



An introduction to Membrane Algorithms

José Antonio Andreu Guzmán

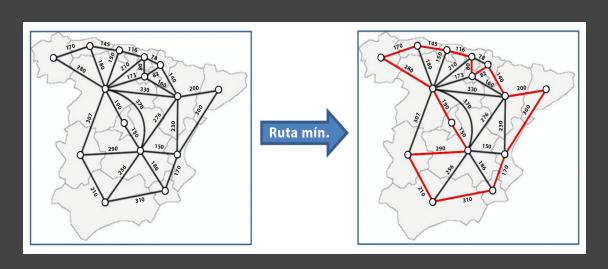
20th Brainstorming Week on Membrane Computing

Metaheuristics

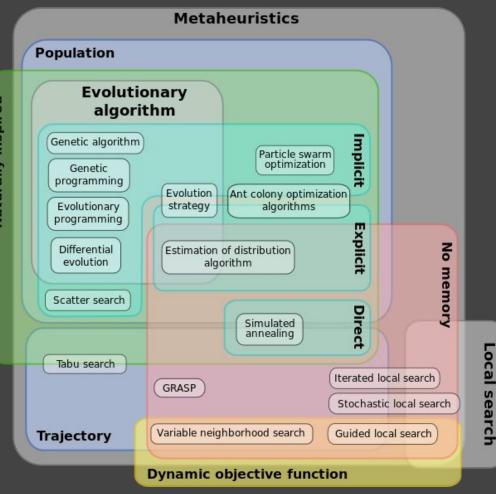
Membrane Computing

Membrane Algorithms

Travelling Salesman Problem

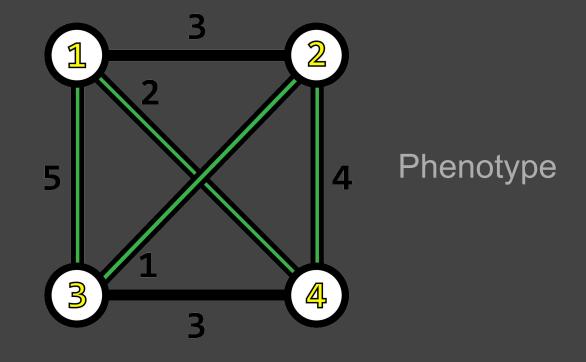


n	O(n!)
1	0.0001 seg
10	6.048 min
100	2.95·10 ¹⁴⁶ años



SOLUTIONS OPERATORS

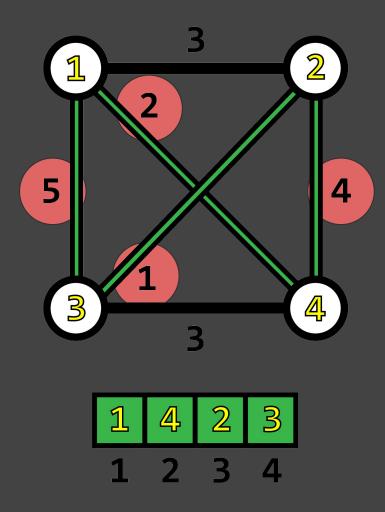
Chromosome



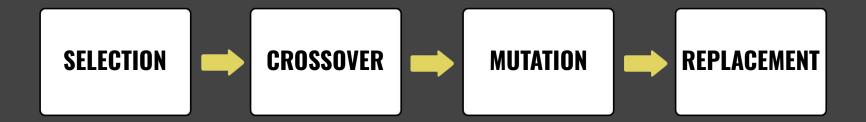


Genotype

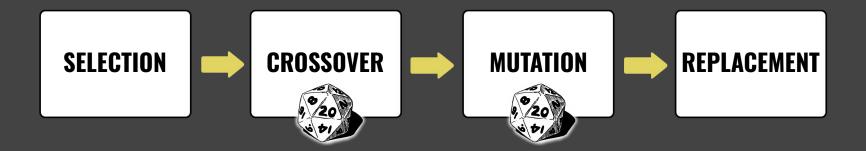
Fitness



Genetic Algorithm operators

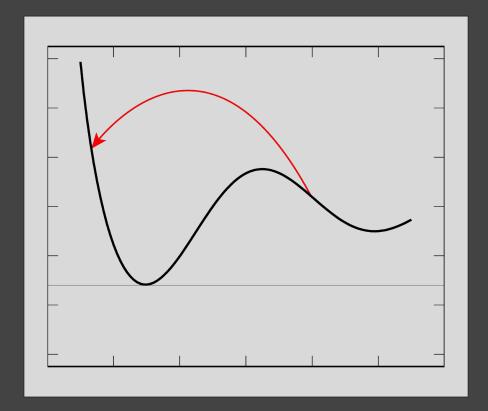


Genetic Algorithm operators



