Rubik's Cube Solver using Thistlewaiste's ES Strategy "An Evolutionary Approach for Solving"
"the Rubik-Cube Incorporating Exact Methods" by Nail El-Sourani, Sascha Hauke, and Markus
Borschbach, University of Applied Sciences, Germany Advised by Dr. Amitabha Mukherjee, IIT Kanpur

> study by 'Anurag Misra' IIT Kanpur, Uttar Pradesh

## Abstract

The research paper "An Evolutionary Approach for Solving the Rubik-Cube Incorporating Exact Methods" by Nail El-Sourani, Sascha Hauke, and Markus Borschbach, University of Applied Sciences, Germany applies group theory and evolutionary strategy to solve Rubik's Cube. Though previously too there have been methods to solve the Rubik's Cube, like Thistlewaiste's and Kociemba's and Rokicki's The nobility with their approach is that it doesnot require terabytes of lookup tables unlike the former algorithms. However, the integrity and success of their method is still needs to tested and verified.

## 1 1. Introduction

Rubik's Cube since introduced in 70's has been the most intriguing puzzles in the history of mankind. It is a challenging task. The size of the solution space because possible number of configurations of the Rubik's cube and an attempt to optimize the solution, i.e. lowest possible number of moves and lowest calculation complexity makes it a very interesting optimization problem. The least possible number of moves to solve the Rubik's Cube, also known as the God's number is however yet unknown. All the algorithms so far have focused on decreasing the upper bound. All the approaches so far are strictly exact methods and require terabytes of pre-calculated lookup

April 19, 2015

tables, reflected by current lowest upper bound of 22 moves(ROKICKI'S
 ALGORITHM).

Evolutionary or Genetic Algorithms have been applied in a large number 13 of problems, like the Traveling Salesman Problem, and sometime deliver 14 better superior solutions than the classical algorithms. This suggests the idea 15 of applying it to solve the Rubik's Cube. All relevant methods so far involve 16 partitioning the solution space into mathematical groups. Thistlewaiste's 17 uses 4, while Kociemba's and Rokicki's uses 2 subgroups. This makes their 18 approach precursor for developing Evolutionary Algorithms; interesting issue 19 is to do it without look-up tables. 20

## 21 2. Thistlewaiste's Algorithm

Among of the one of primary exact methods, it was chosen to work fur-22 ther on developing an Evolutionary Algorithm for solving the Rubik's Cube. 23 TWA divides the problem into four subproblems using the four nested groups: 24 G0=iF, R, U, B, L, D ¿, G1=i F,U,B,D,R2,L2¿, G2=i U,D,R2,L2,F2,B2¿, 25 G3= ;F2,R2,U2,B2,L2,D2;, G4=I i.e. the Solved State. The functional prin-26 ciple is to put the cube into a state such that it can thence be solved only 27 from moves in the sequences generated from G(i+1) group. The state is 28 however achieved by using moves from the G(i) group. Every stage is simply 29 a lookup table showing a transition sequence for each element in the current 30 coset space  $G_{i+1}/G_{i}$  to the next one (i=i+1). These coset spaces are re-31 finements of the earlier solution, limiting the possible configurations of the 32 Rubik's cube in each state. 33

34 EXCERPT TAKEN FROM THE RESEARCH PAPER An Evolutionary

<sup>35</sup> Approach for Solving the Rubiks Cube Incorporating Exact Methods: AL <sup>36</sup> GORITHM

The exact orders for each group are calculated as follows:  $G_0, |G_0| = 4.33 *$ 

 $_{38}$  10<sup>19</sup> represents the order of the Cube Group.

