Computing canonical heights on elliptic curves in quasi-linear time

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Abstract

We introduce an algorithm that can be used to compute the canonical height of a point on an elliptic curve over the rationals in quasi-linear time. As in most previous algorithms, we decompose the difference between the canonical and the naive height into an archimedean and a non-archimedean term. Our main contribution is an algorithm for the computation of the non-archimedean term that requires no integer factorization and runs in quasi-linear time.

1. Introduction

Let $E$ denote an elliptic curve defined over a number field $K$. The canonical height is a quadratic form $\hat{h}: E(K) \otimes \mathbb{R} \to \mathbb{R}$, first constructed by Néron [13] and Tate (unpublished). For several applications, such as computing generators for $E(K)$ and computing the regulator appearing in the conjecture of Birch and Swinnerton-Dyer, one needs to compute $\hat{h}(P)$ for points $P \in E(K)$.

To this end, one typically chooses a Weierstrass equation $W$ for $E$ over $K$ with $O_K$-integral coefficients and decomposes $\hat{h}(P)$ (or $\hat{h}(P) - h(P)$, where $h$ is the naive height on $E$ with respect to $W$) into a sum of local terms, one for each place of $K$. For simplicity, let us assume that $K = \mathbb{Q}$. There are several efficient algorithms for the computation of the contribution at infinity (see § 5). A very simple and efficient algorithm of Silverman [14] can be used to compute the non-archimedean contributions separately. However, in order to determine the non-archimedean places which contribute to $\hat{h}(P)$ (or $\hat{h}(P) - h(P)$), the algorithm of [14] assumes that the prime factorization of the discriminant $\Delta(W)$ is known, which renders this approach inefficient when the coefficients of $W$ are large. This observation motivated Silverman’s article [16], where it is shown how to compute $\hat{h}(P)$ without the need to factorize $\Delta(W)$. Nevertheless, the algorithm of [16] requires the prime factorization of $\gcd(c_4(W), c_6(W))$ in order to find a globally minimal Weierstrass equation for $E$.

In this note, we introduce an algorithm for the computation of $\hat{h}(P)$ that does not require any factorization into primes at all and runs in time that is quasi-linear in the size of the input data and the desired precision of the result. More precisely, let $\|W\|$ denote the largest absolute value of the coefficients of $W$ and let $d$ denote the number of desired bits of precision after the binary point. We denote the time needed to multiply two $d$-bit integers by $M(d)$. The following result is our main theorem. Recall the ‘soft-O’ notation: $f(n) \in \tilde{O}(g(n))$ means that there are constants $c, m > 0$ such that, for $n$ sufficiently large, $|f(n)| \leq cg(n)(\log g(n))^m$. Using fast multiplication algorithms, $M(d) \in O(d)$. We also use the notation $f(n) \ll g(n)$ to express the fact that there is a constant $c > 0$ such that $|f(n)| \leq cg(n)$ for $n$ sufficiently large.
THEOREM 1.1. Let $E$ be given by a Weierstrass equation $W$ with coefficients in $\mathbb{Z}$ and let $P \in E(\mathbb{Q})$. Then we can compute $\hat{h}(P)$ to $d$ bits of (absolute) precision in time
\[
\ll \log(d + h(P))M(d + h(P)) + (\log \log ||W||)^2 M((\log \log ||W||)(\log ||W||)) \\
+ \log(d + \log ||W||)^2 M(d + \log ||W||) \\
\in \tilde{O}(d + h(P) + \log ||W||).
\]

Since the size of the input is measured by $h(P) + \log ||W||$ (the first term gives the size of $P$, the second term gives the size of $W$) and the size of the output is measured by $\log h(P) + d$, this means that we can compute $\hat{h}(P)$ in quasi-linear time.

The strategy of the proof is to first find an algorithm for the computation of the local non-archimedean contributions that does not assume minimality (see Proposition 4.3). Building on this, the non-archimedean contribution to $\hat{h}(P) - h(P)$ can be computed upon observing that it is a sum of rational multiples of logarithms of prime numbers, which can be determined by working globally modulo a suitable power of $\Delta(W)$. Combining this with a complexity analysis of the fastest known algorithm for the computation of the local height at infinity due to Bost and Mestre [6], Theorem 1.1 follows. We note that Marco Caselli, working on his PhD under the supervision of Cremona, is currently extending the Bost–Mestre algorithm to also deal with complex places.

The paper is organized as follows. In §2, we set up some notation and introduce the notion of Kummer coordinates of points on elliptic curves. Heights and their local decompositions are recalled in §3. In §4, we discuss an algorithm that allows us to compute a non-archimedean local summand of $\hat{h}(P) - h(P)$ efficiently, without assuming minimality, and we estimate its running time. Section 5 contains a discussion of the algorithm of Bost–Mestre for the computation of the local height at infinity and of its running time. We then combine the non-archimedean and the archimedean results into an efficient algorithm for the computation of $\hat{h}(P)$ in §6, leading to a proof of Theorem 1.1. Finally, in §7, we discuss the practicality of our algorithm.

2. Kummer coordinates

Let $K$ be a field and consider an elliptic curve $E/K$, given by a Weierstrass equation
\[
W: y^2 + a_1xy + a_3y = x^3 + a_2x^2 + a_4x + a_6,
\]
where $a_1, a_2, a_3, a_4, a_6 \in K$. As usual, let
\[
\begin{align*}
  b_2 &= a_1^2 + 4a_2, \\
  b_4 &= 2a_4 + a_1a_3, \\
  b_6 &= a_3^2 + 4a_6, \\
  b_8 &= a_7^2a_6 + 4a_2a_6 - a_1a_3a_4 + a_2a_3^2 - a_1^2,
\end{align*}
\]
and let
\[
\Delta(W) = -b_2^2b_8 - 8b_6^4 - 27b_6^2 + 9b_2b_4b_6
\]
denote the discriminant of the equation $W$. We define the functions $g$ and $f$ for $P \in E(K) \setminus \{O\}$ by
\[
\begin{align*}
  g(P) &= x(P)^4 - b_4x(P)^2 - 2b_6x(P) - b_8, \\
  f(P) &= 4x(P)^3 + b_2x(P)^2 - 2b_4x(P) + b_6.
\end{align*}
\]
Then, for $P \in E(K) \setminus E[2]$, $x(2P) = g(P)/f(P)$. We now extend this to all $P \in E(K)$. 

