

Artificial Intelligence

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Learning Outcomes of Search Agents

- Informed search
 - Also called blind search
 - The strategies have no additional information about states beyond that provided in the problem definition
 - Depth-first search
 - Breadth-first search
 - Uniform-cost search
- Informed search
 - Also called heuristic search
 - The strategies know whether one non-goal node is “more promising” than another
 - Greedy best-first search
 - A* search
- Reference reading
 - Chapter 3

Admissible Heuristic of A* Search

- The tree-search version of A* is optimal if $h(n)$ is admissible
 - Evaluation function: $f(n) = g(n) + h(n)$
 - $g(n)$: the cost so far to reach the node n
 - $h(n)$: the estimated cost to goal from the node n
 - $f(n)$: the estimated total cost of path through the node n to goal
- An admissible heuristic never overestimates the cost to reach the goal
 - A heuristic $h(n)$ is admissible if for every node n , $h(n) \leq h^*(n)$, where $h^*(n)$ is the true cost to reach the goal state from n .

Domination of Heuristic Functions

- If $h_2(n) \geq h_1(n)$ for all n (both admissible)
 - then h_2 dominates h_1 , which indicates that h_2 is better for search
- Domination translates directly into efficiency
 - A* using h_2 will never expand more nodes than A* using h_1 .

7	2	4
5		6
8	3	1

Start State

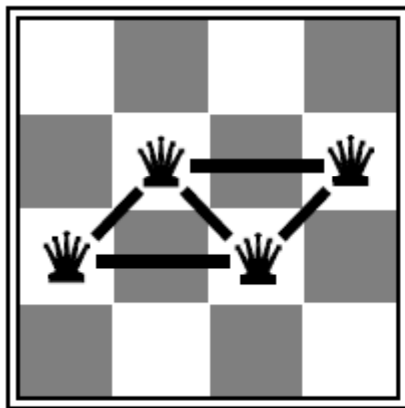
	1	2
3	4	5
6	7	8

Goal State

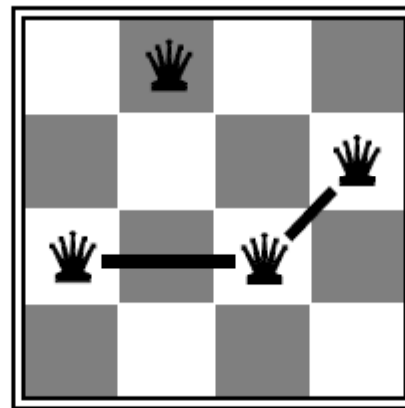
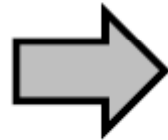
- $h_1(S) = 8$
- $A^*(h_1) = 39,135$ nodes
- $h_2(S) = 18$
- $A^*(h_2) = 1,641$ nodes

N-queens Problem

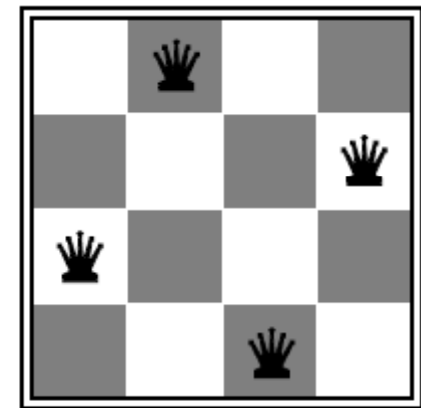
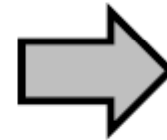
- Put n queens on an $n \times n$ board with no two queens on the same row, column, or diagonal
- Move a queen to reduce number of conflicts



