Exploiting symmetries in polyhedral computations

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Abstract In this note we give a short overview on symmetry exploiting techniques in three different branches of polyhedral computations: The representation conversion problem, integer linear programming and lattice point counting. We describe some of the future challenges and sketch some directions of potential developments.

1 Introduction

Symmetric polyhedra such as the Platonic and Archimedean solids have not only fascinated mathematicians since time immemorial. They occur frequently in diverse contexts of art and science. Less known to a general audience, but of great importance to modern mathematics and its applications, are higher dimensional analogues of these familiar objects. One standard description is as a set of solutions to a system of linear inequalities

\[ P = \{ x \in \mathbb{R}^n : Ax \leq b \}, \]

where \( A \) is a real \( m \times n \) matrix and \( b \in \mathbb{R}^m \). A prominent example is the \( n \)-cube obtained by \( 2n \) inequalities \( \pm x_i \leq 1 \). It has \( 2^n \) vertices (extreme points) with coordinates \( \pm 1 \) and its group of symmetries is the hyperoctahedral group of order \( 2^{2n} n! \).

Linear models, and therefore polyhedra, are used in a wide range of mathematical problems and in applications such as transportation logistics, machine scheduling, time tabling, air traffic flow management and portfolio planning. They are central objects in Mathematical Optimization (Mathematical Programming) and are for instance heavily used in Combinatorial Optimization. Frequently studied symmetric polyhedra have names like “Travelling Salesman”, “Assignment”, “Matching” and “Cut”. For these and further examples we refer to [Sch03] and the numerous references therein. Over the years a rich combinatorial and geometric theory of polyhedra has been developed (see [Zie97], [Gru03]). Symmetry itself is clearly a central topic in mathematics, and through the spread of computer algebra systems like \textsc{GAP} and \textsc{MAGMA}, sophisticated tools from Computational Group Theory are widely used today (see [HEO05]). Nevertheless, although many polyhedral problems are modeled with a high degree of symmetry, standard computational techniques for their solution do not take advantage of them. Even worse, often the used methods are known to work notoriously poorly on symmetric problems.

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In this short survey we describe three main areas of polyhedral computations, in which the rich geometric structure of symmetric polyhedra can potentially be used for improved algorithms:

I: Polyhedral representation conversion using symmetry
II: Symmetric integer linear programming
III: Counting lattice points and exact volumes of symmetric polyhedra

There are multiple strong dependencies among the three topics and each one has its theoretical and algorithmic challenges as well as important applications. Before we take a closeup view on the three topics we give a brief introduction to the different types of polyhedral symmetries and how these can be determined and worked with.

What are polyhedral symmetries?

The symmetries of a polyhedron can be of a purely combinatorial nature or they can also have a geometric manifestation as affine symmetries, that is, as affine maps of $\mathbb{R}^n$ preserving the polyhedron. Among these symmetries are the “more visually accessible” isometries which are composed of translations, rotations and reflections. All of the symmetries of the $n$-cube for example are part of its isometry group. There exists a representation as a linear group in $\text{GL}_n(\mathbb{R})$ and as a finite orthogonal group of isometries. However, if we perturb the defining inequalities a bit, all of these affine symmetries may be lost, while the new polyhedron is still combinatorially equivalent to a cube, sharing all of its combinatorial symmetries. These are defined as automorphisms of the polyhedral face lattice which encodes the combinatorial structure of a polyhedron. We refer to our survey [BDP+12] for further reading on these different types of polyhedral symmetry groups. The study of combinatorial lattices and their automorphisms is itself an active research area (see [MS02]). The same is true for the study of possible isometry groups, respectively of finite orthogonal groups in $O_n(\mathbb{R})$. Their classification becomes in a way impractical for $n \geq 5$ (see [O37135]), despite the classification of finite simple groups (see [CCN+85]). Even less is known about symmetry groups of polyhedra (see [Rob84]). Here, an “implication phenomenon” occurs, which has not much been studied so far. For instance, if a 4-gon has an element of order 4 among its affine symmetries, the 4-gon has to be the affine image of a square (2-cube), with an affine symmetry group of order 8. These kind of implications clearly can potentially be exploited algorithmically, for example when detecting polyhedral symmetries.

