

Incorporating transposons into Genetic Algorithms

by Samuel Santos Hernán

Genetic Algorithms

Genetic Algorithms (GAs) are optimization techniques inspired by **natural evolution**, introduced in the 1960s by John Holland.

- Utilize processes like **selection**, **mutation**, and **crossover** to evolve solutions
- Effective in exploring large solution spaces
- Premature convergence is a significant challenge, leading to loss of diversity



Genes for transposition Structural genes Inverted Repeats Inverted IS

The Role of Transposons

Transposons are **mobile genetic elements** that can enhance genetic diversity within GAs.

- Help maintain genetic variability to prevent **premature convergence**
- Introduce new genetic combinations, promoting exploration of solution spaces
- Two types: DNA transposons (direct movement) and retrotransposons (move via RNA)





The Snake Game Problem

The Snake Game serves as a testbed for evaluating GAs with transposons.

- A snake in a 2D board that aims at eating as many apples as possible while avoiding the walls and its own body, since it grows with every apple eaten.
- In each cycle, the snake can either turn left, keep its current direction, or turn right, relative to its head's current orientation.

 This problem is simple enough to focus on transposons, but still complex enough so as to being able to observe a decent diversity of behaviors and hence to compare the different algorithm approaches.





Research Objectives

This study aims at integrating **transposons** into **GAs** in order to enhance optimization performance.

• Main goal: Improve performance in **GAs** using transposons, and preventing premature convergence in optimization issues.

 Secondary objectives: Analyze effectiveness of transposons and their probability settings



∨ THESIS > _pycache_ **System Architecture Overview** > Documentation ✓ Genes > _pycache_ SnakeDNA.py SnakeGene.py The system consists of several classes designed to simulate genetic Transposon.py operations in the Snake Game. Results SnakeGene Class: Defines an individual gene Scores Transposon Class: Manages transposon mechanisms and (transposon) ✓ Snake mutation probability > _pycache_ **SnakeChromosome Class**: Wraps a list of SnakeGenes and a list of Snake.py Transposons inside a class, and enriches them with several important methods. SnakeFarm.py SnakeGame.py > SnakeGenerations Snake Class: Represents the agent's behavior, with a body and a brain ∨ Utils SnakeFarm Class: Class in charge of executing the Genetic algorithm. > _pycache_ **SnakeGame Class**: Environment for agent interactions Constants.py **Utils.py** Universidad 0 main.py

Evaluating Transposons in GAs

Three configurations were tested to evaluate the performance of transposon-enhanced GAs.

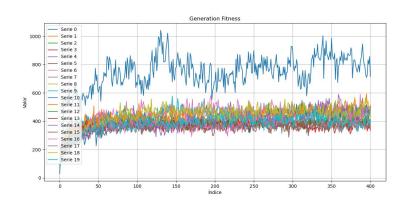
- Traditional GA: Baseline with no transposons
- GA with 20% Transposon Probability: Lower integration rate
- GA with 40% Transposon Probability: Higher integration to assess impact on diversity





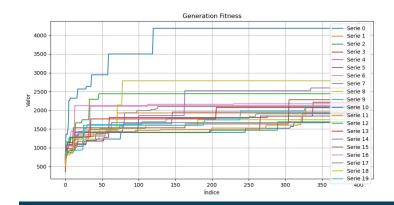


Standard GAs



Mean scores of populations

Mean scores of the populations in the different executions



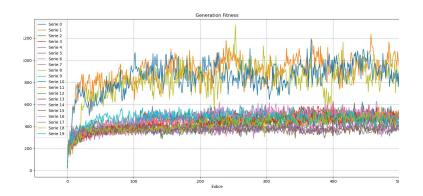
Best scores

Evolution of best scores of the populations in the different executions



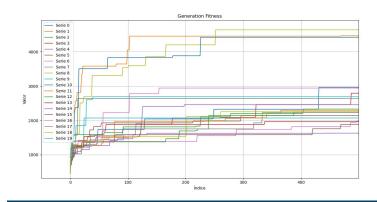
GAs with transposons

(maximum probability of mutation: 20%)



Mean scores of populations

Mean scores of the populations in the different executions



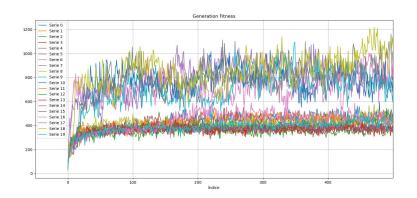
Best scores

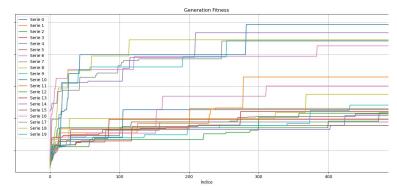
Evolution of best scores of the populations in the different executions



GAs with transposons

(maximum probability of mutation: 40%)





Mean scores of populations

Mean scores of the populations in the different executions

Best scores

Evolution of best scores of the populations in the different executions



Key Findings and Conclusions

The research indicates that incorporating **transposons** into **GAs** can significantly enhance performance.

- Transposons help maintain diversity
- Improved exploration of solution spaces leads to better optimization outcomes
- Demonstrated potential for transposons to enhance traditional GAs effectively



Future Work Directions

Future research will explore additional avenues to further validate findings.

- Implementing different features, such as using A* distances instead of Manhattan, for example.
- Testing in more complex environments to assess **robustness**
- Investigating different types of transposons for varied impacts
- Expanding applications of transposon-enhanced GAs to other optimization problems







Thank you

Code available at: https://github.com/Somersault0023/TFM-UAH-Transposons/tree/main