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Note that $\mathbb{P}^1$ is the Kummer variety $E/\{\pm 1\}$ of $E$. An explicit covering map $E \to \mathbb{P}^1$ is given by

$$
\kappa: \quad E \to \mathbb{P}^1
$$

$$(x : y : 1) \mapsto (x : 1)
$$

$$(0, 1).
$$

We call $(x_1, x_2) \in \mathbb{A}^2(K) \setminus \{(0, 0)\}$ a pair of Kummer coordinates for $P \in E(K)$ if

$$
\kappa(P) = (x_1 : x_2).
$$

The degree 4 homogenizations of $g$ and $f$ are

$$
\delta_1(x_1, x_2) = x_1 - b_4 x_1^2 x_2^2 - 2b_6 x_1 x_2^3 - b_8 x_2^4,
$$

$$
\delta_2(x_1, x_2) = 4x_1^2 x_2 + b_2 x_1^2 x_2^2 + 2b_4 x_1 x_2^3 + b_6 x_2^4,
$$

respectively. For $(x_1, x_2) \in \mathbb{A}^2(K)$, we set

$$
\delta(x_1, x_2) = (\delta_1(x_1, x_2), \delta_2(x_1, x_2)).
$$

It follows that if $(x_1, x_2)$ is a pair of Kummer coordinates for $P \in E(K)$, then $\delta(x_1, x_2)$ is a pair of Kummer coordinates for $2P$.

3. Heights

Let $K$ be a number field and let $E/K$ be an elliptic curve given by a Weierstrass equation $W$, as in (2.1). We denote by $M_K$ the set of places of $K$. For a place $v \in M_K$, we normalize the associated absolute value $| \cdot |_v$ so that it restricts to the usual absolute value on $\mathbb{Q}$ when $v$ is an infinite place and so that $|p|_v = p^{-1}$ when $v$ is a finite place above $p$. We write $n_v = [K_v : \mathbb{Q}_v]$ for the local degree, where $w$ is the place of $\mathbb{Q}$ below $v$. Then we have the product formula

$$
\prod_{v \in M_K} |x|^n_v = 1 \text{ for all } x \in K^x.
$$

The naive height of $P \in E(K) \setminus \{O\}$ with respect to $W$ is given by

$$
h(P) = \frac{1}{[K : \mathbb{Q}]} \sum_{v \in M_K} n_v \log \max\{|x_1|_v, |x_2|_v\},
$$

where $(x_1, x_2)$ is a pair of Kummer coordinates for $P$. Note that $h(P)$ does not depend on the choice of $(x_1, x_2)$, by the product formula.

The limit

$$
\hat{h}(P) = \lim_{n \to \infty} \frac{h(nP)}{n^2}
$$

exists and is called the canonical height (or Néron–Tate height) of $P$.

For the computation of $\hat{h}(P)$, the limit construction is not suitable due to slow convergence and exponential growth of the size of the coordinates. Instead, one decomposes $\hat{h}(P)$ into local terms. We now recall how this can be achieved, following [10]. For $v \in M_K$ and $Q \in E(K_v)$, we set

$$
\Phi_v(Q) = \frac{\max\{|\delta_1(x_1, x_2)|_v, |\delta_2(x_1, x_2)|_v\}}{\max\{|x_1|_v, |x_2|_v\}^4},
$$

where $(x_1, x_2) \in \mathbb{A}^2(K_v) \setminus \{(0, 0)\}$ is a pair of Kummer coordinates for $Q$. Since $\delta_1$ and $\delta_2$ are homogeneous of degree 4, $\Phi_v(Q)$ does not depend on the choice of $(x_1, x_2)$. The function $\Phi_v$ is continuous and bounded on $E(K_v)$, so it makes sense to define

$$
\Psi_v(Q) = -\sum_{n=0}^{\infty} 4^{-n-1} \log \Phi_v(2^n Q),
$$

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which is, likewise, continuous and bounded. Note that for \( P \in E(K) \),
\[
h(2P) - 4h(P) = \frac{1}{[K : \mathbb{Q}]} \sum_{v \in M_K} n_v \log \Phi_v(P),
\]
and Tate’s telescoping trick yields the formula
\[
\hat{h}(P) = h(P) - \frac{1}{[K : \mathbb{Q}]} \sum_{v \in M_K} n_v \Psi_v(P),
\]
which we will use to compute the canonical height.

Remark 1. Several normalizations for the local height on elliptic curves can be found in
the literature (see the discussion in [10]). Our normalization corresponds to that used in [10],
so in particular, our canonical height is twice the canonical height in Silverman’s paper [14]
and in his books on elliptic curves. More precisely, we have
\[
\hat{\lambda}_v(Q) = 2\hat{\lambda}^{\text{SilB}}_v(Q) + \frac{1}{6} \log |\Delta(W)|_v,
\]
where \( \hat{\lambda}^{\text{SilB}}_v \) is the normalization used in Silverman’s book [15, Chapter VI]. The advantages
of our normalizations are discussed in [10]; the crucial advantage of \( \hat{\lambda}^{\text{SilB}}_v \) is its independence
of the chosen Weierstrass equation.

In § 5, we need to know how local heights change under isogenies.

Proposition 3.1 (Bernardi [1]). Let \( E \) and \( E' \) be elliptic curves defined over \( K_v \), given by
Weierstrass equations \( W \) and \( W' \), respectively. Let \( \varphi: E \rightarrow E' \) be an isogeny of degree \( n \). If
\( Q \in E(K_v) \) satisfies \( \varphi(Q) \neq 0 \), then
\[
\hat{\lambda}_v(\varphi(Q)) = n\hat{\lambda}_v(Q) - \log |F_\varphi(Q)|_v - \log |m'(\varphi)|_v,
\]
where
\[
F_\varphi(Q) = \prod_{R \in \ker(\varphi) \setminus \{O\}} (x(Q) - x(R))
\]
and
\[
m'(\varphi) = \lim_{Q \to O} \frac{x(Q)}{x(\varphi(Q))}.
\]
4. Non-archimedean local error functions

In this section, we let $K$ be a non-archimedean local field with normalized additive valuation $v: K \to \mathbb{Z} \cup \{\infty\}$. Let $O$ denote the valuation ring of $K$, let $k$ denote the residue class field of $O$ and let $\pi$ be a uniformizing element of $O$. Consider an elliptic curve $E/K$ given by a Weierstrass equation $W$, as in (2.1), with coefficients in $O$.

Given $P \in E(K)$, we choose a pair of Kummer coordinates $(x_1, x_2)$ for $P$ and define

$$\varepsilon(x_1, x_2) = \min\{v(\delta_1(x_1, x_2)), v(\delta_2(x_1, x_2))\} - 4 \min\{v(x_1), v(x_2)\} \in \mathbb{Z}.$$

Note that $\varepsilon$ does not depend on the choice of Kummer coordinates, so we can define $\varepsilon(P) = \varepsilon(x_1, x_2)$ for any such choice. The function $\varepsilon$ is non-negative, bounded and continuous in the $v$-adic topology. Hence we can define

$$\mu(P) = \sum_{n=0}^{\infty} \frac{1}{4^{n+1}} \varepsilon(2^n P) \in \mathbb{R}. \quad (4.1)$$

It follows that $\mu$ is non-negative, bounded and continuous as well. One can show that, in fact, $\mu(P) \in \mathbb{Q}$ (compare Table 1).

