EdMSM: Multi-Scalar-Multiplication for recursive SNARKs and more

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Abstract. The bottleneck in the proving algorithm of most of elliptic-curve-based SNARK proof systems is the Multi-Scalar-Multiplication (MSM) algorithm. In this paper we give an overview of a variant of the Pippenger MSM algorithm together with a set of optimizations tailored for curves that admit a twisted Edwards form. This is the case for SNARK-friendly chains and cycles of elliptic curves, which are useful for recursive constructions.

Accelerating the MSM over these curves on mobile devices is critical for deployment of recursive proof systems on mobile applications. This work is implemented in Go and uses hand-written `arm64` assembly for accelerating the finite field arithmetic (`bigint`). This work was implemented as part of a submission to the ZPrize competition in the open division "Accelerating MSM on Mobile" (`https://www.zprize.io/`). We achieved a 78% speedup over the ZPrize baseline implementation in Rust.

Keywords: elliptic curves · multi-scalar-multiplication · implementation · zero-knowledge proof

1 Introduction

A SNARK is a cryptographic primitive that enables a prover to prove to a verifier the knowledge of a satisfying witness to a non-deterministic (NP) statement by producing a proof π such that the size of π and the cost to verify it are both sub-linear in the size of the witness. Today, the most efficient SNARKs use elliptic curves to generate and verify the proof. A SNARK usually consists in three algorithms *Setup*, *Prove* and *Verify*.

The *Setup* and *Prove* algorithms involve solving multiple large instances of tasks about polynomial arithmetic in $\mathbb{F}_r[X]$ and multi-scalar multiplication (MSM) over the points of an elliptic curve. Fast arithmetic in $\mathbb{F}_r[X]$, when manipulating large-degree polynomials, is best implemented using the Fast Fourier Transform (FFT) [Pol71] and MSMs of large sizes are best implemented using a variant of Pippenger's algorithm [BDLO12, Section 4]. For example, Table 1 reports the numbers of MSMs required in the *Setup*, *Prove* and *Verify* algorithms in the [Gro16] SNARK and the KZG-based PLONK universal SNARK [GWC19]. The report excludes the number of FFTs as the dominating cost for such constructions is the MSM computation ($\sim 80\%$ of the overall time).

Given a set of n elements $G_1, \cdots, G_n$ (bases) in $\mathbb{G}$ a cyclic group (e.g. group of points on an elliptic curve) whose order $\#\mathbb{G}$ has b-bit and a set of n integers $a_1, \cdots, a_n$ (scalars) between 0 and $\#\mathbb{G}$, the goal is to compute efficiently the group element $[a_1]G_1 + \cdots + [a_n]G_n$. In SNARK applications, we are interested in large instances of variable-base MSMs ($n = 10^7, 10^8, 10^9$) — with random bases and random scalars — over the pairing groups $\mathbb{G}_1$ and $\mathbb{G}_2$.

Table 1: Cost of `Setup`, `Prove` and `Verify` algorithms for [Gro16] and PLONK. m = number of wires, n = number of multiplication gates, a = number of addition gates and ℓ = number of public inputs. $\mathtt{M}_{\mathbb{G}}$ = multiplication in $\mathbb{G}$ and `P`=pairing. *Note:* Both Groth16 and PLONK verifiers have a dependency on the number of public inputs ℓ, but for PLONK it is just a polynomial evaluation (FFT).

	`Setup`	`Prove`	`Verify`
[Gro16]	$3n$ $\mathtt{M}_{\mathbb{G}_1}$ m $\mathtt{M}_{\mathbb{G}_2}$	$(3n+m-\ell)$ $\mathtt{M}_{\mathbb{G}_1}$ n $\mathtt{M}_{\mathbb{G}_2}$	3 `P` ℓ $\mathtt{M}_{\mathbb{G}_1}$
PLONK (KZG)	d $(\geq n+a)$ $\mathtt{M}_{\mathbb{G}_1}$ 1 $\mathtt{M}_{\mathbb{G}_2}$	$9(n+a)$ $\mathtt{M}_{\mathbb{G}_1}$	2 `P` 18 $\mathtt{M}_{\mathbb{G}_1}$