It is important to note that the same abstract group can have different affine representations. We think that a key ingredient for future algorithmic improvements will be the use of geometric information coming with the affine representations of polyhedral symmetry groups. By a basic result in representation theory there is an invariant affine subspace $I$ coming with each affine symmetry group. The polyhedron splits nicely into an invariant part $P \cap I$ and symmetric slices orthogonal to it. These lie in fibers (pre-images) of the orthogonal projection onto $I$. In a way, all of the symmetry is within these fibers.

Given a polyhedron with a group of symmetries, we say two vertices (or inequalities) are equivalent, if there exists a group element that maps one to the other. The set of vertices (and the set of facets / defining inequalities) splits into a number of orbits (disjoint sets of equivalent elements). For example, the $n$-cube has only one orbit of vertices and one orbit of facets. The same is true for all Platonic polyhedra and their higher dimensional analogues. In contrast, the Archimedean polyhedra
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like the soccer ball (truncated icosahedron) have more than one orbit of facets, but only one orbit of vertices. In all of these examples, their combinatorial symmetry group is equal to the group of affine symmetries. Its invariant affine subspace is a single point, the barycenter of the vertices.

In general, for a polyhedron \( P \) with a group of affine symmetries, the vertices of the polyhedron split into orbits \( O_1, \ldots, O_l \) and the invariant part \( P \cap \mathcal{I} \) is equal to the convex hull \( \text{conv}\{b_1, \ldots, b_l\} \), with \( b_i = (\sum_{x \in O_i} x)/|O_i| \) being the barycenter of orbit \( O_i \). This is due to two facts: The barycenter map, taking a point to the barycenter of its orbit, is an affine map. And second, the affine image of a convex hull of given points is equal to the convex hull of their affine images.

Thus working with the lower dimensional polyhedron \( P \cap \mathcal{I} \) and its vertices gives us access to vertices of \( P \). Orbits of integral points in \( P \) have barycenters at specific locations in \( P \cap \mathcal{I} \). For instance, if the group acts transitive on the coordinates of \( \mathbb{R}^n \) then orbits have barycenters at integral multiples of \( \left( \frac{1}{n}, \ldots, \frac{1}{n} \right) \). For more general coordinate permutations the barycenters form a scaled copy of a standard lattice (see [HRS12]).

**How to determine and work with symmetries?**

If the symmetries of the polyhedron are not known, the first difficulty is their determination and how to represent them. In general we like to work with as many symmetries as possible. However, the combinatorial symmetries can usually not be found without having full knowledge about the vertex-facet incidences of the polyhedron (see [KS03]). In contrast, the group of affine symmetries can be determined from the vertices or defining inequalities alone, by finding the automorphism group of an edge colored graph. If \( P \) is given as the convex hull of its vertices \( x_1, \ldots, x_k \), for instance, then the affine symmetry group can be obtained from the automorphism group of the complete graph with \( k \) vertices and edge labels \( x_i^t Q^{-1} x_j \), where \( Q = \sum_{i=1}^k x_i x_i^t \). For details and a proof we refer to [BDS09]. For further methods to compute polyhedral symmetry groups we refer to [BDP+12]. Automorphism groups of graphs can be computed with software like [bliss] or [nauty]. Given a polyhedral description, the affine symmetries can conveniently be obtained directly with our software [SymPol], which by now can also be used through [polymake]. For instance, given a polyhedron with its description contained in input-file, simply call:

```
 sympol --automorphisms-only input-file
```