**Remark 2.** If $K$ is the completion of a number field at a non-archimedean place $v$, then $n_v \log \Phi_v(P) = -\varepsilon(P)(\log \#k)$ and $n_v \Psi_v(P) = \mu(P)(\log \#k)$ for $P \in E(K)$, where $\Phi_v$ and $\Psi_v$ are as defined in §3.

If we have bounds for $\varepsilon(P)$ and for the denominator of $\mu(P)$, then we can use (4.1) to compute $\mu(P)$.

**Lemma 4.1.** Assume that $M \geq 2$ and $B$ are non-negative integers such that:

1. $M' \mu(P) \in \mathbb{Z}$ for some $0 < M' \leq M$; and
2. $\max\{\varepsilon(P): P \in E(K)\} \leq B$.

Set

$$m = \left\lfloor \frac{\log(BM^2/3)}{\log 4} \right\rfloor.$$

Then $\mu(P)$ is the unique fraction with denominator $\leq M$ in the interval $[\mu_0, \mu_0 + 1/M^2]$, where

$$\mu_0 = \sum_{n=0}^{m} 4^{-n-1} \varepsilon(2^n P).$$

**Proof.** We know that $\mu(P)$ is a fraction with denominator bounded by $M$. Two distinct fractions with this property have distance greater than $1/M^2$ (here we use $M \geq 2$), so there

---

**Table 1.** Non-zero values of $v(\Delta)$ and upper bounds $\alpha$ for $\mu$ for minimal Weierstrass equations.

<table>
<thead>
<tr>
<th>Type</th>
<th>$v(\Delta)$</th>
<th>$\mu$</th>
<th>$\alpha$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$I_m$</td>
<td>$m \geq 2$</td>
<td>$(i(m-i))/m$, $i = 1, \ldots, m-1$</td>
<td>$m/4$</td>
</tr>
<tr>
<td>III</td>
<td>$\geq 3$</td>
<td>$1/2$</td>
<td>$1/2$</td>
</tr>
<tr>
<td>IV</td>
<td>$\geq 4$</td>
<td>$2/3$</td>
<td>$2/3$</td>
</tr>
<tr>
<td>$I_m^*$</td>
<td>$\geq 6 + m$</td>
<td>$1, (m+4)/4$</td>
<td>$(m+4)/4$</td>
</tr>
<tr>
<td>IV$^*$</td>
<td>$\geq 8$</td>
<td>$4/3$</td>
<td>$4/3$</td>
</tr>
<tr>
<td>III$^*$</td>
<td>$\geq 9$</td>
<td>$3/2$</td>
<td>$3/2$</td>
</tr>
</tbody>
</table>
is at most one such fraction in the given interval. On the other hand, we know that
\[ \mu_0 \leq \mu(P) \leq \mu_0 + \sum_{n \geq m} 4^{-n-1} B = \mu_0 + \frac{B}{3 \cdot 4^{m+1}} \leq \mu_0 + \frac{1}{M^2}. \]
\qed

We now discuss how to bound \( \varepsilon(P) \) and the denominator of \( \mu(P) \).

**Lemma 4.2.** For \( P \in E(K) \):

(i) \( 0 \leq \mu(P) \leq \frac{1}{6} v(\Delta(W)) \);

(ii) \( 0 \leq \varepsilon(P) \leq v(\Delta(W)) \); and

(iii) the denominator of \( \mu(P) \) is bounded from above by \( v(\Delta(W)) \).

**Proof.** Note that if \( v(\Delta(W)) = 0 \) and \( v(x_1, x_2) = 0 \), then also \( v(\delta(x_1, x_2)) = 0 \), since one can express a power of \( \Delta \) as the resultant of \( \delta_1 \) and \( \delta_2 \). Therefore, if the Weierstrass equation \( W \) is minimal, then \( \varepsilon(P) \) vanishes if and only if \( P \) has non-singular reduction; in particular,
\[ \varepsilon(P) = 0 \Rightarrow \varepsilon(2P) = 0. \] (4.2)
One can also prove (4.2) using explicit formulas which do not require minimality of \( W \). If \( W \) is minimal, then, by the above, the value \( \mu(P) \) vanishes if and only if \( P \) has non-singular reduction. More generally, it depends only on the component of the special fiber of the Néron model of \( E \) that \( P \) reduces to (see [14]). Hence the same is true for \( \varepsilon(P) \), since
\[ \varepsilon(P) = 4\mu(P) - \mu(2P). \] (4.3)

For minimal \( W \), the non-zero values that \( \mu \) can take and an upper bound \( \alpha \) are given in Table 1. These are taken from [10, 15].

Let \( v_{\text{min}} \) denote the minimal discriminant of \( E \) over \( O \). In general,
\[ 0 \leq \mu(P) \leq \alpha + \frac{1}{6} (v(\Delta(W)) - v_{\text{min}}), \]
by [10, Proposition 8], and (i) follows from a straightforward computation. Because of (4.3), this also proves (ii). By the proof of [10, Proposition 8], a transformation from one integral Weierstrass equation to another does not change \( \mu(P) \) mod \( Z \), so (iii) follows from Table 1. \qed

Lemmas 4.1 and 4.2 lead to an algorithm for the computation of \( \mu(P) \). A pair \( (x_1, x_2) \) of Kummer coordinates for \( P \) is said to be primitive if \( \min\{v(x_1), v(x_2)\} = 0 \). Recall that \( \pi \) denotes a uniformizer of \( K \).

**Algorithm 1.**

1. Set \( B := v(\Delta) \).
2. If \( B \leq 1 \), then return 0. Otherwise set \( m := \lfloor \log(B^3/3)/\log 4 \rfloor \).
3. Set \( \mu_0 := 0 \). Let \( (x_1, x_2) \) be primitive Kummer coordinates for \( P \) with \( (m+1)B+1 \) \( v \)-adic digits of precision.
4. For \( n := 0 \) to \( m \) do:
   a. Compute \( (x'_1, x'_2) := \delta(x_1, x_2) \) (to \( (m+1)B+1 \) \( v \)-adic digits of precision).
   b. Set \( \ell := \min\{v(x'_1), v(x'_2)\} \).
   c. If \( \ell = 0 \), then return \( \mu_0 \).
   d. Set \( \mu_0 := \mu_0 + 4^{-n-1}\ell \).
   e. Set \( (x_1, x_2) := \pi^{-\ell}(x'_1, x'_2) \).
5. Return the unique fraction with denominator at most \( B \) in the interval \([\mu_0, \mu_0 + 1/B^2]\).