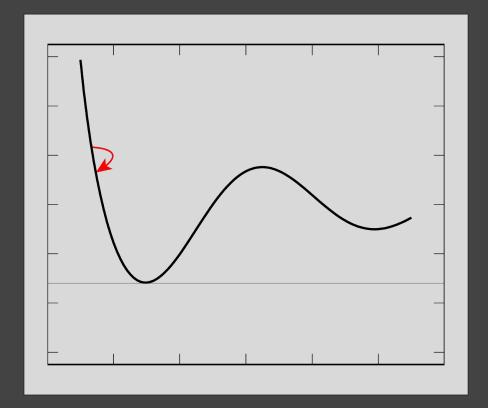
Crossover

- High probabilities
- Exploration > Exploitation

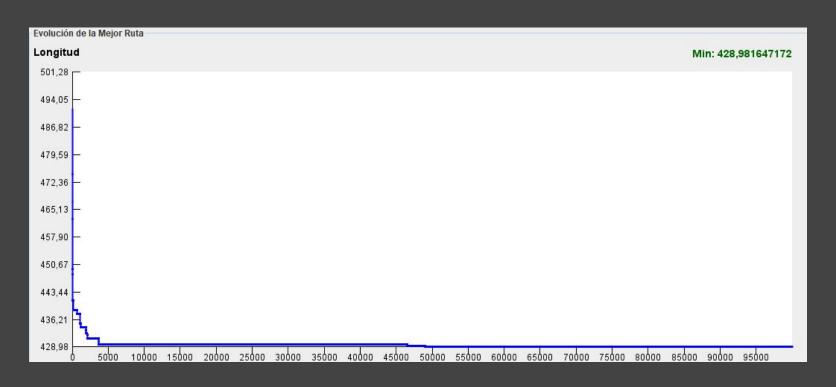


Mutation

- Low probabilities
- Exploration < Exploitation

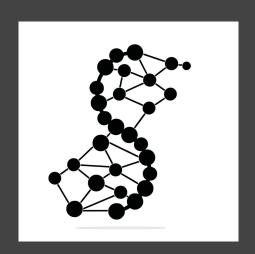


Premature convergence



Now yes! Membrane Algorithms

Metaheuristics





Membrane Algorithms Membrane Computing

```
1

\begin{array}{c}
3 & ac \\
a \rightarrow ab \\
a \rightarrow b\delta \\
c \rightarrow cc
\end{array}

\begin{array}{c}
b \rightarrow d \\
d \rightarrow de \\
(cc \rightarrow c) > (c \rightarrow \delta)
\end{array}

\begin{array}{c}
e \rightarrow e_{out}
\end{array}
```