If the symmetry group of a polyhedron (or parts of it) are given as a *permutation group*, we can use sophisticated tools from *Computational Group Theory*. Each element of the group is then viewed as a *permutation* of the index set \( \{1, \ldots, m\} \) of the input, for instance of \( m \) defining inequalities. In practice, it is necessary to work with a small set of *group generators* if the group is large, and there are advanced heuristics to obtain such sets. Each face (and in particular each vertex) of a polyhedron is determined by a number of inequalities that are satisfied with equality; it can therefore be represented by a subset of \( \{1, \ldots, m\} \). Given generators of a large permutation group and two subsets that represent faces, a typical computational bottleneck is to decide if both are in the same orbit. The fundamental data structures used for this in practice are *bases and strong generating sets* (BSGS, see [Ser03], [HEO05]). Based on them, backtrack searches can be used to perform essential tasks, such as deciding on (non-)equivalence, obtaining stabilizers or fusing and splitting orbits. An elaborate version is the *partition backtrack* introduced by Leon [Leo91]. These backtracking methods work quite well in practice, although
from a complexity point of view the mentioned problems are thought to be difficult (see [Luk93]). Although computer algebra systems like [GAP] and [MAGMA] provide functions to work with permutation groups, for performance reasons it is often desirable to use problem specific code (see for example [KO06]). Nevertheless, all of these approaches, including [GAP] and [MAGMA], rely on efficient implementations of some partition backtrack. Therefore we have created a flexible C++-implementation [PermLib] of Leon’s partition backtrack (see [Reh10a]) that can serve as a basis for the development of algorithms which combine tools from Computational Group Theory and Polyhedral Combinatorics. By now, [PermLib] has successfully been integrated into current versions of [SymPol], [polymake] and [SCIP].

I. Representation Conversion

By a fundamental theorem in polyhedral combinatorics, the Farkas-Minkowski-Weyl theorem, every polyhedron has a second representation as the convex hull of finitely many vertices (extreme points) and, in the unbounded case, some rays (see [Zie97]). Converting representations from inequalities to vertices (and rays), or vice versa, is a frequent task known as representation conversion problem (or convex hull problem). The importance of these conversions is due to the fact that some problems, like the maximization of a nonlinear convex function, are easy to solve in one presentation, but not in the other. Often, vertices represent objects that one would like to classify. These objects can be quite diverse, for instance perfect quadratic forms (see [DSV07]) or the elementary flux modes in biochemical reaction systems (see [SH94]). Representation conversions are also often used to analyze polyhedra in Combinatorial Optimization (see [Sch03]). So far there exists no efficient algorithm for finding all the vertices of a polyhedron. In fact, the existence of such an algorithm appears to be unlikely, as it is NP-complete for polyhedra that are possibly unbounded (see [KBB+08]). Nevertheless, several algorithms and implementations are widely used in practice (see for example [cdd] and [lrs] which are also available through [polymake]).

Quite often one is only interested in one representative for each orbit of vertices (or inequalities) in a representation conversion. For example, when maximizing a nonlinear convex function on a polyhedron, or when vertices and inequalities in one orbit correspond to equivalent objects of some sort. Representation conversion up to symmetries has been considered in different contexts, and depending on the problem, different techniques have been successful. The most successful approaches currently known are the Incidence Decomposition Method and the Adjacency Decomposition Method (see for instance [CR96], [DFPS01], [DFMV03], [DSV07], [DSV09], [DSV10]). Both methods decompose the problem into a number of lower dimensional subproblems. They can be used recursively and can be parallelized (see [CR01], [DI07]). Loosely speaking, the Incidence Decomposition Method fixes an orbit of the input, whereas the Adjacency Decomposition Method fixes an orbit of the output and then lists all “neighboring” orbits. For details we refer to our survey [BDS09]. We note that it is a priori not clear which method works best. We think best results can be achieved by a combination of different algorithms. All methods known so far do not use geometric insights and still rely on subproblem conversions that do not exploit available symmetry.

Our software [SymPol] and the experimental [GAP] package [Polyhedral] provide implementations of decomposition methods. These preliminary tools have already successfully been used in our own work ([DSV07], [BBC+09], [DSV09], [DSV10], [DSE11]), but also by others: For instance, Kumar [Kum11] obtains a
classification of elliptic fibrations that was previously impossible. Jacques Martinet writes in [Mar03b] about the result in [DSV07]: “It seems plainly impossible to classify 8-dimensional perfect lattices.” SymPol can also be used to verify cumbersome calculations in proofs, like the edge-graph diameter analysis of the recently discovered, celebrated counterexamples to the Hirsch conjecture (see [San12], [RS10]). For example, with the call

`sympol --idm-adm-level 0 1 --adjacencies input-file`

where `input-file` contains the 48 vertices of the 5-dimensional Santos prismatoid (see Table 1 in [San12]), SymPol returns a text file with a description of the adjacency graph of facets up to symmetry. Using a visualization tool like Graphviz, the produced text file, say `adjacencies.dot`, can then easily be turned into an image with a command like

`neato -Tpng -o adjacencies.png adjacencies.dot`

From the obtained image (see figure) it is easily verified that the shortest path from facet 1 to facet 12 is of length 6, which is the key calculation in the proof of [San12].

