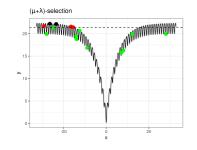
## **Optimization in Machine Learning**

# **Evolutionary Algorithms ES / Numerical Encodings**





#### Learning goals

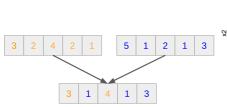
- Recombination
- Mutation
- A few simple examples

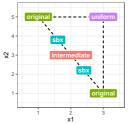
#### RECOMBINATION FOR NUMERIC

Options for recombination of two individuals  $\mathbf{x}, \tilde{\mathbf{x}} \in \mathbb{R}^d$ :

- **Uniform crossover**: Choose gene j of parent 1 with probability p and of parent 2 with probability 1 p
- Intermediate recombination: Offspring is created from mean of two parents:  $\frac{1}{2}(\mathbf{x} + \tilde{\mathbf{x}})$
- Simulated Binary Crossover (SBX): generate two offspring

$$\bar{\mathbf{x}} \pm \frac{1}{2}\beta(\tilde{\mathbf{x}} - \mathbf{x}), \ \bar{\mathbf{x}} = \frac{1}{2}(\mathbf{x} + \tilde{\mathbf{x}}), \ \beta \in [0, 1]$$
 uniformly at random







#### **MUTATION FOR NUMERIC**

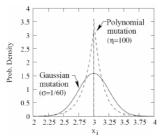
Mutation: Individuals get modified

**Example** for  $\mathbf{x} \in \mathbb{R}^d$ :

 Uniform mutation: Select random gene x<sub>j</sub> and replace it by uniformly distributed value (within feasible range).

• Gauss mutation:  $\mathbf{x} \pm \mathcal{N}(\mathbf{0}, \sigma \mathbf{I})$ 

• Polynomial mutation: Use a different distribution instead of normal distribution

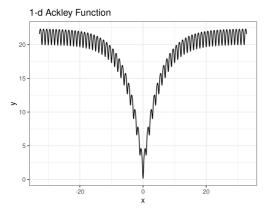


Source: K. Deb, D. Deb. Analysing mutation schemes for real-parameter genetic algorithms, 2014



#### **EXAMPLE OF AN EVOLUTIONARY ALGORITHM**

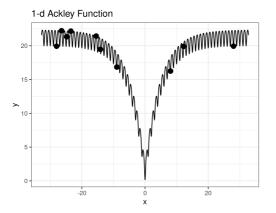
(Simple) EA on 1-dim Ackley function on [-30, 30]. Usually, for optimizing a function  $f: \mathbb{R}^d \to \mathbb{R}$ , individuals are encoded as real vectors  $\mathbf{x} \in \mathbb{R}^d$ .





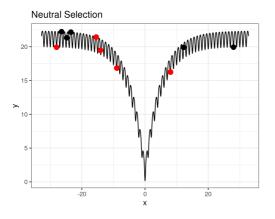
#### **EXAMPLE OF AN EVOLUTIONARY ALGORITHM**

Random initial population with size  $\mu=$  10



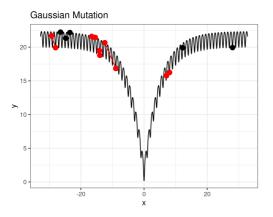


We choose  $\lambda=5$  offsprings by neutral selection (red individuals).



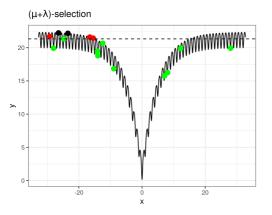


Use Gaussian mutation with  $\sigma = 2$ , but without recombination.



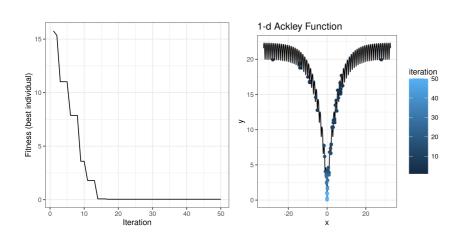


Use  $(\mu + \lambda)$  selection. Selected individuals are marked in green.





#### After 50 iterations:





#### **EXAMPLE 2: GRID OF BALLS**

Consider a grid in which n balls with random radius are placed.





**Aim:** Find the circle with the largest possible radius in the grid that does **not** intersect with the other existing circles.

- What is the fitness function?
- How is the population defined?

Implementation: https://juliambr.shinyapps.io/balls/

#### **EXAMPLE 2: GRID OF BALLS / 2**

In our example, the chromosome of an individual is the center of a circle, so the chromosomes are encoded as 2-dimensional real vectors  $\mathbf{x} = (x_1, x_2) \in \mathbb{R}^2$ .



The population  $P \subset \mathbb{R}^2$  is given as a set of circle centers.

The fitness function evaluates an individual  $\mathbf{x} \in P$  based on the distance to the nearest neighboring gray circle k.

$$f(\mathbf{x}) = \min_{k \in Grid} distance(k, \mathbf{x}),$$

where the distance is defined as 0 if a circle center is within the radius of a circle of the grid.

This function is to be maximized: we are looking for the largest circle that does not touch any of the gray circles.