



ITMO UNIVERSITY

Saint Petersburg, Russia

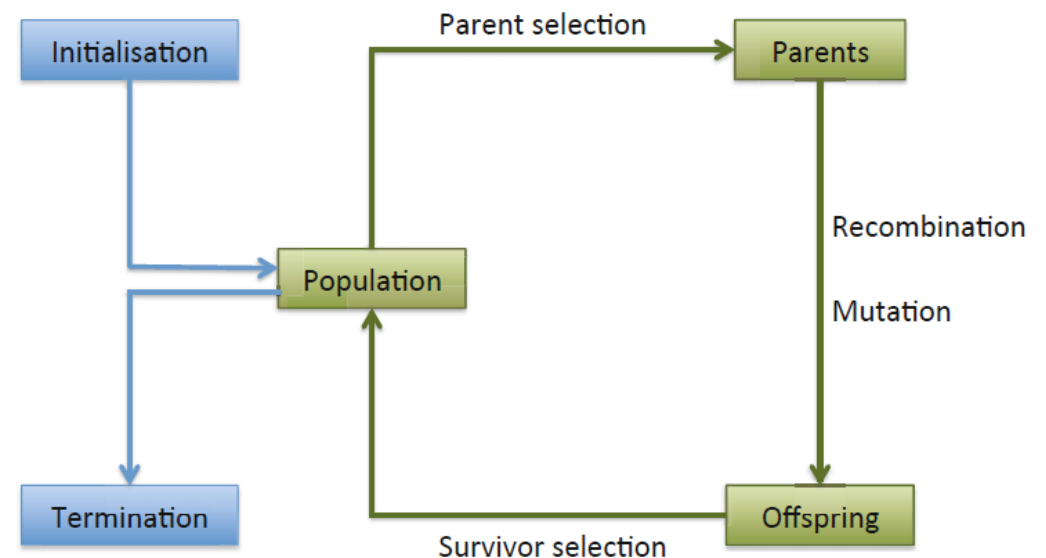
# Lecture 2: Components of Evolutionary Algorithms

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# What are Evolutionary Algorithms?

- ✓ Nature inspired stochastic optimization algorithms
- ✓ Population of individuals within problem's environment
- ✓ Evolution is based on two forces:
  - Variation (recombination, mutation)
  - Selection
- ✓ Family of generate and test methods



# Components of EAs

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- ✓ Representation of individuals
- ✓ Population of individuals
- ✓ Evaluation function (fitness function)
- ✓ Parent selection mechanism
- ✓ Variation operators (recombination, mutation)
- ✓ Survivor selection mechanism
- ✓ Terminate conditions

# Representation

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- ✓ Bridge between **original problem context** and **problem solving space**
- ✓ Objects forming possible solution within original problem context are referred to such terms as:
  - Phenotype
  - Candidate solution
  - Individual
- ✓ Encoded individuals within EA are called:
  - Genotype
  - Chromosome
- ✓ Elements of genotype can be called **genes**, alleles.

# Representation

- ✓ Mapping from phenotype to genotype is **encoding**
- ✓ Mapping from genotype to phenotype is **decoding**

Example: find integer  $x$ , that optimizes  $x^2$

Original optimization problem

EA solution space

|    |    |           |
|----|----|-----------|
| 13 | ←→ | 0 1 1 0 1 |
| 24 | ←→ | 1 1 0 0 0 |
| 8  | ←→ | 0 1 0 0 0 |

Encoding

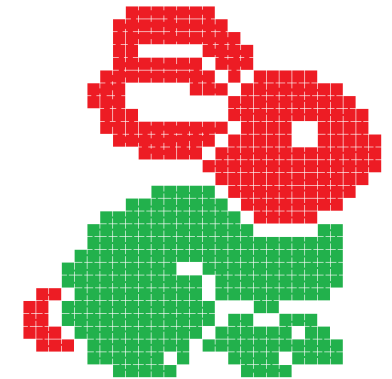


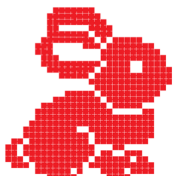
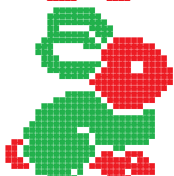
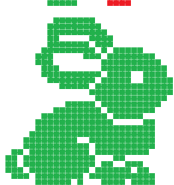
# Evaluation function (fitness function)

- ✓ Fitness function defines requirements to which population should adapt
- ✓ Fitness function is a procedure, that assigns a quality measure for solutions (genotypes)
  - For phenotypes it called **objective function**
- ✓ Naturally fitness function should be maximized
- ✓ However, there are no problems to reverse it to minimization