Let us make a remark on the increasing importance of mathematical software in general: As sophisticated computational tools become an increasingly important basis for high-level mathematical research, their creation also becomes an increasingly important service to the mathematical community. More and more mathematicians use computers in the creative process and to verify standard parts of difficult larger proofs (see e.g. [Hal05], [CK09], [Hal11]). Timothy Gowers [Gow00] even guesses that at the end of the 21st century, computers will be better than humans in proving theorems. Although we would not go as far as Gowers, we are convinced that in the future, parts of proofs will routinely be performed by computers. With a symbiosis of human and computer reasoning we will see substantial advances in mathematical problems. In this way reliable mathematical software becomes an increasingly important part of mathematics itself.

**Challenges**

One of the most challenging polyhedral conversion problems arises in conjunction with lattice sphere packing problem, a classical problem in the Geometry of Numbers. Since its solution up to dimension 8 almost 80 years ago, it is still open in dimension $n \geq 9$, with the exception of dimension 24 (see [Mar03b], [CK09]). One way to approach this problem is via finding the vertices of a locally polyhedral
object known as Ryshkov polyhedron (see Sch09a for details). The currently open 9-dimensional case leads to a challenging representation conversion problem of a 45-dimensional Ryshkov polyhedron. Main difficulties here come from faces that carry the symmetries of the exceptional Weyl group $E_8$. We think that this problem is a particular nice test case, as all finite rational matrix groups appear as stabilizers of faces in the Ryshkov polyhedron. So in a way, this challenging representation conversion problem gives a universal test case for any future algorithmic advances.

II. Integer Linear Programming

Linear programming is the task of maximizing (or minimizing) a linear function on a polyhedron given by linear inequalities. It serves as a fundamental basis for theory and computations in Mathematical Optimization (see Tod02). In Integer linear programming vectors to be optimized are restricted to integers (number of goods, etc.) or even to 0/1 entries (encoding a simple yes-no-decision). Integer linear programming is widely used in practical applications. In fact, “the vast majority of applications found in operations research and industrial engineering involve the use of discrete variables in problem formulation” (from a book review of Wol98).

In many of these problems the involved polyhedra have symmetries (see Sch03). From a complexity point of view, integer programming is NP-complete (see Kar72), whereas linear programming can be solved in polynomial time. For fixed dimension, polynomial time algorithms are known for integer linear programming (see Len83).

A linear programming problem $\text{max } c^T x$ with $x \in P$ is invariant with respect to a linear symmetry group $\Gamma \leq \text{GL}_n(\mathbb{R})$, if the polyhedron $P$ and the utility vector $c \in \mathbb{R}^n$ are preserved by it, that is, if $\Gamma P = P$ and $\Gamma c = c$. Any solution to the linear program, its orbit and the barycenter of its orbit lie in the same hyperplane orthogonal to the utility vector and therefore have the same utility value. Due to the convexity of polyhedra, the barycenter is also a feasible solution. As it lies in the invariant linear subspace $I$ of $\Gamma$, the linear programming problem always has a solution attained within $I$. Thus it is possible to solve the lower-dimensional linear program $\text{max } c^T x$ with $x \in P \cap I$. Such symmetry reductions are often referred to as “dimension reduction” or “variable reduction”. The symmetries of an integer linear program are more restrictive, as also $\mathbb{Z}^n$ has to be left invariant by the group $\Gamma$.

Exploiting symmetries in integer programming is much more difficult than in linear programming. In fact, symmetries are rather problematic, as standard methods like branch-and-bound or branch-and-cut (see Sch80) have to solve many equivalent sub-problems in such cases. In contrast to linear programming, it is not possible to simply consider the intersection with the invariant affine subspace, as integral solutions can lie outside. Nevertheless, in recent years it has been shown that it is possible to exploit symmetries in integer programming; see for example Mar03a, Fri07, BM08, KP08, OLR09, LMT09. These specific methods fall into two main classes: They either modify the standard branching approach, using isomorphism tests or isomorphism free generation to avoid solving equivalent subproblems; or they use techniques to cut down the original symmetric problem to a less symmetric one, which contains at least one element of each orbit of solutions. For further reading we refer to the excellent survey Mar10.

