

Computational Intelligence

Subject11: Genetic Algorithm



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Agenda

- ▶ Evolutionary approaches
- ▶ Genetic algorithm
- ▶ Genetic programming
- ▶ Evolution strategy



Evolutionary Approaches

- ▶ Genetic Algorithm
 - ▶ A metaheuristic inspired by the process of natural selection
- ▶ Genetic Programming
 - ▶ A technique of evolving programs to find a fit for a particular task
- ▶ Evolutionary Strategy
 - ▶ An optimization technique based on the ideas of evolution
- ▶ Evolutionary Programming
 - ▶ One of the major evolutionary algorithm paradigms where the goal is to optimize numerical parameters

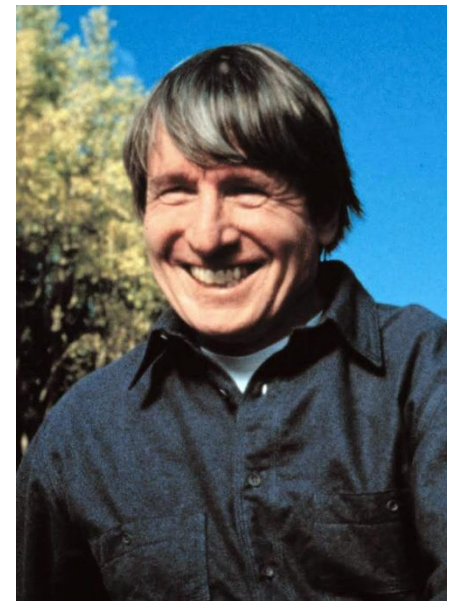
Evolutionary Approaches

- ▶ Differential Evolution
 - ▶ A method to optimize a problem by iteratively improving the candidate solution
- ▶ Cultural Algorithm
 - ▶ A branch of EC with a knowledge component along with the population
- ▶ Cooperative Coevolution
 - ▶ Dividing a large problem into subcomponents and solving them independently
- ▶ Memetic Algorithm
 - ▶ An extension of the traditional genetic algorithm

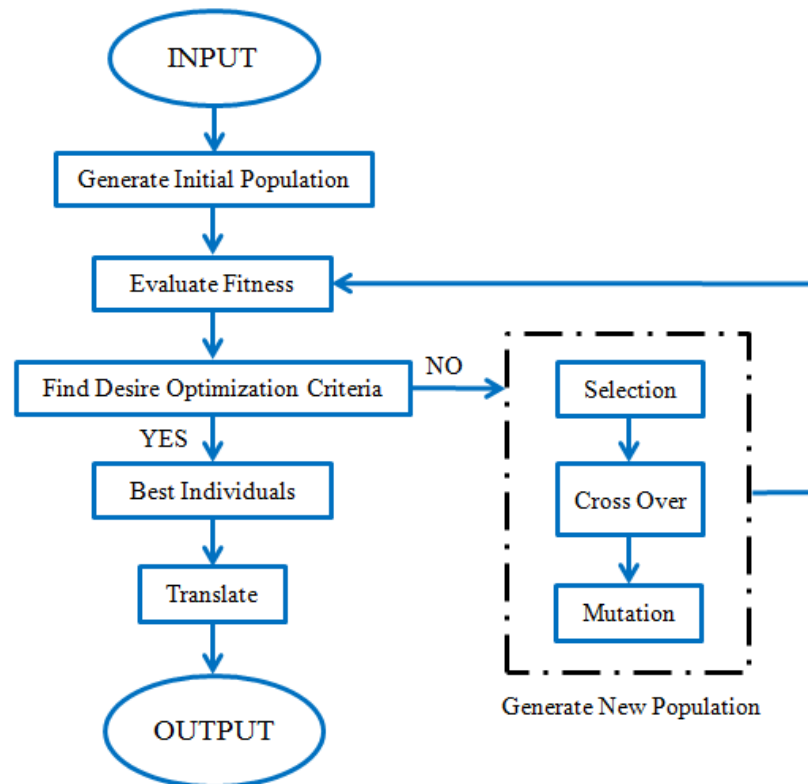
Genetic Algorithm

Brief history

- ▶ **Alex Fraser**, simulation of artificial selection of organisms
- ▶ **Hans-Joachim Bremermann**, adopted a population of solution to optimization problems
 - ▶ Recombination, Mutation, and Selection
- ▶ **Barricelli**, simulated the evolution to play game
- ▶ **Lawrence J. Fogel** and evolutionary programming
- ▶ **John Henry Holland**, the pioneer in what became known as genetic algorithms