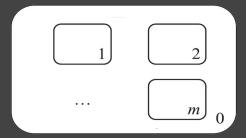
Membrane structure

Objects

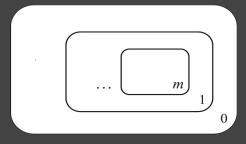
Communication rules

Evolution rules

OLMS



NMS



Membrane structure

1 2 3 4

4 3 2 1

Objects

1 3 2 4

Communication rules

4 2 3 1

1 4 2 3

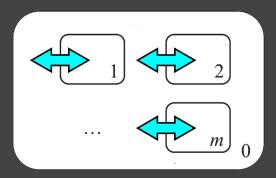
Evolution rules

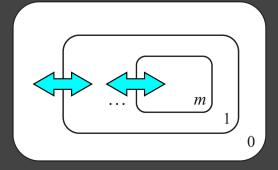
Membrane structure

Objects

Communication rules

Evolution rules



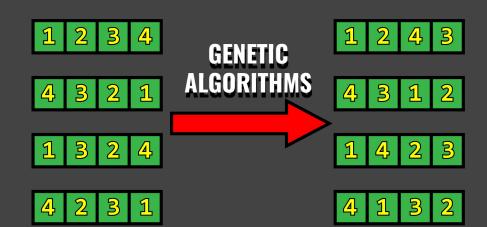


Membrane structure

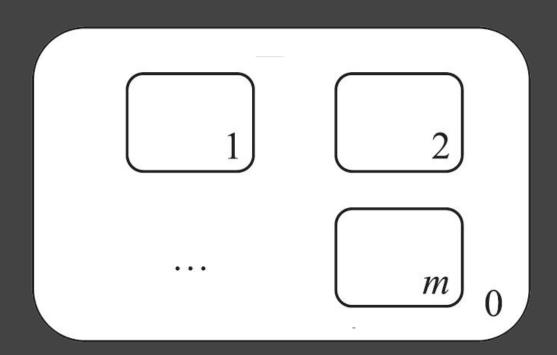
Objects

Communication rules

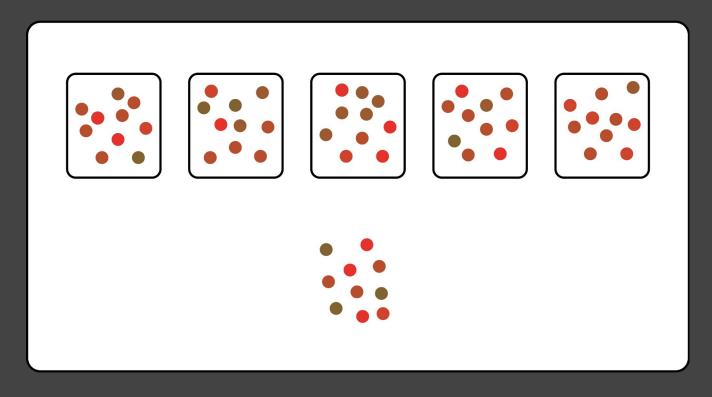
Evolution rules



One Level Membrane System



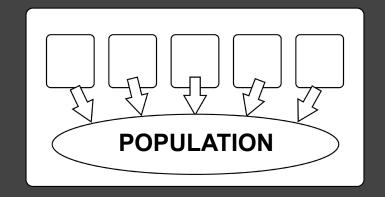
Population of chromosome objects

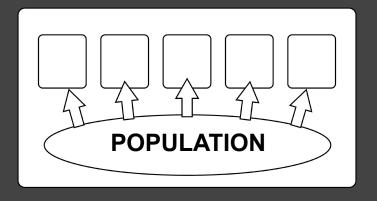


Communication rules

Send-out rule: Proportional

Send-in rule: Copy





Evolution rule: Genetic Algorithm

Genetic Algorithm Performance with Different Selection Strategies in Solving TSP

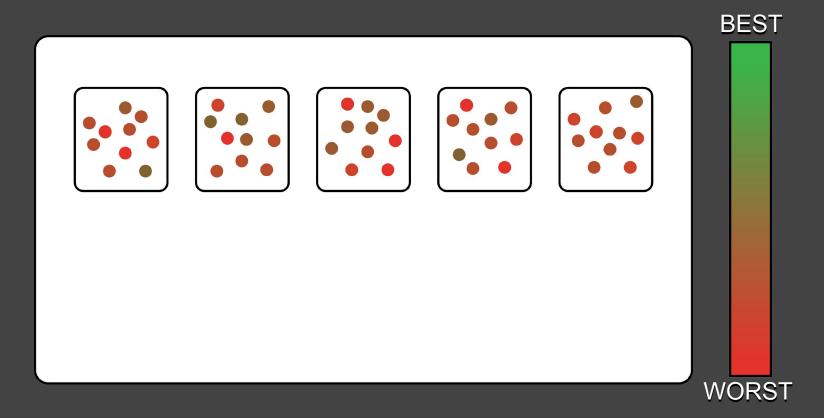
Noraini Mohd Razali, John Geraghty

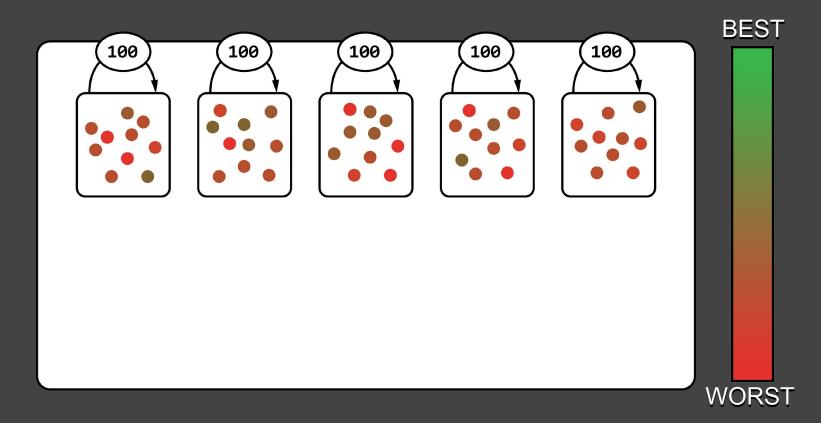
