# 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 $\mathbb{Q}(a_1, \ldots, a_n)$ where the extensions 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.

## 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,[1] a C library for exact computation in $\mathbb{R}$ and $\mathbb{C}$. Numbers are represented as elements of fields $\mathbb{Q}(a_1, \ldots, a_n)$ 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
```

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[1]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/.

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 $\mathbb{Q}$.

Such examples are within the scope of the expression simplification tools in computer algebra systems like Mathematica and Maple.[2] The key difference is that we work with more structured representations; we also handle numerical evaluation and predicates rigorously. Our approach is inspired by various earlier implementations of $\overline{\mathbb{Q}}$ and by theoretical work on transcendental fields.

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*).

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

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

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

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

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

*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 $p, q, r, s \in \mathbb{Q}[X_1, \ldots, X_n]$, we can consequently decide whether $ps \equiv qr \bmod I$. Indeed, we can also decide whether $q, s \not\equiv 0 \bmod I$ 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 $K = \mathbb{Q}(a_1, \ldots, a_n)$ where the numbers $a_1, \ldots, a_n$ 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 $f$.

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[2]Open source systems tend to perform much worse: SymPy, for example, fails to simplify examples 3–5. Maple curiously also fails to simplify example 3.

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

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 [11], 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* $z_1, \ldots, z_n$ *are linearly independent over* $\mathbb{Q}$, *then* $\mathbb{Q}(z_1, \ldots, z_n, e^{z_1}, \ldots, e^{z_n})$ *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 $a_1, \ldots, a_n$, 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 $z \to \overline{z}$.

- 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 $a_1, \ldots, a_n$ 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 $P(a) = 0$ with $P \in \mathbb{C}[X]$ together with an isolating complex ball for a root.

- *Transcendental*: $a$ is a symbolic transcendental (or conjecturally transcendental) constant ($\pi$, $\gamma$, etc.) or function ($e^z$, $\log(z)$, $z^w$, $\Gamma(z)$, $J_\nu(z)$, 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 presently supports extension numbers of the first three types. We represent algebraic and transcendental extensions in the usual way as symbolic expressions $f(z_1, \ldots, z_p)$. The arguments $z_1, \ldots, z_p$ are real or complex numbers which may belong to different fields, say $z_1 \in \mathbb{Q}(b_1, \ldots, b_r)$, $z_2 \in \mathbb{Q}(c_1, \ldots, c_s)$, 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 extension numbers: 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 not feasible to find the ideal $I$ necessary to define a formal field isomorphic to $\mathbb{Q}(a_1, \ldots, a_n)$. 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 typically 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 $I_{\text{red}}$.

The algorithm uses a *work parameter* $W$. This can be taken as a numerical precision in bits for step (c), say with $W_{\min} = 64$ and $W_{\max} = 4096$ and an implied doubling of $W$ on each iteration. We explain the meaning of "heuristics with strength $W$" in step (d) below in section 2.3. If we take $W_{\max} = \infty$ to force a True/False answer, then termination when $z = 0$ is conditional on the asymptotic completeness of the methods to find relations in (d).

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 $a_1, \ldots, a_n$, or we may attempt to prove $\mu(p) = 0$ directly and take $J = \langle p \rangle$. The latter is sometimes easier. For example, if $a_1, \ldots, a_n$ are algebraic extension numbers, it is often cheaper to compute the minimal polynomial specifically for $z$ than to compute all of $I$.

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 $K = \mathbb{Q}(a_1, \ldots, a_n)$.

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

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

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

**Output:** True (implying $z = 0$), False ($z \neq 0$), or Unknown.

---

1. For $W = W_{\min}, \ldots, W_{\max}$, do:

   (a) If $p \equiv 0 \bmod I_{\mathrm{red}}$, return True.

   (b) If it can be certified that $I_{\mathrm{red}} = I$, 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 $I_{\mathrm{red}} \leftarrow I_{\mathrm{red}} \cup J$. (See Algorithm 2.)

