

# Calcium: computing in exact real and complex fields

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## ABSTRACT

Calcium is a C library for real and complex numbers in a form suitable for exact algebraic and symbolic computation. Numbers are represented as elements of fields  $\mathbb{Q}(a_1, \dots, a_n)$  where the extension numbers  $a_k$  may be algebraic or transcendental. The system combines efficient field operations with automatic discovery and certification of algebraic relations, resulting in a practical computational model of  $\mathbb{R}$  and  $\mathbb{C}$  in which equality is rigorously decidable for a large class of numbers.

## CCS CONCEPTS

• **Computing methodologies** → **Representation of exact numbers; Hybrid symbolic-numeric methods; Exact arithmetic algorithms; Symbolic and algebraic manipulation.**

## KEYWORDS

Exact real arithmetic, Number fields, Transcendental numbers

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## 1 INTRODUCTION

A field  $K$  is said to be *effective* if its elements can be enumerated and the operations  $\{+, -, \cdot, /, =\}$  are computable. Examples include the rationals  $\mathbb{Q}$ , finite fields  $\mathbb{F}_q$ , and the algebraic numbers  $\overline{\mathbb{Q}}$ .

The fields of real and complex numbers  $\mathbb{R}$  and  $\mathbb{C}$  are notably non-effective, even when restricted to so-called *computable numbers* (a real number  $x$  is said to be *computable* if there is a program which, given  $n$ , outputs some  $x_n \in \mathbb{Q}$  with  $|x - x_n| < 2^{-n}$ ). The problem is that equality is only semi-decidable: we can in general prove  $x \neq y$ , but not  $x = y$ , as a consequence of the halting theorem. Nevertheless, as the example of  $\overline{\mathbb{Q}}$  shows, we can hope for an equality test at least for *some* numbers within a suitable algebraic framework.

This paper presents Calcium,<sup>1</sup> a C library for exact computation in  $\mathbb{R}$  and  $\mathbb{C}$ . Numbers are represented as elements of fields  $\mathbb{Q}(a_1, \dots, a_n)$  where the extension numbers  $a_k$  are defined symbolically. The system constructs fields and discovers algebraic relations automatically, handling algebraic and transcendental number fields in a unified way. It is capable of deciding equality for a wide class of numbers which includes  $\overline{\mathbb{Q}}$  as a subset. We show a few basic examples, here using a Python wrapper:

```
>>> (pi**2 - 9) / (pi + 3)
0.141593 {a-3 where a = 3.14159 [Pi]}
>>> phi = (sqrt(5)+1)/2; (phi**100 - (1-phi)**100)/sqrt(5)
3.54225e+20 {354224848179261915075}
>>> i**i - exp(pi / (sqrt(-2)**sqrt(2))**sqrt(2))
0
>>> log(sqrt(2)+sqrt(3)) / log(5+2*sqrt(6))
0.500000 {1/2}
>>> erf(4*atan(ca(1)/5) - atan(ca(1)/239)) + erfc(pi/4)
1
>>> -1e-12 < exp(pi*sqrt(163)) - 262537412640768744 < -1e-13
True
```

In the first example, the field is  $\mathbb{Q}(a)$  where  $a = \pi$ , and the element is  $a - 3$ . The numerical approximation ( $x \approx 0.141593$ ) is computed to desired precision on demand, for example when printing or evaluating a numerical predicate. Examples 2–5 were chosen so that the field of the result simplifies to  $\mathbb{Q}$ .

The basic idea in this work is to combine a multivariate fraction field representation, automatic relation discovery, ideal reduction (where the ideal may be incomplete), and rigorous numerical certification using ball arithmetic. This approach is inspired by earlier implementations of  $\overline{\mathbb{Q}}$  (including Sage [30] and Magma [3]) and by theoretical work on transcendental fields (notably Richardson’s algorithm [26]). The main improvement over previous software is that we handle both algebraic and transcendental numbers in a rigorous and structured way; compared to symbolic computation systems like Mathematica and Maple, our approach provides faster field arithmetic and more reliable results.

This paper is structured as follows. Section 2 presents our high-level strategy for exact computation, described at a general level without reference to low-level implementation details. Section 3 discusses the architecture of Calcium. Section 4 relates our strategy to earlier work and presents some benchmark results.

## 2 COMPUTING IN SUBFIELDS OF $\mathbb{C}$

To simulate  $\mathbb{R}$  and  $\mathbb{C}$ , we may start with  $\mathbb{Q}$  and lazily extend the field with new numbers  $a_k$  as they arise in computations. A general way to compute in such extension fields of  $\mathbb{Q}$  is in terms of quotient rings and their fields of fractions (henceforth *formal fields*).

<sup>1</sup>Pronounced “kalkium” to distinguish it from the chemical element. Calcium is free and open source (LGPL 2.1+) software. The source repository is <https://github.com/fredrik-johansson/calcium> and the documentation is available at <http://fredrikj.net/calcium/>.

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In the following, we assume that  $a_1, \dots, a_n$  is a finite list of complex numbers. We let  $X_1, \dots, X_n$  denote independent formal variables, we let  $\mu : \mathbb{Q}[X_1, \dots, X_n] \rightarrow \mathbb{C}$  denote the evaluation homomorphism induced by the map  $X_k \mapsto a_k$ , and we define

$$I := \ker \mu = \{f \in \mathbb{Q}[X_1, \dots, X_n] : f(a_1, \dots, a_n) = 0\}$$

as the ideal of all algebraic relations among  $a_1, \dots, a_n$  over  $\mathbb{Q}$ .

**THEOREM 1.** *Assume that  $I$  is known (in the sense that an explicit list of generators  $I = \langle f_1, \dots, f_m \rangle$  is known). Then*

$$K := \mathbb{Q}(a_1, \dots, a_n) \cong K_{\text{formal}} := \text{Frac}(\mathbb{Q}[X_1, \dots, X_n]/I)$$

is an effective subfield of  $\mathbb{C}$ .

**PROOF.** The isomorphism is obvious. Decidability of “=” in the formal field follows from the fact that we can compute a Gröbner basis for  $I$ . Given formal fractions  $\frac{p}{q}$  and  $\frac{r}{s}$  with  $p, q, r, s \in \mathbb{Q}[X_1, \dots, X_n]$ , we can consequently decide whether  $ps \equiv qr \pmod{I}$ . Indeed, we can also decide whether  $q, s \not\equiv 0 \pmod{I}$  and thereby ensure that the fractions define numbers in the first place.  $\square$

Some easy special cases are worth noting:

- The trivial field  $K = \mathbb{Q}$  (take  $n = 0$ ).
- Transcendental number fields  $K = \mathbb{Q}(a_1, \dots, a_n)$  where the numbers  $a_1, \dots, a_n$  are algebraically independent over  $\mathbb{Q}$ .
- Algebraic number fields  $K = \mathbb{Q}(a) \cong \mathbb{Q}[X]/\langle f(X) \rangle$  where  $a$  is an algebraic number with minimal polynomial  $f$ .