Abstract—A genetic algorithm (GA) has several genetic operators that can be modified to improve the performance of particular implementations. These operators include parent selection, crossover and mutation. Selection is one of the important operations in the GA process. There are several ways for selection. This paper presents the comparison of GA performance in solving travelling salesman problem (TSP) using different parent selection strategy. Several TSP instances were tested and the results show that tournament selection strategy outperformed proportional roulette wheel and rankbased roulette wheel selections, achieving best solution quality with low computing times. Results also reveal that tournament and proportional roulette wheel can be superior to the rankbased roulette wheel selection for smaller problems only and become susceptible to premature convergence as problem size increases.

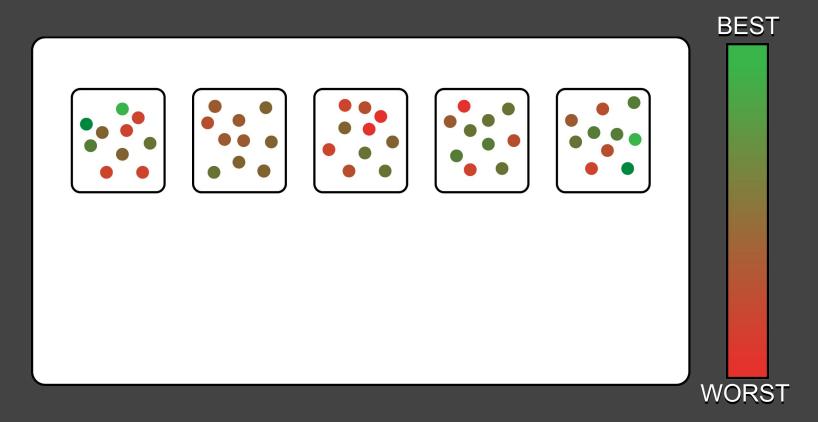
Index Terms— Genetic algorithm, Selection, Travelling salesman problem, Optimization