2. Return Unknown.

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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 $a_1, \ldots, a_n$ is a tuple $(m_1, \ldots, m_n)$ with some $m_k \neq 0$ such that

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

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 $W \to \infty$ 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).

Algorithm 2 will only find relations that are expressible in terms of the given $a_1, \ldots, a_n$. For example, to add the relation for a square root extension $a_k = \sqrt{z}$ in step A, we need to be able to express $z$ in terms of $K' = \mathbb{Q}(a_1, \ldots, a_{k_1}, a_{k+1}, a_n)$ 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 $(\sqrt{z})^2 \to z$. One possibility is that we always adjoin $z$ (or $b_1, \ldots, b_m$ such that $z \in \mathbb{Q}(b_1, \ldots, b_m)$) to the field where we create $\sqrt{z}$. An alternative is to modify Algorithm 2 so that it can 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

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**Algorithm 2:** Construct ideal of algebraic relations.
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**Input:** Extension numbers $a_1, \ldots, a_n$, a work parameter $W$.
**Output:** A reduction ideal $I_{\mathrm{red}} \subseteq I$ for $\mathbb{Q}(a_1, \ldots, a_n)$.

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Initialize $I_{\mathrm{red}} \leftarrow \{\}$. Depending on $W$, run a subset of A-F:

A **Direct algebraic relations**. For absolute or relative algebraic extensions $a_k$, add the defining relations to $I_{\mathrm{red}}$.

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 $I_{\mathrm{red}}$.

C **Log-linear relations**. Let $L$ denote the set of extension numbers of the form $a_k = \log(z_k)$, along with $\pi i$ if available. Use LLL with precision $W$ to search for relations $\sum_j m_j \log(z_j) = 0$ or $m_0(2\pi i) + \sum_j m_j \log(z_j) = 0$. 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 $I_{\mathrm{red}} \leftarrow I_{\mathrm{red}} \cup \langle m_0(2\pi i) + m_1 a_1 + \ldots + m_n a_n \rangle$.

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 $I_{\mathrm{red}}$ with relations resulting from functional equations and connection formulas such as $\Gamma(z+1) = z\Gamma(z)$ or $\mathrm{erf}(z) = -\mathrm{erf}(-z) = -i\,\mathrm{erfi}(iz)$. 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.

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relation that we will not be able to prove (say, $m\pi + ne = 0$ with $m, n \in \mathbb{Z}$).[3]

## 2.4  Choosing extension numbers

We have so far assumed that the extension numbers $a_1, \ldots, a_n$ 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: $\mathbb{Q}(-\frac{5}{3}\sqrt{8}) \to \mathbb{Q}(\sqrt{2})$, $\mathbb{Q}(e^{-\pi}) \to \mathbb{Q}(e^{\pi})$.

- *Pruning*: removing redundancy. Ex.: $\mathbb{Q}(-\sqrt{2}, \sqrt{2}) \to \mathbb{Q}(\sqrt{2})$.

- *Unification*: replacing extensions by a common generator. Ex.: $\mathbb{Q}(\sqrt{2}, \sqrt{3}) \to \mathbb{Q}(\sqrt{2} + \sqrt{3})$, $\mathbb{Q}(\pi^{1/2}, \pi^{1/3}) \to \mathbb{Q}(\pi^{1/6})$.

- *Specialization*: simplifying special cases. Ex.: $\mathbb{Q}(e^0) \to \mathbb{Q}$, $\mathbb{Q}(e^{\log(z)}) \to \mathbb{Q}(z)$ and $\mathbb{Q}(\log(e^z)) \to \mathbb{Q}(z, \pi, i)$.

- *Atomization*: rewriting an extension in terms of more "atomic" parts. Ex.: $\mathbb{Q}(\sqrt{2} + \sqrt{3}) \to \mathbb{Q}(\sqrt{2}, \sqrt{3})$, $\mathbb{Q}(e^{x+y}) \to \mathbb{Q}(e^x, e^y)$, and $\mathbb{Q}(\log(xy)) \to \mathbb{Q}(\log(x), \log(y), \pi, i)$.

