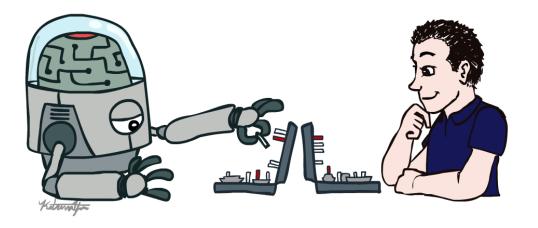
#### Lecture 04

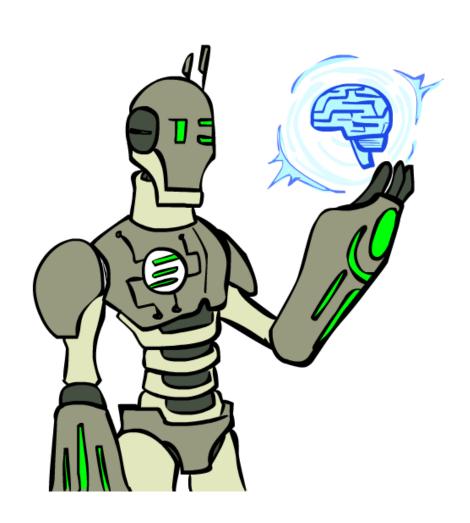
#### **Ashis Kumar Chanda**

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### Today



- Local search
- Hill climbing problem
- 8 Queen problem
- Variation of hill climbing problem

#### Another Approach: Local Search

- Now we'll look at search where the path is irrelevant;
  the goal state itself is the solution
  - State space : set of "complete" configurations
  - o Goal: some configuration satisfying.
  - o Example: 8 queen problem, integrated circuit design.

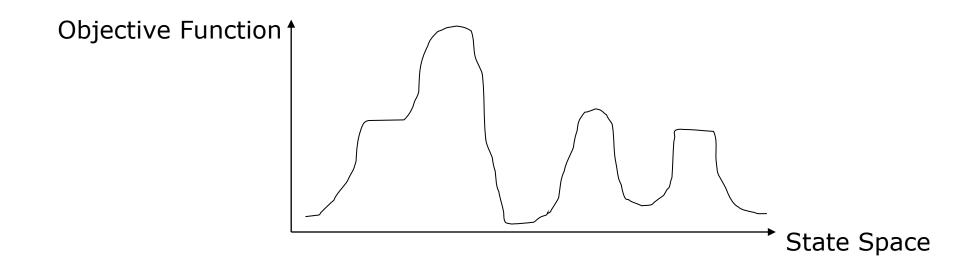
#### Another Approach: Local Search

#### Local search algorithm

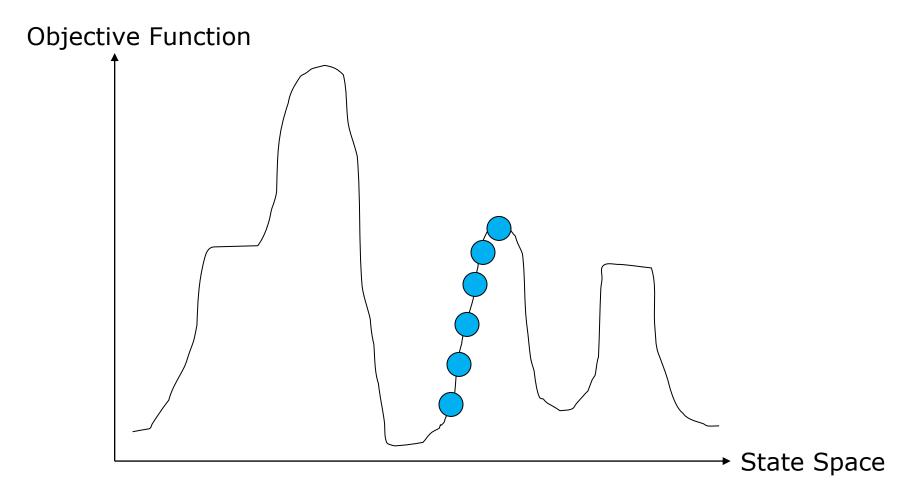
- Searching from a start state to neighboring states, without keeping track of the paths.
- Keep a single "current" state, try to improve it in next state.
- Might never explore a portion of the search space where a solution actually resides.

#### Local Search Example

- o Problem: Hill climbing search
- Each point (state) in the landscape has an "elevation" defined by the objective function.
- The aim is to find the highest peak.



# Hill Climbing Problems



If elevation corresponding to cost, then the aim is to find the lowest valley – a global minimum. We call it gradient descent.

### Exploring the Landscape

- Local Maxima: peaks that aren't the highest point in the space
- Plateaus: the space has a broad flat region that gives the search algorithm no direction (random walk).
- Ridges: flat like a plateau, but with drop-offs to the sides; steps to the North, East, South and West may go down. Ridges result in a sequence of local maxima.

