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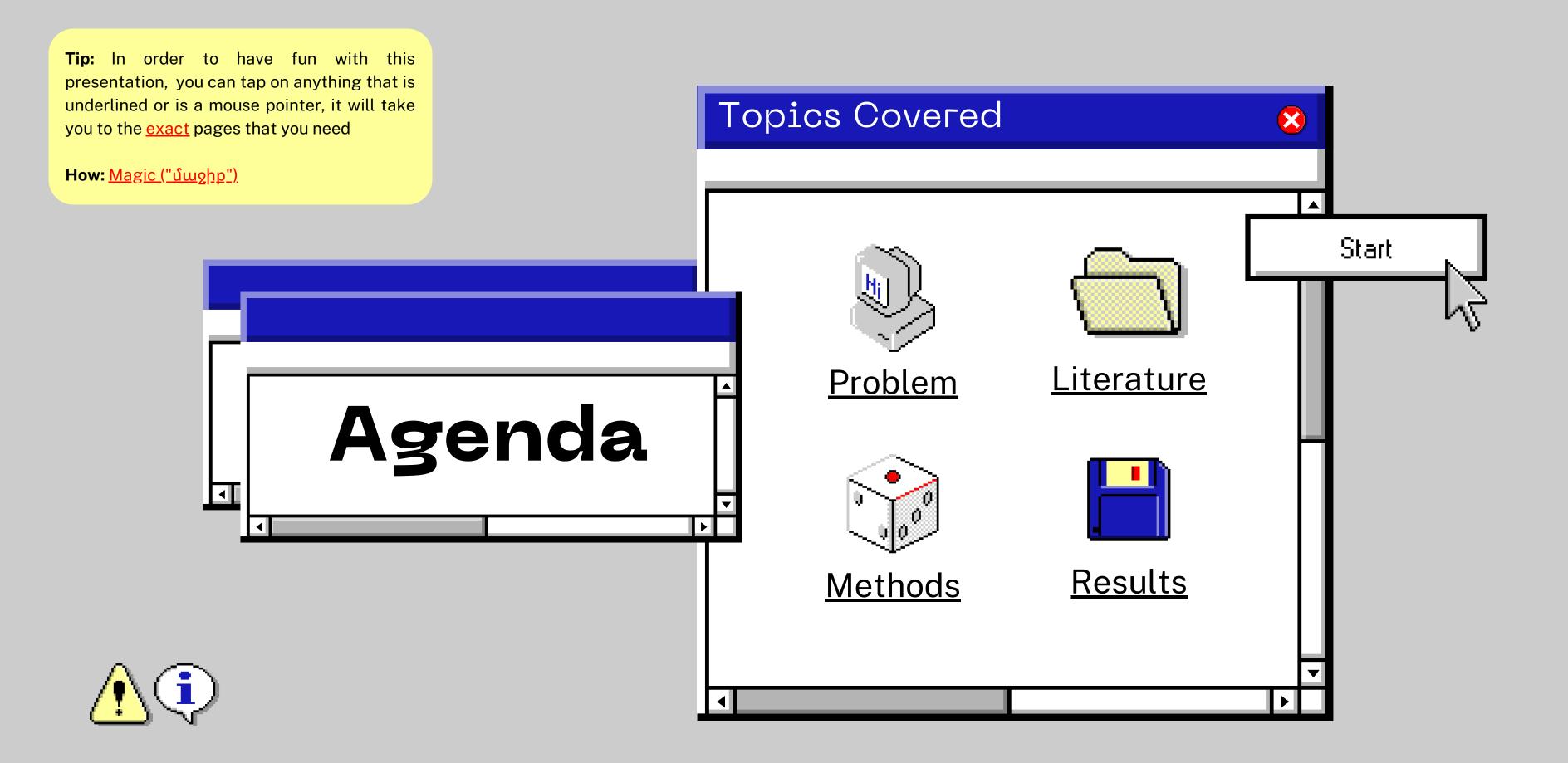