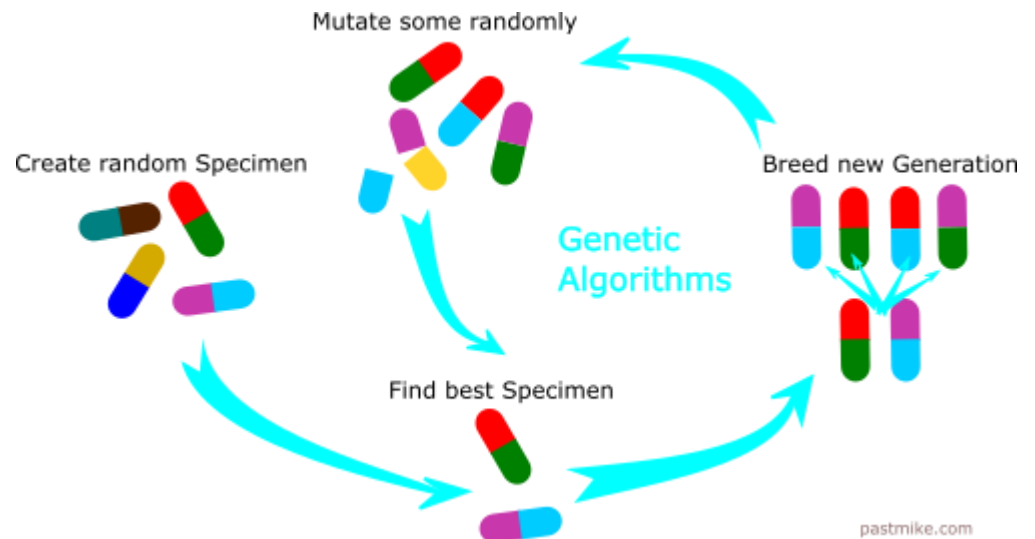
Genetic Algorithm



Genetic Algorithm

Applications of GA

- ▶ Search
- ▶ Optimization
- ▶ Machine Learning
- ▶ Robotics
- ▶ Control
- ▶ Scheduling



Genetic Algorithm

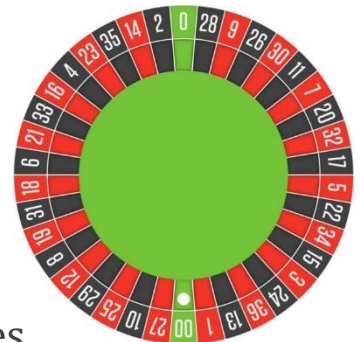
Standard Genetic Algorithm

- ▶ Chromosomes contain binary genes
- ▶ Constant size of population during the process
- ▶ Fitness proportionate selection
- ▶ Single-point crossover
 - ▶ High crossover probability (+95%)
- ▶ Bit Flip mutation
 - ▶ Low mutation probability (chromosome's length)

Genetic Algorithm

Standard Genetic Algorithm

- ▶ Selection method - Fitness proportionate selection
 - ▶ AKA Roulette Wheel Selection
 - ▶ A proportion of the wheel is assigned to each of the chromosomes
 - ▶ Based on their fitness value
 - ▶ Optimized chromosomes → More chance of selection



$$P_i = \frac{f_i}{\sum f_i}$$

No.	Chromosome	Fitness	Chance
1	01101	169	14.4
2	11000	576	49.2
3	01000	64	5.5
4	10011	361	30.9

