

## 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

**F(solution) = objective function to maximize/minimize**

$$F(x) = 2^{(-2(x-0.1)/(0.9)^2 * ((\sin(5\pi x))^6$$

F(x,y,z) = some complex but differentiable equation (classic optimization)  
(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/parameter representation!!**

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?*

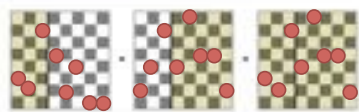
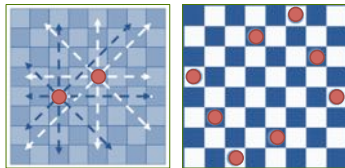
## Some “Simple” Examples

- 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 kill each other

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



