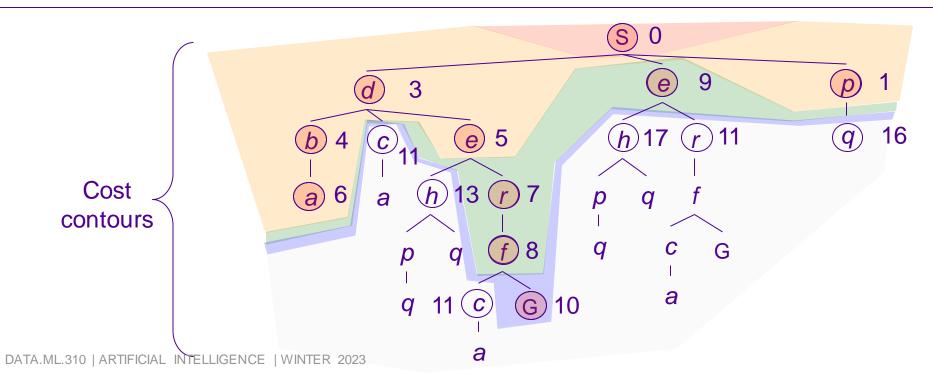


Uniform Cost Search 2,(a),

Strategy: expand a cheapest node first:

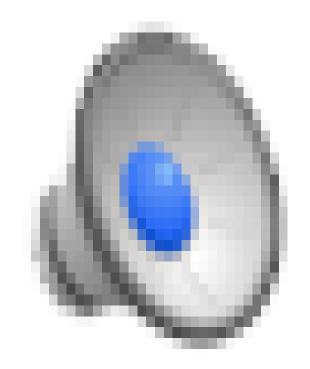
Fringe is a priority queue (priority: cumulative cost)







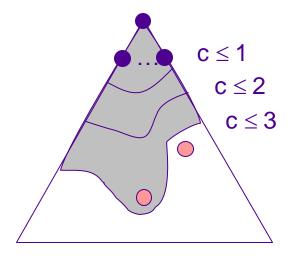
Demo Contours UCS Pacman Small Maze

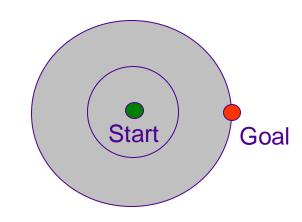




Uniform Cost Issues

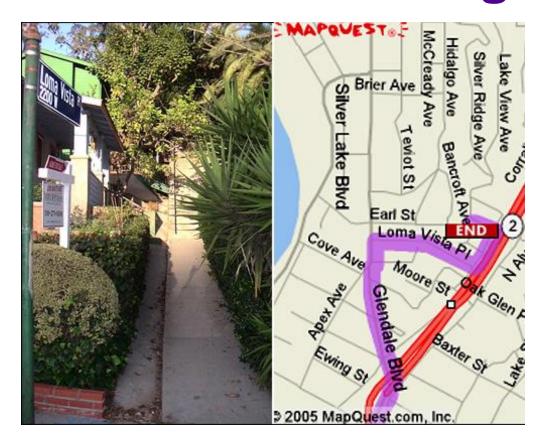
- Remember: UCS explores increasing cost contours
- The good: UCS is complete and optimal!
- The bad:
 - Explores options in every "direction"
 - No information about goal location
- We'll fix that soon!







Search Gone Wrong

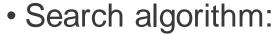






Recap: Search

- Search problem:
 - States (configurations of the world)
 - Actions and costs
 - Successor function (world dynamics)
 - Start state and goal test
- Search tree:
 - Nodes: represent plans for reaching states
 - Plans have costs (sum of action costs)



- Systematically builds a search tree
- Chooses an ordering of the fringe (unexplored nodes)
- Optimal: finds least-cost plans