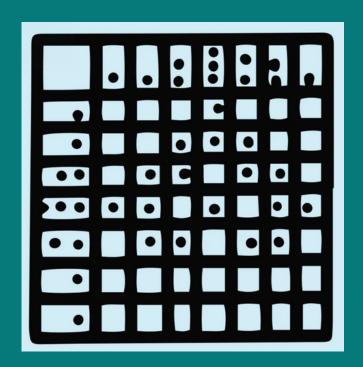


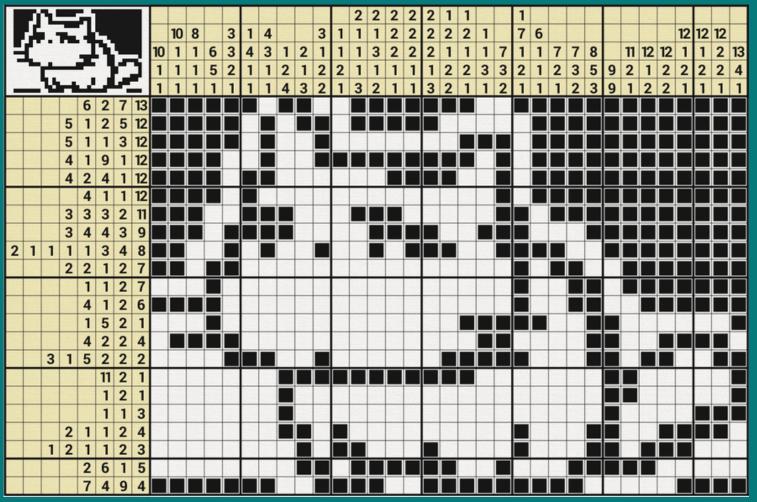


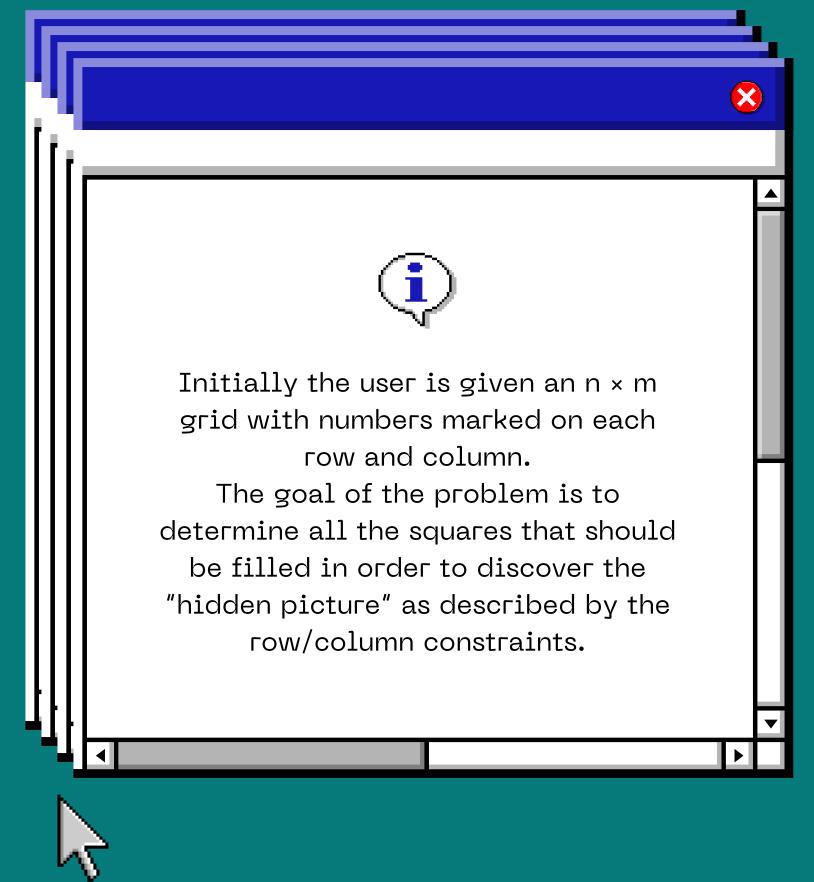


PROBLEM

Nonogram is a grid-based logic puzzle.



