The general case is a *mixed field* in which the extension numbers may be algebraic or transcendental and algebraically dependent or independent in any combination.

**EXAMPLE 1.**  $\mathbb{Q}(\log(i), \pi, i) \cong \text{Frac}(\mathbb{Q}[X_1, X_2, X_3]/I)$  where  $I = \langle 2X_1 - X_2X_3, X_3^2 + 1 \rangle$ .

Theorem 1 solves the arithmetic part of computing in finitely generated subfields of  $\mathbb{C}$ , at least up to practical issues such as the complexity of multivariate polynomial arithmetic and Gröbner basis computations. The crucial assumption made in Theorem 1, however, is that the ideal  $I$  is known. In general, finding  $I$  is an extremely hard problem. For example, although  $\mathbb{Q}(\pi) \cong \mathbb{Q}(X_1)$  and  $\mathbb{Q}(e) \cong \mathbb{Q}(X_2)$ , it is an open problem to prove  $\mathbb{Q}(\pi, e) \cong \mathbb{Q}(X_1, X_2)$ . There are specific instances where we can prove algebraic independence (the Hermite-Lindemann-Weierstrass theorem, Baker’s theorem, transcendence of isolated numbers such as  $\Gamma(\frac{1}{4})$ , results for  $E$ -functions [12], etc.), but we typically only have conjectures. Most famous (and implying  $\mathbb{Q}(\pi, e) \cong \mathbb{Q}(X_1, X_2)$  as a special case) is:

**CONJECTURE 2 (SCHANUEL’S CONJECTURE).** *If  $z_1, \dots, z_n$  are linearly independent over  $\mathbb{Q}$ , then  $\mathbb{Q}(z_1, \dots, z_n, e^{z_1}, \dots, e^{z_n})$  has transcendence degree at least  $n$  over  $\mathbb{Q}$ .*

Thus, in general, we can only determine  $I$  conjecturally. We address this limitation below in section 2.2.

## 2.1 Defining extension numbers

We stress that we cannot simply input  $I$  as a way to *define* the extension numbers  $a_1, \dots, a_n$ , since this does not give enough information about the embedding (that is,  $\mu$ ) in  $\mathbb{C}$ . We are not interested in computing in an abstract algebraic structure but in a concrete model of  $\mathbb{R}$  and  $\mathbb{C}$  where we can do (at least) the following:

- Evaluate the complex conjugation map  $z \rightarrow \bar{z}$ .
- Evaluate numerical ordering relations ( $x < y$ ,  $|x| < |y|$ , etc.).
- Exclude singularities (e.g. division by zero) and choose well-defined branches of multivalued functions.

We therefore need a symbolic way to define extension numbers  $a_1, \dots, a_n$  so that they are explicitly (numerically) computable, and we need to construct  $I$  from this symbolic data rather than vice versa. The following types of extensions are useful:

- *Absolute algebraic:*  $a$  is a fixed algebraic constant  $a \in \overline{\mathbb{Q}}$ , for example  $i$ ,  $\sqrt{2}$ ,  $e^{2\pi i/3}$ , or  $[a^5 - a - 1 = 0; a \approx 1.17]$ . Such a constant can be defined canonically by its minimal polynomial over  $\mathbb{Q}$  together with an isolating ball for a root.
- *Relative algebraic:*  $a$  is defined by an equation  $a^{m/n} = c$  with  $c \in \mathbb{C}$ , or more generally  $P(a) = 0$  with  $P \in \mathbb{C}[X]$ , together with an isolating complex ball for a root.
- *Transcendental* (perhaps conjecturally):  $a$  is a symbolic constant ( $\pi$ ,  $\gamma$ , etc.) or function ( $e^z$ ,  $\log(z)$ ,  $z^w$ ,  $\Gamma(z)$ ,  $J_V(z)$ , etc.) evaluated at some point.
- *Black-box computable:*  $a$  is defined by a program for numerical evaluation in ball arithmetic. (We will not be able to prove algebraic relations except self-relations like  $a - a = 0$ .)

Calcium represents algebraic and transcendental extensions in the usual way as symbolic expressions  $f(z_1, \dots, z_p)$ . The arguments  $z_1, \dots, z_p$  are real or complex numbers which may belong to different fields, say  $z_1 \in \mathbb{Q}(b_1, \dots, b_r)$ ,  $z_2 \in \mathbb{Q}(c_1, \dots, c_s)$ , etc.

Each extension number defines a computable number through recursive numerical evaluation of the symbolic function and its arguments in arbitrary-precision ball arithmetic. This is at least true in principle assuming that we can decide signs at discontinuities. An important improvement over many symbolic computation systems is that we exclude non-numerical extensions: for example, when adding  $\log(z)$  as an extension number, we must be able to prove  $z \neq 0$ . We fix principal branches of all multivalued functions.

This is only a starting point: we can imagine other classes of extensions (periods, solutions of implicit transcendental equations, etc.). The main point is not the precise internal classification but the logical separation between field elements and extension numbers.

## 2.2 Working with an incomplete ideal

As already noted, it is often infeasible to find an ideal  $I$  defining a formal field isomorphic to  $\mathbb{Q}(a_1, \dots, a_n)$ . Even in cases where all relations in  $I$  in principle can be determined, they may be costly to compute explicitly, for instance when they involve algebraic extensions of even moderately high degree.

Fortunately, it is usually sufficient to construct a partial ideal  $I_{\text{red}} \subseteq I$ . We call this the *reduction ideal* since it helps keeping expressions partially reduced (allowing for efficient computations) even if  $I_{\text{red}} \neq I$ . The reason why we do not need to ensure  $I_{\text{red}} = I$  is that we can use the evaluation map  $\mu$  (implemented in ball arithmetic) as a witness of nonvanishing for particular field elements. Algorithm 1 provides a template for evaluating predicates, given a possibly incomplete reduction ideal  $I_{\text{red}}$ .

Algorithm 1 depends on a *work parameter*  $W$ . We can take  $W$  to be a numerical precision for step (c), say with  $W_{\text{min}} = 64$  bits and  $W_{\text{max}} = 4096$  bits and an implied doubling of  $W$  each iteration. We explain “heuristics with strength  $W$ ” in step (d) below in section 2.3.

If we set  $W_{\max} = \infty$  to force a True/False answer, then termination when  $z = 0$  is conditional on the completeness of the methods to find relations in (d).

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**Algorithm 1:** Test if  $z = 0$ .

