

ARTIFICIAL INTELLIGENCE

Edge Matching Puzzle

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The Puzzle and the Challenge



- Published by **Christopher Monckton**.
- Released on July 2007.
- Offers \$ 2M for the first completed solution to the puzzle.
- No complete solution has been provided yet!!!
- Further details at <http://www.eternity-puzzle.com/>



Puzzle Overview

- Consists of 256 tiles.
- Each tile is colored with 4 different patterns.
- Tiles are to be placed on a 16X16 grid such that the edges of the adjacent tiles should match.



The Difficulties and Challenges

- Huge search space ($256! \times 4^{256}$ i.e. 1.15×10^{661} possibilities.)
- Recursive construction of the solution consumes lot of time.
- Problem of Recursive back-tracking.
- Effective branching factor for A* algorithm is about 381. compared to chess which has only about 100.
- No efficient algorithm can be designed - NP completeness.
- Successful attempts have matched as much as 396 to 459 edges out of 480.



Previous Works(The Recursive Back-Tracking and GA)

Doc Smiths C++ solver

- Uses recursive Back-Tracking algorithm
 - Looks for suitable tile to be placed.
 - If not found, Removes last placement and repeats.

Joel's JAVA implementation

- Uses same algorithm as Doc smith's solver

Results:(When run on a dual core 2.00Ghz Intel with 1 GB of RAM)

Code	6X6	8X8	10X10
Doc Smith	891msec	91mins	10+hrs
Joel	790msec	107mins	10+hrs

Munoz's Evolutionary Algorithm

- transformed the constraints into optimization objectives and used the optimization power of Genetic Algorithms

(We're mainly following his works and paper)



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The Genetic Algorithm

¹ Simulation of the process of evolution on Earth.

Steps Involved:

Initialization

- Creating bunch of "organisms" with unique set of "genes" i.e. generating a "generation".

Simulation:

- Every organism is allotted **fitness value** depending on some criterion.
- Best organisms from the generation chosen depending on fitness value.
- Mutation, Crossover, Reproduction.
- Next generation of organisms created.
- Search for the **Perfect organism**
 - If found **Return;**
 - if not found **Continue;**

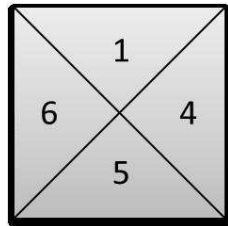
¹courtesy: <http://www-cs-students.stanford.edu/~jl/Essays/ga.html>



Applying Genetic algorithm

Tile Representation:

- Colour of the edge of the tile is replaced with specific numbers notations.



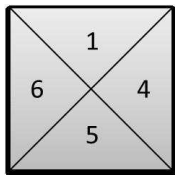
Applying Genetic algorithm

Tile Representation:

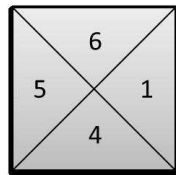
Rotations of tiles

Orientation	Rotation(clockwise)
1	original
2	90°
3	180°
4	270°

Example:



rotation



ori..
1 4 5 6 1

ori..
6 1 4 5 2



Terminologies:

- **Individuals**

Represented as 2D matrix with each entry as tile number and its orientation.

For ex 113:3 represents tile no 113 with rotation 3.

- Retains the properties just like genes in chromosomes.
- Properties like fitness are associated with Individuals.

- **Generation**

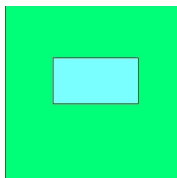
A set of certain number of individuals.

- Crossover, mutation, Elitism applied on one generation to get next one.

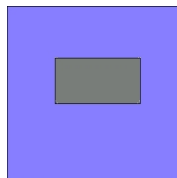


- Crossover

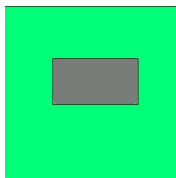
Copying of the properties of two parent individuals into the offspring.



Parent 1



Parent 2

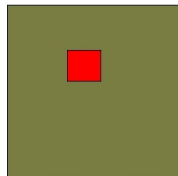
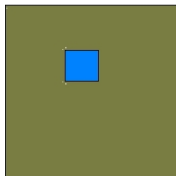


Child



- Mutation

Properties of the individual are modified to create better offspring.



Mutation can be carried out by rotating, swapping etc.

- Elitism

A certain no of best individuals are selected to retain their properties as it is for next generation.



- Fitness

The property of the individual to be optimized.

$$\text{Normal Fitness: } 1 - \left(\frac{\text{score}}{480} \right)$$

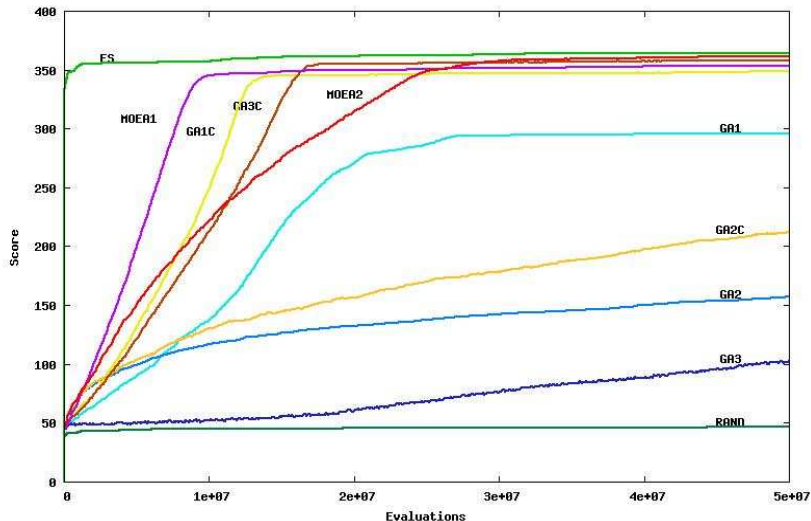
in case of multiple objectives:

$$\text{Combined Fitness: } 1 - \left(\frac{1}{k} \cdot \sum_{i=1}^k \frac{\text{objective}_i}{\text{Maxobjective}_i} \right)$$

where k is number of objectives



Results: Previous Works



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²courtesy: Paper in "Evolutionary Genetic Algorithms in a Constraint Satisfaction Problem: Puzzle Eternity II"



Problems faced

- Problem of incompleteness - Problem of Local maxima.
- optimization of the Genetic operations.
- Fitness Criteria.
- Selection criteria.



Results

Evaluatons	Elitism	Crossover	Mutation	BestFitness
1000000000	10	179	50	292
1000000000	10	179	50	273
100000000	0	1	0	48
100000000	0	0	1	90
100000000	1	0	0	NoCompl
100000000	1	1	0	159
100000000	1	1	1	262
100000000	1	1	2	268
100000000	1	2	1	254
100000000	2	1	1	257
100000	10	14	12	10
10	2	2	10	1
10000	2	2	10	2



- Evolutionary Genetic Algorithms in a Constraint Satisfaction Problem: Puzzle Eternity II

by "Jorge Munoz, German Gutierrez, and Araceli Sanchis", University Carlos III of Madrid Avda. de la Universidad 30, 28911 Legan's, Spain (2009)

- Solving the Eternity II Puzzle Using Evolutionary Computing Techniques

(A thesis) by "Papa Ousmane Niang", Concordia University, Montreal, Quebec, Canada (2010)

- Codes of Jorge Munoz

Codes that used for paper "Evolutionary Genetic Algorithms in a Constraint Satisfaction Problem: Puzzle Eternity II"

