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Harnessing parallel disks to solve Rubik's cube[☆]

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ABSTRACT

The number of moves required to solve any configuration of Rubik's cube has held a fascination for over 25 years. A new upper bound of 26 is produced. More important, a new methodology is described for finding upper bounds. The novelty is two-fold. First, parallel disks are employed. This allows 1.4×10^{12} states representing *symmetrized cosets* to be enumerated in seven terabytes. Second, a faster table-based multiplication is described for symmetrized cosets that attempts to keep most tables in the CPU cache. This enables the product of a symmetrized coset by a generator at a rate of 10 million moves per second.

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1. Introduction

Twenty-five years ago, [Frey and Singmaster \(1982\)](#) conjectured at the end their book, *Cubik Math*, that “God's number” is in the low 20's:

No one knows how many moves would be needed for “God's Algorithm” assuming he always used the fewest moves required to restore the cube. It has been proven that some patterns must exist that require at least seventeen moves to restore but no one knows what those patterns may be. Experienced group theorists have conjectured that the smallest number of moves which would be sufficient to restore any scrambled pattern – that is, the number of moves required for “God's Algorithm” – is probably in the low twenties.

This conjecture remains unproven today. At the time of this conjecture, the best known bounds were a lower bound of 17 and an upper bound of 52 ([Frey and Singmaster, 1982](#)). The current best lower bound is 20 ([Reid, 1995b](#)). In [Kunkle and Cooperman \(2007\)](#), the authors demonstrated a new upper bound of 26.

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Note that in all cases, we consider a *move* to be any quarter or half turn of a face of the cube, also known as the *face-turn metric*. We do not consider the alternative *quarter-turn metric*, which defines a half-turn to be two moves.

We present a new, algebraic approach based on the analysis of cosets for mathematical groups. Rubik's cube can be viewed as a mathematical group with each group element a permutation of the facelets of Rubik's cube. A *facelet* is one of the 54 tiles, with 9 facelets per face. The 18 natural moves are taken as the generators of the group. Since the generators always fix in place the center facelet of a face, Rubik's cube can be viewed as a permutation group acting on the 48 movable facelets. Multiplication of moves or generators is simply composition of moves. Hence, multiplication is identified with permutation multiplication for permutation groups.

Like previous efforts, we decompose Rubik's group G using a subgroup H . As is well-known in group theory, the elements of Rubik's group G are partitioned into *cosets* $Hg = \{hg : h \in H\}$, where $g \in G$. Hence, there are $|G|/|H|$ cosets. The set of cosets of H in G is denoted as G/H .

This allows a divide-and-conquer strategy. The number of configurations of Rubik's cube is $|G| \approx 4.3 \times 10^{19}$. Given an unsolved state or permutation $g \in G$, one can identify the coset Hg containing g . One searches for a shortest sequence of generators τ such that the product $g\tau \in H$. One then searches for a shortest sequence of generators of H , σ , such that $g\tau\sigma$ is the identity permutation (the home position). After finding upper bounds on the length of τ and on the length of σ , the sum of the two is then an upper bound for the number of moves to solve Rubik's cube. Traditionally, one chooses a particular subgroup R , such that $|R| \approx 2 \times 10^{10}$ and $|G|/|R| \approx 2.2 \times 10^9$. This makes the search feasible in RAM for both problems (after the number of states is reduced by symmetries).

Unlike previous efforts, we choose the *square subgroup* Q , the group generated by squares of generators (or 180° turns), as our subgroup. Since $|Q| \approx 6.6 \times 10^5$, $|G|/|Q| \approx 6.5 \times 10^{13}$ (or 1.4×10^{12} after reduction by symmetries).

1.1. Paper organization

The paper is organized as follows. In the rest of the introduction, we provide an overview of the three primary contributions of this work: a new upper bound on solutions to Rubik's Cube; the use of parallel disk-based computation; and methods for producing small, fast tables for group multiplication. Then, Section 2 briefly reviews some related work in these areas. Section 3 presents background and some basic concepts. In particular, this includes the definitions of symmetrized group element and symmetrized coset. Section 4 describes the fast multiplication algorithm, along with the perfect hash function. Section 5 shows that all elements of the square subgroup are solvable in 13 moves. Section 6 shows that all cosets are within 16 moves of the trivial coset. Finally, Section 7 presents methods for further reducing the upper bound on solutions, providing the final upper bound of 26 moves.