---

**Input:** Extension numbers  $a_1, \dots, a_n$ , an element  $z \in \mathbb{Q}(a_1, \dots, a_n)$  represented by a formal fraction  $p/q$  with  $p, q \in \mathbb{Q}[X_1, \dots, X_n]$  (such that  $\mu(q) \neq 0$ ), a reduction ideal  $I_{\text{red}} \subseteq I$ , and work limits  $W_{\min}, W_{\max}$ .  
**Output:** True (implying  $z = 0$ ), False ( $z \neq 0$ ), or Unknown.

- (1) For  $W = W_{\min}, \dots, W_{\max}$ , do:
    - (a) If  $p \equiv 0 \pmod{I_{\text{red}}}$ , return True.
    - (b) If it can be certified that  $I_{\text{red}} = I$ , return False.
    - (c) Using ball arithmetic with strength  $W$ , compute an enclosure  $E$  with  $\mu(p) \in E$ . If  $0 \notin E$ , return False.
    - (d) Using heuristics with strength  $W$ , attempt to find and prove a new set of relations  $J$  with  $J \subseteq I$ , and set  $I_{\text{red}} \leftarrow I_{\text{red}} \cup J$ . (See Algorithm 2.)
  - (2) Return Unknown.
- 

Step (b) is applicable, for example, in simple algebraic or transcendental number fields such as  $\mathbb{Q}(\sqrt{2})$  and  $\mathbb{Q}(\pi)$ .

In step (d), we may try to find a general relation for  $a_1, \dots, a_n$ , or we may attempt to prove  $\mu(p) = 0$  directly and take  $J = \langle p \rangle$ . The latter is sometimes easier. For example, if  $a_1, \dots, a_n \in \overline{\mathbb{Q}}$ , it is often cheaper to compute the minimal polynomial specifically for  $z$  than to compute all of  $I$ . In the current Calcium implementation, (d) consists of proving  $z = 0$  through a direct  $\mathbb{Q}$  calculation, ensuring termination over  $\mathbb{Q}$  (see section 3.3); updating the preexisting  $I_{\text{red}}$  with other types of relations is a possible future improvement.

It is an implementation detail whether we cache the updated ideal  $I_{\text{red}}$  for future use in the same field after exiting Algorithm 1.

### 2.3 Constructing the ideal

We will now describe a practical strategy to construct a reduction ideal  $I_{\text{red}}$  for a given field  $K = \mathbb{Q}(a_1, \dots, a_n)$ .

Which relations are interesting to include, and when? The minimalist solution is that we set  $I_{\text{red}} = \{ \}$  when we construct  $K$ , and only populate  $I_{\text{red}}$  lazily in Algorithm 1. The maximalist solution is to ensure  $I_{\text{red}} = I$  up front. There is a tradeoff: on one hand, we want to capture as much of the true ideal  $I$  as possible so that testing equality is trivial and so that there is minimal expression swell in computations. On the other hand, we do not want to waste time finding potentially useless relations and computing Gröbner bases every time we construct a field. As in Algorithm 1, it is useful to make the effort dependent on a work parameter  $W$  controlling numerical precision, choice of heuristics, and so forth.

Algorithm 2 implements a smorgasbord of methods for finding relations, most of which involve searching for (linear) integer relations. We recall that an integer relation between complex numbers  $a_1, \dots, a_n$  is a tuple  $(m_1, \dots, m_n)$  with some  $m_k \neq 0$  such that

$$m_1 a_1 + \dots + m_n a_n = 0, \quad m_i \in \mathbb{Z}.$$

The LLL algorithm can be used to compute a basis matrix for all integer relations among a finite list of numbers; see for example Algorithm 7.13 in [16]. More precisely, LLL finds a basis of candidate

relations which may or may not be correct. We are guaranteed to find all integer relations as  $W \rightarrow \infty$  where  $W$  is the numerical precision, but we have to use exact computations to certify or reject the relations obtained at a fixed finite  $W$ . Since the certifications can be expensive, it is useful to make them dependent on  $W$  (for example, limiting bit sizes of field elements in recursive computations).

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**Algorithm 2:** Construct ideal of algebraic relations.

---

**Input:** Extension numbers  $a_1, \dots, a_n$ , a work parameter  $W$ .  
**Output:** A reduction ideal  $I_{\text{red}} \subseteq I$  for  $\mathbb{Q}(a_1, \dots, a_n)$ .

Initialize  $I_{\text{red}} \leftarrow \{ \}$ . Depending on  $W$ , run a subset of A-F (more expensive steps may be skipped when  $W$  is small):

- A **Direct algebraic relations.** For absolute or relative algebraic extensions  $a_k$ , add the defining relations to  $I_{\text{red}}$ .
  - B **Vieta's formulas.** For algebraic extensions  $a_k$  that are conjugate roots of the same polynomial, add the interrelations defined by Vieta's formulas to  $I_{\text{red}}$ .
  - C **Log-linear relations.** Let  $L$  denote the set of extensions of the form  $a_k = \log(z_k)$ , along with  $\pi i$  if available. Use LLL with precision  $W$  to search for relations  $2\pi i m_0 + \sum_j m_j \log(z_j) = 0$ . Attempt to certify each candidate relation:
    - Compute an enclosure of  $\frac{1}{2\pi i} \sum_j m_j \log(z_j)$  and verify that it contains a unique integer.
    - Attempt to prove  $\prod_j z_j^{m_j} = 1$  using Algorithm 1 (using exact recursive computations in the fields of the arguments  $z_j$ ).
    - If both certification steps succeed, update the ideal with  $I_{\text{red}} \leftarrow I_{\text{red}} \cup \langle 2\pi i m_0 + m_1 a_1 + \dots + m_n a_n \rangle$ .
  - D **Exp-multiplicative relations.** Let  $E$  denote the set of extension numbers of the form  $a_k = \sqrt{z_k}$ ,  $a_k = z_k^{m/n}$ ,  $a_k = z_k^{w_k}$ ,  $a_k = e^{z_k}$  or  $a_k \in \overline{\mathbb{Q}}$ . Search for potential multiplicative relations  $\prod_j a_j^{m_j} = 1$  using LLL applied to  $\log(E)$  and certify the candidate relations through exact recursive computations similarly to the log-linear case.
  - E **Special functions.** Update  $I_{\text{red}}$  with relations resulting from functional equations and connection formulas such as  $\Gamma(z+1) = z\Gamma(z)$  or  $\text{erf}(z) = -\text{erf}(-z) = -i \text{erfi}(iz)$ . Candidate relations can be found by numerical comparison of function arguments and certified through exact recursive computations.
  - F **Algebraic interrelations.** Use resultants or LLL (followed by certification using resultants) to search for linear (or bilinear, etc.) relations among algebraic extensions.
- 

Algorithm 2 will only find relations that are expressible in terms of the given  $a_1, \dots, a_n$ . For example, to add the relation for a square root extension  $a_n = \sqrt{z}$  in step A, we need to be able to express  $z$  in terms of  $K' = \mathbb{Q}(a_1, \dots, a_{n-1})$  as a formal fraction  $f/g$  with  $\mu(f/g) = z$ . The relation is then  $\langle g^2 X_k^2 - f^2 \rangle$ . If  $z$  cannot be expressed in  $K'$ , then  $a_k$  behaves like a transcendental number within the present field. However, it is usually desirable to make  $z$  part of the field so that ideal reduction automatically produces  $(\sqrt{z})^2 \rightarrow z$ . One possibility is that we always adjoin  $z$  (or  $b_1, \dots, b_m$  such that  $z \in \mathbb{Q}(b_1, \dots, b_m)$ ) to the field where we create  $\sqrt{z}$ . An alternative is to modify Algorithm 2 to append new extension numbers to the existing field whenever it may help simplifications.