LITERATURE

NP-Complete problem tried to be solved by:



Converting
Nonogram problem
to Integer Linear
Programming (ILP)
problem

Definition of puzzle can be used by CPLEX

Can be used only for black and white nonograms

Simple Depth First Search (Brute Force) algorithm filling and emptying cells

Recursive brute force trying every conceivable configuration of cells

The recursive algorithm requires exponential time to implement

Developing a
heuristic using
Simulated Annealing
(SA) to explore
various search
spaces

Starting with high temperature, error determined by adding up constraint violations

Not complete, can get stuck at local minimum when temperature is not high enough













METHODS

Constraint Satisfaction Problem (CSP)

Builds valid configurations of each row and tries to combine them to meet the column restrictions. Pre-processes the data to remove incorrect subtrees determined from previous iterations.

Genetic Algorithm (GA)

Generates a fitness function, starts running the algorithm on a random image, and optimizes it through the fitess function, takes into consideration the mutations and the cross over



Constraint Satisfaction Problem



Step 1:

Analyzing the layout of the board to identify the accuracy of the filling. The conjunction of the leftmost and rightmost complete run. From leftmost filling with the length of the first constraint till the last. For rightmost, the same from right to left. Taking the intersection of two runs to fill the squares.

Step 2:

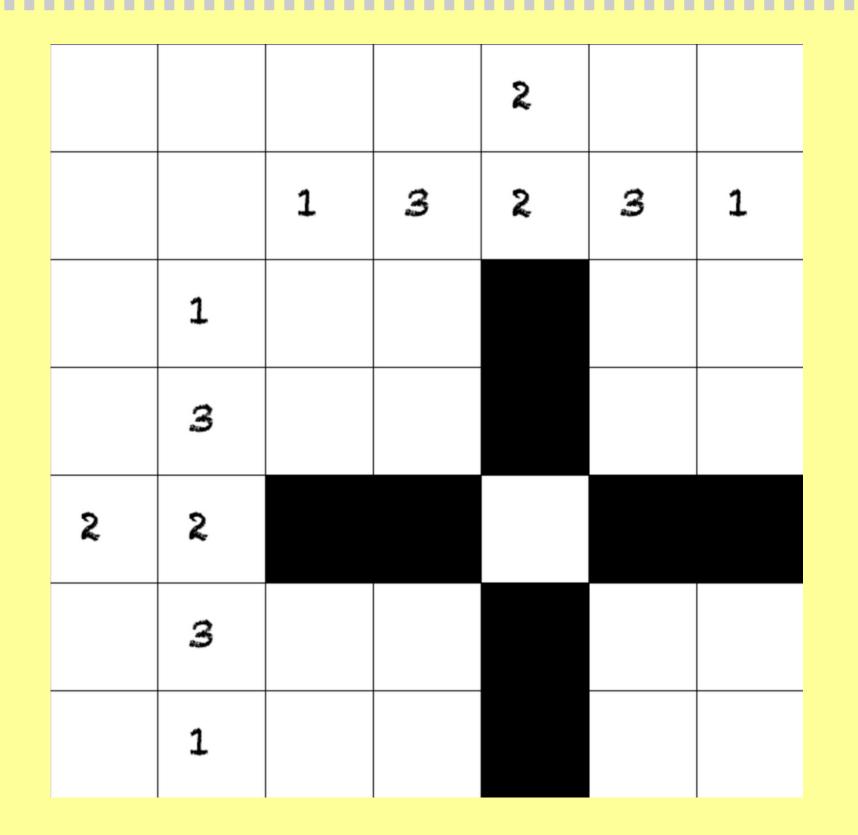
Creating all eligible row configurations by gradually increasing # of white spaces between runs till the last cell, calculating all combinations. The state space is built by the permutations. Check for node violation; if depicted, that node and its subtree are removed from the state space

Step 3:

The solution from created tree is identified by a DFS-style backtracking search algorithm. The leaf nodes contain the goal state (define the configurations for last row). The algorithm keeps # of column constraints, and evaluates combinations of current, next rows. Prunes any subtree violating col constraint.