1.2. Upper bounds

We choose this very small subgroup for three reasons: in the limit, with a subgroup equal to the identity of the entire group, this method would produce exactly optimal results; the square subgroup is the only non-trivial subgroup that preserves all 48 symmetries of the cube; and, the use of a small subgroup allows us to more efficiently prove nearly optimal upper bounds on any coset (see Section 7).

The search for the worst case in G/Q produces 17 symmetrized cosets (equivalence classes of cosets under symmetries of the cube) that each require 16 moves to return to the subgroup Q . Any element in the subgroup Q can be "solved", using only generators of Q , in 15 moves. However, by allowing one to use any generators of Rubik's group G , an easy computation shows that 13 moves always suffice to solve for any element of Q . This immediately produces a bound of $16 + 13 = 29$ moves.

We further reduce the upper bound from 29 to 26 in two steps. First, a refinement strategy is used to analyze all group elements that are members of a symmetrized coset at level 9. The refinement computation shows that the optimal solution for any such group element is bounded above by only 20 moves, instead of the expected $13 + 9 = 22$ moves, giving elements at the furthest level a bound of $(16 - 9) + 20 = 27$. Finally, that bound of 27 is reduced to 26 by directly analyzing the 17 worst case symmetrized cosets.

1.3. Parallel disks

We briefly review the use of the parallel disks in proving that all cosets are within 16 moves of the square subgroup. The computation was parallelized using a library developed by the second author called TOP-C (Task Oriented Parallel C/C++) (Cooperman, 1996). Initially, the computation was carried out in 63 cluster hours using a high-end SAN (storage area network). Later, we duplicated the computation in 183 cluster hours using only the local disks of a commodity cluster.

To find bounds on the length of solutions among the symmetrized cosets, we use a parallel disk-based version of breadth-first search. A standard algorithm maintains a current *frontier* corresponding to all states at a given level ℓ . For each state of the frontier, one makes all moves (applies all generators) to produce a list of neighbors. Those neighbors that have already been seen at level ℓ or $\ell - 1$ are eliminated. The remaining neighbors form the next frontier at level $\ell + 1$.

The primary difficulty with creating a disk-based version of this algorithm is in the duplicate detection phase, where we must compare newly generated states to those previously seen. This typically makes use of random access data structures, such as hash tables, which are not efficient on disk. To solve this problem we delay duplicate detection and use a method similar to that of bucket sort, essentially breaking the task into RAM-sized pieces that can make use of random access efficiently.

A second issue arises when the search frontier exceeds the capacity of disk, as it does in our case. We solve this problem using an *implicit open list*. This method works by encoding the search frontier in the hash table and reconstructing it with an inverse hash function, instead of explicitly saving the non-duplicate states.

Further background on disk-based computation can be found in Section 2, and details of our method can be found in Section 6.

1.4. Small fast multiplication tables

One additional technique we employ is a fast multiplication routine that accepts the hash index of a symmetrized coset and a generator, and directly produces the hash index of the product. This allows us to avoid the costly step of working with the original representation of a symmetrized coset. While the fast multiplication has some complexity, it is conceptually simple, and based on the group theoretic decomposition of a group G into a subgroup H such that H further decomposes into a product QN , for N a normal subgroup of H . The definition of normal subgroup is given in Section 3.1 and further details of the fast multiplication are provided in Section 4.

1.5. Extensions to previous work

This paper is an extension of the work presented in Kunkle and Cooperman (2007), which originally proved 26 moves sufficient for Rubik's cube. Along with more general enhancements, we provide three specific contributions over that previous work.

First, we provide a more thorough analysis of the square subgroup, including the distribution of depths of the elements and a visual representation of the group structure.

Second, we include additional experimental results for the disk-based computation in Section 6. These new results compare the results using two different cluster computing architectures, using either global shared disk or locally attached disks. We find that the relatively cheap local disk architecture can be used to efficiently perform computations that previously were performed on a much larger supercomputer.

Finally, we present an analysis of three techniques for refining the upper bound on the group radius, including: optimal solvers; image intersection; and projection. We also provide experimental results detailing additional computations that correct an omission in the previous result, which did not provide shortened solutions for all of the necessary group elements.

