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Abstract. We analyse and compare the complexity of several algorithms for computing modular polynomials. Under the assumption that rounding errors do not influence the correctness of the result, which appears to be satisfied in practice, we show that an algorithm relying on floating point evaluation of modular functions and on interpolation has a complexity that is up to logarithmic factors linear in the size of the computed polynomials. In particular, it obtains the classical modular polynomial \( \Phi_\ell \) of prime level \( \ell \) in time
\[
O \left( \ell^2 \log^3 \ell M(\ell) \right) \subseteq O \left( \ell^3 \log^{4+\varepsilon} \ell \right),
\]
where \( M(\ell) \) is the time needed to multiply two \( \ell \)-bit numbers.

Besides treating modular polynomials for \( \Gamma_0(\ell) \), which are an important ingredient in many algorithms dealing with isogenies of elliptic curves, the algorithm is easily adapted to more general situations. Composite levels are handled just as easily as prime levels, as well as polynomials between a modular function and its transform of prime level, such as the Schl"afli polynomials and their generalisations.

Our distributed implementation of the algorithm confirms the theoretical analysis by computing modular equations of record level around 10000 in less than two weeks on ten processors.

1. Definitions and main result

Modular polynomials, in their broadest sense, are bivariate polynomials with a pair of modular functions as zero. Given any two modular functions \( f \) and \( g \) for arbitrary congruence subgroups, the function fields \( \mathbb{C}(f) \) and \( \mathbb{C}(g) \) are finite extensions of \( \mathbb{C}(j) \), so that there are two polynomials relating \( f \) resp. \( g \) to \( j \). Taking the resultant of these polynomials with respect to \( j \), one sees that a polynomial relationship between \( f \) and \( g \) exists. In practice, one is rather interested in the minimal polynomial of \( f \) over \( \mathbb{C}(g) \), say, that will be called the modular polynomial of \( f \) with respect to \( g \). If the functions satisfy some conditions on the rationality and integrality of their \( q \)-expansion coefficients, the modular polynomial has rational integral coefficients.

Different modular polynomials parameterise moduli spaces related to elliptic curves. Let \( \Gamma = \text{SL}_2(\mathbb{Z})/\{\pm 1\} = \text{PSL}_2(\mathbb{Z}) \) be the full modular group, and \( \mathbb{C}_\Gamma = \mathbb{C}(j) \) the field of modular functions invariant under \( \Gamma \); \( j \) itself parameterises isomorphism classes of elliptic curves. Of special interest for applications is the congruence
subgroup
\[ \Gamma^0(\ell) = \left( \begin{array}{cc} 1 & 0 \\ 0 & \ell \end{array} \right)^{-1} \Gamma \left( \begin{array}{cc} 1 & 0 \\ 0 & \ell \end{array} \right) \cap \Gamma = \left\{ \left( \begin{array}{cc} a & b \\ c & d \end{array} \right) \in \Gamma : \ell \mid b \right\} \]

for \( \ell \) prime; its modular polynomial parameterises isomorphism classes of elliptic curves together with an isogeny of degree \( \ell \). This is heavily used in Atkin’s and Elkies’s improvements to Schoof’s algorithm for counting points on elliptic curves \[8\]. Other applications include the computation of the endomorphism ring of an elliptic curve \[25, 17\] or of an isogeny between two elliptic curves \[19\].

More general modular polynomials occur in modern complex multiplication constructions for elliptic curves \[14, 13, 11\]; they are discussed in Section 4. When counting points on an elliptic curve by the Schoof–Elkies–Atkin algorithm, the modular polynomials are usually supposed to be available after a precomputation phase. This makes sense in situations where the algorithm is carried out many times for curves of about the same size, as is typical in cryptography. When establishing point counting records \[12\], however, one quickly realises that the computation of the modular polynomials itself becomes a bottleneck that may even dominate from a complexity theoretic point of view.

In this article, after surveying in Section 2 the state of the art as it appears in the literature, I present an algorithm based on evaluation and interpolation that computes modular polynomials in time essentially linear in the bit size of the polynomial. As it is based on floating point computations, one has to assume that rounding errors do not disturb the output, a heuristic that is supported by the implementation described in Section 5. A precise analysis of rounding errors to obtain a rigorously proved result appears to be out of reach. Note, however, that the correctness of the output may be checked probabilistically. For instance, one may instantiate the variable \( j \) occurring in the modular polynomial by the \( j \)-invariant of an elliptic curve over a finite field and verify that the specialised polynomial has the expected splitting behaviour.

Under the assumption that it is sufficient to use a floating point precision of \( O(n) \) for a polynomial whose largest coefficient has \( n \) digits, the following holds:

**Result 1.1** (heuristic). Let \( \Gamma' \subseteq \Gamma \) be a congruence subgroup, \( f \) a modular function for \( \Gamma' \) and \( \Phi(X) \in \mathbb{C}(j)[X] \) the characteristic polynomial of \( f \) with respect to the function field extension \( \mathbb{C}(j)/\mathbb{C}(j) \). Assume that \( \Phi \) has coefficients in \( \mathbb{Z}[j] \), and that the largest integer coefficient occurring in it has \( n \) digits. Suppose that a system of representatives of \( \Gamma' \backslash \Gamma \) is known and that \( f \) can be evaluated at precision \( O(n) \) in time \( O(\log n M(n)) \), where \( M(n) \leq O(n \log^{1+\varepsilon}) \) is the time needed to multiply two numbers with \( n \) digits. Then there is an algorithm that computes \( \Phi \) in time

\[ O \left( \deg_X \Phi \deg_j \Phi \left( \log^2 \deg_X \Phi \deg_j \Phi \right) + \log n \right) M(n) \).

In particular, the classical modular polynomial \( \Phi_\ell \) such that \( \Phi_\ell(j(z), j(\ell z)) = 0 \) for \( \ell \) prime is obtained in time

\[ O \left( \ell^2 \log^2 \ell M(\ell \log \ell) \right) \subseteq O \left( \ell^3 \log^{4+\varepsilon} \ell \right). \]

If \( \Phi \) is a dense polynomial and all of its coefficients (or at least a constant fraction of them) are of bit size about \( n \), then this algorithm is linear in the size of the polynomial, except for the logarithmic factors.

