

Evolutionary Algorithms [Wikipedia]

An evolutionary algorithm (EA) is a subset of evolutionary computation, a generic **population-based metaheuristic optimization algorithm**.

An EA uses mechanisms inspired by biological evolution, such as **reproduction, mutation, recombination, and selection**. Candidate solutions to the optimization problem play the role of **individuals** in a population, and **the fitness function determines the quality of the solutions** (see also loss function).

Evolution of the population then takes place after the **repeated application of the above operators**.

Genetic Algorithm

Components

Table 1.1: Simple Genetic Algorithm

Representation	Bit strings
Recombination	1-point crossover
Mutation	Bit flip
Parent Selection	Fitness proportional
Survival Selection	Generational

Chromosome: Representation

Individuals / chromosomes can be represented as a string of values.

Binary Representation Issues

Table 1.2: Chromosome *A*

Variable/Gene = 512									
1	0	0	0	0	0	0	0	0	0

Table 1.3: Chromosome *A*

Variable/Gene = 0									
0	0	0	0	0	0	0	0	0	0

Small, one bit, change can lead to drastic fitness changes. Solution: **use Gray Code**.

Gray Code

The reflected binary code (RBC), also known as reflected binary (RB) or Gray Code (named after Frank Gray – Bell Labs), is an **ordering of the binary numeral system such that two successive values differ in only one bit** (binary digit).

Table 1.4: Gray Code

Decimal	Binary	Gray	Decimal	Binary	Gray
0	0000	0000	7	0111	0100
1	0001	0001	8	1000	1100
2	0010	0011	9	1001	1101
3	0011	0010	10	1010	1111
4	0100	0110	11	1011	1110
5	0101	0111	12	1100	1010
6	0110	0101	13	1101	1011

Hamming Distance

The **Hamming distance** between **two equal-length strings** of symbols is the **number of positions at which the corresponding symbols are different**.

Gray code: subsequent numbers \rightarrow **Hamming distance of 1**

Integer Representation

- Some problems naturally map (8-queens) to integer representations
 - solution is an assignment variable = integer value
- Unrestricted: any value is permissible
- Restricted: only values from a certain set / domain
 - for example $\{0, 1, 2, 3\}$ for $\{\text{North, East, South, West}\}$
- Consider:
 - is there any relationship between values (e.g. 3 is more like 4 than 789 \rightarrow ordinal relationship) or not ($\{\text{North, East, South, West}\}$)
 - Choose your recombination / mutation strategy accordingly

Permutation Encoding

In permutation encoding, **every chromosome is a string of numbers that represent a position in a sequence**.

Permutation encoding is useful for **ordering problems**. For some types of crossover and mutation corrections must be made to leave the chromosome consistent (i.e. have valid sequence in it) for certain problems.

Genetic Algorithm: Other Selection Mechanisms

Elitism

Elitism (Elitist selection) is a strategy in genetic algorithms (in practice: evolutionary algorithms in general) where one or more most fit individuals (the elites) in every generation, are inserted into the next generation **without undergoing any change**.

This strategy usually speeds up the convergence of the algorithm.

Steady State Selection

In every generation, few chromosomes are selected (good - with high fitness) for creating a new offspring.

Then some (bad - with low fitness) chromosomes are removed and the new offspring is placed in their place.

The rest of population survives to new generation.

Fitness Proportional Selection: Issues

Fitness Proportional Selection approach has issues:

- outstanding individuals take over the entire population quickly; this is called **premature convergence**
- potential solutions
- when **fitness values are all very close together**, there is almost **no selection pressure** [= **selection is almost uniformly random**] → performance increases slowly
- Potential solutions:
 - windowing (slide and subtract some value based on history)
 - Goldberg's sigma scaling: $f'(x) = \max(f(x) - (f_{avg} - 2 \times \sigma_f), 0.0)$
 - **ranking**

Rank Based Selection

In rank selection, the **selection probability does not depend directly on the fitness**, but on **the fitness rank of an individual within the population**. The exact fitness values themselves do not have to be available, but only a sorting of individuals according to quality.

- Linear ranking:

$$P(R_i) = \frac{1}{n} \left(sp - (2sp - 2) \frac{i - 1}{n - 1} \right) \quad 1 \leq i \leq n, \quad 1 \leq sp \leq 2 \quad \text{with } P(R_i) \geq 0, \quad \sum_{i=1}^n P(R_i) = 1$$

sp – selection pressure (the degree to which the better individuals are favored: the higher the selection pressure, the more the better individuals are favored) which can take values between 1.0 (no selection pressure) and 2.0 (high selection pressure)

- Exponential ranking

Replace worst

- In this scheme, the worst k members of the population are selected for replacement
- This can lead to rapid improvements in the average population fitness, but can also cause premature convergence
 - it will focus on most fit individual
- Typically used with large populations

Age-Based Replacement

- Rather than look at fitness of the individual, pick the oldest (in iterations) first to replace
- A FIFO queue will be needed

Other Recombination Methods

“Cut and Crossfill”

1. Pick a random crossover point
2. Cut both parents in two segments after this position
3. Copy the first segment of Parent 1 into Child 1 and the first segment of Parent 2 into Child 2
4. Scan Parent 2 from left to right and fill the second segment of Child 1 with values from Parent 2, skipping those that are already contained in it
5. Do the same for Parent 1 and Child 2

Multiparent Recombination

- The idea: recombine genes of more than 2 parents
- Some strategies:
 - Allele frequency-based: ρ -sexual voting
 - segmentation and recombination of parents-based: the diagonal crossover
 - Based on numerical operations on real-values alleles

Other Recombination Options

- Integer representation:
 - same approaches as for binary
- Floating-point representation:
 - simple recombination
 - simple arithmetic recombination
 - whole arithmetic recombination
 - * Child 1: $\alpha \times x + (1 - \alpha) \times y$ and Child 2: $\alpha \times y + (1 - \alpha) \times x$

Integer representation

Same approaches as for binary

Floating-point representation:

- Simple recombination
- Simple arithmetic recombination
- Whole arithmetic recombination

Child 1: αx

Other Mutation Mechanisms

Scramble mutation

A subset of genes is chosen and their values randomly shuffles/scrambled.

Inversion mutation

A subset of genes, a substring is selected and inverted.

Interchange (order changing) mutation

Randomly select two positions of the chromosome and interchange the values.

Selected Problems

8-Queens Problem

Artificial Neuron (Perceptron)

A (single-layer) perceptron is a model of a biological neuron. It is made of the following components:

- inputs x_i - numerical values representing information
- weights w_i

Genetic Programming

Traditional Programming vs ...

Traditional Programming:

Input Data + Program (Rules) = Output

...:

Input Data + Output = Program (Rules)