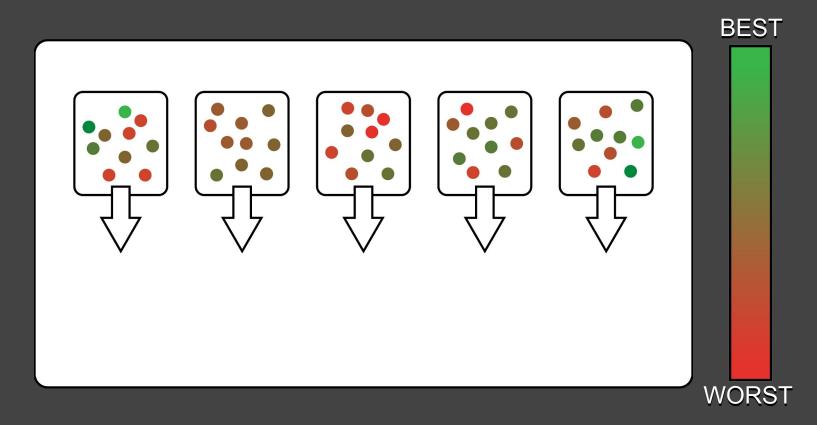
The different selection strategy used in the GA process will significantly affect the performance of the algorithm differently. This study is intended to examine the performance of GA when using different selection strategy specifically in solving the travelling salesman problem (TSP). TSP is a classical example of a NP-hard combinatorial optimization problem. Many production and scheduling problems can be reduced to a simple concept that there is a salesman who must travel from city to city, visiting each city exactly once and returning to the home city [2]. It is possible for the salesman to select the orders of the cities visited so that the total distances travelled in his tour is as small as possible which will apparently save him time and money [2]. Although TSP is conceptually simple, it is difficult to obtain an optimal solution. The main difficulty of this problem is the enormous number of possible tours; (n-1)!/2 for symmetric n cities tour. As the number of cities in the problem increases the numbers of permutations of valid