2. Related work

One approach to finding bounds on solutions to Rubik's cube would be to produce the entire Cayley graph for the corresponding group. This approach was used to show that 11 moves suffice for Rubik's

$2 \times 2 \times 2$ cube (Isaacs, 1981). This smaller cube problem has also proven useful for the testing of newer methods, such as a graph representation using only two bits per element, described in Cooperman et al. (1990). For the full $3 \times 3 \times 3$ Rubik's cube, these methods are not feasible, since it has over 4.3×10^{19} states.

The first lower bound, of 17, was shown using a simple counting argument, i.e. the number of possible states achievable in 16 moves is less than the total number of possible states, so there must be some states requiring at least 17 moves (Frey and Singmaster, 1982).

It was then conjectured that a specific cube position, *super-flip*, would be a position requiring a near-maximum number of moves to solve. The super-flip position is of interest because it is the only element, other than the identity, in the center of Rubik's group. A solution of length 20 was found by Winter (1992), which was later proven optimal by Reid (1995b).

The first published upper bound was 52. Discovered by Thistlethwaite (Frey and Singmaster, 1982), it was based on solving the cube in a series of four steps, corresponding to a chain of subgroups of length four. The four steps were proven to have worst case lengths of 7, 13, 15, and 17, for the total of 52.

This algorithm was improved by Kociemba (2007) to use a subgroup chain of length two. Reid (1995a) proved the worst case for the two steps was 12 and 18, for a total upper bound of 30. Further analysis showed that the worst case never occurs, and so a bound of 29 was shown. This bound was further refined by Radu (2006) to 27, which was the best upper bound before our working showing 26 moves suffice (Kunkle and Cooperman, 2007). Recently, between the acceptance and publication of this paper, Rokicki (2008b) reduced the upper bound to 23 using 7.8 core-years of CPU time, an extension of the method described in Rokicki (2008a).

Besides work into methods with provable worst cases, several optimal solvers with no worst case analysis have been developed. The method developed by Kociemba and analyzed by Reid has a natural extension that guarantees optimal solutions. Korf (1997) used similar techniques to optimally solve ten random cube states, one in 16 moves, three in 17 moves, and the remaining six in 18 moves.

One key to our result is the use of disk-based methods for performing search and enumeration, and specifically the disk-based breadth-first search described in Section 6. Several disk-based search methods have been introduced in recent years. The common difficulty each of these methods must overcome is the significant latency of disk, which disallows the random access patterns typically used when detecting duplicate states.

Korf (2004) used sorting-based delayed duplicate detection to solve sliding tile puzzle and Towers of Hanoi type problems. Korf and Schultze (2005) also introduced hash-based delayed duplicate detection, which can be used to avoid external sorting in some applications (we use a hash-based method for our computation). Zhou and Hansen (2004) introduced structured duplicate detection, which can utilize disk without delaying the detection of duplicates. Robinson and Cooperman (2006) introduced tiered duplicate detection as a method to speedup the enumeration of the Baby Monster sporadic simple group, and more recently applied it to the problem of the Fischer₂₃ group (Robinson et al., 2007b). A comparative analysis of each of these methods, among other search techniques, can be found in Robinson et al. (2007a).

3. Notation and basic concepts

3.1. Group theory definitions

We review the formal mathematical definitions. Recall that a group G is a set with multiplication and an identity e ($eg = ge = g$), inverse ($gg^{-1} = g^{-1}g = e$), and an associative law ($(gh)k = g(hk)$). A permutation of a set Ω is a one-to-one and onto mapping from Ω to Ω . Composition of mappings provides the group multiplication, and the group inverse is the inverse mapping. A permutation group G is a subset of the permutations of a set Ω with the above operations. A subgroup $H < G$ is a subset H that is closed under group operations. A group G has generators $S \subseteq G$, written $G = \langle S \rangle$, if any element of G can be written as a product of elements of S and their inverses. The order of the group is the number of elements in it, $|G|$.