 $G_1$ : The first coset space G1/G0 contains all Cube states, where the edge orientation does not matter. This is due to the impossibility of flipping edge cubies when only using moves from G1. As there are  $2^{11}$  possible edge orientations, $|G_1/G_0| = 2^{11} = 2048$ 

43 The order of  $|G_1|$  is  $|G_1||G_0||G_1/G_0| = 2.11 * 10^{16}$ .

- (2)  $G_2$  Using only moves from  $G_2$ , no corner orientations can be altered (elim-
- <sup>45</sup> inating 3<sup>7</sup> states). Additionally, no edge cubies can be transported to or from
- the middle layer (eliminating 12!/(8!\*4!) states). The coset space  $G_2/G_1$  thus

<sup>47</sup> depicts a reduced puzzle of the order  $|G_2/G_1| = 3^7 * 12!/(8! * 4!) = 1082565$ <sup>48</sup> and  $|G_2||G_1|/|G_2/G_1| = 1.95 * 10^{10}$ 

<sup>49</sup> (4)  $G_3$ : Once in the coset space  $G_3/G_2$ , the Cube can be solved by only <sup>50</sup> using moves from  $G_3$ , here the edge cubies in the L, R layers can not trans-<sup>51</sup> fer to another layer (eliminating  $8!/(4!^*4!)^*2$  states) and corners are put <sup>52</sup> into their correct orbits, eliminating  $8!(4!^*4!)^*3$  states). Thus, $|G_3/G_2| =$ <sup>53</sup> (8!/(4!\*4!)) \* 2 \* 2 \* 3 = 29400 and  $|G_3||G_2|/|G_3/G_2| = 6.63 * 10^5$ .

 $G_4$  as  $G_4$  represents the solved state - obviously  $|G_4| = 1$  which means there 54 exist a mere  $|G_3|$  possible states for which a solution needs to be calculated to 55 transfer from  $G_4/G_3$  to solved state. Most essential to TWA are the groups 56  $G_1, G_2, G_3$  as  $G_0$  simply describing the Cube Group  $G_c$  and  $G_4$  the solved 57 state. To further exemplify how the coset spaces simplify the Rubiks Cube 58 puzzle the following may prove helpful. When looking at the constraints in-59 duced by  $G_2/G_1/G_0$  carefully (combining the constraints induced by  $G_2/G_1$ 60 and  $G_1/G_2$  it is essentially a Rubiks Cube with only 3 colors - counting 61 two opposing colors as one. This representation can be reached for each 62 distinct coset space by examining and applying its effect to the complete 63 Rubiks Cube puzzle. While solving the Rubiks Cube in a divide and conquer 64 manner, breaking it down into smaller problems (by generating groups and 65 coset spaces) is effective, there exists one major flaw. Final results obtained 66 by concatenating shortest subgroup solution do not necessarily lead to the 67 shortest solution, globally. 68

# <sup>69</sup> 3. The Thistlethwaite ES - An Evolution Strategy Based on the <sup>70</sup> Thistlethwaites Algorithm

In the classic TWA the order of each subproblem get reduced from stage
to stage. This algorithm present a 4-phase ES. Each phase here has a fitness
function.

A scrambled Cube is duplicated times and the main loop is entered with 74 a fitness function. Mutation sequences are generated using the group  $G_0$ 75 started using a fitness function  $phase_0$ . As soon as Cubes which solve  $phase_0$ 76 have been evolved, the phase transition begins. During phase transition, from 77 those  $phase_0$ -solving Cubes, a random Cube is chosen and duplicated. This 78 is repeated times and yields in the first population after the phase transition. 79 Now the phase-counter is increased by one, and the main ES loop is entered 80 again. This process is repeated until  $phase_4$  is solved (i.e.  $phase_5$  is reached), 81 presenting a solution sequence to the originally scrambled Cube. In order to avoid the TWES getting stuck in local optima an upper bound for calculated
generations is introduced. As soon as this upper bound is reached, the chain
is terminated,

Fitness Function translating the TWA into an appropriate Fitness Function for an Evolutionary Algorithm essentially forces the design of four distinct sub-functions. As each subgroup of  $G_0$  has different constraints, custom methods to satisfy these constraints are proposed.