$h = 5$



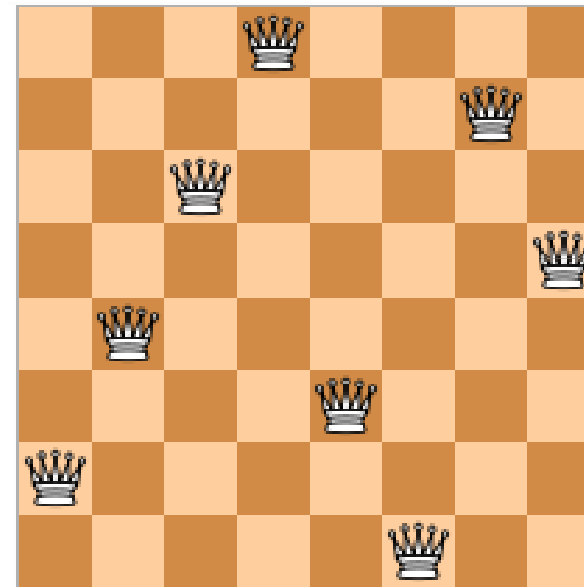
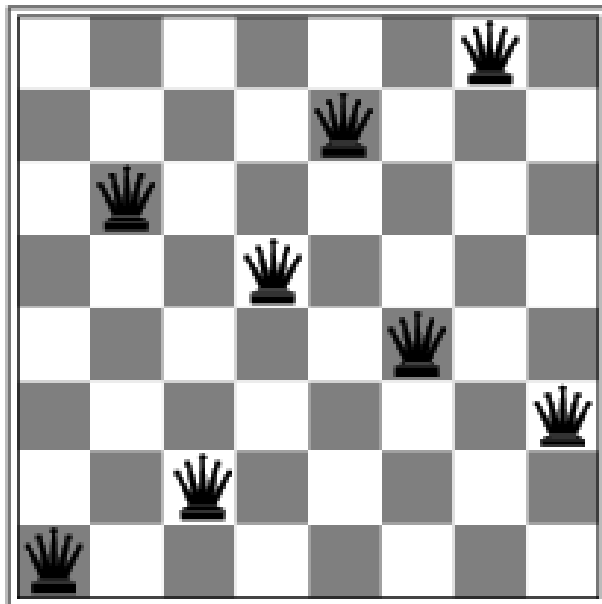
$h = 2$



$h = 0$

8-Queens Problem

- Place 8-queens on a chessboard such that no queen attacks any other
 - A queen attacks any piece in the same row, column, or diagonal