- *Function replacement*: rewriting a function in terms of a different function or combination of functions. Ex.: $\mathbb{Q}(\sin(x)) \to \mathbb{Q}(e^{ix}, i)$, $\mathbb{Q}(e^{x+yi}) \to \mathbb{Q}(e^x, \cos(y), \sin(y), i)$.

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 $z_1 \in K_1 = \mathbb{Q}(a_1, \ldots, a_n)$ and $z_2 \in K_2 = \mathbb{Q}(b_1, \ldots, b_m)$, 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 [20, 5]. 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

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[3]We can imagine an optional "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 $\mathrm{Frac}(\mathbb{Q}[X_1, \ldots, X_n]/I)$.

### 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 $\mathbb{Q}[X_1, \ldots, X_n]/I$: 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 $I$.[4] 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 $\mathbb{Q}[X_1, \ldots, X_n]/I$ even if they are coprime in $\mathbb{Q}[X_1, \ldots, X_n]$, 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 [19] 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 $\mathbb{Q}[X_1, \ldots, X_n]$ 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 $a_1 \succ a_2 \succ \ldots \succ a_n$ for lexicographic elimination. The notion of complexity is somewhat arbitrary, but typically for any symbolic function $f$ and any $z$, we want $f(z) \succ z$. For a discussion of the problem of ordering symbolic expressions, see [20, 5].

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 it configurable or context-dependent. A

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[4]For the present discussion, it does not matter whether we have $I_{\mathrm{red}} = I$.

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 [14] for arbitrary-precision ball arithmetic, Antic [13] for arithmetic in algebraic number fields, and Flint [12] 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.

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## 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 $f(x_1, \ldots, x_n)$ where $x_k$ are `ca_t` arguments and $f$ is a builtin symbol (`Pi`, `Exp`, etc.). A `ca_field_t` object represents a field $\mathbb{Q}(a_1, \ldots, a_n)$ as an array of pointers to the `ca_ext_t` objects $a_1, \ldots, a_n$. 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. Presently, 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. 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 nearly always 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. We anticipate that applications using Calcium from a high-level language will prefer to return True/False and throw an exception for Unknown. The included Python wrapper does precisely this. We illustrate matrix inversion:

```
>>> ca_mat([[1, pi], [0, 1/pi]]).inv()           # nonsingular
[1, -9.86960 {-a^2 where a = 3.14159 [Pi]}]
[0,      3.14159 {a where a = 3.14159 [Pi]}]
>>> ca_mat([[pi, pi**2], [pi**3, pi**4]]).inv()   # singular
   ...
ZeroDivisionError: singular matrix
>>> ca_mat([[1, 0], [0, 1-exp(exp(-10000))]]).inv() # unknown
   ...
ValueError: failed to prove matrix singular or nonsingular
>>> ca_mat([[pi, pi**2], [pi**3, pi**4]]).det()
0
>>> ca_mat([[1, 0], [0, 1-exp(exp(-10000))]]).det()
0e-4342 {-a+1 where a = 1.00000 [Exp(1.13548e-4343 {b})],
                 b = 1.13548e-4343 [Exp(-10000)]}
>>> _ == 0
   ...
ValueError: unable to decide predicate: equal
```

(The third matrix can be inverted by raising the precision limit.)

Like IEEE 754 floating-point arithmetic, `ca_t` also supports *nonstop computing* and allows representing non-finite limiting values. To this end, the `ca_t` type actually represents a set $\mathbb{C}^{**} \supset \mathbb{C}$ comprising numbers as well as various special values: *unsigned infinity* ($\tilde{\infty}$), *signed infinities* ($c \cdot \infty$), *undefined* ($\mathfrak{u}$) and *unknown* (?). Formally, $\mathbb{C}^* = \mathbb{C} \cup \{\tilde{\infty}\} \cup \{c \cdot \infty : |c| = 1\} \cup \{\mathfrak{u}\}$ and $\mathbb{C}^{**} = \mathbb{C}^* \cup \{?\}$. The sets $\mathbb{C}^*$ and $\mathbb{C}^{**}$ are easily implemented on top of $\mathbb{C}$: the `ca_t` type encodes special values by two bits in the field pointer. Unlike IEEE 754, we disambiguate two NaN types (mathematical and computational indeterminacy $\mathfrak{u}$ and ?), we do not distinguish between $-0$ and $+0$, and complex infinities are represented in polar rather than rectangular form (we take $\infty + 2i = \infty$).