We will not attempt to prove the completeness of Algorithm 2 for any particular sets of numbers here (see section 4 for a few remarks). We are constrained by the requirement that potential relations have to be certifiable: it makes no sense to look for a hypothetical relation that we will not be able to prove (say,  $m\pi + ne = 0$  with  $m, n \in \mathbb{Z}$ ).<sup>2</sup>

## 2.4 Choosing extension numbers

We have so far assumed that the extension numbers  $a_1, \dots, a_n$  are given. We usually have a great deal of freedom to choose the form of extension numbers to represent a given field  $K$ . The following are some possible transformations that either generate a new representation of  $K$  itself, generate a larger field  $K' \supseteq K$ , or generate a subfield or overlapping field:

- *Normalization*: replacing an extension number by a simpler (by some measure) generator of the same field. Example:  $\mathbb{Q}(-\frac{5}{3}\sqrt{8}) \rightarrow \mathbb{Q}(\sqrt{2})$ ,  $\mathbb{Q}(e^{-\pi}) \rightarrow \mathbb{Q}(e^\pi)$ .
- *Pruning*: removing redundancy. Ex.:  $\mathbb{Q}(-\sqrt{2}, \sqrt{2}) \rightarrow \mathbb{Q}(\sqrt{2})$ .
- *Unification*: replacing extensions by a common generator. Ex.:  $\mathbb{Q}(\sqrt{2}, \sqrt{3}) \rightarrow \mathbb{Q}(\sqrt{2} + \sqrt{3})$ ,  $\mathbb{Q}(\pi^{1/2}, \pi^{1/3}) \rightarrow \mathbb{Q}(\pi^{1/6})$ .
- *Specialization*: simplifying special cases. Ex.:  $\mathbb{Q}(e^0) \rightarrow \mathbb{Q}$ ,  $\mathbb{Q}(e^{\log(z)}) \rightarrow \mathbb{Q}(z)$  and  $\mathbb{Q}(\log(e^z)) \rightarrow \mathbb{Q}(z, \pi, i)$ .
- *Atomization*: rewriting an extension in terms of more “atomic” parts. Ex.:  $\mathbb{Q}(\sqrt{2} + \sqrt{3}) \rightarrow \mathbb{Q}(\sqrt{2}, \sqrt{3})$ ,  $\mathbb{Q}(e^{x+y}) \rightarrow \mathbb{Q}(e^x, e^y)$ , and  $\mathbb{Q}(\log(xy)) \rightarrow \mathbb{Q}(\log(x), \log(y), \pi, i)$ .
- *Function replacement*: rewriting a function in terms of a different function or combination of functions. Ex.:  $\mathbb{Q}(\sin(x)) \rightarrow \mathbb{Q}(e^{ix}, i)$ ,  $\mathbb{Q}(e^{x+yi}) \rightarrow \mathbb{Q}(e^x, \cos(y), \sin(y), i)$ .

The problem of choosing appropriate extension numbers arises in various situations:

- Evaluating functions and solving equations: for example, given  $z$ , construct a field to represent  $\sqrt{z}$  or  $e^z$ .
- Merging fields, especially for arithmetic: given  $z_1 \in K_1 = \mathbb{Q}(a_1, \dots, a_n)$  and  $z_2 \in K_2 = \mathbb{Q}(b_1, \dots, b_m)$ , compute a field  $K_3$  containing  $z_3 = z_1 \circ z_2$  where  $\circ$  is an arithmetic operation.
- Simplifying a single element (or finite list of elements): given  $z \in K$ , construct  $K' \subseteq K$  with  $z \in K'$  that is better suited for deciding a predicate, user output, numerical evaluation, etc.

We can attempt to set reasonable defaults, but a useful system should probably allow the user to make intelligent choices. It is very difficult to define meaningful canonical forms for general symbolic expressions, and the optimal form often depends on the application [5, 19]. A classical problem is whether it makes sense to expand  $(\pi + 1)^{1000}$  in  $\mathbb{Q}(\pi)$  or whether the result should be represented in  $\mathbb{Q}((\pi + 1)^{1000})$  (in Calcium, this is configurable). Although atomization intuitively simplifies extensions, having more variables can slow down the task of constructing the ideal and performing operations in the formal field, and in any case the choice of “atoms” is often somewhat arbitrary.

We follow a conservative approach in Calcium so far: merging fields simply takes the union of the generators, evaluating functions or creating algebraic numbers only normalizes or specializes in trivial cases, and automatic pruning is mainly done to demote

<sup>2</sup>We can imagine a “nonrigorous mode” similar to the algorithm in [1] which looks for numerical integer relations and uses them to simplify symbolic expressions without guaranteeing correctness.

rational numbers to  $\mathbb{Q}$ . In the future, we intend to implement different behaviors and make them configurable, allowing the user to choose different “flavors” of arithmetic (for example, always unifying algebraic numbers to a single extension, always separating complex numbers into real and imaginary parts, etc.).

Algorithm 2 is notably missing heuristics for trigonometric functions and complex parts (real part, imaginary part, sign, absolute value). We can write trigonometric functions and their inverses in terms of complex exponentials and logarithms, and complex parts in terms of algebraic operations and recursive complex conjugation or separation of real and imaginary parts, but this is not always appropriate, particularly when we end up using complex extensions to describe a real field. We leave this problem for future work.

## 2.5 On formal field arithmetic

We conclude this section with some practical comments about implementing formal fields  $\text{Frac}(\mathbb{Q}[X_1, \dots, X_n]/I)$ .

**2.5.1 Normal forms of fractions.** When computing in formal fraction fields, we face a difficulty which does not arise when merely considering quotient rings  $\mathbb{Q}[X_1, \dots, X_n]/I$ : a formal fraction  $p/q$  need not be in a joint canonical form even if  $p$  and  $q$  are in canonical form with respect to  $I$ .<sup>3</sup> A simple example is that  $1/\sqrt{2} = \sqrt{2}/2$ . This is harmless for deciding equality since reduction by  $I$  will give a zero numerator of  $p/q - r/s = (ps - rq)/(pq)$  for equivalent fractions  $p/q$  and  $r/s$ . However,  $p$  and  $q$  can have nontrivial common content in  $\mathbb{Q}[X_1, \dots, X_n]/I$  even if they are coprime in  $\mathbb{Q}[X_1, \dots, X_n]$ , and failing to remove such content can result in expression swell. This problem manifests itself, for example, in Gaussian elimination.