The time bound of \( O(\log n M(n)) \) for evaluating modular functions at precision \( n \) is motivated by an algorithm due to Dupont; it relies on Newton iterations on an
expression involving the arithmetic-geometric mean and reaches this complexity for
a wide class of modular functions, including those built from Dedekind’s \( \eta \) function
and in particular \( k, k' \) and \( j \) [6]. Alternatively, one
may use an algorithm relying on multievaluation of \( q \)-expansions as described in
[10, Section 6.3]. Its complexity, worse by a factor of \( \log n \), is still sufficient for
the running time of Result 1.1 to hold. The bound \( M(n) \in O(n \log^{1+\epsilon}) \) may be
achieved by the classical Schönhage–Strassen algorithm [30] or the more recent and
asymptotically faster algorithm due to Fürer [18].

The roots of \( \Phi \), that is, the algebraic conjugates of \( f \), are given by the
\( f(M_\nu z) \) with \( \Gamma' \) running through a system of representatives of the residue classes \( \Gamma' \backslash \Gamma \) (see [6]), so that

\[
(1.1) \quad \Phi(X) = \prod_{\nu=1}^{[\Gamma:\Gamma']} (X - f(M_\nu z)) = X^{[\Gamma:\Gamma']} + \sum_{r=1}^{[\Gamma:\Gamma']} c_r X^{[\Gamma:\Gamma']-r},
\]

where the coefficients \( c_r \) of \( \Phi \) are the elementary symmetric functions of the conju-
gates \( f(M_\nu z) \).

By the \( q \)-expansion principle (cf. [6, §3]), a sufficient condition for the \( c_r \) to be
polynomials in \( \mathbb{Z}[j] \) is that \( f \) has rational integral \( q \)-coefficients, is holomorphic in
\( \mathbb{H} = \{ z \in \mathbb{C} : \Im z > 0 \} \) and all conjugates \( f \left( \frac{\alpha z + b}{cz + d} \right) \) with \( \begin{pmatrix} a & b \\ c & d \end{pmatrix} \in \Gamma \) have integral
algebraic \( q \)-coefficients.

2. APPROACHES TO COMPUTING MODULAR POLYNOMIALS FOR \( \Gamma^0(\ell) \)

The case of \( \Gamma^0(\ell) \) with \( \ell \) prime, which is the most important one for applications,
is also the simplest one from a theoretical point of view: The residue classes \( \Gamma^0(\ell) \backslash \Gamma \) are represented by
\[
T^\nu = \begin{pmatrix} 1 & \nu \\ 0 & 1 \end{pmatrix}, \nu = 0, \ldots, \ell - 1, \text{ and } S = \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix}.
\]
(The literature often concentrates on \( \Gamma_0(\ell) = \begin{pmatrix} \ell & 0 \\ 0 & 1 \end{pmatrix}^{-1} \Gamma \begin{pmatrix} \ell & 0 \\ 0 & 1 \end{pmatrix} \cap \Gamma \). Since
\( \Gamma_0(\ell) = S^{-1} \Gamma_0(\ell) S \), a modular function \( f \) for \( \Gamma_0(\ell) \) yields the function \( f(Sz) \) for
\( \Gamma_0(\ell) \), and \( f(Sz) \) is actually a conjugate of \( f \). So both result in the same modular polynomial \( \Phi \). Or phrased differently: \( \mathbb{C}_{\Gamma_0(\ell)} \backslash \mathbb{C}_\Gamma \) is not normal, and another one of its embeddings is \( \mathbb{C}_{\Gamma_0(\ell)} \backslash \mathbb{C}_{\Gamma'} \).

Different functions have been suggested in the literature:

- \( j(z/\ell) \), leading to the so-called “classical” or “traditional” modular polyno-
  mials;
- \( \eta(z)^{2s} = \left( \frac{\eta(z/\ell)}{\eta(z)} \right)^{2s} \) for \( s = 12/\gcd(12, \ell - 1) \), yielding the so-called
  “canonical” modular polynomials;
- functions that are moreover invariant under the Frick–Atkin–Lehner invo-
  lution \( z \mapsto \frac{-z}{\ell} \); in particular, functions of the form
  \[
  T_r(\eta(z)\eta(\ell z)) \quad \frac{\eta(z)}{\eta(z)}
  \]
evaluated in \( \frac{-z}{\ell} \) with \( T_r \) the \( r \)-th Hecke operator and linear combinations
thereof as described in [27, Ch. 5.3].
All these functions lead to modular polynomials with coefficients in \( \mathbb{Z} \).

1. Heights of modular polynomials. The (logarithmic) height of a modular polynomial with coefficients in \( \mathbb{Z} \) is defined as the logarithm of the largest absolute value of the coefficients. It provides a lower bound on the arithmetic precision required to compute the polynomial. For the classical polynomial between \( j(z) \) and \( j(z/\ell) \), the height is given by

\[
6(\ell + 1) (\log \ell + O(1)) \leq O(\ell \log \ell)
\]

according to [4]. For alternative modular polynomials, one observes a similar growth of the coefficients with \( \ell \), but the constant is usually smaller. It should be possible to adapt the argumentation in [4] to obtain a similar bound for further classes of modular polynomials.

2. Using \( q \)-expansions. The system of representatives of \( \Gamma_0(\ell) \backslash \Gamma \) and thus the conjugates of the modular function \( f \) being explicitly known, one may use [13] to compute the modular polynomial \( \Phi \), if only a procedure for recognising its coefficients \( c_\nu \) as polynomials in \( \mathbb{Z}[j] \) is available. The straightforward and most popular approach is to use \( q \)-expansions: The expansion of \( j \) is of the form \( q^{-1} + \cdots \), so that knowing the expansion \( c_\nu = c_{r,k} q^{-k} + \cdots \), one deduces that \( c_\nu \) has leading term \( c_{r,k} j^k \) as a polynomial in \( j \); one subtracts the \( q \)-expansion of this term and continues in the same vein. In fact, only the non-positive \( q \)-powers in the series expansions of the \( c_\nu \) are needed. Several variants of the algorithm are described in [25] with a brief complexity analysis in [8, Section 3]. We shall describe only the algorithm with the best complexity and give a detailed analysis of its running time.