Edge Tables	Size	Inputs	Output
Table Aut	1564 × 18 × 1B	$r_{1,e}, s$	$\alpha \in A$ for $\overline{\alpha(r_1s)}$ a canonical coset rep. of H in E (We choose α such that $\overline{\alpha(r_1s)} = \min_{\beta \in A} \overline{\beta(r_1s)}$.)
Table 1a (coset rep.)	1564 × 18 × 2B	$r_{1,e}, s$	$H\overline{\alpha(r_1s)} \in E/H$ for α defined in terms of r_1 and s by Table Aut. (Note that $H^A = H$.)
Table 1b (N)	1564 × 18 × 2B	$r_{1,e}, s$	$\bar{r}_2 \stackrel{\text{def}}{=} n' \overline{\alpha(r_1s)} \in N$, where $h' \stackrel{\text{def}}{=} \alpha(r_1s) \overline{\alpha(r_1s)}^{-1} \in H$ and $h' = \bar{q}' n'$ for $\bar{q}' \in Q, n' \in N$ for α defined in terms of r_1 and s by Table Aut
Table 2	2048 × 18 × 2B	$r_{2,e}, s$	$r_2^s \in N$
Table 5	2048 × 48 × 2B	$n_e \in N, \alpha$	$\alpha(n) \in N$, where α is the output of Table Aut n is defined by $n = r_2^s$ (output of Table 2 for edges)
Logical op's		$r_{2,e}, r'_{2,e}$	$r_2 r'_2 \in N$ (using addition mod 2 on packed fields)

Fig. 2. Edge tables for fast multiplication of symmetrized coset by generator.

Corner Tables	Size	Inputs	Output
Table 1a	420 × 18 × 2B	$r_{1,c}, s$	$H\bar{r}_1s \in C/H$ for \bar{r}_1s a canonical rep. of a coset of C/H
Table 1b	420 × 18 × 2B	$r_{1,c}, s$	$\bar{r}_2 \stackrel{\text{def}}{=} n' \overline{\alpha(r_1s)} \in N$, where n' is defined by setting $h \stackrel{\text{def}}{=} r_{1s} \bar{r}_1s^{-1} \in H$ and uniquely factoring $h = \bar{q}n'$ for $\bar{q} \in Q, n' \in N$
Table 2	2187 × 18 × 2B	$r_{2,c}, s$	$r_2^s \in N$
Table 4a (coset rep.)	420 × 48 × 2B	$H\bar{r}_1, c\bar{s} \in C/H, \alpha$	$H\overline{\alpha(\bar{r}_1s)} \in C/H$, where $H\bar{r}_1s$ is the output of Table 1a, and α is the output of Table Aut on edges
Table 4b (N)	420 × 48 × 2B	$H\bar{r}_1, c\bar{s} \in C/H, \alpha$	$n^{\alpha(\bar{r}_1s)} \in N$, where $H\bar{r}_1s$ is the output of Table 1a, and n is defined by setting $h = \alpha(\bar{r}_1s) \overline{\alpha(\bar{r}_1s)}^{-1} \in H$, and uniquely factoring h into qn for $q \in Q, n \in N$
Table 5	2187 × 48 × 2B	$n_c \in N, \alpha$	$\alpha(n) \in N$, where α is the output of Table Aut on edges, and n defined by computing $\bar{r}_2 = n' \overline{\alpha(r_1s)}$ (as in Table 1b), and r_2^s computed as in Table 2, and $\bar{r}_2 r_2^s$ computed by logical op's on corners
Logical op's		$r_{2,c}, r'_{2,c}$	$r_2 r'_2 \in N$ (using addition mod 3 on packed fields)

Fig. 3. Corner tables for fast multiplication of symmetrized coset by generator.

where α is chosen as in Eq. (4), \bar{r}_1s defined by Table 1a, $\overline{\alpha(\bar{r}_1s)}$ defined by Table 4a,

$$n^{\alpha(\bar{r}_1s)} \text{ defined by Table 4b, } \bar{r}_2 \text{ defined by Table 1b, } r_2^s \text{ by Table 2, etc.} \tag{5}$$

Note that the choice of $\alpha \in A$ for edges above depends on there being a unique such automorphism that minimizes $\overline{\alpha(r_1s)}$. In fact, this is not true in about 5.2% of cases for randomly chosen r_1 and s . These unusual cases can be easily detected at run-time, and additional tie-breaking logic is generated. We proceed to describe tables for fast multiplication for the common case of unique $\alpha \in A$ minimizing $\overline{\alpha(r_1s)}$, and discuss the tie-breaking logic later.