In special cases, it is possible to find content by computing polynomial GCDs over an algebraic number field instead of over  $\mathbb{Q}$ . Monagan and Pearce [18] provide an algorithm that solves the general problem of simplifying fractions modulo an arbitrary (prime) ideal. Their algorithm uses Gröbner bases over modules. We have not yet implemented this method in Calcium, and only remove content in  $\mathbb{Q}[X_1, \dots, X_n]$  from formal fractions (except in the special case of simple algebraic number fields, where we compute a canonical form by rationalizing the denominator).

**2.5.2 Orderings.** The choice of *monomial order* (*lex*, *deglex*, *degrevlex*, etc.), for multivariate polynomials in formal fields can have a significant impact on efficiency and simplification power. Closely related is the *extension number order*: we typically want to sort the extension numbers in order of decreasing complexity  $a_1 > a_2 > \dots > a_n$  for lexicographic elimination. Calcium will thus, for example, automatically rewrite  $\log(4) + \log(2)$  as  $3 \log(2)$  by eliminating the higher-complexity extension  $\log(4)$ . The notion of complexity is somewhat arbitrary, but typically for any symbolic function  $f$  and any  $z$ , we want  $f(z) > z$ . For a discussion of the problem of ordering symbolic expressions, see [5, 19].

Overall, *lex* monomial ordering often seems to perform best due to its tendency to completely eliminate extension numbers of higher complexity, and it is used by default in Calcium, although degree orders sometimes lead to cheaper Gröbner basis computations and overall simpler polynomials. Calcium currently uses a hardcoded comparison function for extension numbers, but we intend to make

<sup>3</sup>For the present discussion, it does not matter whether we have  $I_{\text{red}} = I$ .

it configurable; a more sophisticated system might use heuristics to choose an appropriate extension number order and monomial order (including weighted and block orders) for each extension field.

### 3 ARCHITECTURE OF CALCIUM

In this section, we describe the design of Calcium as a library and discuss certain low-level implementation aspects.

We chose to implement Calcium as a C library to minimize dependencies. Calcium includes a simple, unoptimized Python wrapper (using ctypes) intended for easy testing.

Calcium depends on Arb [15] for arbitrary-precision ball arithmetic, Antic [14] for arithmetic in algebraic number fields, and Flint [13] for rational numbers, multivariate polynomials and other functionality such as factoring and LLL. A central idea behind Calcium is to leverage these libraries for fast in-field arithmetic combined with rigorous evaluation of numerical predicates. At present, we use a naive implementation of Buchberger’s algorithm for Gröbner basis computation, which can be a severe bottleneck.

#### 3.1 Numbers and context objects

The main types in Calcium are context objects (`ca_ctx_t`) and numbers (`ca_t`). The context object is the parent object for a “Calcium field”, representing a lazily expanding subset of  $\mathbb{C}$ . It serves two purposes: it holds a cache of extension numbers and fields, and it specifies work limits and other settings. Examples of configurable parameters in the context object include: the maximum precision for numerical evaluation, precision for LLL, the degree of algebraic number fields, use of Gröbner bases, use of Vieta’s formulas, the maximum  $N$  for in-field expansion of  $(x + y)^N$ . The user may create different contexts configured for different purposes.

The main Calcium number type, `ca_t`, holds a pointer to a field  $K$  and an element of  $K$ . As in GMP, Flint and Arb, `ca_t` variables have mutable semantics allowing efficient in-place operations. Internally, `ca_t` uses one of three possible storage types for field elements:

- A Flint `fmpq_t` if  $K = \mathbb{Q}$ .
- An Antic `nf_elem_t` if  $K = \mathbb{Q}(a)$ ,  $a \in \overline{\mathbb{Q}}$  is a simple algebraic number field. There are two storage sub-types: Antic uses a specialized inline representation for quadratic fields.
- A rational function `fmpz_mpoly_q_t` (implemented as a pair of Flint multivariate polynomials `fmpz_mpoly_t`) if  $K$  is a generic (multivariate or non-algebraic) field. Arithmetic in this representation relies on the Flint functions for multivariate arithmetic, GCD, and ideal reduction. Some functions also use Flint’s multivariate polynomial factorization.

Caching field data in a context object rather than storing the complete description of a field in each `ca_t` variable is essential for performance: creating new fields can be expensive; repeated operations and creation of elements within a field should be cheap.

Calcium is threadsafe as long as two threads never access the same context object simultaneously. The user can most easily ensure this by creating separate context objects for each thread. For fine-grained parallelism, it is most convenient to convert elements to simpler types such as polynomials. Some of the underlying polynomial and matrix operations are parallelized in Flint.

#### 3.2 Fields and extension numbers

Separate types are used internally in the recursive construction of fields. A `ca_ext_t` object defines an extension number. This can be an algebraic number (see below) or a symbolic constant or function of the form  $f(x_1, \dots, x_n)$  where  $x_k$  are `ca_t` arguments and  $f$  is a builtin symbol (Pi, Exp, etc.). A `ca_field_t` object represents a field  $\mathbb{Q}(a_1, \dots, a_n)$  as an array of pointers to the `ca_ext_t` objects  $a_1, \dots, a_n$ . Field objects also store computational data such as the reduction ideal. Unlike field elements, fields and extension numbers are in principle immutable, but cached data may be mutated internally: for example, extension numbers cache Arb enclosures and update this data when the internal working precision is increased.

The `ca_ctx_t` context object stores `ca_ext_t` and `ca_field_t` objects without duplication in hash tables for fast lookup. The context object holds on to all data until it is destroyed by the user. For applications where memory usage could become an issue, an improvement would be to add automatic garbage collection.

#### 3.3 Canonical algebraic numbers

Calcium contains a type `qqbar_t` which represents an algebraic number by its minimal polynomial over  $\mathbb{Q}$  together with an isolating complex interval for a root. Elements of  $\overline{\mathbb{Q}}$  are thus represented canonically, whereas a `ca_t` allows many different representations.

The `qqbar_t` type is used internally to represent absolute algebraic extension numbers and as a fallback to simplify or test equality of algebraic numbers when Algorithm 2 fails to find a sufficient reduction ideal involving the generators of the field. We thus have a complete test for equality in  $\overline{\mathbb{Q}}$ .