$\mathbb{C}^*$ is a *singularity closure* of $\mathbb{C}$ in which we can extend partial functions $f : \mathbb{C} \setminus \mathrm{Sing}(f) \to \mathbb{C}$ to total functions $f : \mathbb{C}^* \to \mathbb{C}^*$. For example: $1/0 = \tilde{\infty}$, $\log(0) = -\infty$, and $0/0 = \infty - \infty = \mathfrak{u}$. The definitions are simply a matter of convenience (this particular choice of singularity closure is largely copied from Mathematica).

$\mathbb{C}^{**}$ 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 $\tilde{\infty}$. Logical predicates on $\mathbb{C}$ extend to logical predicates on $\mathbb{C}^*$ ($\infty = \infty$, $\mathfrak{u} = \mathfrak{u}$ and $\mathfrak{u} \neq 3$ are all True) while predicates on $\mathbb{C}^{**}$ are tripled-valued (? = 3, ? = $\infty$, ? = $\mathfrak{u}$ and ? = ? are all Unknown).

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 essentially an attempt to unify several existing paradigms for exact computation: *effective real numbers*, *symbolic expressions*, *(embedded) number fields*, and *(embedded) 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 [33, 34, 21, 35]. 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 [30], SageMath [31], Nemo/Hecke [10], etc.) are designed for computation in definite algebraic structures, favoring strong data invariants. *Symbolic* ones (Mathematica, Maple, Maxima [17], SymPy [18], 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 is an attempt to provide a more algebraic package for functionality for real and complex numbers previously only found in symbolic systems. The algebraic approach has benefits for performance and correctness, although we lose some flexibility: we notably give up most superficial manipulation of rational expressions (*expand, combine, apart, factor*, etc.), for better or worse.

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* → Calcium → *expr* can be a useful part in the toolbox of such a simplifier, but will often have to be applied selectively. In fact, Calcium grew out of code to manipulate and test symbolic expressions in FunGrim [15], with the view of having a middle layer between symbolic expressions and polynomial and ball arithmetic.

## 4.3 Algebraic numbers

Computing in $\overline{\mathbb{Q}}$ is a well-studied problem which admits multiple approaches [8, 35]. A generally useful principle is to rely on arithmetic in fixed number fields for efficiency. Calcium is partly inspired by Sage's `QQbar`, which uses a hybrid representation: an algebraic number exists either as an element of a number field $\mathbb{Q}(a)$ or an unevaluated symbolic expression. Numerical values are tracked rigorously using interval arithmetic. A comparison that cannot be resolved numerically forces a simplification to an absolute field.

Sage's approach has two problems: unevaluated symbolic expressions fail to capture arithmetic simplifications, and combining extensions to a single absolute number field can be costly. Calcium's multivariate representation often avoids costly simplifications. A simple test case is to compute $x = \sqrt{2} + \sqrt{3} + \ldots + \sqrt{p_n}$ and check $x - (x-1) - 1 = 0$. For $n = 7$ this takes 2200 s in Sage and 0.003 s in Calcium, including the time to set up and clear a context object (Calcium takes 0.00007 s with the fields already cached).[5]

Calcium is also inspired by the *algebraically closed field* in Magma, which uses multivariate quotient rings $\mathbb{Q}[X_1, \ldots, X_n]/I$ that grow automatically [28, 29]. Magma's `ACF` is not actually a satisfactory implementation of $\overline{\mathbb{Q}}$ for our purposes because it does not

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[5]For a more complex test problem that Calcium handles easily where Sage struggles unless carefully guided, see: `https://ask.sagemath.org/question/52653`

define the embedding of polynomial roots into $\mathbb{C}$: choices that depend on permuting roots are made arbitrarily by the system and cannot be predicted by the user. On the other hand, Magma uses more sophisticated methods in its ideal construction than we currently do for algebraic extensions; an interesting future project would be to integrate some of its techniques with our system.

