

CS 440: Introduction to Artificial Intelligence

Lecture 11

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Recap— Search

- ▶ Initial state
- ▶ Possible actions in each state
- ▶ Transition model:
Takes state and action and gives new state
- ▶ Goal test
Describes whether state is what you want
- ▶ Path cost
Says how easy or hard action sequence is

Recap— Search

Tips on formalizing search problems

- ▶ Represent states simply and precisely
- ▶ Make actions as simple and flexible as possible
- ▶ Count on search strategy to make good choices
- ▶ Analyze complexity and improve if necessary

A* Search

Practical variant of heuristic search

- ▶ Go-to-method for informed search
- ▶ Assumes you have a good way to measure progress

Basic idea: Explore the search node that looks most promising

- ▶ Measure progress by actual cost already incurred
- ▶ *plus* estimate of cost remaining

Thought Experiment

Compare

- ▶ One search problems where solution has depth d
- ▶ Two search problems where solution has depth $d/2$

Alternative Search Algorithm

Suppose you can construct a unique goal state.

Suppose you can apply actions backwards.

- ▶ Conduct two searches in parallel
- ▶ Search A starts from start, looks for goal forward
- ▶ Search B starts from goal, looks for start backward
- ▶ Solution happens when A and B meet

(Demo)

Limits of admissibility

- ▶ Lower bounds on solution can be weak
- ▶ A* explores too much of the search space
- ▶ May be better to give up optimality
- ▶ Settle for solution that's “good enough”

Strategy 1: Hill climbing

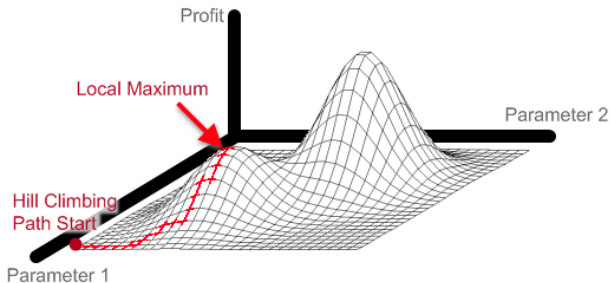
- ▶ Construct random costly solution, then loop:
- ▶ Construct children of current state
- ▶ If children are more costly return current
- ▶ Otherwise set current to best child and repeat

Assumptions

- ▶ Assumes it's easy to construct costly solution
- ▶ Assumes it's easy to modify solutions
- ▶ Also works with continuous spaces
- ▶ Sample or use analytic gradient

Hill Climbing weaknesses

The problem with hill climbing is that it gets stuck on "local-maxima"



from http://www.maxdama.com/2008_07_01_archive.html

Hill Climbing weaknesses

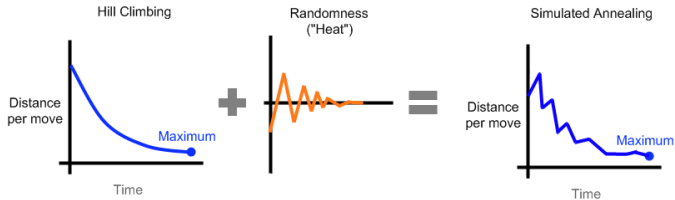
- ▶ Plateau problem
Can't tell locally which way to go
because all solutions look equally good
- ▶ Ridge problem
Higher than surrounding areas but has slope upward
Probes may not detect upward direction

see http://wwwic.ndsu.edu/juell/vp/cs724s00/hill_climbing/index.html

Strategy 2: Simulated annealing

- ▶ Include a little randomness to avoid getting stuck
- ▶ More randomness early
when you're likely to be further from good solution
- ▶ Mirrors process of metalwork
- ▶ Get metal in low-energy state by heating it up
and cooling it slowly

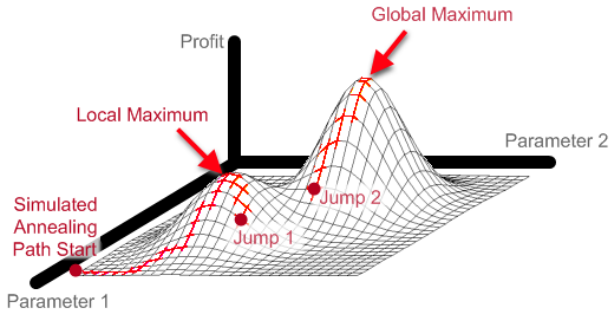
Simulated annealing



from http://www.maxdama.com/2008_07_01_archive.html

Simulated annealing

Simulated Annealing can escape local minima with chaotic jumps



from http://www.maxdama.com/2008_07_01_archive.html

Simulated annealing

- ▶ Construct random costly solution, then loop:
- ▶ Construct children of current state
- ▶ With probability t pick random child
- ▶ Otherwise pick least costly child
- ▶ Decrease t

Example search: Image synthesis

- ▶ Given raw image data
- ▶ Find representation in terms of polygons that approximates the image closely
- ▶ Demo at <http://alteredqualia.com/visualization/evolve/>

Strategy 3: Genetic Search

- ▶ Assumes it's easy to construct costly solution
- ▶ Assumes you can construct new solution from pieces of old ones
- ▶ Assumes there are many ways to change solution
- ▶ Assumes attributes of solutions are often largely independent

Genetic search

- ▶ Create population: set of costly solutions then loop:
- ▶ Rank population by fitness
- ▶ Create next generation
 - ▶ Pick two parents as a function of fitness
 - ▶ Construct child by mixing parents' features
 - ▶ Mutate child
- ▶ Repeat
- ▶ Return best solutions found after fixed time

Genetic Search Visualization

- ▶ <http://www.glauserweb.ch/gentore.htm>
- ▶ <http://www.rennard.org/alife/english/gavgb.html>