We now show that the algorithm is correct and estimate its running time.
Proposition 4.3. Algorithm 1 computes $\mu(P)$. Its running time is
\[
\ll (\log v(\Delta)) M((\log v(\Delta))v(\Delta)(\log \#k))
\]
as $v(\Delta) \to \infty$, with an absolute implied constant.

Proof. If $B = v(\Delta) \leq 1$, then $\mu = \varepsilon = 0$, by Table 1. Otherwise, the loop in step 4 computes the sum in Lemma 4.1 (where now $M = B \geq 2$). When $\ell = 0$ in step 4c, then $\varepsilon(2^n P) = 0$ for all $n' \geq n$ by (4.2), and hence the infinite sum defining $\mu$ is actually a finite sum and equals $\mu_0$. (This step could be left out without affecting the correctness or the worst-case complexity of the algorithm.) Lemma 4.2 shows that $B$ is an upper bound for $\varepsilon$ and that $M = B$ is an upper bound for the denominator of $\mu$. So the algorithm computes $\mu(P)$, provided the precision of $(m+1)B+1$ $v$-adic digits is sufficient. For this, note that the precision loss at each duplication step is given by $\varepsilon(2^n P)$ and is therefore bounded by $B$. So, after at most $m+1$ steps in the loop, the resulting $(x_1, x_2)$ still has at least one digit of precision.

It remains to estimate the running time. We assume that elements of $O$ are represented as truncated power series in $\pi$, whose coefficients are taken from a complete set of representatives for the residue classes. Operations on these coefficients can be performed in time $\ll M(\log \#k)$. Then steps b to e in the loop take negligible time compared with step a, which involves a fixed number of additions and multiplications of elements given to a precision of $(m+1)B+1$ digits, leading to a complexity of
\[
\ll M(((m+1)B+1)(\log \#k))
\]
operations for each pass through the loop. The total running time is therefore
\[
\ll (m+1) M(((m+1)B+1)(\log \#k)) \ll (\log v(\Delta)) M((\log v(\Delta))v(\Delta)(\log \#k))
\]
as $v(\Delta) \to \infty$. \hfill \Box

Remark 3. We stress that our algorithm does not require $W$ to be minimal. If we know that $W$ is minimal, then Silverman’s algorithm [14, §5], which only involves the computation of the valuations of a bounded number of polynomials in the coefficients of $W$ and the coordinates of $P$, can be used to compute $\mu(P)$.

5. Archimedean local heights

Let $K$ be an archimedean local field with valuation $v$. The following methods have been proposed for the computation of the local height $\hat{\lambda} = \hat{\lambda}_v$ on an elliptic curve $E/K$, given by a Weierstrass equation (2.1):

- an elegant series approach due to Tate and modified by Silverman [14];
- a more complicated series approach based on theta functions (see [9, Algorithm 7.5.7]);
- an algorithm based on the arithmetic geometric mean (AGM) and 2-isogenies introduced by Bost and Mestre in an unpublished manuscript [6], which currently requires $v$ to be real (see also Bradshaw’s PhD thesis [7]).

Tate’s series converges linearly. The theta series has a better rate of convergence and is also faster in practice than the elliptic integrals arising in the algorithm are computed using the AGM. If $v$ is real and one is interested in high precision, then the method of Bost and Mestre is preferable, as it converges quadratically. We now describe this algorithm and provide a complexity analysis. Let $v$ be real and let $|\cdot|$ denote the usual absolute value on $K = \mathbb{R}$. We want to compute $\hat{\lambda}(P)$ for a point $P \in E(\mathbb{R})$; for simplicity, we only consider the case $2P \neq O$. Note that the function $\mu$ considered in [6] satisfies $\mu = \frac{1}{2}\hat{\lambda}$.
Applying a transformation, we may assume that $E$ is given by a Weierstrass equation

$$W: y^2 = x(x^2 + ux + v),$$

where $u, v \in \mathbb{R}$. If all points of order two on $E$ are real, then we set $E_0 = E$. Otherwise, consider the isogeny $E \to E_0$ defined by

$$(x, y) \mapsto \left(\frac{x^2 + ux + v}{x}, \frac{x^2 - v}{x^2}\right),$$

where now $E_0$ has full 2-torsion over $\mathbb{R}$ and is given by the Weierstrass equation

$$y^2 = x(x^2 - 2ux + u^2 - 4v).$$

By Proposition 3.1, it suffices to compute the local height of the image of $P$ on $E_0$ to obtain $\hat{\lambda}(P)$. For the algorithm, we need a Weierstrass equation

$$W_0: y^2 = x(x + a_0^2)(x + b_0^2)$$

for $E_0$, where $0 < b_0 < a_0 \in \mathbb{R}$. We may assume that $P$ lies on the connected component $E_0^0(\mathbb{R})$ of the identity; if not, we can apply the algorithm to $2P \in E_0^0(\mathbb{R})$ and compute $\hat{\lambda}(P)$ using

$$\hat{\lambda}(2P) = 4\hat{\lambda}(P) - \log |2y(P)|.$$  \hfill (5.2)

We define the AGM sequences $(a_n)$ and $(b_n)$ by

$$a_n = \frac{a_{n-1} + b_{n-1}}{2}, \quad b_n = \sqrt{a_{n-1}b_{n-1}},$$

and we let $M(a_0, b_0)$ denote their common limit. For $n \geq 1$ we recursively define an elliptic curve $E_n$ over the reals by the Weierstrass equation

$$W_n: y^2 = x(x + a_n^2)(x + b_n^2),$$

and we define a 2-isogeny $\varphi_{n-1}: E_n \to E_{n-1}$ by

$$(x, y) \mapsto \left(\frac{x(x + b_n^2)}{x + a_n^2}, \frac{y(x + a_{n-1}a_n)(x + b_{n-1}a_n)}{(x + a_n^2)^2}\right).$$

Then the sequence of curves $(E_n)_n$ converges to a singular cubic curve $E_\infty$ with equation

$$W_\infty: y^2 = x(x + M(a_0, b_0)^2)^2.$$

Moreover, the sequence of isogenies $(\varphi_n)_n$ converges to the identity map on $E_\infty(\mathbb{R})$.

Now let $\hat{\lambda}_n$ denote the local height on $E_n(\mathbb{R})$. Then Proposition 3.1 asserts that

$$\hat{\lambda}_{n-1}(\varphi_{n-1}(P_n)) = 2\hat{\lambda}_n(P_n) - \log(x(P_n) + a_n^2),$$

whenever we have $x(\varphi_{n-1}(P_n)) \neq 0$.