The tables that implement the above formulas follow. While it is mathematically true that we can simplify $\overline{\alpha(\bar{r}_1s)}$ into $\overline{\alpha(r_1s)}$, we often maintain the longer formula to make clear the origins of that expression, which is needed for an implementation. As before, the subscripts e and c indicate the restriction of a permutation to its action only on edges and only on corners. Figs. 2 and 3 describe the following edge tables, among others.

Ideally, one would use only the simpler formula and tables for edges, and copy that logic for corners. Unfortunately, this is not possible. We must choose a representative automorphism $\alpha \in A$ for purposes of computation. We choose α based on the projection $r_{1,e}$ of r_1 into E (action of r_1 on edges). Hence, Tables 1a and 1b for edges take input r_1 and s , then compute α as an intermediate

Edge Tables	Size	Inputs	Output
Table Mult Aut	48 × 48 × 1B	α, β	the product $\alpha\beta \in A$
Table 1c (A)	1564 × 18 × 1B	$r_{1,e}, s$	$\{\beta \in A : \overline{\beta(\alpha(r_1s))} = \overline{\alpha(r_1s)}\}$
Table 3 (N)	2048 × 48 × 2B	$H\overline{r_{1,c}s} \in C/H, \alpha$	$\overline{r'_2} \stackrel{\text{def}}{=} n''\beta(r_1) \in N$, where β taken from Table 1c, and where $h'' \stackrel{\text{def}}{=} \beta(r_1)\overline{\beta(r_1)}^{-1} \in H$, and $h'' = \overline{q''n''}$, for $q'' \in Q, n'' \in N$

Fig. 4. Edge tables for fast multiplication of symmetrized coset by generator, adjusted to break ties.

computation, then return $H\overline{\alpha(r_1s)}$. A similar computation for corners is not possible, because the intermediate value α depends on $r_{1,e}$ and not on the corresponding element of the corner group $r_{1,c}$.

4.4.0.3. *Tie-breakers: When the minimizing automorphism is not unique.* Table Aut in the previous table for edges defines an automorphism α that minimizes $\overline{\alpha(r_1s)}$. Unfortunately, there is not always a unique such α . In such cases, one needs a tie-breaker, since different choices of α will in general produce different encodings (different hash indices).

For each possible value of $\overline{\alpha(r_1s)}$, with α chosen to minimize the expression, we precompute the stabilizer subgroup $B \leq A$ defined by $B = \{\beta \in A : \overline{\beta(\alpha(r_1s))} = \overline{\alpha(r_1s)}\}$ and use the formulas and additional table below to find the unique $\beta \in B$ such that the product $\alpha\beta$ minimizes the edge pair result $(r'_{1,e}, r'_{2,e})$. Where even this is not enough to break ties, we compute the full encoding, while trying all possible tying automorphisms. This latter situation arises only 0.23% of the time, and does not contribute significantly to the time. The tables of Fig. 4 suffice for these computations.

For edges,

$$Q\beta(\alpha(r_{1,e}r_{2,e}s)) = Q\beta(\alpha(r_1s))\beta(\alpha(r_2^s)) = Q\beta(\overline{\alpha(r_1s)}) (\beta (\overline{r_2}\alpha(r_2^s))) = Q\overline{\alpha(r_1s)}\overline{r'_2} (\beta (\overline{r_2}\alpha(r_2^s))),$$

where $\overline{\alpha(r_1s)}$ is defined by Table 1a, α is chosen to minimize $\overline{\alpha(r_1s)}$,

$$\beta \in A \text{ satisfies } Q\beta(\overline{\alpha(r_1s)}) = Q\overline{\alpha(r_1s)}, \beta(\overline{\alpha(r_1s)}) = \overline{\alpha(r_1s)}\overline{r'_2} (r'_2 \text{ defined in Table 3}), \text{ and } \overline{r_2} \text{ defined by Table 1b for edges.} \tag{6}$$

However for corners,

$$Q\beta(\alpha(r_{1,c}r_{2,c}s)) = Q\beta(\alpha(\overline{r_1s}(\overline{r_2}(r_2^s)))) = Q\beta(\alpha(\overline{r_1s}))\beta(\alpha(\overline{r_2}(r_2^s))) = Q\overline{\alpha(\overline{r_1s})}n^{\beta(\overline{\alpha(\overline{r_1s})})}\beta(\alpha(\overline{r_2}(r_2^s))), \text{ where } \alpha \text{ and } \beta \text{ are chosen as in Eq. (6), and other quantities based on the previous Corner Tables using } \alpha\beta. \tag{7}$$

Table 1c is implemented more efficiently by storing the elements of each of the possible 98 subgroups of the automorphism group, and having Table 1c point to the appropriate subgroup $B \leq A$, stabilizing $r_{1,e}, s$.