Let \( f \) be a modular function for \( \Gamma_0(\ell) \) with valuation \( v < 0 \) at infinity; that is, \( f \) admits a \( q \)-expansion of the form

\[
f(z) = \sum_{k=v}^{\infty} a_k q^{k/\ell} \in \mathbb{C}((q^{1/\ell})) \text{ with } q^{1/\ell} = e^{2\pi i z/\ell}.
\]

Suppose that the \( a_k \) are rational integers. (For instance, \( f = j(z/\ell) \) resp. \( f = \pi^2 z^s \) are valid choices with \( \ell \) and \( \ell \) resp. \( v = -\frac{s(\ell-1)}{12\ell} \). Then the \( q \)-expansions of the conjugates \( f_\nu = f(T^\nu z) \) are of the same form with the coefficients \( a_k \) being multiplied by powers of \( q \); that is, \( f \) is of the form \( a_k q^{k/\ell} \) for \( k = v, v + 1, \ldots \). Then the \( q \)-expansion approach.) Denote by \( v_\infty \) the order of \( f_\infty \) at infinity. For \( f = j(z/\ell) \), one has \( f_\infty = j(\ell z) = q^{-\ell} + \cdots \) and \( v_\infty = -\ell \), and for \( f = \pi^2 z^s \), the transformation behaviour of \( \eta \) yields \( f_\infty = \left( e^{2\pi i z} \right)^{2s} = \ell^s q^{\ell^{-1}/12} + \cdots \) with \( v_\infty = \frac{s(\ell-1)}{12} = \frac{\ell^{-1}}{\gcd(12, \ell-1)} \).

In particular, \( v_\infty \) may be positive or negative. If it is negative, the term with lowest \( q \)-exponent occurring in [13] is \( f_\infty = \prod_{\nu=v}^{\ell-1} f_\nu \) of exponent \( \ell v + v_\infty \); so the degree of \( \Phi \) in \( j \) is \( |\ell v + v_\infty| \), reached for \( c_{r,0} \). For instance, the degree in \( j \) is \( \ell + 1 \) for the classical modular polynomial \( \Phi_\ell \). If \( v_\infty \) is positive, the term with lowest \( q \)-exponent is \( \prod_{\nu=0}^{\ell-1} f_\nu \), so the degree \( |\ell v| \) of \( \Phi \) in \( j \) is reached in \( c_\ell \). For instance, the degree in \( j \) of the canonical modular polynomials is \( \frac{s(\ell-1)}{12} \).
We first consider the complexity of determining the partial product
\[ \prod_{\nu=0}^{\ell-1} (X - f_{\nu}) = X^\ell + \sum_{r=1}^{\ell} \tilde{c}_r X^{\ell-r} \in \mathbb{Z}[\zeta][(q^{1/\ell})][X]. \]

One may proceed by computing the Newton sums \( s_r = \sum_{\nu=0}^{\ell-1} f_{\nu}^r \) for \( 1 \leq r \leq \ell \) and use Newton’s recurrence formulæ to deduce the \( \tilde{c}_r \). Following \cite{26}, one obtains with
\[ f' = \sum_{k=rv}^{\infty} a_{k,r} q^{k/\ell} \]
that the valuation of \( \tilde{c}_r \) is also bounded below by \( [rv] \) and that it is also known up to \( q^{[rv]+d} \). Then one shows by induction on Newton’s relation
\[ \tilde{c}_r = \frac{1}{r} \sum_{k=1}^{r} (-1)^{k+1} s_k \tilde{c}_{r-k} \]
that the valuation of \( \tilde{c}_r \) is also bounded below by \( [rv] \) and that it is also known up to \( q^{[rv]+d} \).

If \( v_\infty \geq 0 \), we need only the terms of the \( \tilde{c}_r \) with non-positive exponents; for \( v_\infty < 0 \), we loose some precision, in fact \( |v_\infty| \) terms, through the final multiplication by \( X - f_\infty \). Thus, the value required for \( d \) is given by
\[ d = \ell |v| + \max(0, -v_\infty) = \deg j \Phi. \]

This is an indication that the algorithm given above should be optimal among those relying on \( q \)-expansions: Outputting polynomials with up to \( \deg j \Phi + 1 \) coefficients, it manipulates series with as many terms. Unfortunately, the computation of the \( s_r \) involves decimating the series for \( f' \) by \( \ell \). Eventually, a factor of \( \ell \) is lost in the running time since \( O(\ell d) \) terms of the \( f' \) are needed.

Once the \( q \)-coefficients of \( f \) are known, the \( f' \) and \( s_r \) are thus computed by \( O(\ell) \) multiplications of series with \( O(\ell d) \) terms and the \( \tilde{c}_r \) by \( O(\ell^2) \) multiplications of series with \( O(d) \) terms. Let \( M_q(d) \) be the number of arithmetic operations in \( \mathbb{Z} \) required to multiply two dense \( q \)-expansions with \( d \) terms. As \( M_q \) is at least linear, the total complexity of the series computations becomes
\[ O(\ell M_q(\ell d)). \]

For functions \( f \) related to \( \eta \) quotients, such as \( f = w_2^2 \) or \( f = j(z/\ell) = \frac{w_2^2(z/\ell)}{w_2^2(z/\ell)+16} \), the effort for computing their \( q \)-expansions is negligible, since it corresponds to a constant number of arithmetic operations with series starting from the easily written down series expansion of \( \eta \).

To write the \( c_r \) as polynomials in \( j \), one has to compute the non-positive parts of the \( q \)-expansions of the \( j^k \) for \( 1 \leq k \leq d \). This can be done in time \( O(d M_q(d)) \), which is negligible compared to the previous steps because \( d \in O(\ell) \). Identifying the \( c_r \) as polynomials then corresponds to solving triangular systems of linear equations, requiring altogether \( O(\ell d^2) \) operations with integers.
The total complexity thus becomes
\[ O \left( \ell M q (\ell^2) \right) \]
integer operations.

Let \( n \in O(\ell \log \ell) \) as in Section 2.1 be a bound on the height of the modular polynomial. Then the bit complexity of the algorithm is given by
\[ O \left( \ell^3 \log \ell M(n) \right) \subseteq O \left( \ell^4 \log^{3+\varepsilon} \ell \right) \]

In [3, Appendix A] it is argued that in fact the running time is higher, on grounds that the coefficients of the powers of \( j \) grow faster than \( O(\ell \log \ell) \). However, this objection can be dismissed by carrying out all computations modulo a sufficiently large prime of bit size \( n \in O(\ell \log \ell) \). Alternatively, and probably preferably in practice, one may work modulo small primes and use the Chinese remainder theorem. Both approaches yield the desired complexity.

Remark. If \( \Phi_\ell \) is sought only modulo some prime \( p \), then the complete algorithm can be carried out modulo \( p \) as soon as \( p \) is larger than \( \ell \) to make the divisions in (2.2) possible. The bit complexity becomes
\[ O \left( \ell^3 \log \ell \log p \log \log p \right)^{1+\varepsilon} \]

2.3. Charles–Lauter. In [3], the authors describe an algorithm to compute the classical modular polynomial \( \Phi_\ell \) directly modulo a prime \( p \) without recourse to \( q \)-expansions of modular functions. Instead, they rely on the moduli interpretation of \( \Phi_\ell \), which characterises pairs of \( \ell \)-isogenous elliptic curves. So the algorithm manipulates only elliptic curves and explicit isogenies over extension fields of \( \mathbb{F}_p \).