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Constraint Satisfaction Problem

```
[['0'],
['0', '5'],
['0', '5', '8'],
['0', '5', '8', '9'],
['0', '5', '8', '9', '12'],
['0', '5', '8', '9', '13'],
['0', '5', '8', '9', '14'],
['0', '5', '8', '9', '16'],
['0', '5', '8', '10'],
['0', '5', '8', '10', '12'],
['0', '5', '8', '10', '13'],
['0', '5', '8', '10', '14'],
['0', '5', '8', '10', '15'],
['0', '5', '8', '10', '16'],
['0', '5', '8', '11', '12'],
['0', '5', '8', '11', '12'],
['0', '5', '8', '11', '13'],
['0', '5', '8', '11', '13'],
['0', '5', '8', '11', '14'],
['0', '5', '8', '11', '15'],
['0', '5', '8', '11', '15'],
['0', '5', '8', '11', '16'],
['0', '5', '8', '11', '16'],
['0', '5', '8', '11', '16'],
['0', '5', '8', '11', '16'],
['0', '5', '8', '11', '16'],
['0', '5', '8', '11', '16'],
```

```
sizeCol 5
sizeRow 5
colClues ,1, ,3,2,2, ,3, ,1
rowClues , ,2, ,1,3,2,3,1
```





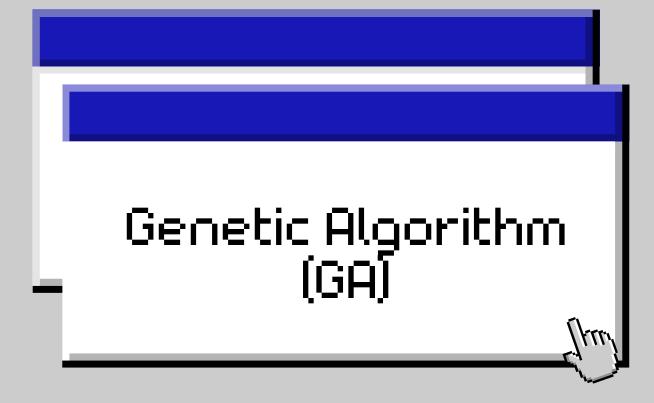
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METHODS

Constraint Satisfaction Problem (CSP)

Builds valid configurations of each row and tries to combine them to meet the column restrictions. Pre-processes the data to remove incorrect subtrees determined from previous iterations.



Generates a fitness function, starts running the algorithm on a random image, and optimizes it through the fitess function, takes into consideration the mutations and the cross over



Genetic Algorithm (GA)



Fitness Function

The fitness function or heuristic function determines how fit an individual is (the ability of an individual to compete with other individuals). It gives a fitness score to each individual.

The probability that an individual will be selected for reproduction is based on its fitness score.

Crossover

For each pair of parents to be mated, a crossover point is chosen at random from within the genes. Offsprings are created by exchanging the genes of parents among themselves until the crossover point is reached. The new offspring are added to the population.