An arithmetic operation in the `qqbar_t` representation involves three steps: resultant computation (using the BFSS algorithm [4]), factoring in  $\mathbb{Z}[x]$  (using the van Hoeij algorithm in Flint), and maintenance of the root enclosure (using interval Newton iteration and other methods based on Arb). Factoring typically dominates, and this is usually much more expensive than a `ca_t` operation in a fixed number field. Nevertheless, `qqbar_t` performs better than `ca_t` in some situations and does not require a context object, making it a useful implementation of  $\overline{\mathbb{Q}}$  in its own right.

#### 3.4 Polynomials and matrices

Calcium provides types `ca_poly_t` and `ca_mat_t` for representing dense univariate polynomials and matrices over  $\mathbb{R}$  or  $\mathbb{C}$ . They support arithmetic, predicates, polynomial GCD and squarefree factorization (using the Euclidean algorithm), matrix LU factorization, rank and inverse (using ordinary and fraction-free Gaussian elimination), determinant and characteristic polynomial, and computing roots or eigenvalues with multiplicities. Most algorithms are basic, and optimization could be an interesting future project.

#### 3.5 Predicates and special values

There are two kinds of predicate functions: structural and mathematical. The structural version of the predicate  $x = y$  for `ca_t` variables asks whether  $x$  and  $y$  contain identically represented elements of the same field. This is cheap to check and gives True/False. The mathematical predicate asks whether  $x$  and  $y$  represent the same complex number. This is potentially expensive, if not undecidable, and gives True/False/Unknown. Applications using Calcium

from a high-level language may prefer to throw an exception for Unknown; the included Python wrapper does precisely this:

```
>>> A = ca_mat([[pi, pi**2], [pi**3, pi**4]])
>>> A.det() == 0 # singular matrix
True
>>> (A + (1 - exp(exp(-1000))))).det() == 0 # nonsingular matrix
False
>>> (A + (1 - exp(exp(-10000))))).det() == 0
NotImplementedError: unable to decide predicate: equal
```

(The last example succeeds when the precision limit is raised.)

The `ca_t` type actually represents a set  $\mathbb{C}^{**} \supset \mathbb{C}$  comprising numbers as well as various special values: *unsigned infinity* ( $\infty$ ), *signed infinities* ( $c \cdot \infty$ ), *undefined* ( $u$ ) and *unknown* ( $?$ ). Formally,  $\mathbb{C}^* = \mathbb{C} \cup \{\infty\} \cup \{c \cdot \infty : |c| = 1\} \cup \{u\}$  and  $\mathbb{C}^{**} = \mathbb{C}^* \cup \{?\}$ .

$\mathbb{C}^*$  is a *singularity closure* of  $\mathbb{C}$  in which we can extend partial functions  $f : \mathbb{C} \setminus \text{Sing}(f) \rightarrow \mathbb{C}$  to total functions  $f : \mathbb{C}^* \rightarrow \mathbb{C}^*$ . For example:  $1/0 = \infty$ ,  $\log(0) = -\infty$ , and  $0/0 = \infty - \infty = u$ . The definitions are simply a matter of convenience.

$\mathbb{C}^{**}$  is a *meta-extension* of  $\mathbb{C}^*$  in which algorithms can be guaranteed to terminate. The meta-value ( $?$ ) represents an undetermined element of  $\mathbb{C}^*$ . For example,  $1/x$  evaluates to  $?$  if Calcium cannot decide whether  $x = 0$  since the value could be either a number or  $\infty$ . Logical predicates on  $\mathbb{C}$  extend to logical predicates on  $\mathbb{C}^*$  ( $\infty = \infty$ ,  $u = u$  and  $u \neq 3$  are all True) while predicates on  $\mathbb{C}^{**}$  are triple-valued ( $? = 3$ ,  $? = \infty$ ,  $? = u$  and  $? = ?$  are all Unknown).

Like IEEE 754 floating-point arithmetic, `ca_t` thus supports *non-stop computing*. Unlike IEEE 754, we disambiguate two NaN types (mathematical and computational indeterminacy  $u$  and  $?$ ), we do not distinguish between  $-0$  and  $+0$ , and complex infinities are represented in polar rather than rectangular form (we take  $\infty + 2i = \infty$ ).

We stress the distinction between numbers and special values: a `ca_t` is explicitly a number, explicitly a singularity (infinity or undefined), or explicitly unknown. It is thus easy to restrict usage strictly to  $\mathbb{C}$ , in contrast to many symbolic computation systems where expressions that represent numbers are syntactically indistinguishable from expressions that are singular or undefined.

## 4 RELATED WORK AND BENCHMARKS

The strategy we have discussed is an attempt to unify several existing paradigms for exact computation: *effective real numbers*, *symbolic expressions*, *formal fraction fields*, and *(embedded) number fields and quotient rings*. The novelty is not the combination of functionality (any general-purpose computer algebra system supports the requisite operations), but the implementation form and interface.

### 4.1 Effective numbers

There are many implementations of “effective” or “computable” numbers which construct a symbolic representation to permit lazy numerical evaluation to arbitrary precision, *without* the ability to decide equality [20, 33, 34]. Our representation is more powerful, but likely inferior if the only goal is numerical evaluation: symbolic computations are often slower, and the rewritten expressions can have worse numerical properties (they will sometimes be better). We rather view Calcium as a second option to try if a direct numerical evaluation fails because it stumbles on an exact comparison.

### 4.2 Symbolic and algebraic systems

Most computer algebra systems arguably belong to one of two paradigms. *Algebraic* systems (Singular [9], Magma [3], Pari/GP [29], Sage [30], Nemo/Hecke [11], etc.) are designed for computation in definite algebraic structures, favoring strong data invariants. *Symbolic* ones (Mathematica, Maple, SymPy [17], etc.), are designed around more heuristic manipulation of free-form symbolic expressions. Roughly speaking, algebraic systems tend to prefer  $(x + 1)^{100}$  in expanded (normal) form and view it as an element of a particular ring such as  $\mathbb{Q}[x]$ , while symbolic systems tend to leave it unexpanded and unassociated with a formal algebraic structure. Calcium takes a more algebraic approach to providing functionality for real and complex numbers previously only found in symbolic systems. This has benefits for performance and correctness, although we lose some flexibility: we give up most superficial manipulation of rational expressions (*expand*, *combine*, *apart*, *factor*, etc.).

Calcium is not a general-purpose expression simplifier like the `simplify` or `FullSimplify` routines in systems like Maple and Mathematica, which combine many heuristics. A roundtrip  $expr \rightarrow \text{Calcium} \rightarrow expr$  can be a useful part in the toolbox of such a simplifier, but will often have to be applied selectively.