# Example

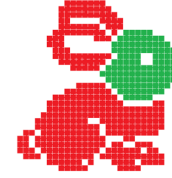
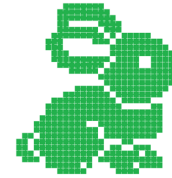
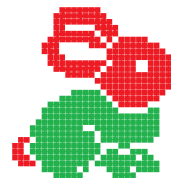
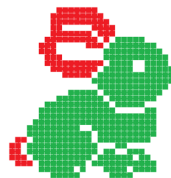
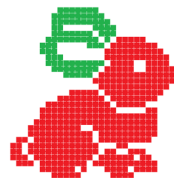
- ✓ Rabbits have 5 parts of body:
  - Ears, head, body, legs, tail
  - Each part can be green or red
  - More green rabbits are better than less green



|       | Phenotype   | Genotype   |   |   |   |   |   |   |
|-------|---|--|---|---|---|---|---|---|
| worst |   | <table><tr><td>0</td><td>0</td><td>0</td><td>0</td><td>0</td></tr></table> | 0 | 0 | 0 | 0 | 0 | 0 |
| 0     | 0   | 0  | 0 | 0 |   |   |   |   |
|       |  | <table><tr><td>1</td><td>0</td><td>1</td><td>0</td><td>0</td></tr></table> | 1 | 0 | 1 | 0 | 0 | 2 |
| 1     | 0   | 1  | 0 | 0 |   |   |   |   |
| best  |  | <table><tr><td>1</td><td>1</td><td>1</td><td>1</td><td>1</td></tr></table> | 1 | 1 | 1 | 1 | 1 | 5 |
| 1     | 1   | 1  | 1 | 1 |   |   |   |   |

# Population

- ✓ Population holds solutions
- ✓ Population is a multiset of genotypes (copies are allowed)
- ✓ May have special structure (grid)
- ✓ Population size is constant
- ✓ Diversity of population – number of different solutions within
- ✓ Initialization: the first population is seeded by random generated individuals





# Parent selection mechanism

- ✓ Parent selection mechanism aimed to select individuals from population, which will generate offspring (children)
- ✓ Parent selection is probabilistic
- ✓ Can be uniform
- ✓ Generally, less quality solution have less chances to be selected

# Variation operators

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- ✓ Variation creates new individuals from old and increases diversity of population
  - Variation operators are stochastic, but some problem specific rules can be used
- ✓ Mutation
  - Applied to one individual and deliver one new individual (mutant)
  - Mutation provides something new
- ✓ Recombination
  - Merges information from two (or more) individuals to create offspring

# Survivor selection mechanism

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- ✓ Mechanism, that selects, which solutions will be selected for the next generation of population
- ✓ Selection is stochastic and based on fitness estimation of individuals
- ✓ Selection saves the initial size of population

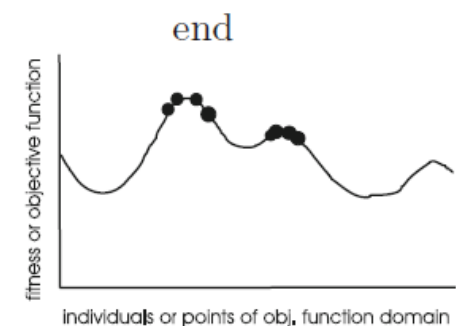
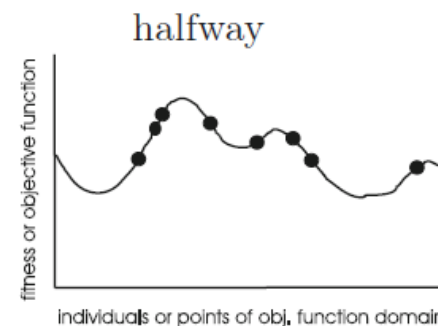
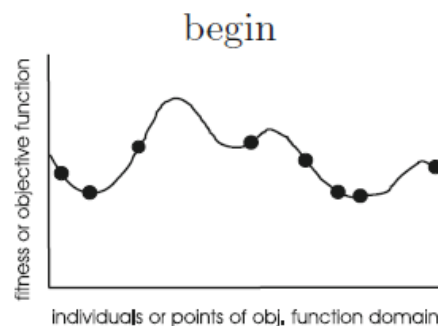
# Terminate conditions

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- ✓ Define rules and conditions, when an algorithm has to stop its evolving process
- ✓ Common terminate conditions:
  - The maximally allowed CPU time elapses.
  - The total number of fitness evaluations reaches a given limit.
  - The fitness improvement remains under a threshold value for a given period of time (i.e., for a number of generations or fitness evaluations).
  - The population diversity drops under a given threshold.

