Introduction to Genetic Algorithms

Bioinspired Algorithms:

- Swarm intelligence
- Evolutionary computing
- Neural Networks
- Ant colony optimization

Evolutionary computing is a form of stochastic search, randomly searching until we find a solution. However, searching an entire solution space is intractable. ECs generally involve populations of solutions over a space of solutions.

Requirements:

- A population of individuals defined by some properties (Genes)
- A way to evaluate an organism's fitness (Fitness Function)
- A method for selecting fitter individuals as parents.
- A way to introduce variety into offspring
 - Copying errors (mutation)
 - Mixing genes from two parents

We must specify what we want evolution to find by defining our fitness function. This function assigns a fitness score to one individual based on its overall performance.

In order for evolutionary computation to be worthwhile it must be easier to define the fitness function than to compute the solution directly.

Evolutionary Computing Techniques:

- heuristic Not guaranteed to find the best solution
- stochastic Random
- Intended for hard optimization problems

EC techniques are examples of metaheuristic approaches.

In contrast to EC approaches, some metaheuristic approaches do not use a population of solutions. for example Hill Climbers:

- Population Set of individuals that the GA is acting on
- Individual A member of the population
- Genotype A string of symbols from some alphabet that encodes a particular solution (Sometimes called genomes or chromosomes)
- Phenotype The actual solution encoded by a genotype

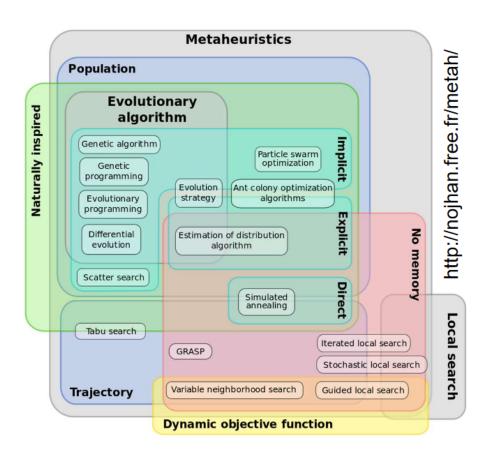


Figure 1: Untitled

- Genotype-phenotype mapping Analogous to development
- Genes (Loci) Chunks of genotype each taking a value the value is called the 'Allele'
- Fitness The value or quality of an individual solution phenotype
- Selection Choosing which current individuals get to reproduce
- Parents Individuals selected to reproduce
- Offspring Individuals that result from reproduction
- Crossover The recombination of the alleles from multiple parents
- Mutation Replacing offspring alleles with random alternatives
- Fitness Landscape An 'evolutionary search space' organising all possible solutions according to the neighbourhood relationships that result from the GA's Genotype structure and Genetic operators with landscape altitude set by the fitness of each solution.

Generating Initial Populations:

We start with a population of individuals with random genotypes. We can then apply a fitness score to each member of the population.

We can pick the next parents proportional to their fitness. We can use roulette base selection to compute the probability that they will be selected based on their fitness.

Alternatively, we can use rank-proportionate selection to choose parents based on how they rank in fitness.

- Sample k unique individuals at random from the population
- Pick a winner to be the parent
- To find N parents run N tournaments

Reproduction:

Once we have selected parents we can make a copy of their genes and put them into the next generation.

- $\bullet\,$ Mutation Each offspring gene is replaced by a random allele with probability m
- Crossover the offspring genotype inherits genes from multiple parents
- Elitism The fittest current individual is copied into the next generation perfectly

Crossover:

• One-Point - Pick a point along the genotype, the offspring inherits the first k genes from parent A and the last genes from parent B

- Two-Point Pick a range to get from one parent and outside the range genes are taken from the other parent
- Uniform Randomly make a decision for each gene to inherit from each parent.

The population size is generally kept constant:

- Generational Reproduction Continue selecting and reproducing parents until we have N new offspring
- Steady-State Reproduction As soon as we have generated a new offspring we pick a member of the current population and kill it replacing it with the new offspring.

Stopping Conditions:

- 1. The perfect solution is found Maximum fitness
- 2. We have run out of time
- 3. We think the GA has finished and things are as good as they are going to get.