Bost and Mestre use (5.3) to give a formula for $\hat{\lambda}(P)$. Note that $\varphi_{n-1}$ maps $E_n(\mathbb{R})$ onto the connected component $E_{n-1}^0(\mathbb{R})$ and that points on $E_{n-1}^0(\mathbb{R})$ always have a unique preimage on $E_n^0(\mathbb{R})$ under $\varphi_{n-1}$. Setting $P_0 = P$, we therefore get a well-defined sequence of preimages $P_n = (x_n, y_n) \in E_n^0(\mathbb{R})$, which converges to a point $P_\infty = (x_\infty, y_\infty) \in E_\infty(\mathbb{R})$. Here $x_n$ can be calculated using

$$x_n = \frac{1}{2}(x_{n-1} - a_{n-1}b_{n-1} + \sqrt{(x_{n-1} + a_{n-1}^2)(x_{n-1} + b_{n-1}^2)}).$$
From (5.3), we deduce that
\[ \hat{\lambda}(P) = \hat{\lambda}_0(P) = \log \lim_{n \to \infty} (x_n + a_n^2)^{2n-1}, \]
and equivalently,
\[ \hat{\lambda}(P) = \log(x_1 + a_1^2) + \sum_{n=1}^{\infty} 2^n \log \frac{x_{n+1} + a_{n+1}^2}{x_n + a_n^2}. \] (5.4)

Because of the quadratic convergence of the AGM, these formulas can be used to compute \( \hat{\lambda}(P) \) to an accuracy of \( 2^{-d} \) in \( \ll \log(d + \|W\|) \) steps. This has already been shown by Bradshaw (see [7, §6.1]). We give a slightly different error estimate. First, note that
\[ a_n - b_n \leq 2^{1-2^n} (a_0 - b_0). \]

Because \( x_n > 0 \) and \( 0 < b_0 < b_n < a_n \), this implies that
\[ \frac{a_n^2 - b_n^2}{x_n + a_n^2} \leq 2^{1-2^n} (a_0 - b_0) \frac{a_n + b_n}{x_n + a_n^2} \leq 2^{2-2^n} \left( \frac{a_0}{b_0} - 1 \right). \] (5.5)

Now set
\[ s_n := 1 - \frac{x_{n+1} + a_{n+1}^2}{x_n + a_n^2} \quad \text{and} \quad \vartheta := \frac{a_0}{b_0} - 1 + \sqrt{\frac{a_0}{b_0} - 1}. \]

Then \( 0 < s_n < 1 \) and \( \vartheta \ll \|W\| \). The sequence \( s_n \) converges rapidly to zero for large \( n \), since (5.5) implies that
\[ s_n = \left| \frac{1}{2} \left( \sqrt{\frac{x_n + b_n^2}{x_n + a_n^2}} + \frac{x_n + ((a_n^2 + b_n^2)/2)}{x_n + a_n^2} \right) - 1 \right| \]
\[ \leq \frac{1}{2} \left( \sqrt{\frac{x_n + b_n^2}{x_n + a_n^2}} - 1 \right) + \frac{1}{2} \left| \frac{x_n + ((a_n^2 + b_n^2)/2)}{x_n + a_n^2} - 1 \right| \]
\[ \leq \frac{1}{2} \sqrt{\frac{a_n^2 - b_n^2}{x_n + a_n^2}} + \frac{1}{4} \left( \frac{a_n^2 - b_n^2}{x_n + a_n^2} \right) \]
\[ \leq 2^{-2^n-1} \sqrt{\frac{a_0}{b_0} - 1} + 2^{-2^n} \left( \frac{a_0}{b_0} - 1 \right) \]
\[ \leq 2^{-2^n-1} \vartheta. \] (5.6)

In particular, \( s_n \leq \frac{1}{2} \) for \( n \geq \log_2(\log_2 \vartheta + 1) + 1 \), so that \( |\log(1 - s_n)| \leq 2s_n \) for such \( n \). We can use this to bound the tail of the series in (5.4). Namely,
\[ \left| \sum_{n=N+1}^{\infty} 2^n \log \frac{x_{n+1} + a_{n+1}^2}{x_n + a_n^2} \right| \leq \sum_{n=N+1}^{\infty} 2^n |\log(1 - s_n)| \]
\[ \leq \vartheta \sum_{n=N+1}^{\infty} 2^{1+n-2^n-1} \]

if \( N \geq \log_2(\log_2 \vartheta + 1) \). For \( n \geq 4 \), \( n - 2^{n-1} \leq -2^{n-2} \), so
\[ \left| \sum_{n=N+1}^{\infty} 2^n \log \frac{x_{n+1} + a_{n+1}^2}{x_n + a_n^2} \right| \leq \vartheta \sum_{n=N+1}^{\infty} 2^{1-2^{n-2}} \leq 2^{2N-1} \vartheta \] (5.7)

follows, provided \( N \geq \max\{3, \log_2(\log_2 \vartheta + 1)\} \).
Having computed \( \hat{\lambda}(P) \) for \( P \in E(\mathbb{R}) \), we get \( \Psi_\infty(P) \) from
\[
\Psi_\infty(P) = \log \max\{1, |x(P)|\} - \hat{\lambda}(P).
\] (5.8)

**Proposition 5.1.** The algorithm above computes \( \Psi_\infty(P) \) to \( d \) bits of precision in time
\[
\ll \log(d + \log \|W\|)^2 M(d + \log \|W\|).
\]

**Proof.** First, suppose that we have already computed \( a_0, b_0 \) and \( x_0 \) and that \( P \) lies on the connected component \( E_0^0(\mathbb{R}) \). By (5.4) and (5.7),
\[
\left| \hat{\lambda}(P) - \log(x_1 + a_1^2) - \sum_{n=1}^{N} 2^n \log \frac{x_{n+1} + a_{n+1}^2}{x_n + a_n^2} \right| \leq 2^{-d}
\]
for
\[
N = \max\{3, \log_2(d + 2 + \log_2 d) + 1\} \ll \log(d + \log \|W\|).
\]
For every \( n \leq N \), we have to apply a fixed number of additions, multiplications and square roots to compute \( a_{n+1}, b_{n+1} \) and \( x_{n+1} \), which can be done to \( d' \) bits of precision in time \( \ll M(d') \), and we have to compute \( \log(1 - s_n) \). Because of precision loss due to the multiplication by \( 2^n \), we need to compute \( \log(1 - s_n) \) to an additional \( n \) bits, so we need an initial precision of
\[
d + N \ll d + \log(d + \log \|W\|)
\]
bits. A logarithm can be computed to \( d' \) bits of precision in time \( \ll (\log d') M(d') \) using one of several quadratically converging algorithms based on the AGM (see [4, Chapter 7]). Therefore, and by (5.6), we can compute \( \log(1 - s_n) \) to \( d + n \) bits of precision in time
\[
\ll \log(d + \log(d + \log \|W\|)) M(d + \log(d + \log \|W\|)).
\]
The computation of \( \log(x_1 + a_1^2) \) to \( d \) bits of precision takes time
\[
\ll \log(d + \log \|W\|) M(d + \log \|W\|).
\]
Hence, given \( a_0, b_0 \) and \( x_0 \) to \( d + N \) bits of precision, we can compute \( \hat{\lambda}(P) \) to \( d \) bits of precision in time
\[
\ll \log(d + \log \|W\|) \cdot (M(d + \log \|W\|) + \log(d + \log(d + \log \|W\|))) M(d + \log(d + \log \|W\|)).
\]
We can then find \( \Psi_\infty(P) \) using (5.8) in time \( \ll \log(d) M(d) \), which is negligible.