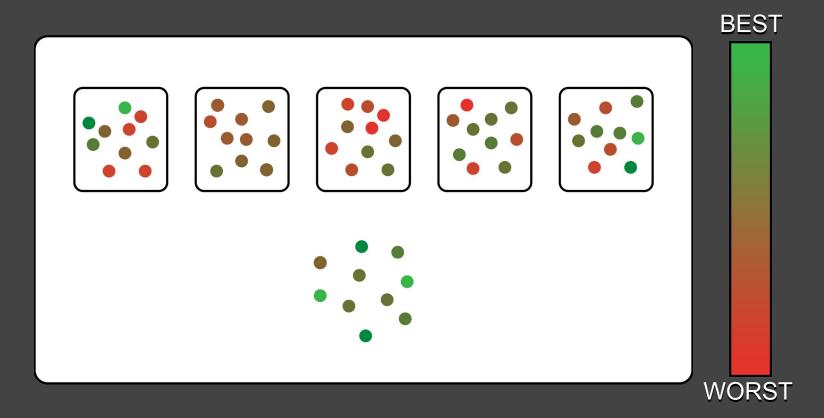
Generate initial population

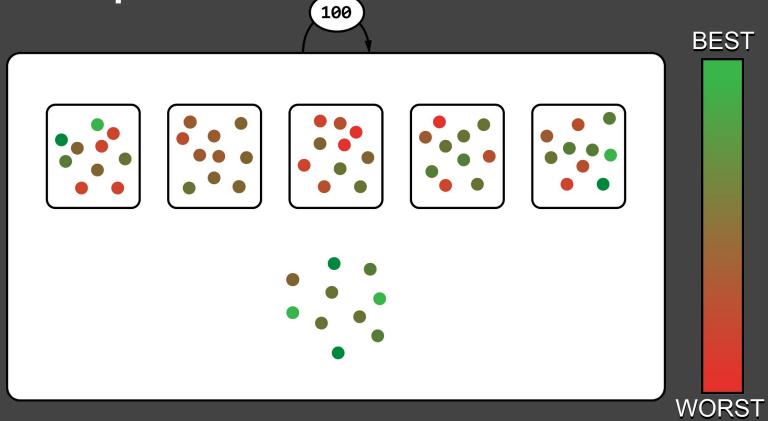


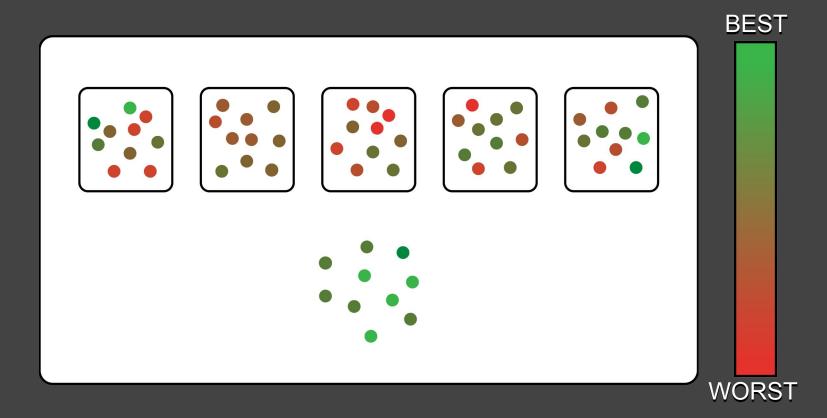


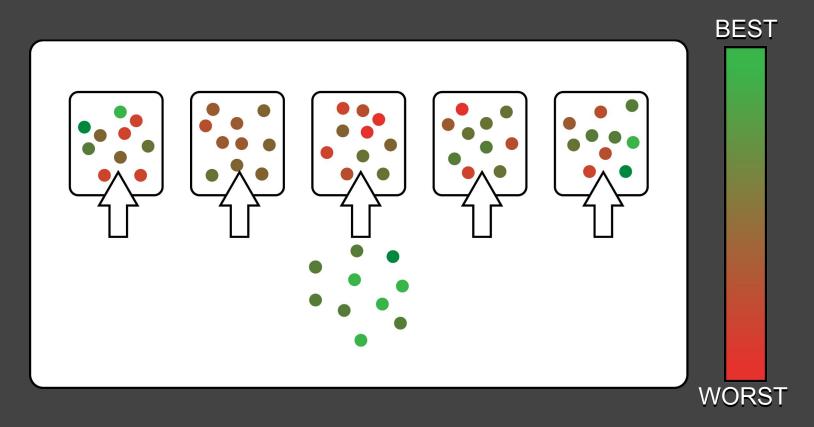


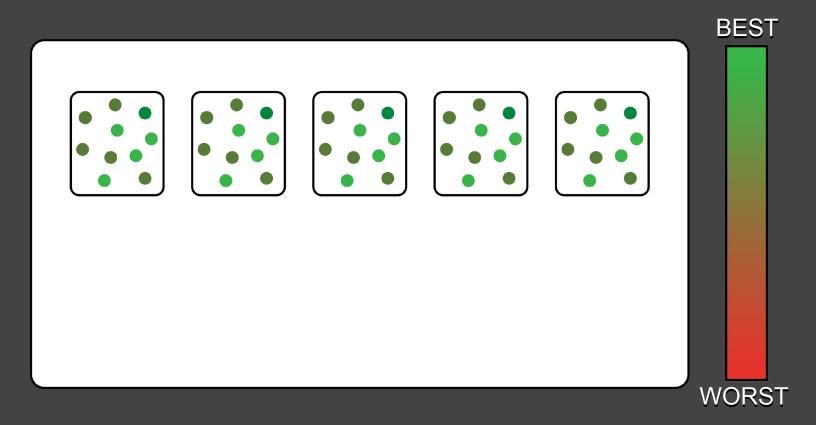


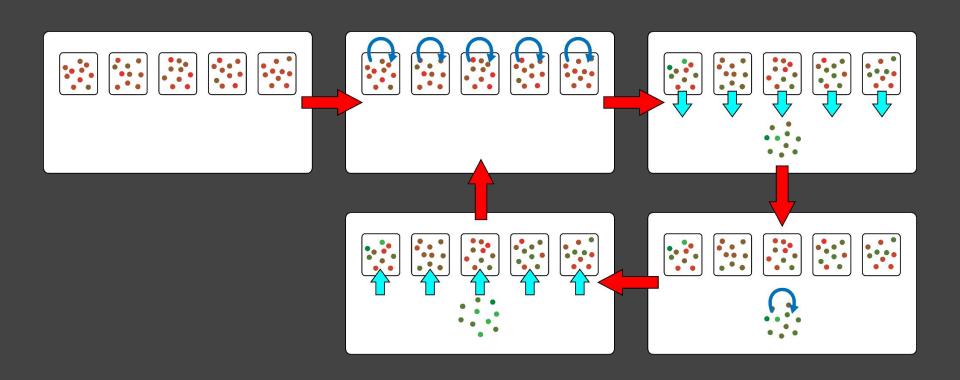




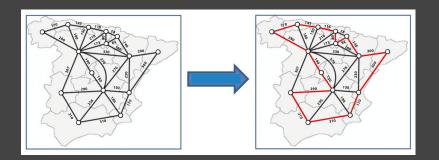


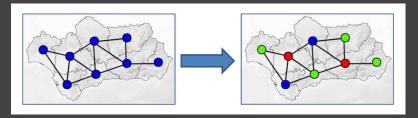






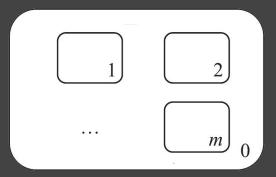
- Problems
- Membrane Structures
- Number of membranes
- Size of population
- Iterations
- Crossover and mutation probabilities
- Communication rules
- Evolution rules
- Develop a framework



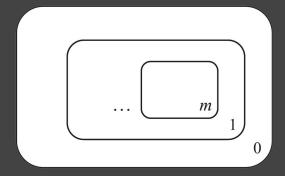


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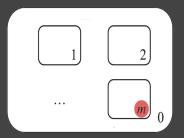


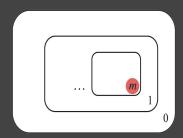


NMS



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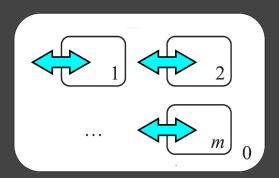


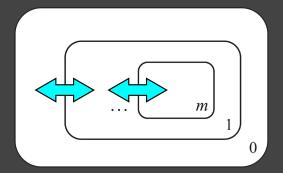
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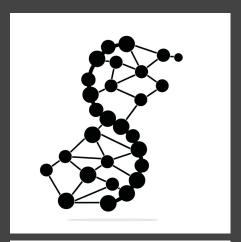


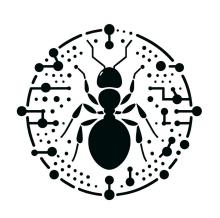
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Papers presented

Towards a General Framework for Membrane Algorithms

A novel solution for GCP based on an OLMS membrane algorithm with dynamic operators

José Antonio Andreu-Guzmán¹ · Luis Valencia-Cabrera¹

Future work

- Different configurations of genetic algorithms in each membrane.
- Incorporation other rules of Membrane Computing.
- Parallelising and optimise with hardware.

Ideas?







Thank you

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