## 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* $\mathbb{L}$ is the closure of $\mathbb{Q}$ with respect to algebraic, exponential and logarithmic extensions.[6]

Richardson [27, 22, 23, 24, 25, 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. 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. We presume that Mathematica and Maple use similar heuristics to ours, but the details are not documented [6], and we do not know to what extent True/False answers are backed up by a rigorous certification in those system.

A practical difficulty when comparing numbers involving elementary functions is that extremely high precision may be needed to distinguish nested exponentials numerically (as an example, consider $\exp(-e^{-e^N}) \neq 1$). This problem can be overcome using asymptotic expansions [32]. We have not yet investigated such methods. The recent work [2] uses irrationality criteria to prove some inequalities, but this is only applicable in very restricted cases.

## 4.5   Miscellaneous examples

This section is short due to page limits. For code and additional benchmarks, we refer to example programs included with Calcium.[7]

### 4.5.1   Exact DFT

As a test of basic arithmetic and simplification, we check $\mathbf{x} - \mathrm{DFT}^{-1}(\mathrm{DFT}(\mathbf{x})) = \mathbf{0}$ where $\mathbf{x} = (x_n)_{n=0}^{N-1}$ and $\mathrm{DFT}(\mathbf{x}) = \sum_{k=0}^{N-1} \omega^{-kn} x_k$ with $\omega = e^{2\pi i/N}$. For this benchmark, we

---

[6]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]

[7]See `http://fredrikj.net/calcium/examples.html` and `http://fredrikj.net/blog/2020/09/benchmarking-exact-dft-computation/` for the complete data for the DFT benchmark.

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

| $x_{n-2}$ | $N$ | Sage $\overline{\mathbb{Q}}$ | Sage `SR` | SymPy | Maple | MMA | Calcium |
|---|---|---|---|---|---|---|---|
| | 6 | 0.020 | 0.047 | fail | 0.016 | 0.078 | 0.00049 |
| | 8 | 0.033 | 0.11 | 1.1 | 0.0060 | 0.057 | 0.00048 |
| $n$ | 16 | 0.15 | 36 | 9.9 | 0.080 | 0.27 | 0.00068 |
| | 20 | 0.22 | 124 | fail | 0.13 | 0.96 | 0.00081 |
| | 100 | 9.2 | fail | fail | 9.1 | $> 60$ | 0.045 |
| | 8 | 5.3 | 0.50 | 2.8 | 0.046 | 0.11 | 0.017 |
| $\sqrt{n}$ | 16 | $> 10^3$ | 46 | 24 | 0.26 | 0.58 | 0.090 |
| | 20 | $> 10^3$ | 154 | fail | 1.1 | 2.3 | 0.17 |
| | 100 | $> 10^3$ | fail | fail | $> 10^3$ | $> 60$ | 38 |
| | 8 | - | 0.20 | 1.8 | 0.044 | 0.29 | 0.0059 |
| $\log(n)$ | 16 | - | 44 | 17 | 0.37 | 0.66 | 0.025 |
| | 20 | - | 136 | fail | 0.74 | 45 | 0.046 |
| | 100 | - | fail | fail | $> 10^3$ | $> 60$ | 26 |
| | 8 | $> 10^3$ | 1.3 | fail | 0.042 | 0.10 | 0.019 |
| $e^{2\pi i/n}$ | 16 | $> 10^3$ | 78 | $> 10^3$ | 0.17 | 0.41 | 0.32 |
| | 20 | $> 10^3$ | 277 | fail | fail | $> 60$ | 1.1 |
| | 100 | $> 10^3$ | fail | fail | $> 10^3$ | $> 60$ | 699* |
| | 8 | - | 0.68 | 17 | 0.072 | 0.21 | 0.0041 |
| $\frac{1}{1+n\pi}$ | 16 | - | 48 | $> 10^3$ | 0.32 | 6.4 | 0.046 |
| | 20 | - | 167 | fail | 2.4 | $> 60$ | 0.12 |
| | 100 | - | fail | fail | $> 10^3$ | $> 60$ | 216 |
| | 8 | - | 0.76 | 22 | 0.074 | 2.6 | 0.082 |
| $\frac{1}{1+\sqrt{n}\pi}$ | 16 | - | $> 10^3$ | $> 10^3$ | 127 | $> 60$ | 8.1 |
| | 20 | - | fail | fail | $> 10^3$ | $> 60$ | 43 |

evaluate the DFT by naive $O(N^2)$ summation (no FFT). We test six input sequences exhibiting both algebraic and transcendental numbers.