Its basic building block is the computation of the instantiated polynomial \( \overline{\Phi} = \Phi_\ell(X, \overline{j}) \) with \( \overline{j} \in \mathbb{F}_{p^2} \) the \( j \)-invariant of a supersingular elliptic curve \( E \). The roots of \( \overline{\Phi} \) are the \( j \)-invariants of the \( \ell+1 \) curves that are \( \ell \)-isogenous to \( E \), and that may be obtained via Vlu’s formulae [31] once the complete \( \ell \)-torsion of \( E \) is known. These are \( \ell^2 \) points defined over an extension of \( \mathbb{F}_{p^2} \) of degree \( O(\ell) \) in the supersingular case; in fact, one expects a degree of \( \Theta(\ell) \) virtually all of the time. Thus writing down the \( \ell \)-torsion points requires a time of \( \Omega(\ell^3 \log p) \).

After having repeated the procedure for \( \deg_j \Phi_\ell = \Theta(\ell) \) different values of \( \overline{j} \), one obtains the coefficients of \( \Phi_\ell \) modulo \( p \) by interpolation. (This is analogous to the floating point approach of Section 3, and more details are given there.) The complexity of the algorithm is thus at least
\[ \Omega \left( \ell^4 \log p \right) \]
and an upper bound of \( O \left( \ell^{4+\varepsilon} \log^{2+\varepsilon} p + \log^{4+\varepsilon} p \right) \) is proved in [3, Theorem 3.2] under the generalised Riemann hypothesis.

Hence, this algorithm is slower by a factor of order \( \ell \) than the one presented in Section 2.2 relying on \( q \)-expansions. This is due to the costly determination of isogenies via their kernels: While the isogenies are defined over \( \mathbb{F}_{p^2} \), their kernels lie in an extension of degree \( O(\ell) \). The \( q \)-expansions of the conjugates of \( j(z/\ell) \), on the other hand, provide a synthetic description of the curves that are \( \ell \)-isogenous to the one with \( j \)-invariant \( j(z) \), and using them it is sufficient to carry out all computations in \( \mathbb{F}_p \) in order to obtain \( \Phi_\ell \mod p \).
The approach for calculating modular polynomials that is described in this section is in fact neither new nor particularly involved. It has been suggested to me by R. Schertz during our work on the class invariants of [14], and later R. Dupont pointed out to me that it can already be found in [1, Chapter 4.5]. However, it has not been noticed before that the algorithm allows to lower the exponent of the computational complexity by 1 and thus to reach an essentially optimal complexity if fast polynomial arithmetic and fast techniques for evaluating modular functions as described in [7, 10] are used.

The basic idea of the evaluation–interpolation approach is to specialise and to compute the identity (1.1) between modular functions in several complex floating point arguments (the evaluation phase); and then to interpolate the coefficients in order to recognise them as polynomials in $j$. Precisely, (1.1) can be rewritten as

$$
\Phi(X, j(z)) = \prod_{\nu=1}^{[\Gamma:\Gamma']} (X - f(M_{\nu}z)) = X^{[\Gamma:\Gamma']} + \sum_{r=1}^{[\Gamma:\Gamma']} c_r(z)X^{[\Gamma:\Gamma']-r}
$$

for all $z$ in the upper complex half plane.

Evaluating the conjugates $f(M_{\nu}z)$ of $f$ in a number of complex arguments $z_k$ with $\Im z_k > 0$, multiplying out the left hand side as a polynomial in $C[X]$ and separating the coefficients according to powers of $X$ yields the values $c_r(z_k)$. Writing $c_r = \sum_{s=0}^{\deg_j \Phi} c_{r,s}j^{\deg_j \Phi-s}$, the $c_r(z_k)$ are actually the values of the polynomial $c_r$ in $j(z_k)$, so that the coefficients $c_{r,s} \in C$ can be retrieved by interpolation as soon as the $c_{r,s}(z_k)$ are known for $\deg_j \Phi + 1$ values of $z_k$. The final step is to round the $c_{r,s}$ to rational integers. The degree of the modular polynomial in $j$ or an upper bound on it may be known beforehand; if it is not, then the algorithm can be repeated with increasing guesses for $\deg_j \Phi$ until rounding to integers succeeds (and doubling the guess at each time results in the same asymptotic complexity as taking the correct value from the beginning).

Let $E(n)$ be the bit complexity for evaluating the modular functions $f$ or $j$ at a precision of $n$ bits (here, $f$ is considered to be fixed, while $n$ tends to infinity with $\ell$). Then the evaluation phase requires $(\ell + 1)(\deg_j \Phi + 1)$ evaluations of $f$ at a cost of

$$
O(\ell \deg_j \Phi E(n))
$$

and the reconstruction of $\deg_j \Phi + 1$ polynomials of degree $\ell + 1$ from their roots. Multiplying complex polynomials by the FFT, this reconstruction step takes

$$
O(\deg_j \Phi \ell \log^2 \ell M(n) + \log n M(n)),
$$

where the (eventually negligible) term $\log n M(n)$ accounts for the computation of a primitive root of unity of sufficiently high order to carry out all the FFTs, and as usual $M(n)$ is the time needed to multiply two $n$-bit numbers.

The interpolation phase consists of $\ell$ interpolations of polynomials of degree $\deg_j \Phi$. Employing again fast algorithms such as [20, Algorithm 10.11], this takes

$$
O(\ell \deg_j \Phi \log^2(\deg_j \Phi) M(n)),
$$

once the roots of unity are available.

It is shown in [8] that among others, Dedekind’s $\eta$ function can be evaluated at precision $n$ in time $E(n) \in O(\log n M(n))$ uniformly for the argument
in $F = \{ z \in \mathbb{H} : |\Re z| \leq \frac{1}{2}, |z| \geq 1 \}$. (The restriction to $F$ is not crucial since we may freely choose our interpolation points.) So in particular, all functions built from $\eta$ such as $j$ and $w_\ell$ satisfy $E(n) \in O(\log n M(n))$. (Alternatively, one may use multievaluation as described in [10, Section 6.3] with a slightly worse amortised cost of $O(\log^2 n M(n))$ per value, which still allows to reach the final complexity of Result 1.1.)

