CS 289 **Evolutionary Computation**



Search, Optimization, Evolution

Problem Solving as Search

- Classic AI way of thinking (e.g McCarthy, Newell and Simon, 1950s)
- Wide domain can be cast this way (planning, theorem-proving, puzzles)
- Methods: optimal/complete search vs local search
- But No Free lunch Theorem
- Good approach when its hard to find a solution, but easy to check if a solution is any good
 - · What's are some examples?

Casting Problems as Search

$$\begin{split} &\textbf{F(solution)} = \textbf{objective function to maximize/minimize} \\ &\textbf{F(x)} = 2^{(-2)(x-0.1)(0.9)v2} * ((Sin(Spix))^6 \\ &\textbf{F(x,y,z)} = some complex but differentiable equation (classic optimization) \\ &\textbf{Optimization} \end{split}$$
(e.g. GPS localization: position relative to a set of reference nodes)

F(v1, v2, v3....vn) = no longer an equation!!! How well a neural network with these weights classifies some images

How well do these feature detector parameters capture objects of interest How well these 1991 stock allocations would have done in 2012 How well do these allocations maximize total agent utilities (class lottery!)

F(circuit/program) = no longer a simple vector/para How well does this circuit solve the required task? How well does a robot with this program navigate

F(fruitfly genome A) = how well does this individual survive compared to its buddies Evolution as a search process over a very complex representation....

Local Search

Local Search

- Many Variants (Hill climbing, Simulated Annealing, K-beam)
- Exploitation vs Exploration
- Representation, Representation, Representation!!
 Solution , Objective, Neighborhood function

Evolutionary Computation

Evolutionary Search is a cooperative local search method Based on the belief that

- = Evolution is a kind of optimization over a very complex landscape
- = And that the genotype-phenotype separation allows it to work across all organisms

Key Features (GA/GP)

- Population (Always have many candidate solutions)
 Representation (Genes/Programs; Fitness function)
- Variation (Nbr function = cross-over between solutions!)
- Selection (Choose best solutions across all new candidates)

What makes this class of methods different from others? When do people use Gas and GPs?

Some "Simple" Examples

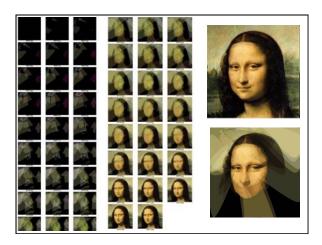
- N-queens
- EvoLISA (image compression)
- Evolving Cellular Automata (Crutchfield, Mitchell)
- · Next lectures!
 - Evolving Lego bridges (Jordan Pollack's Lab)
 - Evolving Robot Bodies and Brains (Pollack and Lipson)
 - Evolving Swarm behaviors.

N-Queens Example On a 8x8 board, place 8 queens, such that they can't How do we cast this as an evolutionary computation?

EvoLISA

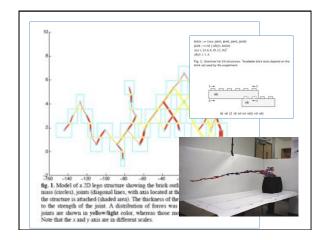
(Roger Asling, 2008)

- · Image Compression
 - Representation: 50 semi-transparent "polygons"
 - DNA = "vector of attributes"
 - Trying to find the best "DNA" to capture a given image
 - Not obvious how to find the right answer, but relatively easy to evaluate a given answer (how well does it match a given image)
 - Compression: Hard to compress, but easy to decompress!
 - Method: Genetic algorithm
 - Population | Variation | Selection
 - Question: Can we evolve a rep. of Mona Lisa?
 - Lets take a look



Evolving "Architectures" in Lego (Pablo Funes and Jordan Pollack, ~1999)

- Next Week's Topic
- EvoCAD
 - Using evolution as a tool for exploring the design space
 - Fitness function: what is the goal of the structure (or could be human directed evolution)
- Bridges, Cranes, and other structures
 - Specific goal (fitness of the structure)
 - Representation: how do I describe the current structure and generate new structures (variation)



Evolving Cellular Automata (Melanie Mitchell and John Crutchfield 1994)

- Evolving a CA to do "computation"
 - E.g. Density, Synchronization, Pattern formation
- Representation: simple and cool idea
 - "rule table" as a "binary string"





Implementation Example

- Problem: Classify Density (if black > ½, then CA turns all black)

 Binary Cellular Automata Ring, N=149, CA Rule radius r=3

 - Claim: no fixed radius rule exists that works for different lattice sizes, but even for a fixed lattice size it is hard to find a rule of low radius (not proven impossible)
- Representation: 128 bit string, population of 100 Candidates
- Variation: Cross-over and Mutation
- Selection: Fitness F100 (performance on 100 randomly generated CAs)
- How it works

 Start with 100 candidate solutions

 Test fitness F100 for all of them
- Test fitness F100 for all of them Pick top 20 (elite set) Generate another 80 by cross-over of randomly selected elite parents, do two mutations in each offspring (did not use the roulette wheel)

- Interesting Results

 Naive methods: Majority, Expand-block

 Sophisticated Method: "particle" method

 Also looked at other "computation" problems (like synchronization)