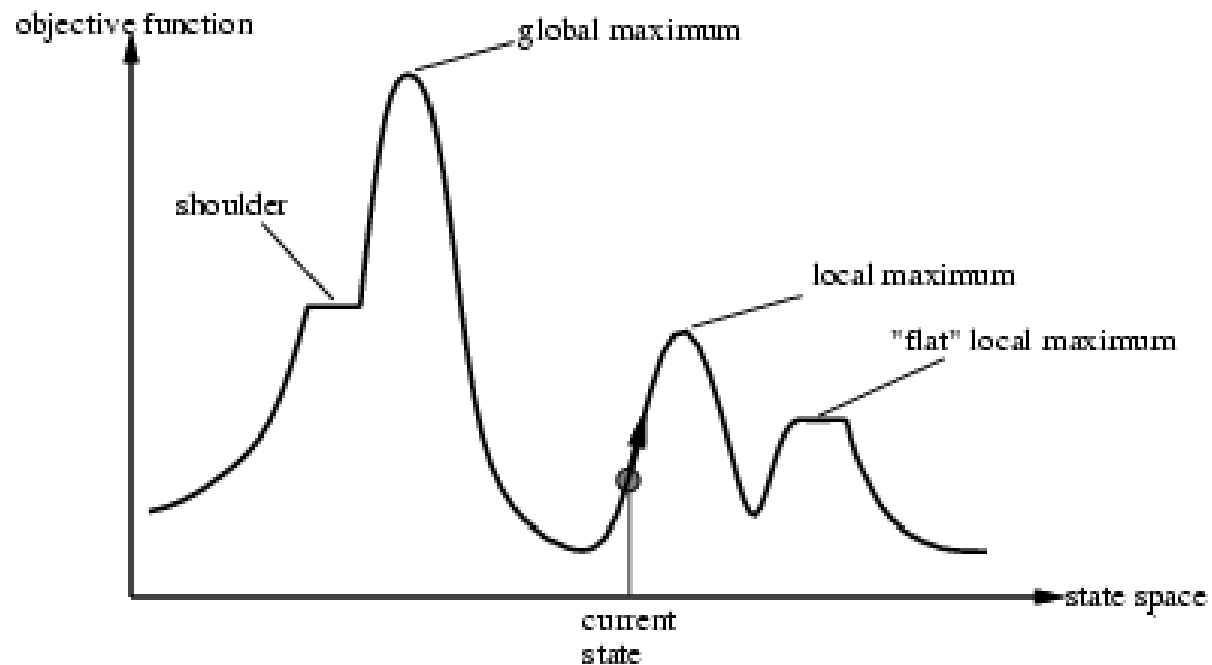


Local Search

- Problem: Romania
 - The search algorithms explore search space systematically
 - When a goal is found, the path to the goal also constitutes a solution to the problem
- Problem: N-Queens
 - The path to the goal is irrelevant
- Local search
 - Operate using a single current node
 - Generally move only to neighbors of that node
 - Useful to find the best state according to an objective function

State-space Landscape

- Location: defined by the state
- Evaluation: defined by the value of the objective function



Hill-climbing Search

- Depending on initial state, can get stuck in local maxima

18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	12	18	13	15	12	14	14
15	14	14	♙	13	16	13	16
♙	14	17	15	♙	14	16	16
17	♙	16	18	15	♙	15	♙
18	14	♙	15	15	14	♙	16
14	14	13	17	12	14	12	18

18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	12	18	13	15	12	14	14
15	14	14	♙	13	16	13	16
♙	14	17	15	♙	14	16	16
17	♙	16	18	15	♙	15	♙
18	14	♙	15	15	14		16
14	14	13	17	12	14	♙	18

Arthur Samuel

- Professor at Stanford University
- Proposed hill-climbing search in 1963
 - Made the first computer checker program on IBM's first commercial
- Applications of hill-climbing search
 - Network flow problem
 - VLSI(Very Large Scale Integration) design



Simulated Annealing Search

- Hill-climbing
 - Never makes downhill moves
 - Get stuck on a local maximum
- Random walk
 - Moving to a successor chosen uniformly at random
 - Is complete but extremely inefficient
- Simulated annealing
 - Escape local maxima by allowing some "bad" moves but gradually decrease their frequency
 - In metallurgy, annealing is the process used to temper or harden metals and glass by heating them to a high temperature and then gradually cooling them
 - The algorithm is quite similar to hill-climbing
 - Instead of picking the best move, it picks a random move
 - If the move improves the situation, it is always accepted

Simulated Annealing Search

18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	12	18	13	15	12	14	14
15	14	14	♔	13	16	13	16
♔	14	17	15	♔	14	16	16
17	♔	16	18	15	♔	15	♔
18	14	♔	15	15	14	♔	16
14	14	13	17	12	14	12	18

18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	12	18	13	15	12	14	14
15	14	14	♔	13	16	13	16
♔	14	17	15	♔	14	16	16
17	♔	16	18	15	♔	15	♔
18	14	♔	15	15	14		16
14	14	13	17	12	14	♔	18

18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	12	18	13	15	12	14	14
15	14	14	♔	13	16	13	16
	14	17	15	♔	14	16	16
17	♔	16	18	15	♔	15	♔
18	14	♔	15	15	14	♔	16
♔	14	13	17	12	14	12	18

18	12	14	13	13	12	14	14
14	16	13	15	12	14	12	16
14	♔	18	13	15	12	14	14
15	14	14	♔	13	16	13	16
	14	17	15	♔	14	16	16
17		16	18	15	♔	15	♔
18	14	♔	15	15	14	♔	16
♔	14	13	17	12	14	12	18

							♔
			♔				
	♔						
			♔				
					♔		
							♔
♔							

Simulated Annealing Search Proposed in 1983

- Scott Kirkpatrick
 - Ph.D. in Physics from Harvard University
- C. Daniel Gelatt, JR
 - PhD in Physics from Harvard University
- Mario P. Vecchi
 - PhD in Electrical Engineering from Massachusetts Institute of Technology