In total, the steps add up to a running time of

$$O\left((\ell \deg j \Phi \log^2 \max(\ell, \deg j \Phi) + \log n)M(n)\right) \subseteq O\left((\deg_X \Phi \deg j \Phi \log^2 \max(\deg_X \Phi, \deg j \Phi) + \log n)M(n)\right),$$

and the latter formulation is valid for arbitrary congruence subgroups $\Gamma'$ in the place of $\Gamma^0(\ell)$.

Here, $n$ must be at least as large as the logarithmic height of $\Phi$, and maybe bigger to account for rounding errors. In the case of the classical modular polynomial $\Phi_\ell$ the bound $O(\ell \log \ell)$ of Section 2.1 yields a complexity of

$$O\left(\ell^2 \log^2 \ell M(\ell \log \ell)\right).$$

This proves Result 1.1.

4. Further kinds of modular polynomials

4.1. Modular functions of composite level. Besides its better complexity compared to the approach using $q$-expansions, the evaluation–interpolation algorithm has the advantage of a great flexibility, which makes it easily adaptable to a variety of different modular polynomials. In fact, the proof of Result 1.1 in Section 3 does not use special properties of $\Gamma^0(\ell)$ and is valid for any congruence subgroup.

An application is given by the computation of the polynomial relationship between $j$ and a modular function $f$ of composite level $N$. At first sight, these polynomials are not of great interest. In the point counting or more generally isogeny computation context, it is more efficient to express an isogeny of composite degree as a composition of prime degree isogenies. But this kind of modular polynomials occurs naturally in the context of $\ell$, in which elliptic curves with complex multiplication are obtained via modular functions $f$ of composite level. Levels $N = pq$ or $N = p^2$ with $p$ and $q$ prime are examined in [14], but more general settings can be devised in a straightforward way. The polynomial relationship between $f$ and $j$ is used to derive from a special value of $f$ the $j$-invariant of the corresponding elliptic curve.

If the $q$-expansion approach were to be pursued to compute this kind of modular polynomials, it would be necessary to somehow obtain the $q$-expansions of the conjugates $f(M_x z)$ for a system of representatives $(M_x)$ of $\Gamma^0 \backslash \Gamma$. This is straightforward only for translations; for all other matrices, it is necessary to take the transformation behaviour of $f$ under unimodular matrices into account, which requires ad hoc computations for each particular function. (This is illustrated by $j(Sz/\ell) = j(\ell z)$ and $w_\ell^2(Sz) = \ell^6 \left( \frac{\eta(z)}{\eta(\ell z)} \right)^{2s}$, which are not derived from $j(z/\ell)$ resp. $w_\ell(z)$ by the same generic transformation.) In the case of $\Gamma^0(\ell)$, only the matrix $S$ asks for special treatment; in the case of composite level, many more
special matrices appear. For instance, the full article [12] is concerned with deriving the conjugates of only the functions \( \left( \frac{\eta(z/p_1)\eta(z/p_2)}{\eta(z)} \right)^s \) for \( p_1, p_2 \) prime and \( s = 24/\gcd(24, (p_1 - 1)(p_2 - 1)) \).

In the evaluation–interpolation approach, the only difference between \( \Gamma^0(\ell) \) and other congruence subgroups \( \Gamma' \) is the enumeration of the system of representatives \( (M_\nu) \). For most interesting \( \Gamma' \) (and in particular for arbitrary \( \Gamma^0(N) \)), this is easy; in any case, this step is independent of the particular function \( f \). Then Result 1.1 applies, and one obtains an algorithm that is essentially linear in its output size.

4.2. Schl"afli equations. In [29], Schl"afli examines transformations of prime level \( \ell \) of special modular functions different from \( j \) that lead to particularly simple modular equations. Weber gives a systematic treatment of them in [32, §§73–74]. Let \( f = \zeta_{48} \gamma_{(z+1)/24} = \zeta_{48}w_2(z+1) \) be one of “the Weber functions”, a modular function of level 48. Let \( \ell \) be a prime not dividing this level. Then \( g(z) = f(z/\ell) \) is the root of a monic polynomial \( \Phi^f(X) \) of degree \( \ell + 1 \) with coefficients in \( \mathbb{Z}[f] \).

The evaluation–interpolation approach allows to easily obtain this polynomial with only minimal modifications to the algorithm. The function \( f \) being modular for \( \Gamma(48) \) and \( \ell \) being different from 2 and 3, a quick computation reveals that \( g \) is modular for \( \Gamma(48) \cap \Gamma^0(\ell) \). The polynomial \( \Phi^f(X) \) is thus given by an equation analogous to (1.1):

\[
\Phi^f(X) = \prod_{\nu=1}^{\ell+1} (X - g(M_\nu z)) = X^{\ell+1} + \sum_{r=1}^\ell c_r X^{\ell+1-r},
\]

where the \( M_\nu \) range over a set of representatives of \( \Gamma(48) \cap \Gamma^0(\ell) \setminus \Gamma(48) \). Such a set may be obtained by multiplying the standard representatives of \( \Gamma^0(\ell) \setminus \Gamma \) from the left by a matrix in \( \Gamma^0(\ell) \) such that they end up in \( \Gamma(48) \). Precisely, a possible set of representatives is given by \( \left( \begin{array}{cc} 1 & 48
\nu \\ 0 & 1 \end{array} \right) \) for \( \nu = 0, \ldots, \ell - 1 \) (corresponding to the translations) and \( \left( \begin{array}{cc} 1 - 48k & 48k \\ -48k & 1 + 48k \end{array} \right) \) with \( k = 48^{-1} \mod \ell \), corresponding to the inversion \( S \).

So the only modification required for the evaluation–interpolation algorithm to work is this adaptation of the matrices \( M_\nu \), and of course \( j \) has to be replaced by \( f \) in the interpolation phase to recover the coefficients \( c_r \) as elements of \( \mathbb{Z}[f] \).

Hence Result 1.1 clearly applies also to the Schl"afli equations, showing that they can be computed in essentially linear time with respect to the output size.