To compute \( a_0, b_0 \) and \( x_0 \) from a given Weierstrass equation, we need to find the roots of at most two polynomials of degree \( \leq 3 \) with real coefficients, transform the corresponding Weierstrass equation and find the image of our point \( P \) under these transformations. The roots of a polynomial of fixed degree \( d' \) bits of precision can be found in time \( \ll M(d') \) (see [4, Theorem 6.4]); the same holds for the evaluation of a polynomial of fixed degree. To counter loss of precision, we start with an initial precision of \( \ll d + \log \|W\| + \log(d + \log \|W\|) \) bits, so we can compute \( a_0, b_0 \) and \( x_0 \) to \( d + N \) bits of precision in time
\[
\ll M(d + \log \|W\| + \log(d + \log \|W\|)),
\]
which is dominated by the complexity of the remaining parts of the algorithm. The logarithmic correction terms coming from (5.2) and from Proposition 3.1 applied to the isogeny (5.1) and to the change of model needed to find \( W_0 \) can be computed to a sufficient number of bits of precision in time \( \ll \log(d + \log \|W\|) M(d + \log \|W\|) \). Hence the result follows. \( \square \)
Remark 4. For large \( n \), computing \( \log(1 - s_n) \) using an AGM-based algorithm might be less efficient than using a power series such as

\[
\log x = 2 \sum_{k=0}^{\infty} \frac{1}{2k+1} \left( \frac{x-1}{x+1} \right)^{2k+1}.
\]

The reason is that, by (5.6), the numbers \( 1 - s_n \) are very close to 1, so only few terms of the power series have to be computed.

6. Computing the canonical height of rational points

We will combine the results of \( \S \S 4 \) and 5 into an efficient algorithm for computing the canonical height of a point \( P \) on an elliptic curve \( E \) over a number field, which will prove Theorem 1.1. For simplicity, we take this number field to be \( \mathbb{Q} \) in the following. We assume that our curve is given by a Weierstrass equation (2.1) \( W \) with coefficients in \( \mathbb{Z} \), but we make no minimality assumption.

One usually computes \( \hat{h}(P) \) using the decomposition (3.2) into local heights \( \hat{\lambda}_v(P) \). The local height \( \hat{\lambda}_\infty(P) \) can be computed using the algorithm of Bost–Mestre discussed in \( \S 5 \) or one of the other approaches mentioned there. If the factorization of \( \Delta(W) \) is known, we can use [14, \S 5] to compute the local heights \( \hat{\lambda}_p(P) \) efficiently. Alternative, but less efficient, algorithms can be found in [17, 18]. If we know that \( W \) is minimal (for which some factorization is required; see the introduction), then we can use [16] to compute \( \sum_p \hat{\lambda}_p(P) \) without factorizing \( \Delta(W) \).

Another approach to computing \( \hat{h}(P) \) without factorization is discussed in [11], but their method does not yield a polynomial-time algorithm.

Our goal is to devise an algorithm for the computation of \( \hat{h}(P) \) that runs in time that is quasi-linear in \( \log ||W||, h(P) \) along with the required precision \( d \), which is measured in bits after the binary point. We note that \( h(P) \) is the logarithm of a rational number, so it can be computed in time \( \ll \log(h(P) + d) \log(h(P) + d) \). In the previous section, we showed that there is a quasi-linear algorithm for the computation of \( \Psi_\infty(P) \) (see Proposition 5.1).

It remains to see how the total contribution

\[
\Psi^f(P) := \sum_p \Psi_p(P) = \sum_p \mu_p(P) \log p
\]

coming from the local error functions at finite places can be computed efficiently; here we write \( \mu_p \) for the local height correction function over \( \mathbb{Q}_p \), as in Definition 4.1.

Fix \( P \in E(\mathbb{Q}) \). We assume that \( (x_1, x_2) \in \mathbb{Z}^2 \) is a primitive (that is, \( \gcd(x_1, x_2) = 1 \)) pair of Kummer coordinates for \( P \). We set \( g_n = \gcd(\delta(x_1^{(n)}, x_2^{(n)})) \), where \( (x_1^{(n)}, x_2^{(n)}) \in \mathbb{Z}^2 \) is a primitive pair of Kummer coordinates for \( 2^n P \). Then the definition of \( \mu_p \) implies that

\[
\Psi^f(P) = \sum_{n=0}^{\infty} 4^{-n-1} \log g_n.
\]

See [12] for a related approach in genus 2. By Lemma 4.2, we know that each \( g_n \) divides \( \Delta(W) \).

The key observation is that \( \Psi^f(P) \) is a rational linear combination of logarithms of positive integers, which can be computed exactly as follows.

Algorithm 2.

1. Set \( (x_1', x_2') := \delta(x_1, x_2), g_0 := \gcd(x_1', x_2') \) and \( (x_1, x_2) := (x_1'/g_0, x_2'/g_0) \).
2. Set \( D := \gcd(\Delta(W), g_0^5) \) and \( B := \lceil \log D/\log 2 \rceil \).
3. If \( B \leq 1 \), then return 0. Otherwise set \( m := \lceil \log(B^5/3)/\log 4 \rceil \).
4. For \( n := 1 \) to \( m \) do:
   a. Compute \((x_1', x_2') := \delta(x_1, x_2) \mod D^{m+1}g_0\).
   b. Set \( g_n := \gcd(D, \gcd(x_1', x_2')) \) and \((x_1, x_2) := (x_1'/g_n, x_2'/g_n)\).
5. Use Bernstein’s algorithm from [2] to compute a sequence \((q_1, \ldots, q_r)\) of pairwise coprime positive integers such that each \( g_n \) (for \( n = 0, \ldots, m \)) is a product of powers of the \( q_i \):
\[ g_n = \prod_{i=1}^{r} q_i^{e_i} \cdot n. \]
6. For \( i := 1 \) to \( r \) do:
   a. Compute \( a := \sum_{n=0}^{m} 4^{-n-1}e_{i,n}. \)
   b. Let \( \mu_i \) be the simplest fraction between \( a \) and \( a + 1/B^4 \).
7. Return \( \sum_{i=1}^{r} \mu_i \log q_i \), a formal linear combination of logarithms.