### 4.3 Algebraic numbers

Computing exactly in  $\overline{\mathbb{Q}}$  is a well-studied problem which admits many different approaches. Here are some examples:

- The Core library [35] represents algebraic numbers as symbolic expressions (DAGs), using interval arithmetic and root separation bounds to resolve exact zero tests.
- Sage’s `QQbar` uses a hybrid representation: an algebraic number exists either as a symbolic expression or as an element of a number field  $\mathbb{Q}(a)$ ; a comparison that cannot be resolved numerically forces a simplification to an absolute field.
- The *algebraically closed field* in Magma [27, 28] uses lazily extended multivariate quotient rings  $\mathbb{Q}[X_1, \dots, X_n]/I$ , building on an idea that first appeared in [10].

Calcium’s model borrows the idea of a hybrid representation from Sage (with arithmetic in fixed number fields for efficiency [8]), but more closely resembles Magma in the use of multivariate quotient rings to handle multiple extensions. It differs from Magma in some respects, including the use of fraction field arithmetic (allowing polynomial denominators) and the use of ball arithmetic and numerical embeddings rather than modular computations.<sup>4</sup>

In comparisons with Sage, we have found that Calcium performs much better, often due to the superiority of a multivariate representation (providing arithmetic simplifications such as  $x - (x - y) = y$  and avoiding the degree explosion of coercing to a univariate field). A simple test is to let  $x = \sqrt{2} + \sqrt{3} + \dots + \sqrt{17}$  (summing over prime numbers) and check  $x - (x - 1) - 1 = 0$ ; this takes 17 s in Sage 9.2 and 0.002 s in Calcium 0.3, including the time to set up and clear a context object (Calcium takes 0.00006 s with the fields cached).<sup>5</sup>

<sup>4</sup>Magma’s ACF is not actually a satisfactory implementation of  $\overline{\mathbb{Q}}$  for our purposes as it does not define the embedding of roots into  $\mathbb{C}$ : choices that depend on permuting roots are made arbitrarily by the system and cannot be predicted by the user.

<sup>5</sup>A more difficult example problem is to test the equality of two large (7000 operations) constant expressions in <https://ask.sagemath.org/question/52653>. Calcium solves this problem in 8 seconds while Sage does not finish in several hours.

We have not attempted a detailed comparison with other algebraic number implementations since our main interest is to incorporate transcendental numbers, as discussed next.

#### 4.4 Elementary numbers

The next interesting structure after  $\overline{\mathbb{Q}}$  is the field of *elementary numbers*. To be precise, there are two common definitions of such a field: the *exp-log field*  $\mathbb{E}$  is the closure of  $\mathbb{Q}$  with respect to exponential and logarithmic extensions  $e^z$  and  $\log(z)$  ( $z \neq 0$ ), while the *Liouvillian field*  $\mathbb{L}$  also allows algebraic extensions.<sup>6</sup>

Richardson [21–26] has constructed a decision procedure for equality in  $\mathbb{E}$  and  $\mathbb{L}$  using computations in towers of extensions over  $\mathbb{Q}$ , which always succeeds if Schanuel’s conjecture is true and will loop forever when given a counterexample. The surface part of such a decision procedure is essentially Algorithm 1, in which we iteratively attempt to either prove inequality or find an algebraic relation that implies equality. (This kind of gradual ideal construction with an asymmetric zero test is also central to [10, 28], and appears in other contexts; see for example Appendix D in [32].)

Assuming Schanuel’s conjecture, it can be shown that any relation between elementary numbers must result from a combination of log-linear relations, exp-multiplicative relations, and relations resulting from the identical vanishing of algebraic functions (for example,  $\sqrt{(\log(2))^2} - \log(2) = 0$ , due to the identical vanishing of  $\sqrt{x^2} - x$  on the local branch). Algorithm 2 is inspired by Richardson’s algorithm, but incomplete: it will find logarithmic and exponential relations, but only if the extension tower is flattened (in other words, we must avoid extensions such as  $e^{\log(z)}$  or  $\sqrt{z^2}$ ), and it does not handle all algebraic functions.

Much like the Risch algorithm, Richardson’s algorithm has apparently never been implemented fully. Mathematica and Maple presumably use similar heuristics to ours, but the details are not documented [6], and we do not know to what extent those systems certify results rigorously. Indeed, we can find examples where many computer algebra systems miscompute. For example, let  $X = 2 \log(\sqrt{2} + \sqrt{3}) - \log(5 + 2\sqrt{6})$  ( $= 0$ ) and construct the  $2 \times 2$  matrix  $A_1 = (0, X; 0, 0)$  as well as  $A_2 = (0, X + e^{-1000}; 0, 0)$ . Maple2020 and Sage 9.2 erroneously compute  $\text{rank}(A_1) = 1$ , while Mathematica 12.2 erroneously computes  $\text{rank}(A_2) = 0$ . The former error is a typical result of assuming that nontrivial expressions are nonzero, the latter a typical result of assuming that numerically small values are zero. Calcium computes the correct rank of either matrix.

A practical difficulty when comparing elementary numbers is that high precision may be needed to distinguish nested exponentials numerically (see the example in section 3.5). This problem can be solved using asymptotic expansions [31] and in some cases using irrationality criteria [2]. We have not investigated such methods.

#### 4.5 Miscellaneous examples

For code and additional benchmarks, we refer to example programs included with Calcium (<http://fredrikj.net/calcium/examples.html>).

**4.5.1 Exact DFT.** As a test<sup>7</sup> of basic arithmetic and simplification, we check  $\mathbf{x} - \text{DFT}^{-1}(\text{DFT}(\mathbf{x})) = \mathbf{0}$  where  $\mathbf{x} = (x_n)_{n=0}^{N-1}$  and  $\text{DFT}(\mathbf{x}) =$

**Table 1: Time (s) for exact DFT and zero test.**

$x_{n-2}$	$N$	Sage $\overline{\mathbb{Q}}$	Sage SR	SymPy	Maple	MMA	Calcium
$n$	6	0.0070	0.44	fail	0.016	0.078	0.00014
	8	0.018	0.11	1.1	0.0060	0.057	0.00016
	16	0.090	58	9.9	0.080	0.27	0.00033
	20	0.14	172	fail	0.13	0.96	0.00045
$\sqrt{n}$	100	8.2	fail	fail	9.1	>60	0.044
	8	1.6	0.52	2.8	0.046	0.11	0.0060
	16	>10 <sup>3</sup>	68	24	0.26	0.58	0.037
	20	>10 <sup>3</sup>	208	fail	1.1	2.3	0.064
$\log(n)$	100	>10 <sup>3</sup>	fail	fail	>10 <sup>3</sup>	>60	17
	8	-	0.48	1.8	0.044	0.29	0.0044
	16	-	53	17	0.37	0.66	0.021
	20	-	188	fail	0.74	45	0.043
$e^{2\pi i/n}$	100	-	fail	fail	>10 <sup>3</sup>	>60	26
	8	>10 <sup>3</sup>	1.1	fail	0.042	0.10	0.0069
	16	>10 <sup>3</sup>	106	>10 <sup>3</sup>	0.17	0.41	0.17
	20	>10 <sup>3</sup>	329	fail	fail	>60	0.24
$\frac{1}{1+n\pi}$	100	>10 <sup>3</sup>	fail	fail	>10 <sup>3</sup>	>60	86*
	8	-	0.68	17	0.072	0.21	0.0035
	16	-	62	>10 <sup>3</sup>	0.32	6.4	0.043
	20	-	219	fail	2.4	>60	0.12
$\frac{1}{1+\sqrt{n}\pi}$	100	-	fail	fail	>10 <sup>3</sup>	>60	202
	8	-	0.76	22	0.074	2.6	0.072
	16	-	>10 <sup>3</sup>	>10 <sup>3</sup>	127	>60	7.2
	20	-	fail	fail	>10 <sup>3</sup>	>60	62