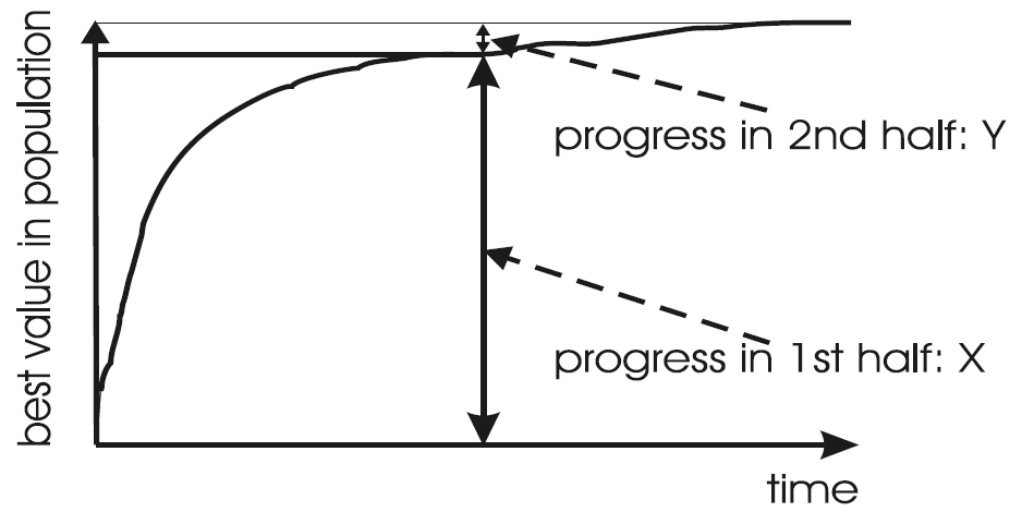
# Behavior of EAs

- ✓ Evolutionary process is a trade-off between **exploration** and **exploitation**
- ✓ Exploration – generation of something new, observation of untested regions of search space
- ✓ Exploitation – concentration on specific regions of good solutions, trying to improve them



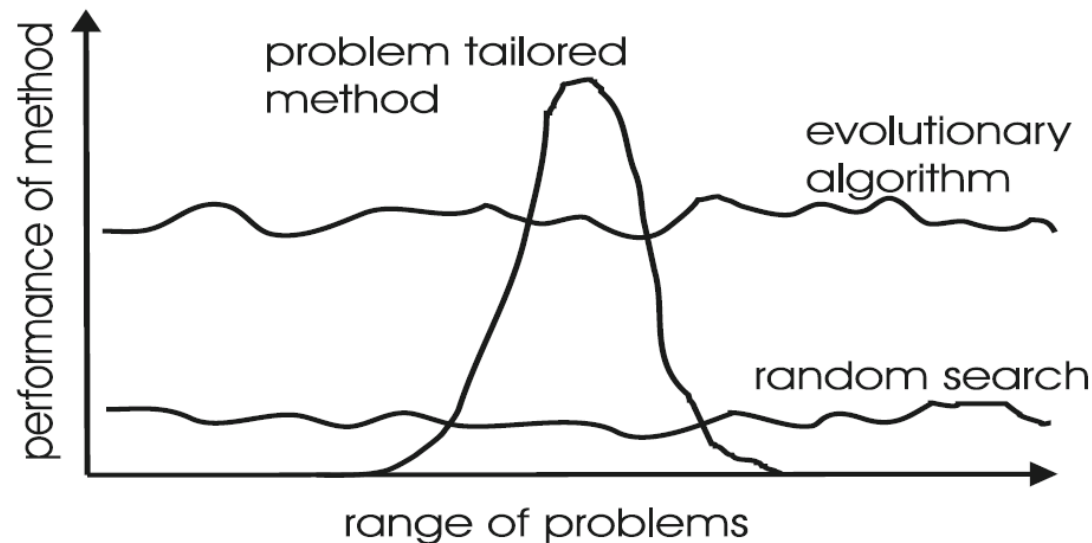
# Anytime behavior

- ✓ Anytime behavior is typical for many iterations based algorithms
- ✓ This means, that algorithm can be stopped anytime, and algorithm will have a solution



# View of performance

- ✓ It is assumed, that EAs are better than random search approach in average
- ✓ But in many cases, problem specific algorithms are more appropriate





# Thank you for your attention!

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