EXCERPT FROM THE RESEARCH PAPER: An Evolutionary Approach 90 for Solving the Rubiks Cube Incorporating Exact Methods: ALGORITHM 91  $G_0$  to  $G_1$  To reach  $G_1$  from any scrambled Cube, we have to orient all 92 edge pieces right while ignoring their position. The fitness function for this 93 phase simply increases the variable  $phase_0$  by 2 for each wrong oriented edge. 94 Furthermore, we add the number of moves that have already been applied to 95 the particular individual in order to promote shorter solutions. Finally, we 96 adjust the weight between w(number of wrong oriented edges) and c(number 97 of moves applied to current Cube individual). This will be done similarly in 98 all subsequent phases. 99

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<sup>101</sup>  $phase_0 = 5 * (2w) + c$  With a total of 12 edges which can all have the <sup>102</sup> wrong orientation this gives max2w= 24. The Cube has been successfully <sup>103</sup> put into  $G_1$  when  $phase_0 = c$ .Reaching  $G_1$  is fairly easy to accomplish, thus <sup>104</sup> making the weight-factor 5 a good choice.

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 $G_1$  to  $G_2$  In order to fulfill  $G_2$  the 8 corners have to be oriented correctly. 106 Edges that belong in the middle layer get transferred there. Tests with the 107 Thistleth-waite ES showed it somewhat problematic to do this in one step. 108 Oftentimes, the algorithm would get stuck in local optima. To solve this, the 109 process of transferring a Cube from  $G_1$  to  $G_2$  has been divided into two parts. 110 First, edges that belong into the middle layer are transferred there. Second, 111 the corners are oriented the right way. The first part is fairly easy and the 112 fitness function is similar to that from  $phase_0$  except for w(number of wrong 113 positioned edges), i.e. edges that should be in the middle layer but are not. 114  $phase_1 = 5(2w) + c$ . In the second part, for each wrong positioned corner, 4 115 penalty points are as-signed as they are more complex to correct than edges. 116 Obviously, in order to put the Cube from  $G_1$  to  $G_2$  both phases described 117 here have to be fulfilled, which yields:  $phase_2 = 10(4v) + phase_1$  where v 118 represents the number of wrong oriented corners. The weighing factor is in-119 creased from 5 to 10 to promote a successful transformation into  $G_2$  over a 120

<sup>121</sup> short sequence of moves.

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 $G_2$  to  $G_3$  We now have to put the remaining 8 edges in their correct orbit. The same is done for the 8 corners which also need to be aligned the right way. Thus, the colors of two adjacent corners in one circuit have to match on two faces. In  $G_3$  the Cube will only have opposite colors on each face. Let x (number of wrong colored facelets) and y (number of wrong aligned corners), then  $phase_3 = 5(x + 2y) + c$ .

<sup>129</sup> An Evolutionary Approach for Solving the Rubiks Cube  $G_3$  to  $G_4(solved)$ <sup>130</sup> The Cube can now be solved by only using half-turns. For the fitness function <sup>131</sup> we simply count wrong colored facelets. Let z be the number of wrong colored <sup>132</sup> facelets, then  $phase_4 = 5z + c$ .

To summarize, 5 different fitness functions are needed for the Thistlethwaite 133 ES.  $phase_i$  is solved if  $phase_i = c$ , i=0, ..., 4 and with the properties of nested 134 groups we can conclude, given the above, a solved Cube implies:  $phase_i = c$ . 135 Fulfilling the above equation satisfies the constraints induced by the groups 136 G0,...,G4, with the final fitness value c describing the final solution sequence 137 length. The maximum sequence length (s) needed to transform the Cube 138 from one subgroup to another is given by Thistlethwaite. Those lengths are 139 7,13,15,17 (the sum of which is 52, hence 52 Move Strategy). 140

### <sup>141</sup> 4. Conclusion

Verification of the claims of the testbenchs and results given in the research paper is still under process. No result can therefore be quoted hencefar. We expect however, the algorithm delivers as promised. Work is currently under progress. Code would be uploaded on github, and link shared.

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