Taking specifics of the function \( f \) into account, a more efficient algorithm can be obtained, gaining a constant factor. Weber shows in [12, p. 266] that only every 24-th coefficient in \( f \) of the \( c_r \) may be non-zero. Precisely, \( f^g^t \) having a non-zero coefficient implies \( \ell t + k \equiv \ell + 1 \mod 24 \). Hence the number of evaluations can be reduced by a factor of about 24. (In [10 Ch. 4.5], similar sparse modular equations are studied, in which one out of eight coefficients is non-zero. The Borwein’s suggest in this chapter the evaluation–interpolation approach while profiting of this sparseness.) Weber shows that the \( \Phi^f(X) \) are symmetric, which could also be taken into account during the interpolation phase.
Weber derives the Schl"afli equations in a very compact form. He considers the new functions

\[ A = \left( \frac{1}{g} \right)^r + \left( \frac{g}{f} \right)^r \quad \text{and} \quad B = (fg)^s + \left( \frac{2}{\ell} \right) \left( \frac{f}{2g} \right)^s \]

with \( r, s \) such that \( 12((\ell - 1)r, 12((\ell + 1)s) \) and \( \frac{(\ell - 1)r}{12} \equiv \frac{(\ell + 1)s}{12} \pmod{2} \). Depending on \( \ell \pmod{24} \), these conditions are satisfied by some \( r \) and \( s \) such that \( 2r|\ell + 1 \) and \( 2s|\ell - 1 \). Weber \cite[p. 268]{weber} then shows that

\[ \Phi_{\ell} = (fg)^{(\ell+1)/2} \left( A^{(\ell+1)/(2r)} - B^{(\ell-1)/(2s)} + \sum_{\alpha=0}^{(\ell+1)/(2r)-1} \sum_{\beta=0}^{(\ell-1)/(2s)-1} c_{\alpha,\beta} A^\alpha B^\beta \right). \]

In this form, the polynomial has about \( \ell^2 / (4rs) \) coefficients, which is often less than the roughly \( \ell^2 / 24 \) coefficients if it is written as an equation between \( f \) and \( g \). However, the more compact form appears to be less suited to the evaluation–interpolation algorithm: There is no easy way, in the spirit of \( \Phi_{\ell} \), of obtaining its values when interpreting it as a univariate polynomial in \( A(z) \) with coefficients in \( \mathbb{Z}[B(z)] \). So instead of performing \( O(\ell) \) interpolations of polynomials of degree \( O(\ell) \), one would apparently have to obtain all coefficients at once by solving a linear system with \( O(\ell^2) \) unknowns, resulting in a complexity that is worse than that of Result \( \Phi_{\ell} \).

### 4.3. Generalised Schl"afli equations

The Schl"afli equations of the previous paragraph can obviously be generalised to further modular functions: Given a modular function \( f \) and a prime \( \ell \), one may want to compute the polynomial relating \( f(z) \) and \( f(z/\ell) \).

In \cite[Ch. 5]{weber}, a theory analogous to Weber’s is developed for the functions \( \Phi_{\ell} \). The derived polynomials are used to obtain symbolically minimal polynomials of special algebraic values of \( \eta \) quotients.

In the context of a \( p \)-adic algorithm for the computation of class polynomials and ultimately elliptic curves with complex multiplications, another application is developed in \cite[Ch. 6.8]{weber}. Let \( \tau \) be a quadratic integer. Then the singular value \( j(\tau) \) lies in the ring class field for the order \( \mathcal{O} = [1, \tau]_\mathbb{Z} \) and is the \( j \)-invariant of an elliptic curve with complex multiplication by \( \mathcal{O} \). In \cite{weber}, the authors describe an algorithm to compute the canonical lift of \( j(\tau) \) to the \( p \)-adic numbers at arbitrary precision by some kind of Newton iteration. The main algorithmic step consists of applying an isogeny to \( j(\tau) \) (or more precisely to the elliptic curve with this \( j \)-invariant) that corresponds to a smooth principal ideal and that can thus be realised by composing isogenies of manageable prime degrees. In \cite{weber}, the algorithm is generalised to other class invariants. Let \( f \) be a modular function for \( \Gamma^0(N) \) such that \( f(\tau) \) is a \( \ell \)-class invariant in the sense that it also lies in the ring class field of \( \mathcal{O} \). Let \( \ell \) be a prime number not dividing \( N \) that splits as \( (\ell) = \mathfrak{P} \mathfrak{P}^{*} \) in \( \mathcal{O} \). As in the case of \( j(\tau) \), one needs to “apply an isogeny” and compute \( f(\tau^{-1}) \). This value is given as a root of the modular polynomial \( \Psi \) between \( f(z) \) and \( f(z/\ell) \), specialised with \( f(\tau) \) in the place of \( f(z) \). (The modular polynomial \( \Phi \) between \( f(z) \) and \( j(z) \), treated in Section \( 4.1 \), also plays a role as it helps to choose whenever \( \Psi \) has multiple roots: \( f(\tau^{-1}) \) is also a root of \( \Phi \) specialised with \( j(\tau^{-1}) \) in the place of \( j(z) \).)

A quick computation shows that \( g(z) = f(z/\ell) \) is modular for \( \Gamma^0(\ell) \cap \Gamma^0(N) = \Gamma^0(\ell N) \), so that \( \Phi_{\ell} \) yields the minimal polynomial of \( g \) over \( \mathbb{Q}_{\Gamma^0(N)} \) if the \( M_{\ell} \) are
chosen as a system of representatives of \( \Gamma^0(\ell N) \setminus \Gamma^0(N) \). Such a system is obtained precisely as for the Schl"afli equations by multiplying the standard representatives of \( \Gamma^0(\ell) \setminus \Gamma \) by matrices in \( \Gamma^0(\ell) \) so that they end up in \( \Gamma^0(N) \). For instance, a set of representatives is given by \( \begin{pmatrix} 1 & N\nu \\ 0 & 1 \end{pmatrix} \) for \( \nu = 0, \ldots, \ell - 1 \) and \( \begin{pmatrix} 1 - Nk & Nk \\ -Nk & 1 + Nk \end{pmatrix} \) with \( k = N^{-1} \mod \ell \). So at first sight, the evaluation–interpolation algorithm seems to apply.

The problem lies in recognising the coefficients \( c_r \) as polynomials in \( f \). In fact, the \( c_r \) are modular for \( \Gamma^0(N) \), so unless \( C_{\Gamma^0(N)} = C(f) \) (otherwise said, \( f \) is a *Hauptmodul* for \( \Gamma^0(N) \)), there is no reason that they are rational in \( f \). Generically, one expects \( C_{\Gamma^0(N)} = C(j, f) \) and may hope (under suitable integrality and rationality conditions) to recover the \( c_r \) as elements of \( \mathbb{Z}[j, f] \).

The interpolation phase turns out to take negligible time; thus a simple quadratic algorithm using iterated differences is employed instead of a quasilinear one.