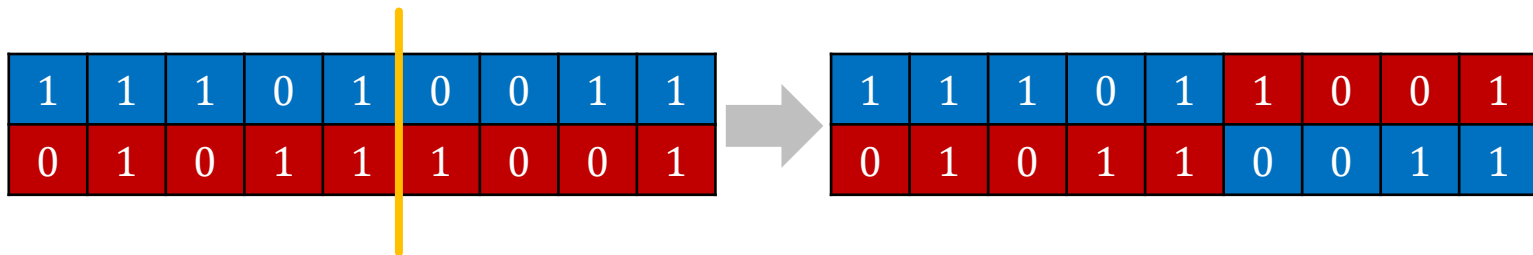
Genetic Algorithm

Standard Genetic Algorithm

► Recombination/Crossover

Single-point crossover

- A random point (Crossover Point) is picked
- Swapping items from the selected point
- **Outcome:** two offspring, each with some data from both parents



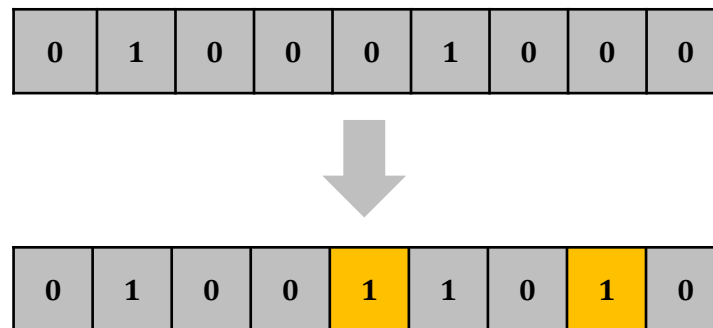
Genetic Algorithm

Standard Genetic Algorithm

- ▶ Mutation

 - Bit Flip (binary)**

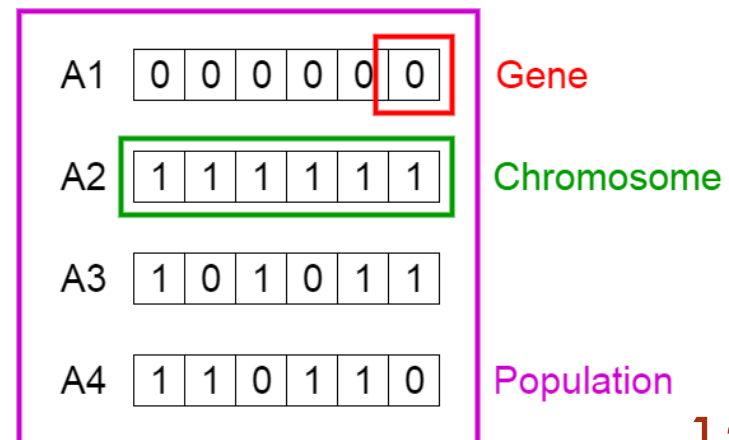
 - ▶ Reverse the values of some genes



Genetic Algorithm

Implementation

- ▶ Init population: available solutions to solve a problem
- ▶ In the SGA, chromosomes are binary arrays
- ▶ In the extended version, they can be:
 - ▶ Graph
 - ▶ Tree
 - ▶ Struct
 - ▶ Node
 - ▶ Array members
 - ▶ etc.