The naive algorithm uses a double-and-add strategy to compute each $[a_i]G_i$ then adds them all up, costing on average $3/2 \cdot b \cdot n$ group operations ($+$). There are several algorithms that optimize the total number of group operations as a function of n such as Strauss [Str64], Bos–Coster [dR95, Sec. 4] and Pippenger [Pip76] algorithms. For large instances of a variable-base MSM, the fastest approach is a variant of Pippenger's algorithm [BDLO12, Sec. 4]. For simplicity, we call it the bucket method. In this paper we are interested in the bucket-method MSM on inner curves of 2-chains and 2-cycles of elliptic curves. We've chosen to test and benchmark our results on mobile devices because it is critical for deployment of recursive proof systems on mobile applications.

Organization of the paper. Section 2 provides definitions of 2-chains and 2-cycles of elliptic curves and some results we prove. In section 3, we explain the bucket method and provide its complexity analysis. The core of the paper are sections 4 and 5. Section 4 provides our optimizations to the bucket method for both generic elliptic curves and the twisted Edwards curves. We prove that inner 2-chains and 2-cycles always fall into the second more optimized case. Finally, section 5 reports on our implementation of the bucket method alongside our optimizations. We choose to tailor the implementation to the widely used BLS12-377 curve and to benchmark our results on a mobile widely available mobile device.

2 2-chain and 2-cycle of elliptic curves

2.1 2-chains

Following [EG22], a 2-chain of elliptic curves is a set of two curves as in Definition 2.

Definition 1. A 2-chain of elliptic curves is a list of two distinct curves $E_1/\mathbb{F}_{p_1}$ and $E_1/\mathbb{F}_{p_2}$ where p_1 and p_2 are large primes and $p_1 \mid \#E_2(\mathbb{F}_{p_2})$. SNARK-friendly 2-chains are composed of two curves that have highly 2-adic subgroups of orders $r_1 \mid \#E_1(\mathbb{F}_{p_1})$ and $r_2 \mid \#E_2(\mathbb{F}_{p_2})$ such that $r_1 \equiv r_2 \equiv 1 \mod 2^L$ for a large integer $L \geq 1$. This also means that $p_1 \equiv 1 \mod 2^L$.

In a 2-chain, the first curve is denoted the *inner curve*, while the second curve whose order is the characteristic of the inner curve, is denoted the *outer curve* (cf. Fig. 1).

Inner curves from polynomial families. The best elliptic curves amenable to efficient implementations arise from polynomial based families. These curves are obtained by parameterizing the Complex Multiplication (CM) equation with polynomials $p(x), t(x), r(x)$ and $y(x)$. The authors of [EG22] showed that the polynomial-based pairing-friendly Barreto–Lynn–Scott families of embedding degrees $k = 12$ (BLS12) and $k = 24$ (BLS24) [BLS03]

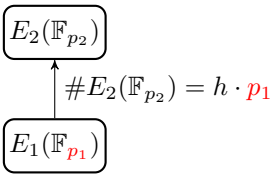


Figure 1: A 2-chain of elliptic curves.

are the most suitable to construct inner curves in the context of pairing-based SNARKs. These curves require the seed x to satisfy $x \equiv 1 \bmod 3 \cdot 2^L$ to have the 2-adicity requirement with respect to both r and p.

A particular example of an efficient 2-chain for SNARK applications is composed of the inner curve BLS12-377 [BCG+20] and the outer curve BW6-761 [EG20].

We prove useful results (Prop. 1 and Lemma 1) that will be needed later to optimize the MSM computation.

Proposition 1 ([EG22, Sec. 3.4])**.** *All inner BLS curves admit a short Weierstrass form* $Y^2 = X^3 + 1$.