5. Implementation

The evaluation–interpolation algorithm has been implemented in C by the author for different kinds of modular polynomials. One big advantage of the algorithm (besides its superior complexity) is that it is rather straightforward to implement, especially in the context of complex multiplication (cf. [10]), where the major building blocks such as the evaluation of modular functions and computations with polynomials over floating point numbers are already in use. The author’s implementation relies on the existing libraries *gmp* [22] with an assembly patch for AMD64 by P. Gaudry [21] for fast basic multiprecision arithmetic, *mpfr* [23] and *mpc* [16], which provide elementary operations with arbitrary precision floating point real and complex numbers with exact rounding. A library for fast polynomial operations via Karatsuba, Toom–Cook and the FFT has been written for the occasion [9]. All timings mentioned in the following have been obtained on Opteron processors clocked at 2.4 GHz. The heights are given as logarithms in base 2, and the arithmetic precision is given in bits.

5.1. Details of the implementation.

5.1.1. Details slowing the implementation down asymptotically. Of the different techniques for evaluating modular functions described in [10], the asymptotically fast ones are not beneficial in the present context. For computing univariate class polynomials, they do not pay off at degrees below 100 000. Such high degrees are for the time unachievable for bivariate modular polynomials, since the number of coefficients would make it impossible to store the result. Hence all examples have been computed via the sparse representation of the \( \eta \) function as a \( q \)-series.

The interpolation phase turns out to take negligible time; thus a simple quadratic algorithm using iterated differences is employed instead of a quasilinear one.

5.1.2. Details speeding the implementation up. In order to work with real instead of complex numbers as much as possible, purely imaginary values are chosen for the \( z_k \). Then the \( q(z_k) \) are real, and so are the values \( j(z_k) \) and thus, by (3.1), the

---

The text continues with further details on implementation and comparisons.
\( \Phi(X, j(z_k)) \). While the values of the conjugates \( f(M_\nu z_k) \) still have to be computed as complex numbers, roughly half of them suffice for functions \( f \) for \( \Gamma_0(\ell) \) with rational \( q \)-coefficients: \( f(z_k) \) is real, \( f(z_k - r) \) is the complex conjugate of \( f(z_k + r) \) for \( r \in \mathbb{Z} \), and finally \( f(-1/z_k) \) has to be real again to obtain a real polynomial in the end. When reconstructing \( \Phi(X, j(z_k)) \) from its roots, the complex conjugate roots can be grouped so that only real arithmetic is required.

Also the interpolation phase uses only real numbers. It is furthermore helped by choosing the \( z_k \) such that the \( j(z_k) \) lie in a simple arithmetic progression, for instance, \( j(z_k) = 1728 + k \) for \( k \geq 1 \).

5.1.3. Parallelisation. Another nice feature of the evaluation–interpolation algorithm is that it can be parallelised almost arbitrarily, in principal up to distributing at the level of each modular function evaluation. A more natural, coarser parallelisation of the evaluation phase, which needs almost no communication, is amply sufficient in practice: Each processor treats a different \( z_k \) and computes an approximation \( \Phi(X, j(z_k)) \) by evaluating the conjugates \( (f(M_\nu z_k))_\nu \) and reconstructing the polynomial from these roots. The result is written into a file on a shared hard disk. The interpolation phase has been too fast to warrant a parallel implementation. If it were to become a bottleneck, a natural way of distributing the work would be to have each processor compute the coefficients of a different polynomial \( c_r \in \mathbb{Z}[j] \).

Implementing the communication via files solves a second problem. During the evaluation phase, the matrix \( (c_r(z_k))_{k,r} \) is computed row-wise; for the interpolation, it is accessed column-wise. However, the matrix is roughly as big as the final modular polynomial and for the largest computed examples (see Section 5.2.1) does not fit into main memory any more. Writing each row into a different file during the evaluation and reading all files in parallel during the interpolation can be seen as a means of transposing the matrix on disk without having to keep it in main memory.

5.1.4. Arithmetic precision. Except for the classical modular polynomials, no upper bound on the height of the polynomials is known, and even in the classical case, the bound is not completely explicit, but contains an \( O(1) \) term (see Section 2.1). In the implementation, an approximate bound is obtained by carrying out all computations first at very low precision (100 bits). Due to the numeric instability of interpolation, the resulting coefficients are wrong already in the first digit, but one obtains an idea of the height of the correct polynomial (actually, the height at low precision turns out to be larger than the real one). This estimate is multiplied by a factor (between 1.1 and 2 depending on the function and the level), determined experimentally, to deal with rounding errors.

In the context of class polynomial computations, it has been observed in [10] that virtually no rounding errors occur: Evaluating modular functions via the \( q \)-development of \( \eta \) as well as multiplication of polynomials is rather uncritical from a numerical point of view. It suffices to add a few bits to the target precision. This should also hold for the evaluation phase in our context. In practice, it can be observed, however, that a significantly higher precision is needed than just the height of the result. This can only be due to the interpolation phase, which implicitly handles Vandermonde matrices, that are known to be ill conditioned. It is an open
problem to choose the interpolation points \( j(z_k) \) so as to minimise the rounding errors.

5.2. Examples.

5.2.1. Modular polynomials for \( \Gamma^0(\ell) \). For achieving the point counting record for elliptic curves with the Schoof–Elkies–Atkin algorithm \([12]\), the polynomials have been systematically computed with the function

\[
 f_{\ell,r} = \frac{T_r(\eta(z)\eta(\ell z))}{\eta(z)\eta(\ell z)}
\]

(evaluated in \(-1/z\)) suggested in \([27]\). Here,

\[
 T_r(f) = \frac{1}{r} \sum_{\nu=0}^{r-1} f \left( \frac{z + 24\nu}{r} \right) + f(rz)
\]

is the \( r \)-th Hecke operator for a prime \( r \geq 5 \) such that \( 24 | (r-1)(\ell+1) \), \( r \) is a quadratic residue modulo \( \ell \) and \( \ell \) is a quadratic residue modulo \( r \). (The unusual factor 24 in front of \( \nu \) takes care of the fact that the exponents in the \( q \)-expansion of \( \eta \) are not integral due to the factor \( q^{1/24} \); it eliminates 24\(^{th}\) roots of unity.)

It is not advisable to obtain the values of \( f_{\ell,r} \) via its \( q \)-expansion. First of all, the expansion is dense and thus much slower to evaluate than that of \( \eta \). But more importantly, \( q \) tends to infinity when \( z \) approaches the cusp 0 of the fundamental domain for \( \Gamma^0(\ell) \), so that the \( q \)-series converges arbitrarily slowly. This is not an issue when relying on the evaluation of \( \eta \) only, since its known transformation behaviour under unimodular matrices allows to transform the arguments into the fundamental domain for \( \Gamma \), whose only cusp is at infinity and corresponds to \( q = 0 \). So the implementation evaluates \([5,1]\) numerically with \( 2r + 4 \) evaluations of \( \eta \) per value of \( f_{\ell,r} \). The resulting dependence of the running time on \( r \) can be seen clearly in the following examples.