Genetic Algorithm

Implementation

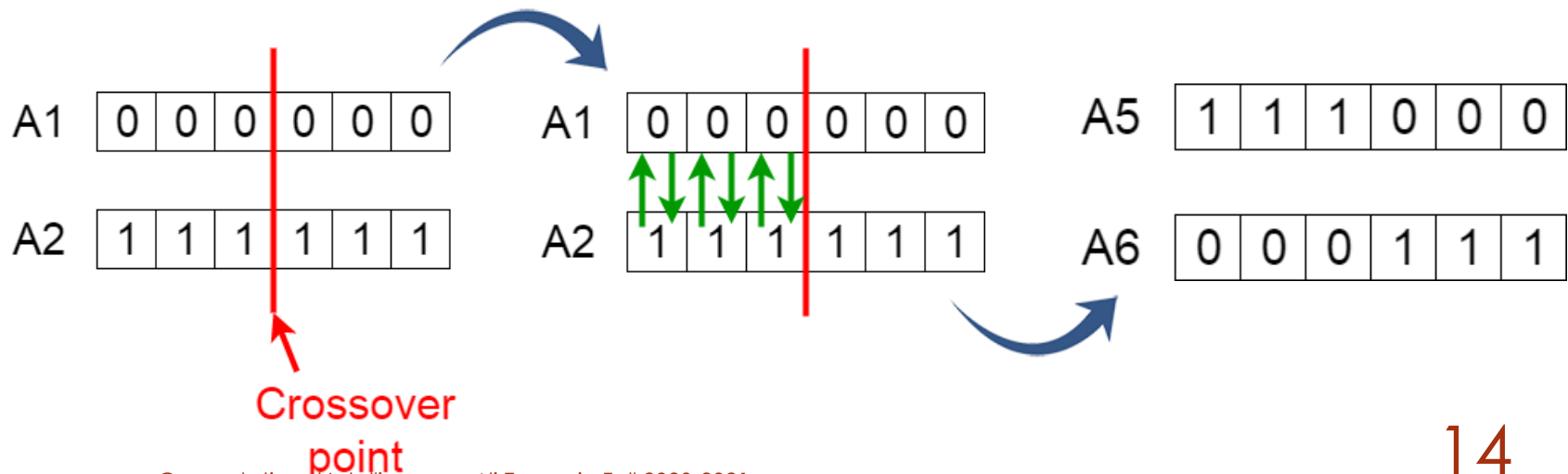
- ▶ Fitness function: How good a chromosome is compared to others?
- ▶ Consider a weight/score to each chromosome
- ▶ Select two chromosomes holding the greatest scores



Genetic Algorithm

Implementation

- ▶ In recombination, randomly choose a break point
- ▶ Start replacing the items from the given point
- ▶ Add new items to the main population



Genetic Algorithm

Implementation

- ▶ In Mutation, reverse some random genes
- ▶ Add new items to the main population
- ▶ Check the Termination condition: no new offspring

Before Mutation

A5

1	1	1	0	0	0
---	---	---	---	---	---

After Mutation

A5

1	1	0	1	1	0
---	---	---	---	---	---

Genetic Algorithm

Pseudo Code:

START

Generate the initial population

Compute fitness

REPEAT

 Selection

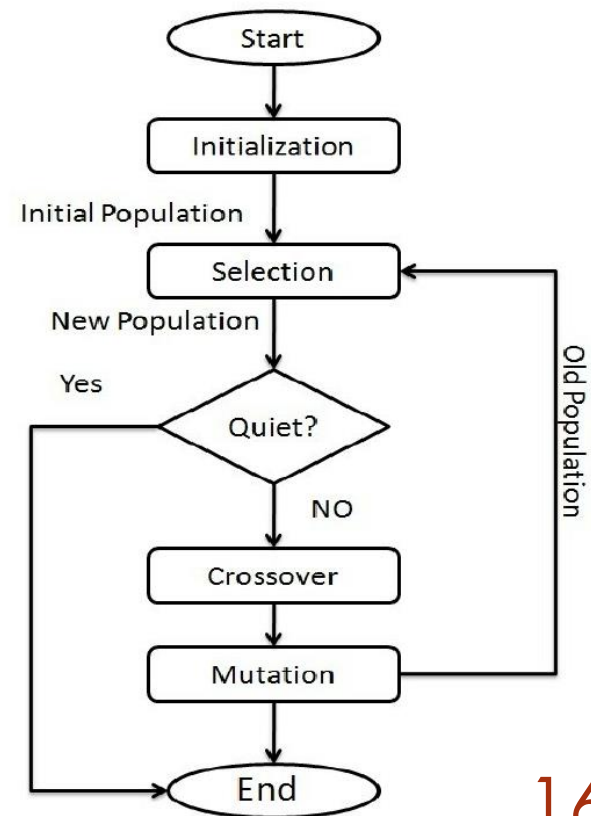
 Crossover

 Mutation

 Compute fitness

UNTIL the population is converged

STOP



Genetic Algorithm

Sample Code:

```
class Individual {  
    int fitness = 0;  
    int[] genes = new int[5];  
    int geneLength = 5;  
  
    public Individual() {  
        Random rn = new Random();  
        //Set genes randomly for each individual  
        for (int i = 0; i < genes.length; i++) {  
            genes[i] = Math.abs(rn.nextInt() % 2);  
            fitness = 0;  
        }  
    }  
}
```