$\sum_{k=0}^{N-1} \omega^{-kn} x_k$  with  $\omega = e^{2\pi i/N}$ . For this benchmark, we evaluate the DFT by naive  $O(N^2)$  summation (no FFT). We test six input sequences exhibiting both algebraic and transcendental numbers. For example, the second test with  $N = 8$  has  $\mathbf{x} = [\sqrt{2}, \sqrt{3}, \dots, \sqrt{9}]$ .

Table 1 shows timings for Calcium 0.3, Sage 9.2 (both QQbar and the symbolic ring SR), SymPy 1.5, Maple 2020, and Mathematica 12.2 (MMA). MMA was run on a faster computer in the free Wolfram Cloud, with a 60 s timeout. Others were interrupted after 10<sup>3</sup> s.

In most cases, we see order-of-magnitude speedups over the competing systems. All timings were done with empty caches; most systems (including Calcium) run faster a second time, but comparisons are difficult since the systems use caches differently.

SymPy fails to prove equality unless  $n = 2^k$ , and Sage’s SR fails except for  $n = 2^k, 3, 5, 6, 10, 12, 20$ ; Maple (with `simplify()`) fails on the fourth test sequence for large  $n$ . The test case marked (\*) only succeeds if we manually disable Gröbner bases in Calcium.

**4.5.2 Conjugate logarithms.** Example 1 in [1] asks for simplifying  $C_1 = -\frac{1}{8}i\pi \log^2(\frac{2}{3} - \frac{2i}{3}) + \frac{1}{8}i\pi \log^2(\frac{2}{3} + \frac{2i}{3}) + \frac{1}{12}\pi^2 \log(-1 - i) + \frac{1}{12}\pi^2 \log(-1 + i) + \frac{1}{12}\pi^2 \log(\frac{1}{3} - \frac{i}{3}) + \frac{1}{12}\pi^2 \log(\frac{1}{3} + \frac{i}{3})$ . Calcium evaluates this to  $\frac{1}{96}(4 \log(\frac{1-i}{3})\pi^2 - 8 \log(\frac{-1-i}{3})\pi^2 - 5\pi^3 i)$ , eliminating redundant logarithms. Calcium does not discover  $C_1 = -\frac{1}{48}\pi^2 \log(18)$  (the output of Mathematica’s `FullSimplify`) since it does not rewrite the extension numbers, but it proves equality when this form is given ( $C_1 + \frac{1}{48}\pi^2 \log(18)$  evaluates to 0). Calcium takes 0.008 s, or 0.00008 s when the fields are already cached; Mathematica’s `Simplify` takes 0.015 s and leaves  $C_1$  unsimplified but proves  $C_1 + \frac{1}{48}\pi^2 \log(18) = 0$  in 0.02 s, while `FullSimplify` takes 0.03 s.

Maple’s `simplify` returns  $-\frac{1}{48}\pi^2(\log(2) + 2\log(3))$  in 0.000024 s. Indeed, this is a trivial computation if we (like Maple) atomize the logarithms. Since this is not yet implemented in Calcium, we see the result of the slower, generic approach using integer relations.

<sup>6</sup>Clearly  $\overline{\mathbb{Q}} \neq \mathbb{E}, \mathbb{E} \subseteq \mathbb{L}$  and  $\overline{\mathbb{Q}} \subseteq \mathbb{L}$ , but it is unknown if  $\mathbb{E} = \mathbb{L}$  and if  $\overline{\mathbb{Q}} \subset \mathbb{E}$ . [7]

<sup>7</sup>Source code: <https://fredrikj.net/blog/2020/09/benchmarking-exact-dft-computation/>

## 5 FUTURE WORK IDEAS

We have already discussed several ideas for future work. Perhaps the most important topics are: new classes of extension numbers; simplification, normalization and context-dependent rewriting of extension numbers; improved numerical algorithms and methods for working with equivalence classes of formal fractions; ideal construction. We conclude by elaborating on the last point.

The usual bottlenecks in constructing ideals (and often in Calcium as a whole) are: searching for integer relations with LLL, proving integer relations through recursive computations, and computing Gröbner bases. Algorithm 2 could be improved in many ways, most notably through preprocessing to avoid redundant work and reduce the dimension or improve numerical conditioning of LLL matrices. Some preprocessing strategies are discussed in [1].

The PSLQ algorithm is often claimed to be superior to LLL for integer relations (see for example [1]), but this is not obviously true with modern floating-point LLL implementations. One benefit of LLL is that we obtain a matrix of all integer relations at once, whereas PSLQ has to be run repeatedly to eliminate relations one by one. We invite further comparison of these algorithms. In some cases, purely symbolic methods should be superior to either.

An explicit Gröbner basis computation can sometimes be avoided by setting up extension numbers and relations appropriately. This is exploited in Magma's algebraically closed field [28]. We have also observed empirically that many calculations in Calcium work perfectly well (and faster) without computing a Gröbner basis, presumably because the constructed ideal basis is sufficiently triangular to be effective for reductions (in some cases, we found that it suffices to compute the Hermite normal form of the LLL output matrices); a better understanding of this phenomenon would be welcome.

A glaring problem is that when introducing  $n$  extension numbers, say by adding  $\sqrt{2} + \sqrt{3} + \sqrt{5} + \dots$ , we construct all the intermediate fields  $\mathbb{Q}(a_1)$ ,  $\mathbb{Q}(a_1, a_2)$ ,  $\dots$ ,  $\mathbb{Q}(a_1, \dots, a_n)$ , from scratch. This is doing nearly  $n$  times more work than should be needed. One possible solution is to let the user write down a list of extension numbers and create  $\mathbb{Q}(a_1, \dots, a_n)$  at once for computations. Another solution is to take advantage of the data that has already been computed for  $\mathbb{Q}(a_1, \dots, a_{n-1})$  to generate the data for  $\mathbb{Q}(a_1, \dots, a_n)$ . This seems hard to solve efficiently and in such a way that the system does not behave unpredictably depending on the order of computations.

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