<table>
<thead>
<tr>
<th>( \ell )</th>
<th>2039</th>
<th>( \ell )</th>
<th>2017</th>
</tr>
</thead>
<tbody>
<tr>
<td>( r )</td>
<td>5</td>
<td>( r )</td>
<td>61</td>
</tr>
<tr>
<td>degree in ( j )</td>
<td>136</td>
<td>degree in ( j )</td>
<td>156</td>
</tr>
<tr>
<td>estimated height</td>
<td>5816</td>
<td>estimated height</td>
<td>6211</td>
</tr>
<tr>
<td>precision</td>
<td>6397</td>
<td>precision</td>
<td>6832</td>
</tr>
<tr>
<td>height</td>
<td>5040</td>
<td>height</td>
<td>5311</td>
</tr>
<tr>
<td>time for evaluation</td>
<td>5900 s</td>
<td>time for evaluation</td>
<td>74000 s</td>
</tr>
<tr>
<td>time for interpolation</td>
<td>110 s</td>
<td>time for interpolation</td>
<td>130 s</td>
</tr>
</tbody>
</table>

The largest computed polynomial has a level of 10079. The computation takes less than two weeks on a small cluster with ten processors; as a compressed text file, the polynomial fills about 16 Gb of space. This illustrates that the limiting factor for the algorithm becomes not so much its running time, but rather the space requirements of its output, as can be expected for algorithms that have an essentially linear time complexity with respect to their output size.
It would be natural to try to linearly combine functions \( f_{\ell,r} \) with different \( r \) to reduce the pole order at infinity, or even to find a function on \( X_0(\ell) \) with minimal pole order. As a result, the degree of the modular polynomial in \( j \) (and thus the number of interpolation points) and the size of its coefficients (and thus the required precision) could be reduced. A high price, however, would have to be paid by an increased running time for the evaluation. As high performance clusters are becoming increasingly available, while bandwidth and storage space appear to be the bottleneck for modular polynomial computation, this might, however, be the road to follow for constructing a manageable database of modular polynomials up to a high level.

5.2.2. \textit{Schläfli equations.} As explained in Section 4.2, a trivial modification to the enumeration of the \( \ell + 1 \) matrices \( M_\nu \) adapts the code to computing the Schläfli equations between \( f(z) \) and \( f(z/\ell) \). It is then imperative to change the interpolation code so as to take advantage of the sparseness of the polynomial, in which only every 24-th coefficient is non-zero. This reduces the number of evaluations and (after a suitable transformation) the degree of the interpolated polynomials roughly by a factor of 24. The latter is also important since the interpolation phase becomes numerically more stable.

\[
\begin{array}{|c|c|}
\hline
\ell & 10079 \\
\hline
r & 5 \\
\hline
degree in \( j \) & 672 \\
\hline
estimated height & 31865 \\
\hline
precision & 35051 \\
\hline
height & 28825 \\
\hline
time for evaluation & 10 000 000 s \approx 120 d \\
\hline
time for interpolation & 56 000 s \approx 16 h \\
\hline
\end{array}
\]

5.2.3. \textit{Generalised Schläfli equations.} As argued in Section 4.3, the minimal polynomial of \( g(z) = f(z/\ell) \) over \( \mathbb{C}(f) \) for some function \( f \) for \( \Gamma^0(N) \) need not be of degree \( \ell + 1 \) any more, in which case it is a priori unclear how to obtain it by evaluation–interpolation. But besides the functions \( f \) that are Hauptmoduln for \( \Gamma^0(N) \), a few others may be handled by this approach. Namely, let

\[
\nu_{p_1,p_2} = \frac{\eta(z/p_1)\eta(z/p_2)}{\eta(z)\eta(z/(p_1p_2))}
\]

with \( p_1, p_2 \) prime and \( 24|(p_1-1)(p_2-1) \) be the functions of level \( N = p_1p_2 \) suggested as class invariants in [14]. These functions are invariant under the Fricke–Atkin–Lehner involution. Now, \( X_0^+(N) \) is of genus 0 for \( N = 35 \) and \( N = 39 \) [28], and apparently, the double \( \eta \) quotients are Hauptmoduln in this case. The degree of the modular equation of transformation level \( \ell \) thus remains \( \ell + 1 \) (an observation made for \( N = 35 \) in [28, Ch. 7.4]).
For instance, with \( f = w_{3,13} \) and \( g(z) = f(z/ℓ) \), one obtains the following polynomials \( \Phi^{3,13}_w \):

\[
\begin{align*}
\Phi^{3,13}_2 &= g^3 + f^3 - g^2 f^2 - g f + 2(g^2 f + g f^2) \\
\Phi^{3,13}_5 &= g^6 + f^6 - g^5 f^5 - g f + 35g^3 f^3 \\
& \quad + 5(g^5 f^4 + g^4 f^5 + g^3 f^6 + g^2 f^7 + g f^8 + f^9 - g f^3 - g^3 f^4 - g^3 f^2 f^5 + g^3 f + g f^2) \\
& \quad + 10(-g^5 f^2 - g^2 f^5 + g^4 f^4 + g^3 f^2 + g^2 f^4 - g^4 f - g f^4 + g^2 f^2) \\
\Phi^{3,13}_7 &= g^8 + f^8 - g^7 f^7 - g f \\
& \quad + 7(g^7 f^6 + g^6 f^7 - g^5 f^6 + g^5 f^4 + 4 g^6 f^4 + g^6 f^2 + g^2 f^6) \\
& \quad + 7(g^4 f^2 + g^2 f^4 - g^3 f - g f^3 + g^2 f + g f^2) \\
& \quad + 14(-g^7 f - g f^7 + g^5 f^4 + g^4 f^5 + g^3 f^3 + g^3 f^5 + g^4 f^3 + g^3 f^4 + g^3 f) \\
& \quad + 21(-g^5 f^4 - g^4 f^5 + g^3 f + g f^3 - g^3 f^5 + g^3 f^2 + g^3 f^3) \\
& \quad + 21(g^5 f^2 + g^3 f + g f^5 + f^4 - g f^4 - g^2 f^2 - g^2 f^3) \\
& \quad + 42(g^6 f^6 + g^2 f^2) - 182g^4 f^4
\end{align*}
\]

Like the Schläfli equations, the polynomials are symmetric in \( f \) and \( g \), but as can be seen from these and further examples, they are no more sparse.

References


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