Proof. Let $E : Y^2 = X^3 + b$ be a BLS curve over $\mathbb{F}_p$ parametrized by polynomials in x [BLS03]. Let g neither a square nor a cube in $\mathbb{F}_p$. One choice of $b \in \{1, g, g^2, g^3, g^4, g^5\}$ gives a curve with the correct order (i.e. $r \mid \#E(\mathbb{F}_p)$) [Sil09, §X.5]. For all BLS curves, $x-1 \mid \#E(\mathbb{F}_p)$ and $3 \mid x-1$ (which leads to all involved parameters being integers) [BLS03]. If, additionally, $2 \mid x-1$ then $2, 3 \mid \#E(\mathbb{F}_p)$ and the curve has points of order 2 and 3. A 2-torsion point is $(x_0, 0)$ with x_0 a root of $x^3 + b$, hence $b = (-x_0)^3$ is a cube. The two 3-torsion points are $(0, \pm\sqrt{b})$ hence b is a square. This implies that b is a square and a cube in $\mathbb{F}_p$ and therefore $b = 1$ is the only solution in the set $\{g^i\}_{0 \le i \le 5}$ for half of all BLS curves: those with odd x.

An inner BLS curve has a seed x such that $x \equiv 1 \mod 3 \cdot 2^L$ for some large integer L [EG22]. This means that $2 \mid x - 1$ and hence that the curve is always of the form $Y^2 = X^3 + 1$. □

Lemma 1. *All inner BLS curves admit a twisted Edwards form* $ay^2 + x^2 = 1 + dx^2y^2$ *with* $a = 2\sqrt{3} - 3$ *and* $d = -2\sqrt{3} - 3$ *over* $\mathbb{F}_p$*. If further* $-a$ *is a square, the equation becomes* $-x^2 + y^2 = 1 + d'x^2y^2$ *with* $d' = 7 + 4\sqrt{3} \in \mathbb{F}_p$.

Proof. Proposition 1 shows that all inner BLS curves are of the form $W_{0,1} : y^2 = x^3 + 1$. The following map

$$\begin{aligned} W_{0,1} &\to E_{a,d} \\ (x, y) &\mapsto \left(\frac{x+1}{y}, \frac{x+1-\sqrt{3}}{x+1+\sqrt{3}}\right) \end{aligned}$$

defines the curve $E_{a,d} : ay^2 + x^2 = 1 + dx^2y^2$ with $a = 2\sqrt{3} - 3$ and $d = -2\sqrt{3} - 3$. The inverse map is

$$\begin{aligned} E_{a,d} &\to W_{0,1} \\ (x, y) &\mapsto \left(\frac{(1+y)\sqrt{3}}{1-y} - 1, \frac{(1+y)\sqrt{3}}{(1-y)x}\right) \end{aligned}$$

If $-a$ is a square in $\mathbb{F}_p$, the map $(x, y) \mapsto (x/\sqrt{-a}, y)$ defines from $E_{a,d}$ the curve $E_{-1,d'}$ of equation $-x^2 + y^2 = 1 + d'x^2y^2$ with $d' = -d/a = (2\sqrt{3}+3)/(2\sqrt{3}-3) = 7 + 4\sqrt{3}$.

These maps work only if $\sqrt{3}$ is defined in $\mathbb{F}_p$, that is 3 is a quadratic residue. This is always the case in $\mathbb{F}_p$ on which an inner BLS curve is defined. Let $\left(\frac{3}{p}\right)$ be $3^{\frac{p-1}{2}} \mod p$,

the Legendre symbol. The quadratic reciprocity theorem tells us that $\left(\frac{3}{p}\right)\left(\frac{p}{3}\right) = (-1)^{\frac{p-1}{2}}$. We have $p \equiv 1 \mod 4$ from the 2-adicity condition, so $\left(\frac{3}{p}\right) = \left(\frac{p}{3}\right)$. Now $\left(\frac{p}{3}\right) \equiv p \mod 3$ which is always equal to 1 for all BLS curves ($x \equiv 1 \mod 3$ and $x - 1 \mid p - 1$). More generally one can prove that when $p = 2$ or $p \equiv 1$ or $11 \mod 12$ then 3 is a quadratic residue in $\mathbb{F}_p$. For inner BLS, we have $p \equiv 1 \mod 3 \cdot 2^L$ with $L \gg 2$. □