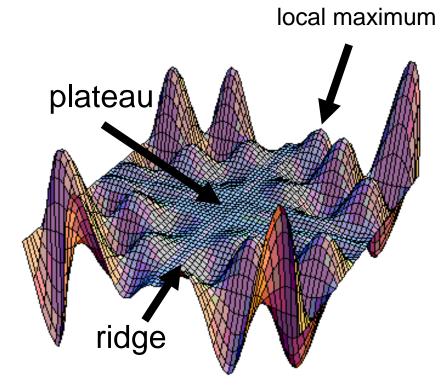
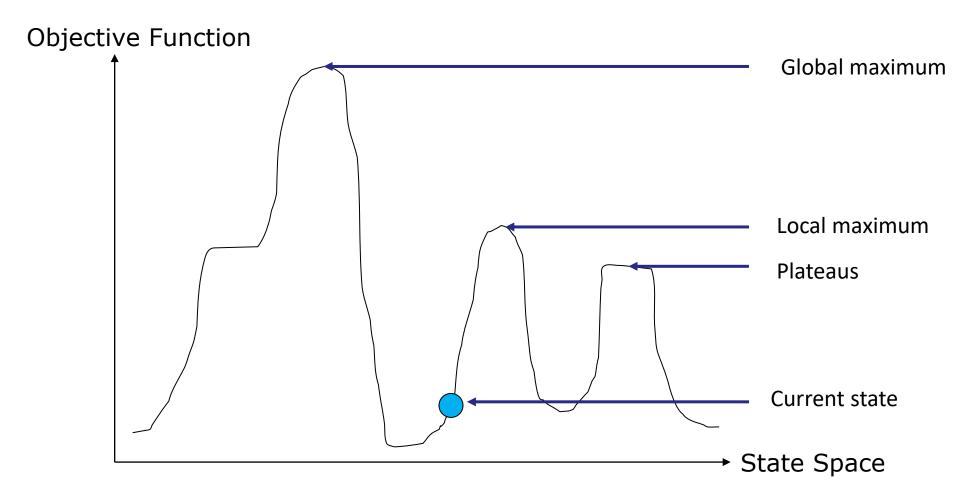
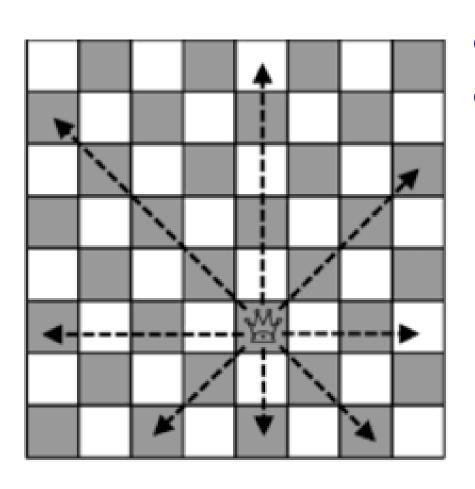


Image from: http://classes.yale.edu/fractals/CA/GA/Fitness/Fitness.html

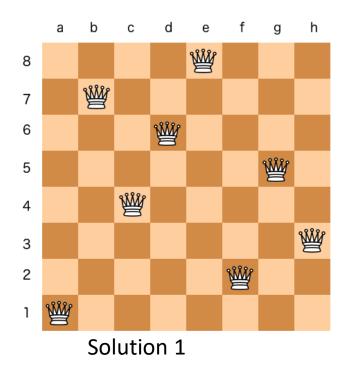
# Hill Climbing Problems





- 8 Queen problem
- A queen attacks any piece in the same row, column, or diagonal.

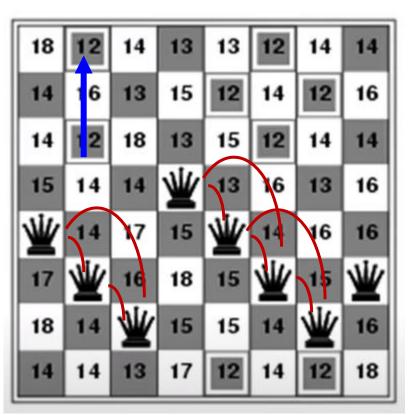
- Problem: put 8 queens on an 8 × 8 board with no two queens on the same row, column, or diagonal.
- o path is irrelevant, just need to reach goal.





- The heuristic cost function h is the number of pairs of queens that are attacking each other.
- This will be zero only for solutions.

- Here, heuristic cost estimate h = 17
- There are 17 ways of queen attacks for the current board.
- The board shows the value of h for each possible successor obtained by moving a queen within its column.



## Hill-climbing variations

#### Random-restart hill-climbing

- o If at first you don't succeed, try, try again...
- o Tries to avoid getting stuck in local maxima.
- o It conducts a series of hill-climbing searches from randomly generated initial states, until a goal is found.

#### Stochastic hill-climbing

- Chooses at random from among the uphill moves
- The probability of selection can vary with the steepness of the uphill move.
- This usually converges more slowly.

#### Simulated annealing

o Generate a random move.

#### Simulated Annealing

- Imagine the task of getting a ping-pong ball into the deepest crevice in a bumpy surface.
- We can bounce the ball out of the local minimum.
- But not hard enough to dislodge it from global minimum.
- o The trick is:
  - o Start by shaking hard (i.e., at a high temperature)
  - o Then, gradually reduce the intensity of the shaking (i.e., lower temperature)

### Simulated Annealing

- Moves are selected at random
- If a move is an improvement, accept
- o Otherwise, accept with probability less than 1.
- Probability gets smaller as time passes and by the amount of "badness" of the move.

## Simulated Annealing

A "bad" move from A to B is accepted with a probability
 (f(B)-f(A)/T)

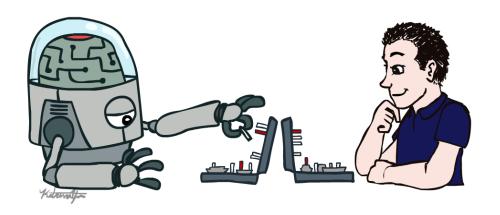
- The higher the temperature, the more likely it is that a bad move can be made.
- As T tends to zero, this probability tends to zero, and SA becomes more like hill climbing
- o If T is lowered slowly enough, SA is complete and admissible.

### Local search and optimization

- Local search= use single current state and move to neighboring states.
- Advantages:
  - Use very little memory
  - Find often reasonable solutions in large or infinite state spaces.

#### Next class?

- Adversarial search
- Game theory



# Thanks!