As many real world applications can be modeled as (mixed) integer programming problems, a variety of professional software packages are available. Two of the leading ones, CPLEX and Gurobi, have by now included some techniques to avoid or use symmetry. Unfortunately it is publicly not known what exactly is done.
None of the known methods uses the rich geometric properties of the involved symmetric polyhedra. Using the fact that solutions are “near” the linear invariant subspace, it is possible to do better. For the special case of a one dimensional invariant subspace, with the full symmetric group $S_n$ acting transitively on the coordinates in $\mathbb{R}^n$, this is shown in [BHJ11]. We have highly promising results with a generalization to arbitrary symmetries in [HRS12]. In particular for direct products of symmetric groups, we not only beat state-of-the-art professional solvers, but even solve a challenging, previously unsolved benchmark problem from [MIPLIB] (instance toll-like).

The main ingredient is the observation that any feasible integer linear programming problem with a non-trivial affine symmetry group contains certain core-sets of integral vectors that can be used as a kind of test-set, that is, if none of the points from the core-set is contained in the feasible region, no integral point is. Assume $\Gamma \leq \text{GL}_n(\mathbb{R})$ is a linear representation of a given symmetry group and $I$ denotes its invariant linear subspace containing the utility vector $c$. Then we say an integral point $z$ in a fiber (pre-image) of the orthogonal projection onto $I$ is in the core-set of the fiber if the convex hull of its orbit $\Gamma z$ does not contain any integral points aside those of the orbit itself. Then, by definition, representatives of each orbit $\Gamma z$ in the core-set can be used as a test-set for feasibility of a fiber.

Using the Flatness Theorem (from the Geometry of Numbers), it can be shown that core-sets are finite for irreducible groups. For direct products of symmetric groups acting on some of the coordinates, the test-set containing only representatives of orbits even reduces to a single point. Besides that not much is known so far about core-sets. Nevertheless, we think that they will serve as a powerful tool in the design of new algorithms for symmetric integer linear programs.

### Challenges

Challenging examples of symmetric integer linear programs can be found in benchmark libraries like [MIPLIB]. These problems come from diverse contexts and have not been chosen to be particular symmetric. Nevertheless many symmetries can be found and exploiting them algorithmically, beating state-of-the-art commercial solvers, remains a challenging test case for future advances.

Some particular symmetric integer linear programming problems coming from difficult combinatorial problems in mathematics have been collected (and worked on) by Francois Margot at `symlp` (see for instance also [Mar03a] and [BM08]). As these problems have been intensively worked on, improving on the currently known results is certainly a hard problem. So this gives a very good benchmark for future improvements as well.

### III. Lattice Point Counting and Exact Volumes

Often it is desirable to know how many integral solutions there are to a system of linear inequalities. Such problems occur frequently in Combinatorics (see [Sta97]) but also in disciplines such as representation theory (Kostka and Littlewood-Richardson coefficients, see [BZ01] and [KT99]), in statistics (contingency tables, see [DS98]), in voting theory (see [WP07] and [GL11]), and even in compiler optimization (see `Graphite`). We refer to [Del05] for an overview. Counting lattice points is moreover intimately related to integer linear programming (see [Las09]).
By a breakthrough result of Barvinok [Bar94] in the 1990s, counting lattice points inside a rational polyhedron can be done in polynomial time for fixed dimension. His ideas are based on evaluating “short rational generating functions” and on constructing unimodular triangulations. His algorithm has been implemented in [LattE] and [barvinok]. The same applies to a slightly more general setting, in which one considers a one-parameter family of dilations $\lambda P$, with $P$ a rational polyhedron and $\lambda$ an integer. By a theory initiated by Ehrhart in the 1960s (see [Ehr67], [BR07]), it is possible to obtain the number of integral points in the dilate $\lambda P$ by a quasi-polynomial in $\lambda$, with its degree equal to the dimension of $P$. A quasi-polynomial $p$ is determined by a finite number of polynomials $p_i$, $i = 0, \ldots, k$, via the setting $p(\lambda) = p_i(\lambda)$ for all $\lambda$ congruent to $i$ mod $k$. In case of $P$ being integral, the Ehrhart quasi-polynomial simply is a polynomial in $\lambda$. In general, the quasi-polynomial can also be computed in polynomial time by Barvinok methods (see [Bar08]). Often, the main interest is only in the leading coefficient of the Ehrhart quasi-polynomial, which is the volume of $P$. Computing the volume itself is already a $\#P$-hard problem (see [DF88], [BW91]).