2.2 2-cycles

Definition 2. A 2-cycle of elliptic curves is a list of two distinct prime-order curves $E_1/\mathbb{F}_{p_1}$ and $E_1/\mathbb{F}_{p_2}$ where p_1 and p_2 are large primes, $p_1 = \#E_2(\mathbb{F}_{p_2})$ and $p_2 = \#E_1(\mathbb{F}_{p_1})$. SNARK-friendly 2-cycles are composed of two curves that have highly 2-adic subgroups, i.e. $\#E_1(\mathbb{F}_{p_1}) \equiv \#E_2(\mathbb{F}_{p_2}) \equiv 1 \mod 2^L$ for a large integer $L \geq 1$. This also means that $p_1 \equiv p_2 \equiv 1 \mod 2^L$.

This notion was initially introduced under different names, for example *amicable pairs* (or equivalently *dual elliptic primes* [Mih07]) for 2-cycles of ordinary curves, and *aliquot cycles* for the general case [SS11]. Some examples of SNARK-friendly 2-cycles include MNT4-MNT6 curves [BCTV14], Tweedle curves [BGH19] and Pasta curves [Hop20].

In particular a 2-cycle is a 2-chain where both curves are inner and outer curves with respect to each other (cf. Fig. 2). This means that both curves in a 2-cycle admit a twisted Edwards form following the same reasoning as in subsection 2.1. In the sequel we will focus on the case of BLS12 inner curves that form a 2-chain but we stress that these results apply to 2-chain inner curves from other families (e.g. BLS24 and BN [AHG22]) and to 2-cycles as well.

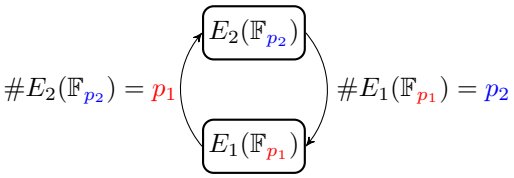


Figure 2: A cycle of elliptic curves.

3 The bucket method

The high-level strategy of the bucket-method MSM can be given in three steps:

- Step 1: reduce the b-bit MSM to several c-bit MSMs for some fixed $c \leq b$
- Step 2: solve each c-bit MSM efficiently
- Step 3: combine the c-bit MSMs into the final b-bit MSM

3.1 Step 1: reduce the b-bit MSM to several c-bit MSMs

1. Choose a window $c \leq b$
2. Write each scalar $a_1, \cdots, a_n$ in binary form and partition each into c-bit parts

$$a_i = \underbrace{(a_{i,1}, a_{i,2}, \cdots, \underbrace{a_{i,b/c}}_{c\text{-bit}})_2}_{b\text{-bit}}$$

3. Deduce b/c instances of c-bit MSMs from the partitioned scalars

$$\begin{aligned} T_1 &= [a_{1,1}]G_1 + \cdots + [a_{n,1}]G_n \\ &\vdots \\ T_j &= [a_{1,j}]G_1 + \cdots + [a_{n,j}]G_n \\ &\vdots \\ T_{b/c} &= [a_{1,b/c}]G_1 + \cdots + [a_{n,b/c}]G_n \end{aligned}$$

Cost of Step 1 is negligible.

3.2 Step 3: combine the c-bit MSMs into the final b-bit MSM

Algorithm 1 gives an iterative way to combine the small MSMs into the original MSM.

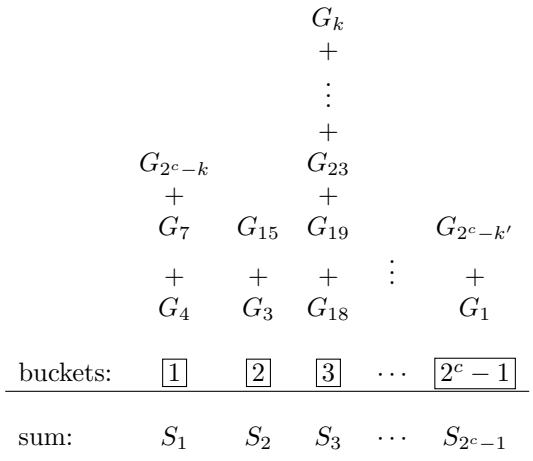
Algorithm 1: Step 3
Output: $T = [a_1]G_1 + \cdots + [a_n]G_n$

```
T ← T_1;
for i from 2 to b/c do
    T ← [2^c]T;                    // Double c times
    T ← T + T_i;                   // Add
return T;
```

Cost of Step 3: $(b/c-1)(c+1) = b - c + b/c - 1$ group operations.