4.5. Optimizations

In the discussion so far, we produce the encoding or hash index of a group element based on an encoding of the action of the group element on edges, along with an encoding of the action of the group element on corners. We can cut this encoding in half due to parity considerations.

Consider the action of Rubik’s cube on the 12 edge cubies and the 8 corner cubies, rather than on the facelets. We define the *edge parity* of a group element to be the parity (even or odd) in its action on edge cubies. (Recall that the parity of a permutation is odd or even according to whether the permutation is expressible as an odd or even number of transpositions.) The *corner parity* is similarly defined.

The edge and corner parity of a symmetrized coset, Hg^A , are well-defined, and are the same as the edge and corner parity of g . This is so because $H = QN$, and elements of Q and N have even edge parity and even corner parity. Parity is unchanged by the action of an automorphism.

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Square Generators			All Generators		
Dist.	Sym. Elts.	Elts.	Dist.	Sym. Elts.	Elts.
0	1	1	0	1	1
1	1	6	1	1	6
2	2	27	2	2	27
3	5	120	3	5	120
4	18	519	4	18	519
5	56	1932	5	62	2124
6	162	6484	6	214	8188
7	482	20310	7	693	27636
8	1258	55034	8	1871	78644
9	2627	113892	9	4093	174521
10	4094	178495	10	5394	233504
11	4137	179196	11	2774	116010
12	2231	89728	12	620	22228
13	548	16176	13	4	24
14	114	1488			
15	16	144			
Total	15752	663552	Total	15752	663552

Fig. 6. Distribution of elements in the square subgroup.

backwards searches required anywhere from milliseconds to a few hours in the worst case. Overall, this optimization took less than one day, requiring no parallelization.

This showed that all elements of the square subgroup have a solution of 13 or fewer moves. Fig. 6 shows the distribution of distances of (symmetrized) elements in the square subgroup, using either just the square generators or the full set of generators.

6. Cosets are within 16 moves of the trivial coset

We constructed the symmetrized Schreier coset graph, with respect to the square subgroup, through breadth-first search. However, the computation is significantly more complex than typical breadth-first search due to the scale of the computation: approximately 1.4 trillion symmetrized cosets. To handle this large scale computation we developed fast multiplication of symmetrized cosets by generators (see Section 4), and developed new methods for disk-based search and enumeration.

Robinson et al. (2007a) presents a comparative analysis of several different methods for parallel disk-based search and enumeration. The two methods from that comparison which we use here are *hash-based delayed duplicate detection* (hash-based DDD) with an *implicit open list*.

The primary data structure is an almost perfectly dense hash array of approximately 1.5×10^{12} entries, corresponding to the range of hash indices for all symmetrized cosets. The hash function is provided by our method of fast multiplication. Each entry of the table holds a four bit value describing the depth at which the corresponding symmetrized coset occurs in the breadth-first search, for a total array size of 685 GB.

Note that we could have used only two bits per state, encoding which of the states are on the current search frontier, instead of representing the exact depth of each element. We chose to use the more detailed representation to allow for efficient post-processing of the data (such as extracting all of the states at a given level).

Algorithm 1 describes how this array is used to perform the search. It uses an iterative process with two phases, *generating* and *merging*.

The generating phase produces new states, buffering them to disk, and continues until either an entire level of the breadth-first search is complete, or until available disk space is exhausted. The merge phase reads the buffered states, and uses the hash values to do duplicate detection in RAM.