We compare with Sage's `QQbar` (algebraics only), Sage's `SR` (using GiNaC), SymPy, Maple, and Mathematica (MMA). MMA was run on a faster computer in the free Wolfram Cloud, with a 60 s timeout. Other systems were interrupted after $10^3$ s. Table 1 shows select timings; 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.

# 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 [29]. 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 $\sqrt{2} + \sqrt{3} + \sqrt{5} + \ldots$, we construct all the intermediate fields $\mathbb{Q}(a_1)$, $\mathbb{Q}(a_1, a_2)$, $\ldots$, $\mathbb{Q}(a_1, \ldots, a_n)$, 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 $\mathbb{Q}(a_1, \ldots, a_n)$ at once for computations. Another solution is to take advantage of the data that has already been computed for $\mathbb{Q}(a_1, \ldots, a_{n-1})$ to generate the data for $\mathbb{Q}(a_1, \ldots, a_n)$. This seems hard to solve efficiently and in such a way that the system does not behave unpredictably depending on the order of computations.

# References

[1] David H. Bailey, Jonathan M. Borwein, and Alexander D. Kaiser. Automated simplification of large symbolic expressions. *Journal of Symbolic Computation*, 60:120–136, January 2014.

[2] Hans-J Boehm. Towards an API for the real numbers. In *Proceedings of the 41st ACM SIGPLAN Conference on Programming Language Design and Implementation*, pages 562–576, 2020.

[3] Wieb Bosma, John Cannon, and Catherine Playoust. The Magma algebra system I: The user language. *Journal of Symbolic Computation*, 24(3-4):235–265, September 1997.

[4] Alin Bostan, Philippe Flajolet, Bruno Salvy, and Éric Schost. Fast computation of special resultants. *Journal of Symbolic Computation*, 41(1):1–29, January 2006.

[5] Jacques Carette. Understanding expression simplification. In *Proceedings of the 2004 international symposium on Symbolic and algebraic computation - ISSAC '04*. ACM Press, 2004.

[6] Jacques Carette. Zero equivalence in computer algebra systems. Mathematics Stack Exchange, `https://math.stackexchange.com/q/3607862`, 2020.

[7] Timothy Y. Chow. What is a closed-form number? *The American Mathematical Monthly*, 106(5):440–448, May 1999.

[8] Henri Cohen. *A Course in Computational Algebraic Number Theory*. Springer Berlin Heidelberg, 1996.

[9] Wolfram Decker, Gert-Martin Greuel, Gerhard Pfister, and Hans Schönemann. Singular 4-1-2 — A computer algebra system for polynomial computations. `http://www.singular.uni-kl.de`, 2019.

[10] Claus Fieker, William Hart, Tommy Hofmann, and Fredrik Johansson. Nemo/Hecke: computer algebra and number theory packages for the Julia programming language. In *Proceedings of the 42nd International Symposium on Symbolic and Algebraic Computation*, ISSAC '17, pages 157–164. ACM, 2017.

[11] S Fischler and T. Rivoal. Effective algebraic independence of values of E-functions, 2019.

[12] William B. Hart. Fast library for number theory: An introduction. In *Mathematical Software – ICMS 2010*, pages 88–91. Springer Berlin Heidelberg, 2010.