**Proposition 6.1.** The preceding algorithm computes \( \Psi^{\ell}(P) \) in time
\[ \ll (\log \log D)^2 M((\log \log D)(\log D)) + M(h(P))(\log h(P)). \]

**Proof.** We note that if \( B \leq 1 \) in step 3, then either \( g_0 = 1 \) and \( \Psi^{\ell}(P) = 0 \), or \( D \in \{2, 3\} \). In the latter case, \( g_0 \) is a power of \( p = 2 \) or \( 3 \) and \( v_p(\Delta(W)) = 1 \), which would imply that \( \varepsilon_p(P) = 0 \), so \( g_0 = 1 \), and we get a contradiction.

Let \( p \) be a prime. If \( p \nmid g_0 \), then \( \varepsilon_p(P) = 0 \) and therefore \( \mu_p(P) = 0 \). So we now assume that \( p \) divides \( g_0 \). We have \( v_p(\Delta(W)) = v_p(D) \leq B \). We see that \( p^{(m+1)}v_p(D)+1 \) divides \( D^{m+1}g_0 \), so computing modulo \( D^{m+1}g_0 \) will provide sufficient \( p \)-adic accuracy so that \( v_p(g_n) = \varepsilon_p(2^nP) \) for all \( n \leq m \) (compare the proof of Proposition 4.3 above). (One could replace \( D^{m+1}g_0 \) by \( D^{m+1-n}g_0 \) in step 4a.) Since all the \( g_n \) are power products of the \( q_i \), there will be exactly one \( i = i(p) \) such that \( p \mid q_{i(p)} \); let \( b_p = v_p(q_{i(p)}) \). Then
\[ \sum_{n=0}^{m} 4^{-n-1}\varepsilon_p(2^nP) = \sum_{n=0}^{m} 4^{-n-1}v_p(g_n) = b_p \sum_{n=0}^{m} 4^{-n-1}e_{i(p),n} = b_pa, \]
so
\[ \mu_p(P) = \sum_{n=0}^{\infty} 4^{-n-1}\varepsilon_p(2^nP) = b_p a + \sum_{n=m+1}^{\infty} 4^{-n-1}\varepsilon_p(2^nP), \]
where the last sum is in \([0, 1/B^4]\) (this follows from \( 0 \leq \varepsilon_p \leq B \); see Lemma 4.2 and the definition of \( m \), and compare the proof of Lemma 4.1). We know that the denominator of \( \mu_p(P) \) is at most \( B \) (see Lemma 4.2), so the denominator of \( \mu_p(P)/b_p \) is at most \( B^2 \), since \( b_p \leq v_p(D) \leq B \). On the other hand, \( a \leq \mu_p(P)/b_p \leq a + 1/(b_pB^4) \leq a + 1/B^4 \), which implies that \( \mu_p(P)/b_p \) is the unique fraction in \([a, a + 1/B^4]\) with denominator at most \( B^2 \), so \( \mu_p(P)/b_p = \mu_i(p) \), by Step 6b. Now
\[ \sum_{p} \mu_p(P) \log p = \sum_{p} \mu_i(p)b_p \log p = \sum_{i=1}^{r} \mu_i \sum_{p | q_i} b_p \log p = \sum_{i=1}^{r} \mu_i \log q_i. \]

It remains to estimate the running time. The computation of \( \delta(x_1, x_2) \) can be done in time \( \ll M(h(P)) \); the same is true for the division in step 1. The gcd in step 1 can be computed in time \( \ll M(h(P)) \log h(P) \). The computations in steps 2 and 3 take negligible time compared with step 4. Each pass through the loop in step 4 takes time \( \ll M((m + 2) \log D)(\log (m + 2) \log D) \), so the total time for the loop is
\[ \ll m M((m \log D)) \log (m \log D) \ll (\log \log D)^2 M((\log \log D)(\log D)). \]

The algorithm in [2] computes suitable \( q_i \) for a pair \( a, b \) of positive integers in time \( \ll (\log ab)(\log \log ab)^2 \). We iterate this algorithm, applying it first to \( g_0 \) and \( g_1 \), then to each
of the resulting \( q_i \) and \( g_2 \) and so on. Note that \( g_n \leq D \) for all \( n \). Because there are always \( \ll \log D \) terms in the sequence of \( q_i \), this leads to a contribution of \( \ll \log D (\log \log D)^3 \) for step 5. This is dominated by the time for the loop. The remaining steps take negligible time. \( \square \)

In practice, the efficiency of this approach can be improved as follows.

- We trial factorize \( \Delta(W) \) up to some bound \( T \) to split off the contributions of all sufficiently small primes \( p \). We can then compute the corresponding \( \mu_p \) using the algorithm of Proposition 4.3 or the algorithm of \( [14] \) (see Remark 3). In step 3, we can then set \( B := \lfloor \log D' / \log T \rfloor \), where \( D' \) is the unfactorized part of \( D \). Note that, in practice, the trial division can take quite a bit more time than it saves, in particular, when the equation has large coefficients, so this modification should be used with care.

- We update our list of ‘building blocks’ \( q_i \) after each pass through the loop in step 4 using the new \( g_n \); we do the computation modulo suitable powers of the \( q_i \) instead of modulo \( D^{m+1}g_0 \). We can also use separate values of \( B \) and \( m \) for each \( q_i \), which will usually be smaller than those given above.

- In this way, we can integrate steps 4, 5 and 6 into one loop.

- We can replace \( B^5 \) in the definition of \( m \) by \( 2B^4 \). Then \( \mu_p(P) \leq b_p a + 1/(2B^3) \) and \( a \leq \mu_p(P)/b_p \leq a + 1/(2b_p B^3) \). If \( \mu_p(P)/b_p = r/s \) with \( s \leq Bb_p \), then we have the bound \( a \leq r/s \leq a + 1/(2sB^2) \). There can be at most one fraction \( r/s \) with \( s \leq B^2 \) satisfying this; if \( r'/s' \) is another such fraction, then

\[
\frac{1}{ss'} \leq \left| \frac{r}{s} - \frac{r'}{s'} \right| \leq \frac{1}{2 \min\{s, s'\} B^2},
\]

which leads to the contradiction \( \max\{s, s'\} \geq 2B^2 \). We can then find \( \mu_i = \mu_p(P)/b_p \) as the first convergent \( r/s \) of the continued fraction expansion of \( a \) that is \( \geq a \) and satisfies \( r/s \leq a + 1/(2sB^2) \).