More genes with the value of 1
represent more fitness

```
}  
// Calculate fitness  
public void calcFitness() {  
    fitness = 0;  
    for (int i = 0; i < 5; i++) {  
        if (genes[i] == 1) {  
            ++fitness;  
        }  
    }  
}  
}
```

Genetic Algorithm

Sample Code:

// Initialize population

```
public void initializePopulation(int size) {  
    for (int i = 0; i < individuals.length; i++) {  
        individuals[i] = new Individual();  
    }  
}
```

// Selection

```
void selection() {  
    //Select the most fittest individual  
    fittest = population.getFittest();  
    //Select the second most fittest individual  
    secondFittest = population.getSecondFittest();  
}
```

Genetic Algorithm

Sample Code:

```
void crossover() {  
    Random rn = new Random();  
    //Select a random crossover point  
    int crossOverPoint = rn.nextInt(population.individuals[0].geneLength);  
    //Swap values among parents  
    for (int i = 0; i < crossOverPoint; i++) {  
        int temp = fittest.genes[i];  
        fittest.genes[i] = secondFittest.genes[i];  
        secondFittest.genes[i] = temp;  
    }  
}
```

Genetic Algorithm

```
void mutation() {  
    Random rn = new Random();  
    //Select a random mutation point  
    int mutationPoint = rn.nextInt(population.individuals[0].geneLength);  
    //Flip values at the mutation point  
    if (fittest.genes[mutationPoint] == 0) {  
        fittest.genes[mutationPoint] = 1;  
    } else {  
        fittest.genes[mutationPoint] = 0;  
    }  
    mutationPoint = rn.nextInt(population.individuals[0].geneLength);  
    if (secondFittest.genes[mutationPoint] == 0) {  
        secondFittest.genes[mutationPoint] = 1;  
    } else {  
        secondFittest.genes[mutationPoint] = 0;  
    }  
}
```

Genetic Algorithm

Sample Application – Travelling Salesman Problem:

Tehr	Esfh	Mshh	Shrz	Tbrz	Zhdn
------	------	------	------	------	------

- ▶ Init population:
 - ▶ Sample: Tbrz→Tehr→Mshh→Esfh→Shrz→Zhdn
- ▶ **Chromosomes:** all possible variations of the array
- ▶ **Fitness:** distance travelled
- ▶ **Selection:** random, FPS, rank, etc.
- ▶ **Crossover:** SPC or TPC
- ▶ **Mutation:** Relocation, Insertion, etc.



Genetic Algorithm

Sample Application

- ▶ Aircraft and train scheduling
- ▶ Memory allocation in RAM
- ▶ Robots antenna to receive signal
- ▶ Medical business to find proper remedy
- ▶ Application scheduling in a computer
- ▶ etc.