3.3 Step 2: solve each c-bit MSM T_j efficiently

1. For each T_j, accumulate the bases G_i inside buckets
   Each element $a_{i,j}$ is in the set $\{0, 1, 2, \cdots 2^c - 1\}$. We initialize $2^c - 1$ empty buckets (with points at infinity) and accumulate the bases G_i from each T_j inside the bucket corresponding to the scalar $a_{i,j}$.



Cost: $n - (2^c - 1) = n - 2^c + 1$ group operations.

2. Combine the buckets to compute T_j
   This step is also a c-bit MSM of size $2^c - 1$ but this time the scalars are ordered and known in advance $S_1 + [2]S_2 + \cdots + [2^c - 1]S_{2^c-1}$, thus we can compute this instance efficiently as follows

$$
\begin{array}{cccccccccccccc}
 & & S_{2^c-1} & & & & & & & & & & & \\
+ & & S_{2^c-1} & + & S_{2^c-2} & & & & & & & & & \\
 & & \vdots & & & & & & & & & & & \\
+ & & S_{2^c-1} & + & S_{2^c-2} & + & \cdots & + & S_3 & + & S_2 & & & \\
+ & & S_{2^c-1} & + & S_{2^c-2} & + & \cdots & + & S_3 & + & S_2 & + & S_1 \\
\hline
 & [2^c-1]S_{2^c-1} & + & [2^c-2]S_{2^c-2} & + & \cdots & + & [3]S_3 & + & [2]S_2 & + & S_1 & &
\end{array}
$$

Cost: $2(2^c - 2) + 1 = 2^{c+1} - 3$ group operations.

Cost of Step 2: $n - 2^c + 1 + 2^{c+1} - 3 = n + 2^c - 2$ group operations.

Combining Steps 1, 2 and 3, the expected overall cost of the bucket method is

Total cost: $\frac{b}{c}(n + 2^c) + (b - c - b/c - 1) \approx \frac{b}{c}(n + 2^c)$ group operations.

Remark 1 (On choosing c). The theoretical minimum occurs at $c \approx \log n$ and the asymptotic scaling looks like $\%_0(b\frac{n}{\log n})$. However, in practice, empirical choices of c yield a better performance because the memory usage scales with 2^c and there are fewer edge cases if c divides b. For example, with $n = 10^7$ and $b = 256$, we observed a peak performance at $c = 16$ instead of $c = \log n \approx 23$.

4 Optimizations

4.1 Parallelism

Since each c-bit MSM is independent of the rest, we can compute each (Step 2) on a separate core. This makes full use of up to b/c cores but increases memory usage as each core needs $2^c - 1$ buckets (points). If more than b/c cores are available, further parallelism does not help much because m MSM instances of size n/m cost more than 1 MSM instance of size n.

4.2 Precomputation

When the bases $G_1, \cdots, G_n$ are known in advance, we can use a smooth trade-off between precomputed storage vs. run time. For each base G_i, choose k as big as the storage allows and precompute k points $[2^c - k]G, \cdots, [2^c - 1]G$ and use the bucket method only for the first $2^c - 1 - k$ buckets instead of $2^c - 1$. The total cost becomes $\approx \frac{b}{c}(n + 2^c - k)$. However, large MSM instances already use most available memory. For example, when $n = 10^8$ our implementation needs 58GB to store enough BLS12-377 curve points to produce a Groth16 [Gro16] proof. Hence, the precomputation approach yield negligible improvement in our case.

4.3 Algebraic structure

Since the bases $G_1, \cdots, G_n$ are points in $\mathbb{G}_1$ (or $\mathbb{G}_2$), we can use the algebraic structure of elliptic curves to further optimize the bucket method.