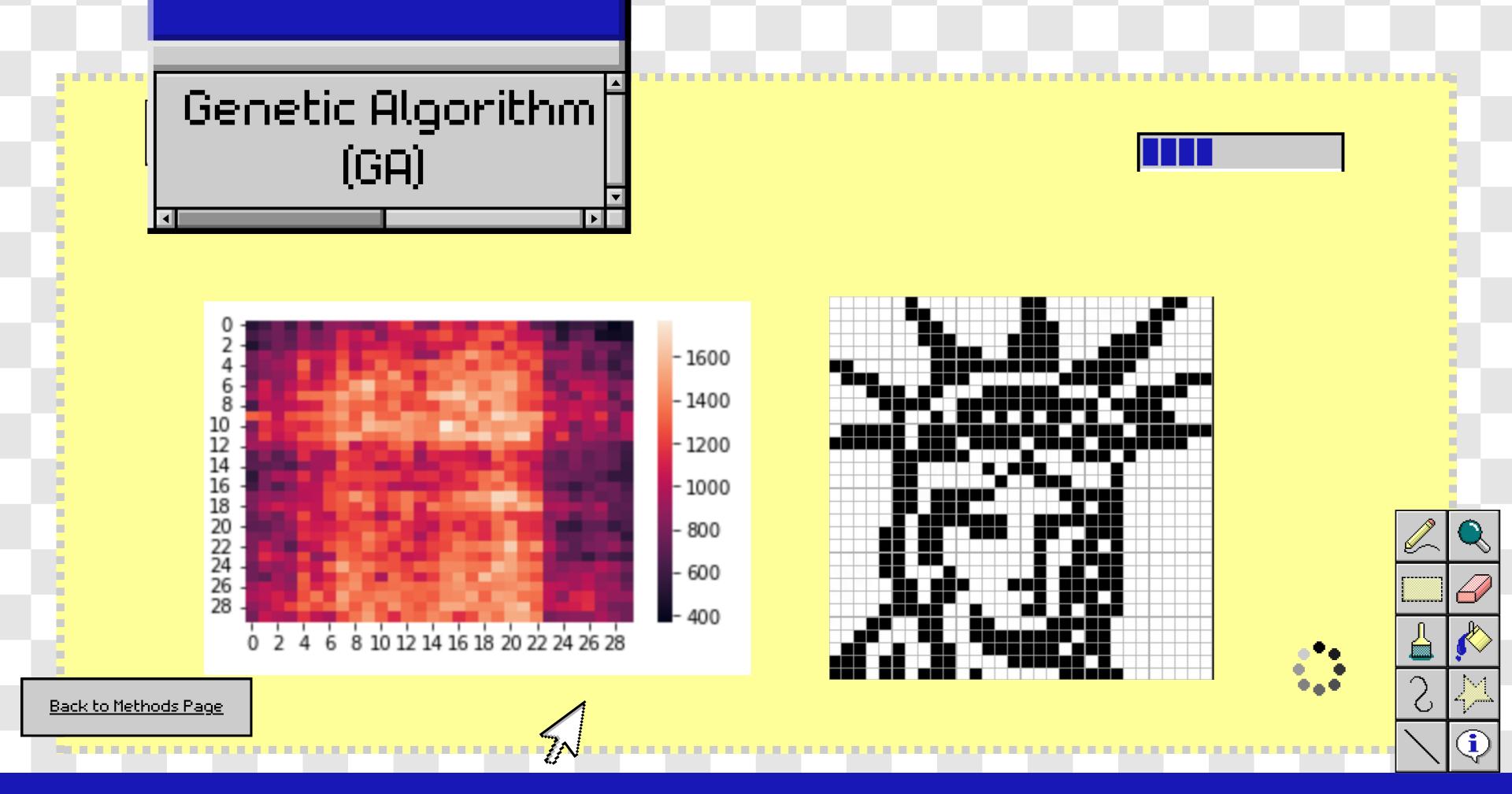
Mutation

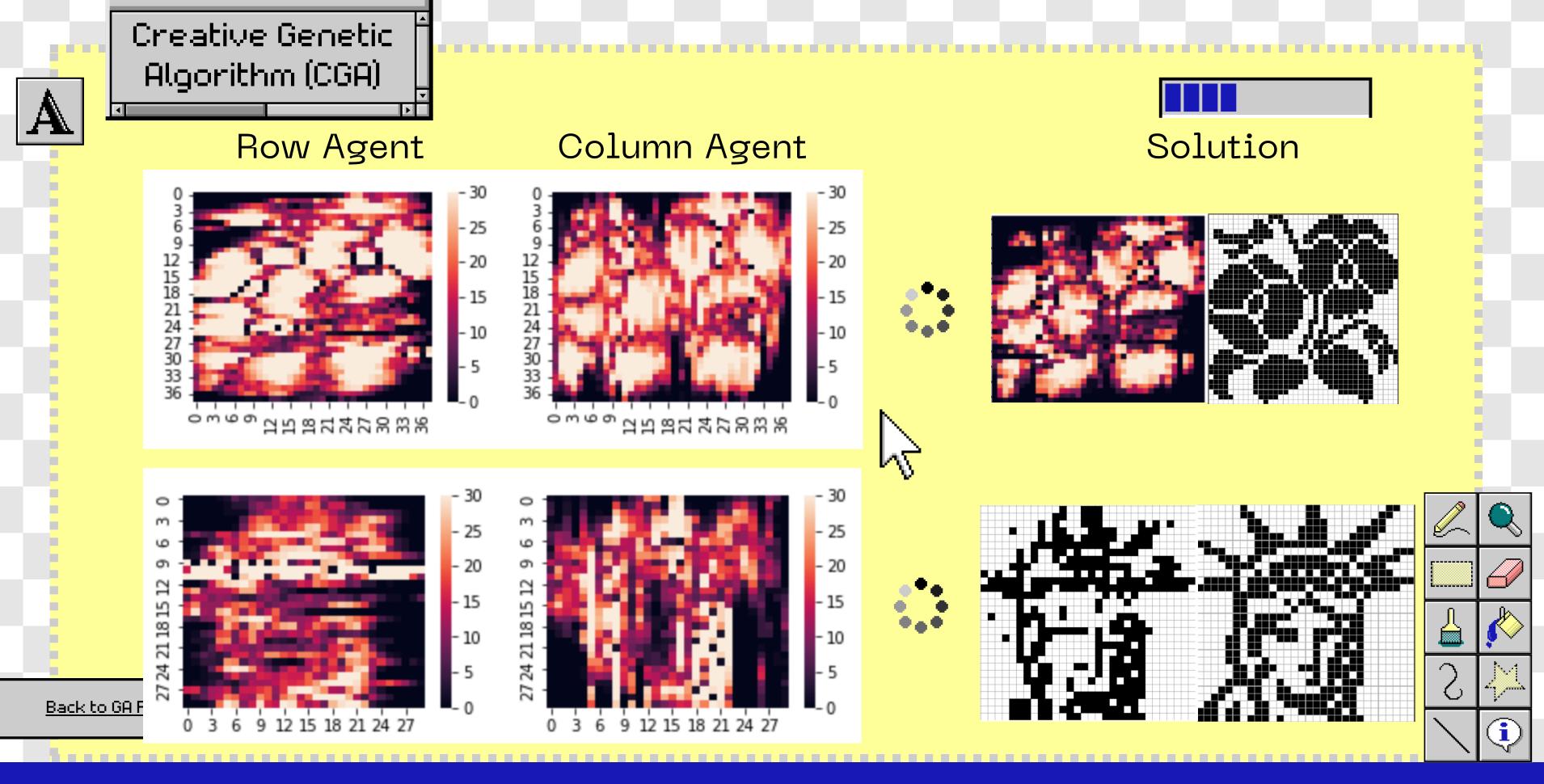
In certain new offspring formed, some of their genes can be subjected to a mutation with a low random probability. This implies that some of the bits in the bit string can be flipped.

Mutation occurs to maintain diversity within the population and prevent premature convergence.

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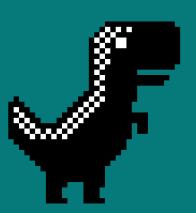






RESULTS







P1: 0.004364967346191406 | P2: 7.82012939453125e-05 | P3: 0.0014328956604003906

Total: 0.005876064300537109



Total: 136.0588545693465925