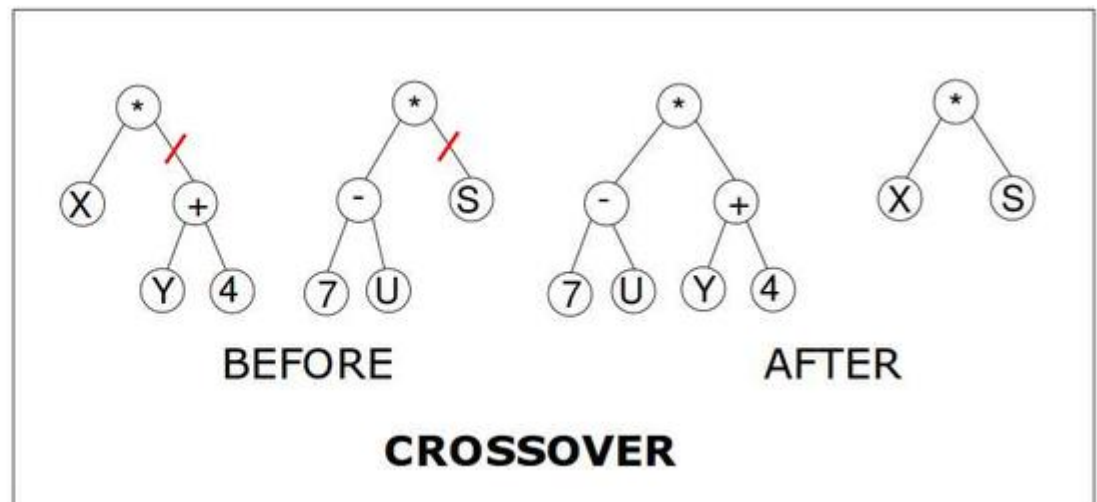
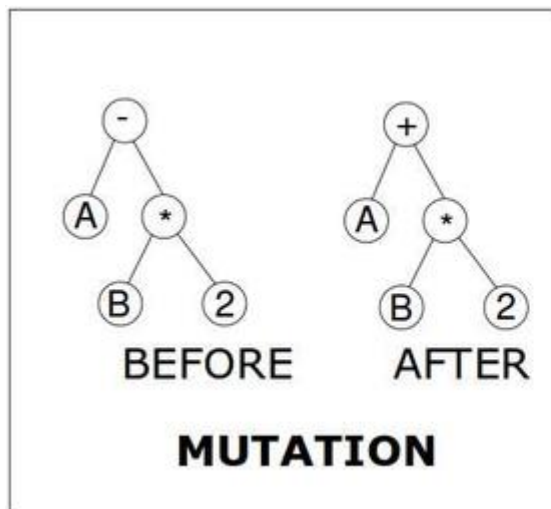


Genetic Programming

- ▶ A technique of evolving programs
- ▶ Used as an automatic programming tool
- ▶ Starting from a population of random programs, fit for a particular task
- ▶ Searching for an **optimal** or at least suitable program
 - ▶ Generally using Hill-climb algorithm
- ▶ **Selection** of the fittest programs
- ▶ Swapping random parts of selected pairs (**Crossover**)
- ▶ Substitution of some random part of a program with some others (**Mutation**)

Genetic Programming

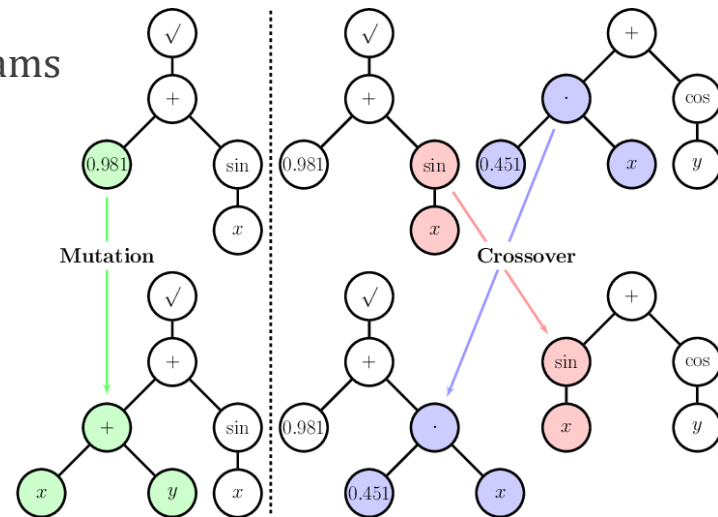
Codes are usually implemented in a tree structure



Genetic Programming

Genetic Algorithms vs. Genetic Programming

- ▶ GP applies GA to a population of computer programs
- ▶ GP utilizes tree structures
- ▶ GP: creating entire software programs
- ▶ GA: creating shorter pieces of code



Evolution Strategy

Study Yourself

- ▶ An optimization technique based on ideas of evolution
- ▶ Has Mutation and Selection, but not Crossover
- ▶ The most basic and canonical version: Gaussian ES
- ▶ Other approaches:
 - ▶ Covariance Matrix Adaptation
 - ▶ Natural Evolution Strategies
- ▶ Applications:
 - ▶ Deep Reinforcement Learning
 - ▶ Deep Learning



What's Next?

► Swarm Intelligence



Questions?