Algorithm 1 Construct Symmetrized Schreier Coset Graph

```

1: Initialize array of symmetrized cosets with all levels set to unknown (four bits per coset).
2: Add trivial coset to array; set level  $\ell$  to 0.
3: while previous level had produced new neighbors, at next level do
4:   {Generate new elements from the current level}
5:   Let a segment be those nodes at level  $\ell$  among  $N$  consecutive elements of the array.
6:   Scan array starting at beginning.
7:   while we are not at the end of the array, extract next segment of array and do
8:     for each node at level  $\ell$  (representing a symmetrized coset) do
9:       for each generator do
10:        Compute product by fast multiplication.
11:        Compute hash index of product.
12:        Save hash index in bucket  $b$ , where  $b$  is the high bits of the hash index. Note, we only save
           the low order bits of the hash index not encoded by the bucket number. (This value fits
           in four bytes.)
13:        If bucket  $b$  is full, transfer it (write it) to a disk file for bucket  $b$ .
14:      end for
15:    end for
16:    Transfer all buckets to corresponding disk files.
17:  end while
18:  {Now merge buckets into array of symmetrized cosets.}
19:  for each bucket  $b$  on disk do
20:    Load portion of level array corresponding to bucket  $b$  into main memory.
21:    for each buffered element on disk for this bucket do
22:      Read value into RAM and delete from disk (in large chunks).
23:      Look up corresponding level value in array.
24:      If a value already exists for the element, it is a duplicate. Otherwise, set its level to  $\ell$ .
25:    end for
26:    Write portion of level array back to disk.
27:  end for
28:  Increment level  $\ell$ .
29: end while

```

We used up to 7 terabytes of storage at any given time, as a buffer for newly generated states in the breadth-first search.

The fast multiplication algorithm allowed us to multiply a symmetrized coset by a generator at a rate of approximately 5 million times per second.

The computation using shared disk required 63 cluster hours, or approximately 8000 CPU hours.

For the distributed disk architecture, we use a recently purchased cluster at Northeastern University, called TeraCluster. Here, we used 30 nodes, using one computing core per node. For external storage, we used up to 80 GB per node, on a locally attached disk (or up to 2.3 TB of aggregate storage).

We used just one core per node because we found experimentally that executing more than one task simultaneously on a single node could significantly reduce performance. We hypothesize that this occurs when the two tasks disrupt the large contiguous disk accesses of the other, reducing them to smaller, less efficient disk operations. This was not seen in the shared disk architecture, where the SAN was composed of hundreds or thousands of disks. As future work, we intend to examine the use of multiple disks per node, in hopes that each additional disk could be used by a separate task.

The fact that we were using fewer CPUs in the distributed disk architecture is somewhat offset by the fact that the fast multiplication algorithm was twice as fast, performing approximately 10 million multiplications per second per computing core. This speed differential is likely due in part to the different cache architecture of the two CPUs.

Algorithm 2 Refinement by Image Intersection

Input: a subgroup Q of a group G , a coset Qg ; a desired upper bound u ; and a set of words w_1, w_2, \dots in generators of G such that $Qgw_i = Q$.

Output: a demonstration that all elements of Qg have solutions of length at most u or else a subset $S \subseteq Qg$ such that all elements of $Qg \setminus S$ are known to have solutions of length at most u .

- 1: Let $k = u - \text{len}(g)$. Let $U_0 = \{q \in Q \mid \text{dist}(q) > k\} \subseteq Q$. Then $(Q \setminus U_0)g$ is the subset of elements in the coset Qg which are known to have solutions of length at most u . The set U_0g is the “unknown set”, for which we must decide if they have solutions of length u or less.
- 2: For each $i \geq 1$, let $U_i = U_{i-1} \setminus \{q \in U_{i-1} \mid \text{dist}(qgw_i) \leq u - \text{len}(w_i)\}$. (Note that $qgw_i \in Q$). By $\text{dist}(qgw_i)$, we mean the shortest path in the full set of generators of G . If $\text{dist}(qgw_i) \leq u - \text{len}(w_i)$, then qg has a solution of length at most u . The solution for qg is given by a path length $\text{len}(w_i)$ followed by a path of length $u - \text{len}(w_i) = \text{dist}(qgw_i)$.
- 3: If $U_i = \emptyset$ for some $i \leq j$, then we have shown that all elements of Qg have solutions of length at most u . If $U_j \neq \emptyset$, then we have shown that all elements of $(Q \setminus U_j)g$ have solution length at most u .

typically only use optimal solvers when we wish to refine the bound only on a small number of cosets and/or when there are few elements to consider per coset.

We made use of two solvers previously developed by the community. The first is Cube Explorer, developed by Kociemba (2007). This is currently the most efficient, and most developed solver for Rubik’s cube.

