

# Evolutional Computing Repository

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## Description

This repository contains the implementation of genetic algorithms for optimization, including selection, crossover, and mutation methods. The main script, `MainOptimizationScript.py`, is responsible for executing the optimization process and evaluating the algorithm's performance. Additional scripts and libraries provide modular implementations of key components.

## Repository Structure

```
EvolutionalComputing/
├── MainOptimizationScript.py      # Main script for optimization
├── Playground.py                 # Auxiliary script for testing and
experimentation
├── Experiments_1A/              # Scripts for running specific experiments
│   ├── ExperimentPopulationSize.py # Experiment: Varying population sizes
│   ├── ExperimentCrossoverRate.py  # Experiment: Varying crossover rates
│   ├── ExperimentMutationRate.py   # Experiment: Varying mutation rates
│   └── ExperimentElitismProportion.py # Experiment: Varying elitism proportions
├── Library/                     # Library with selection, crossover, and
mutation methods
│   ├── CrossoverMethods.py        # Crossover methods
│   ├── MutationMethods.py         # Mutation methods
│   ├── SelectionMethods.py        # Selection methods
│   └── __pycache__/               # Compiled files
├── Results/                     # Directory for storing results
└── _compiled/                   # Directory for compiled files
```

## Main Scripts

### MainOptimizationScript.py

This is the main script that implements the genetic algorithm. It includes:

- **Algorithm Configuration:** Parameters such as population size, number of generations, selection, crossover, and mutation methods.
- **Fitness Functions:** Implementation of fitness functions like Base, Ackley, Drop-Wave, and Levi.
- **Performance Metrics:** Success rate, best fitness, population diversity, and execution time.
- **Stopping Criteria:** Configurable stopping criteria:
  - Fixed number of generations (`GENERATION_COUNT`).
  - Discovery of a solution with a desired fitness (`TARGET_FITNESS`).
  - No improvement over a specified number of generations (`NO_IMPROVEMENT_LIMIT`).
- **Visualizations:** Convergence, population diversity, and optimal points distribution plots.

## Key Methods

- `evaluate_fitness`: Evaluates the fitness of a chromosome based on the selected function.
  - `single_optimization`: Executes a single optimization.
  - `multiple_optimization`: Executes multiple optimizations and calculates aggregated metrics.
  - `elitism_optimization`: Implements the elitism method for population evolution.
  - `maintain_diversity`: Applies strategies to maintain population diversity.
  - `plot_convergence_curve`: Plots the convergence curve.
  - `plot_population_diversity`: Plots the population diversity.
  - `plot_optimal_points`: Plots the distribution of optimal points.
  - `save_results`: Saves results, configuration, and visualizations to a timestamped folder.
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## Playground.py

This script is used for testing and experimenting with different configurations and methods of the genetic algorithm. It allows developers to:

- Test new fitness functions.
  - Experiment with different selection, crossover, and mutation methods.
  - Debug and validate the behavior of the genetic algorithm components.
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## Experiments\_1A/

### ExperimentPopulationSize.py

This script evaluates the impact of varying **population sizes** on the algorithm's performance. It analyzes how population size affects convergence, diversity, and execution time.

### ExperimentCrossoverRate.py

This script analyzes the effect of varying **crossover rates** on the algorithm's performance. It evaluates how crossover impacts diversity and convergence.

### ExperimentMutationRate.py

This script analyzes the effect of varying **mutation rates** on the algorithm's performance. It evaluates how mutation impacts diversity and convergence.

### ExperimentElitismProportion.py

This script evaluates the impact of different **elitism proportions** on the optimization process. It helps determine the most efficient proportion of elites for specific problems.

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## Library/

### SelectionMethods.py

Implements various selection methods for choosing parents in the genetic algorithm:

- **Tournament Selection:** Selects the best individual from a randomly chosen subset of the population.
- **Roulette Wheel Selection:** Selects individuals with a probability proportional to their fitness.

### CrossoverMethods.py

Implements crossover methods for combining parent chromosomes to produce offspring:

- **Single-Point Crossover:** Splits parent chromosomes at a random point and swaps segments.
- **Arithmetic Crossover:** Combines parent chromosomes using a weighted average.

### MutationMethods.py

Implements mutation methods for introducing random changes to chromosomes:

- **Random Mutation on Individual Genes:** Randomly alters genes in a chromosome based on a mutation rate.

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## How to Use

1. Configure the parameters in the `MainOptimizationScript.py` file.
2. Run the script to perform optimizations:

```
python MainOptimizationScript.py
```

3. To run specific experiments, navigate to the `Experiments_1A` folder and execute the desired script:

```
python Experiments_1A/ExperimentPopulationSize.py
```

4. Visualize the results in the generated plots and metrics displayed in the console.

## Requirements

- Python 3.11 or higher
- Required libraries:
  - `numpy`
  - `matplotlib`
  - `scipy`

Install the dependencies with:

```
pip install numpy matplotlib scipy
```

## Contact

For questions or suggestions, contact Gabriel Fernandes.