

Lesson 10: Genetic Algorithm

Reviewer

Key Terms

- **Mutation**
- **Gene**
- **Chromosome**
- **Crossover**
- **Population**
- **Fitness**

What is Genetic Algorithm?

Genetic algorithms (GAs) and **genetic programming (GP)** are branches of **evolutionary computing**, a subset of artificial intelligence where solutions evolve over time to fit a given set of parameters or solve specific problems.

Process Flow

1. **Start**
2. **Initialization**
3. **Evaluation**
4. **Selection**
5. **Crossover**
6. **Mutation**
7. **Generation**
8. Check: **End?**
 - If No: Loop back
 - If Yes: **Stop**

Understanding Genetic Algorithms

A **genetic algorithm** or **GA** is a search technique used in computing to find true or approximate solutions to **optimization and search problems**.

The techniques are inspired by **natural evolution** such as **inheritance, mutation, selection**, and **crossover**.

GAs are implemented by having an array of bits or characters to represent the **chromosomes**.

Detailed Process Flow

1. **Start**
2. **Initialization:** Create **Initial Population**
3. Loop:
 - (a) **Selection**
 - (b) **Crossover**
 - (c) **Mutation**
 - (d) Create **New Population**
4. Check Termination (Quiet?):
 - If No: Return to **Old Population** and repeat loop.
 - If Yes: **End**

Steps for Genetic Algorithm

1. **Choose initial population**
2. **Evaluate the fitness** of each individual in the population
3. Repeat until termination:
4. **Select best ranking individuals** to reproduce
5. Breed new generation through **crossover** and/or **mutation** and give birth to offspring
6. **Evaluate the individual fitnesses** of the offspring
7. Replace worst ranked part of the population with offspring

Foundation of Genetic Algorithms

Genetic algorithms are based on an analogy with the genetic structure and behavior of **chromosomes** of the population. Following is the foundation of GAs based on this analogy:

1. Individuals in the population **compete for resources and mate**.
2. Those individuals who are successful (**fittest**) then mate to create more offspring than others.
3. Genes from the "**fittest**" parent propagate throughout the generation; that is, sometimes parents create offspring which is **better than either parent**.
4. Thus each successive generation is **more suited for their environment**.

Operators of Genetic Algorithms

1. **Selection Operator:** The idea is to give preference to the individuals with **good fitness scores** and allow them to pass their genes to successive generations.
2. **Crossover Operator:** This represents **mating between individuals**. Two individuals are selected using selection operator and crossover sites are chosen randomly. Then the genes at these crossover sites are exchanged thus creating a **completely new individual (offspring)**.

Example Diagram Description:

- Parent 1: A B | C D | E F G H
- Parent 2: F G | H A | D B E A
- Offspring: F G H B C D E A (Genes exchanged at crossover sites)

3. **Mutation Operator:** The key idea is to **insert random genes** in offspring to maintain the **diversity in the population** to avoid premature convergence.

Example Diagram Description:

- Before Mutation: F G H B C D E A
- After Mutation: F G **M** B C D E **N** (Random genes 'M' and 'N' inserted)

Algorithm Summary

The whole algorithm can be summarized as:

1. **Randomly initialize populations**
2. **Determine fitness** of population
3. Until convergence repeat:
 - (a) **Select parents** from population
 - (b) **Crossover** and generate new population
 - (c) Perform **mutation** on new population
 - (d) **Calculate fitness** for new population