[13] William B. Hart. ANTIC: Algebraic number theory in C. *Computeralgebra-Rundbrief: Vol. 56*, 2015.

[14] Fredrik Johansson. Arb: Efficient arbitrary-precision midpoint-radius interval arithmetic. *IEEE Transactions on Computers*, 66(8):1281–1292, August 2017.

[15] Fredrik Johansson. FunGrim: A symbolic library for special functions. In *Mathematical Software – ICMS 2020*, pages 315–323. Springer, 2020.

[16] Manuel Kauers. *Algorithms for Nonlinear Higher Order Difference Equations*. PhD thesis, RISC, Johannes Kepler University, Linz, 2005. `http://www.algebra.uni-linz.ac.at/people/mkauers/publications/kauers05c.pdf`.

[17] Maxima. Maxima, a computer algebra system. version 5.44.0, 2020. `http://maxima.sourceforge.net/`.

[18] Aaron Meurer et al. SymPy: symbolic computing in Python. *PeerJ Computer Science*, 3:e103, January 2017.

[19] Michael Monagan and Roman Pearce. Rational simplification modulo a polynomial ideal. In *Proceedings of the 2006 international symposium on Symbolic and algebraic computation - ISSAC '06*. ACM Press, 2006.

[20] Joel Moses. Algebraic simplification: a guide for the perplexed. *Communications of the ACM*, 14(8):527–537, August 1971.

[21] Norbert Müller. The iRRAM: Exact arithmetic in C++. In *Computability and Complexity in Analysis*, pages 222–252. Springer Berlin Heidelberg, 2001.

[22] Daniel Richardson. The elementary constant problem. In *Papers from the international symposium on Symbolic and algebraic computation - ISSAC '92*. ACM Press, 1992.

[23] Daniel Richardson. A simplified method of recognizing zero among elementary constants. In *Proceedings of the 1995 international symposium on Symbolic and algebraic computation - ISSAC '95*. ACM Press, 1995.

[24] Daniel Richardson. How to recognize zero. *Journal of Symbolic Computation*, 24(6):627–645, December 1997.

[25] Daniel Richardson. Zero tests for constants in simple scientific computation. *Mathematics in Computer Science*, 1(1):21–37, October 2007.

[26] Daniel Richardson. Recognising zero among implicitly defined elementary numbers. Unpublished preprint, 2009.

[27] Daniel Richardson and John Fitch. The identity problem for elementary functions and constants. In *Proceedings of the international symposium on Symbolic and algebraic computation - ISSAC '94*. ACM Press, 1994.

[28] Allan Steel. A new scheme for computing with algebraically closed fields. In *Lecture Notes in Computer Science*, pages 491–505. Springer Berlin Heidelberg, 2002.

[29] Allan Steel. Computing with algebraically closed fields. *Journal of Symbolic Computation*, 45(3):342–372, March 2010.

[30] The PARI Group, University of Bordeaux. *PARI/GP version 2.11.2*, 2019. `http://pari.math.u-bordeaux.fr/`.

[31] The Sage Developers. *SageMath, the Sage Mathematics Software System (Version 9.0)*, 2020. `https://www.sagemath.org`.

[32] Joris van der Hoeven. Automatic numerical expansions. In *Proc. of the conference "Real numbers and computers", Saint-Étienne, France*, pages 261–274, 1995.

[33] Joris van der Hoeven. Computations with effective real numbers. *Theoretical Computer Science*, 351(1):52–60, 2006.

[34] Joris van der Hoeven. Effective real numbers in Mmxlib. In *Proceedings of the 2006 international symposium on Symbolic and algebraic computation - ISSAC '06*. ACM Press, 2006.

[35] Jihun Yu, Chee Yap, Zilin Du, Sylvain Pion, and Hervé Brönnimann. The design of Core 2: A library for exact numeric computation in geometry and algebra. In *Mathematical Software – ICMS 2010*, pages 121–141. Springer Berlin Heidelberg, 2010.