Combining Propositions 5.1 and 6.1, we finally obtain an efficient algorithm for computing the canonical height \( \hat{h}(P) \) of a point \( P \in E(\mathbb{Q}) \).

**Proof of Theorem 1.1.** The first term is the time needed to compute \( h(P) \). The second term comes from the complexity bound for the computation of \( \Psi^f(P) \) (using \( \log D \ll \log ||W|| \)) from Proposition 6.1. The third term is the bound for the computation of \( \Psi^\infty(P) \) given in Proposition 5.1.

7. **Implementation and Examples**

We have implemented our algorithm using the computer algebra system Magma [5]. In the current implementation, the factorization into coprimes in the algorithm preceding Proposition 6.1 uses a relatively simple algorithm due to Buchmann and Lenstra [8, Proposition 6.5] instead of the faster algorithm of [2] (or of [3]). In practice, the running time of this part of the algorithm appears to be negligible.

Let us compare our implementation to Magma’s built-in command **CanonicalHeight** (version 2.21-2). The latter uses the method of Bost–Mestre for the computation of the archimedean local height. For the finite part of the height, a globally minimal Weierstrass model is computed. The non-archimedean contributions are then computed separately using the algorithm from [14]; the relevant primes are found by factorizing \( \gcd(\delta_1(x_1, x_2), \delta_2(x_1, x_2)) \), where \( (x_1, x_2) \) is a primitive pair of Kummer coordinates for a point \( P \). The same strategy is currently used in Pari/GP. The computer algebra system Sage contains an implementation of, essentially, Silverman’s original algorithm for the computation of canonical heights from [14]; in particular, it factors the discriminant.

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Example 1. Consider the family $E_a$ of curves given by the Weierstrass equation

$$W_a : y^2 = x^3 - ax + a,$$

where $a$ is an integer, and the non-torsion point $P = (1,1)$ on $E_a$. To compute $\hat{h}(P)$, Magma needs to find a globally minimal model for $E_a$, which boils down to deciding whether a sixth power of a prime divides $a$. Hence, for random integers $a$ of large absolute value, the Magma implementation becomes slow. For instance, taking $a$ to be 15340200419838800017985469042480093838944439369612517409395585315152324187358929634112112413446941478

$$= 2^{37} \cdot a' \text{ with } a' \text{ composite},$$

Magma's built-in CanonicalHeight takes about an hour, but our implementation needs only 0.001 s to compute $\hat{h}(P)$ to 30 decimal digits of precision. For these computations, and the ones below, we used a Dell Latitude E7440 Laptop with 8 GB of memory and an i5–4300U CPU with two cores having 1.9 GHz each. For $a$ equal to 1156498338479533988831879398816130438976947840284552529528425205293802195204696

$$39630006485805791442064043481185654247216831550683337015346740796660618513209536238110527284937458088300717019759850,$$

which has 200 digits and factorizes as the product

$$3 \cdot 7 \cdot 11 \cdot 13 \cdot 17 \cdot 19 \cdot 23 \cdot 29 \cdot 31 \cdot 37 \cdot 41 \cdot 43 \cdot 47 \cdot 53 \cdot 59 \cdot 61 \cdot 67 \cdot 71 \cdot 73 \cdot 79 \cdot 83 \cdot 89 \cdot 97 \cdot 101 \cdot 103 \cdot 107 \cdot 109 \cdot 113 \cdot 127 \cdot 131 \cdot 137 \cdot 139 \cdot 149 \cdot 151 \cdot 157 \cdot 163 \cdot 167 \cdot 173 \cdot 179 \cdot 181 \cdot 191 \cdot 193 \cdot 197 \cdot 199 \cdot 211 \cdot 223 \cdot 227 \cdot 229 \cdot 233 \cdot 239 \cdot 241 \cdot 251 \cdot 257 \cdot 263 \cdot 269 \cdot 271 \cdot 277 \cdot 281 \cdot 283 \cdot 293 \cdot 307 \cdot 311 \cdot 313 \cdot 317 \cdot 331 \cdot 337 \cdot 347 \cdot 349 \cdot 353 \cdot 359 \cdot 367 \cdot 373 \cdot 379 \cdot 383 \cdot 389 \cdot 397 \cdot 401 \cdot 409 \cdot 419 \cdot 421 \cdot 431 \cdot 433 \cdot 439 \cdot 443 \cdot 449 \cdot 457 \cdot 461 \cdot 463 \cdot 467 \cdot 473 \cdot 487 \cdot 491 \cdot 499 \cdot 503 \cdot 509 \cdot 521 \cdot 523 \cdot 541 \cdot 547 \cdot 557 \cdot 563 \cdot 569 \cdot 571 \cdot 577 \cdot 587 \cdot 593 \cdot 599 \cdot 601 \cdot 607 \cdot 613 \cdot 617 \cdot 619 \cdot 631 \cdot 641 \cdot 643 \cdot 647 \cdot 653 \cdot 659 \cdot 661 \cdot 673 \cdot 677 \cdot 683 \cdot 691 \cdot 701 \cdot 709 \cdot 719 \cdot 727 \cdot 733 \cdot 739 \cdot 743 \cdot 751 \cdot 757 \cdot 761 \cdot 769 \cdot 773 \cdot 787 \cdot 797 \cdot 809 \cdot 811 \cdot 821 \cdot 823 \cdot 827 \cdot 829 \cdot 833 \cdot 839 \cdot 853 \cdot 857 \cdot 863 \cdot 877 \cdot 881 \cdot 883 \cdot 887 \cdot 907 \cdot 911 \cdot 919 \cdot 929 \cdot 937 \cdot 941 \cdot 947 \cdot 953 \cdot 967 \cdot 971 \cdot 977 \cdot 983 \cdot 991 \cdot 997,$$

which factorizes as the product

$$a = 2^4 \cdot 23 \cdot 71 \cdot a' \text{ with } a' \text{ composite.}$$

Our implementation needs 0.009 s to compute $\hat{h}(P)$; Magma's CanonicalHeight did not terminate in six weeks. For this $a$, the computation of the canonical height of $50P$, which has naive height $h(50P) \approx 1437536.77$, took 0.215 s, whereas it took Magma 2.83 s to even compute $50P$!

For random $a$ having 5000 digits, the computation of $\hat{h}(P)$ to the standard precision of 30 decimal digits usually takes about 0.7 s. Our implementation is usually faster than CanonicalHeight if $a$ has at least 18 decimal digits. Note that in contrast to our implementation, the Magma implementation of the algorithm of Bost–Mestre for $\lambda_\infty$ is written in C.

The family $E_a$ has no special properties or applications that we are aware of; it was merely chosen for testing purposes.

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