Despite the fact that many counting problems have plenty of symmetries, they have not been exploited systematically so far. In other words, exploiting symmetry in lattice point counting, or more generally in Ehrhart theory, is a vastly open subject. For volume computations the situation seems a bit better. For very special volume computations symmetry can be exploited (see [DSV09]). However, there is still a huge potential for improved methods. Many of the difficulties originate from the fact that the “Barvinok methods” used to solve them rely on unimodular triangulations of polyhedral cones that usually do not inherit the symmetry of the polyhedron. New roads will have to be taken here.

In [Sch12] it is shown that it is possible to exploit symmetries by using a decomposition into symmetric slices, together with a weighted Ehrhart theory. The theoretical background and first implementations for such a theory have just recently been developed (see [BBD+11b], [BST12]). The polyhedral decomposition used in [Sch12] is rather special: There is a linear invariant part and symmetric slices orthogonal to it, which are cross-products of regular simplices (simploptopes). A generalization to other decompositions is easily obtained, whenever there is a decomposition into an invariant part and slices orthogonal to it for which the Ehrhart quasi-polynomial is known. Note that the decomposition can easily be obtained in a automated way, as the invariant part is the intersection of the given polyhedron with the affine space fixed by its symmetry group.

For exploiting symmetry in corresponding volume computations, the integration of polynomials over a polyhedron is used. Using Brion-Lawrence-Varchenko theory, this can efficiently be done by integrating sums of powers of linear forms (see [BBD+11a]. The new decomposition approach of [Sch12] also allows to obtain exact volumes that have not been computable before. This is demonstrated on three well studied examples from Social Choice theory, which give the exact likelihood of certain election outcomes with four candidates that were previously known for three candidate elections only (see [GL11]).

**Challenges**

In Social Choice theory we face a large amount of challenging problems related to probability calculations of voting situations with four and more candidates. The only known results in the context of the “polyhedral model” (IAC hypothesis) appear
to be those in [Sch12], which are obtained by exploiting polyhedral symmetry as described above.

A challenging benchmark volume computation that several researchers previously have looked at is the volume of the Birkhoff polytope $B_n$ (also known as perfect matching polytope of the complete bipartite graph $K_{n,n}$). The current known record is the volume of $B_{11}$ due to Beck and Pixton [BP03], using a complex-analytic way to compute the Ehrhart polynomial. The computation of the volume of $B_{11}$ would certainly be quite a computational achievement.

Conclusions

We expect that symmetry exploiting techniques for polyhedral computations can be vastly improved by using geometric properties that come with affine symmetries of polyhedra. Concentrating on improvements in polyhedral computations with affine symmetries is practically no restriction: If a polyhedron is given, either by linear inequalities or vertices and rays, the affine symmetries of the (potentially larger) combinatorial symmetry group are practically the only ones we can compute.

Concluding, for polyhedral representation conversions we see potential in enhancing decomposition methods through the use of geometric information like fundamental domains, classical invariant theory and symmetric polyhedral decompositions. For integer linear programming we expect that a new class of algorithms based on the concept of core points will help to exploit symmetry on difficult symmetric integer linear programming problems. For exact volume computations and counting of lattice points, there is still a lot of potential for new ideas using symmetry. Overall, we think symmetry should be exploitable whenever it is available. For this goal to be reached there seem still quite some efforts necessary though.

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LattE. Lattice point count, volumes and integrals, http://www.math.ucdavis.edu/Latte/
lrs. Lexicographic reverse search, http://cgm.cs.mcgill.ca/~avis/C/lrs.html
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