The second solver we made use of is based on software developed by Winter (1992). This solver uses a method similar to Cube Explorer, but does not make use of some of the more advanced features or the large precomputations performed by Cube Explorer. However, it is a simple C program, which we relatively easily modified to run in parallel on our cluster computer.

7.3. Refinement by image intersection

The method we introduce here, *image intersection*, attempts to overcome the expense of considering each individual element in a coset in isolation. It is typically much more efficient than optimal solvers, especially when there are many elements to solve per coset.

Note that for the coset Qg , there can be many paths in the coset graph from the identity coset to Qg . In terms of group theory, there are multiple words, w_1, w_2, \dots , where each word is a product of generators of Rubik’s group, and $Qw_1 = Qw_2 = \dots = Qg$. Note that in general, the words are distinct group elements: $w_1 \neq w_2$, and $qw_1w_2^{-1} \neq q$, for $q \in Q$. Nevertheless, $w_1w_2^{-1} \in Q$. The different images of Q produced through multiplication with the different words is the key to finding a refined upper bound.

Next, suppose our goal is to demonstrate an upper bound $k + \ell$ for all of the elements in the coset Qg (at depth ℓ), where $\ell \leq k + \ell < d + \ell$.

Let $Q_k \stackrel{\text{def}}{=} \{q \in Q \mid \text{dist}(q) \leq k\}$, the subset of Q at distance from the identity at most k , and let $Q_k g \stackrel{\text{def}}{=} \{qg \mid q \in Q_k\}$. Let w_i be words such that $Qw_1 = Qw_2 = \dots = Qg$. Assume the words are of length ℓ in the generators of Rubik’s group, i.e., they are shortest words for g .

Note that for all elements of $Q_k w_1$, there is an upper bound, $k + \ell$. Similarly, for all elements of $Q_k w_2$, there is an upper bound, $k + \ell$. Therefore, the elements of $Q_k w_1 \cup Q_k w_2$ have an upper bound of $k + \ell$. More compactly,

$$\text{dist}(Q_k w_1 \cup Q_k w_2) \leq k + \ell.$$

More generally, for w_i a word in the generators of G of length $\text{len}(w_i)$, let $x_i = \text{len}(w_i) - \ell$ represent the number of additional moves used by w_i beyond a shortest path. Then

$$\text{dist}(Q_{k-x_1} w_1 \cup Q_{k-x_2} w_2) \leq k + \ell$$

since the length of any word in $Q_{k-x_1} w_1$ is at most $(k - x_1) + \text{len}(w_1) = k + \ell$ and similarly for w_2 .

Depth	Cosets	Elts. Left	Method	Depth	Cosets	Elts. Left	Method
2	3	66756	Starting Elts.	6	38336	37833	Projection
		222	Image Inters.			26065	Image Inters.
3	23	2652	Projection	7	490879	166425	Projection
		1311	Image Inters.			77555	Optimal Solver
		44	Optimal Solver				
4	241	586	Projection	8	6298864	990419	Projection
		482	Image Inters.			152940	Optimal Solver
		293	Optimal Solver				
5	3002	3756	Projection	9	80741117	2752920	Projection
		2891	Image Inters.			0	Optimal Solver

Fig. 8. Number of remaining elements at each depth during refinement.

In some cases, the optimal solver was given an execution time limit, after which it would halt execution on a given element. Due to this, the optimal solver may have failed to refine some elements which it could have at some earlier depths.

The experimental results of this refinement are shown in Fig. 8. The table shows: the total number of cosets at depths 2 through 9; the total number of remaining unrefined elements across all cosets at each depth; and the resulting number of elements left after each refinement technique was used. The decision of which technique to use at each point was based on timing estimates of the various methods.

The entire process was completed over the course of approximately two weeks, though that period includes down time between computations to analyze results and setup the next step. If the three methods of refinement were more tightly integrated and automated, the entire process would likely take just a few days on a cluster of computers.

Lastly, we refined the cosets at depth 16, showing that all elements there can be solved in 26 or fewer moves. We did this by applying an optimal solver to the elements corresponding to the last three levels of the square subgroup across each of the cosets, and verifying that a solution of 26 or fewer moves was found. With 138262 elements at depth 11 or greater in the square subgroup, and 17 cosets at depth 16, this resulted in 2350454 elements to solve. This is roughly equivalent to the last refinement, at depth 9, of the previous series of computations.

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