Non-Adjacent-Form (NAF). Given a point $G_i = (x, y) \in \mathbb{G}_1$ (or $\mathbb{G}_2$), on a Weierstrass curve for instance, the negative $-G_i$ is $(x, -y)$. This observation is well known to speed up the scalar multiplication $[s]G_i$ by encoding the scalar s in a signed binary form $\{-1, 0, 1\}$ (later called 2-NAF — the first usage might go back to 1989 [MO90]). However, this does not help in the bucket method because the cost increases with the number of possible scalars regardless of their encodings. For a c-bit scalar, we always need $2^c - 1$ buckets. That is said, we can use the 2-NAF decomposition differently. Instead of writing the c-bit scalars in the set $\{0, \cdots, 2^c - 1\}$, we write them in the signed set $\{-2^{c-1}, \cdots, 2^{c-1} - 1\}$ (cf. Alg. 2). If a scalar $a_{i,j}$ is strictly positive we add G_i to the bucket $S_{(a_{i,j})_2}$ as usual, and if $a_{i,j}$ is strictly negative we add $-G_i$ to the bucket $S_{|(a_{i,j})_2|}$. This way we reduce the number of buckets by half.

Total cost: $\approx \frac{b}{c}(n + 2^{c-1})$ group operations.

Algorithm 2: Signed-digit decomposition
Input: $(a_0, \cdots, a_{b/c-1}) \in \{0, \cdots, 2^c - 1\}$
Output: $(a'_0, \cdots, a'_{b/c-1}) \in \{-2^{c-1}, \cdots, 2^{c-1} - 1\}$

```
for i from 0 to b/c − 1 do
    if a_i ≥ 2^{c−1} then
        assert i ≠ b/c − 1;                    // No overflow for the final digit
        a'_i ← a_i − 2^c;                      // Force this digit into {−2^{c−1}, ··· , 2^{c−1} − 1}
        a_{i+1} ← a_{i+1} + 1;                 // Lend 2^c to the next digit
    else
        a'_i ← a_i
return (a'_0, ··· , a'_{b/c−1});
```

The signed-digit decomposition cost is negligible but it works only if the bitsize of $\#\mathbb{G}_1$ (and $\#\mathbb{G}_2$) is strictly bigger than b. We use the spare bits to avoid the overflow. This observation should be taken into account at the curve design level.

Curve forms and coordinate systems. To minimize the overall cost of storage but also run time, one can store the bases G_i in affine coordinates. This way we only need the tuples (x_i, y_i) for storage (although we can batch-compress these following [Kos21]) and we can make use of mixed addition with a different coordinate systems.

The overall cost of the bucket method is $\frac{b}{c}(n+2^{c-1})+(b-c-b/c-1)$ group operations. This can be broken down explicitly to:

- Mixed additions: to accumulate G_i in the c-bit MSM buckets with cost $\frac{b}{c}(n-2^{c-1}+1)$
- Additions: to combine the bucket sums with cost $\frac{b}{c}(2^c - 3)$
- Additions and doublings: to combine the c-bit MSMs into the b-bit MSM with cost $b - c + b/c - 1$
  - $b/c - 1$ additions and
  - $b - c$ doublings

For large MSM instances, the dominating cost is in the mixed additions as it scales with n. For this, we use extended Jacobian coordinates $\{X, Y, ZZ, ZZZ\}$ $(x = X/ZZ, y = Y/ZZZ, ZZ^3 = ZZZ^2)$ trading-off memory for run time compared to the usual Jacobian coordinates $\{X, Y, Z\}$ $(x = X/Z^2, y = Y/Z^3)$ (cf. Table 2).

Table 2: Cost of arithmetic in Jacobian and extended Jacobian coordinate systems. **m**=Multiplication and **s**=Squaring in the field.

Coordinate systems	Mixed addition	Addition	Doubling
Jacobian	$7\mathbf{m} + 4\mathbf{s}$	$11\mathbf{m} + 5\mathbf{s}$	$2\mathbf{m} + 5\mathbf{s}$
Extended Jacobian	$8\mathbf{m} + 2\mathbf{s}$	$12\mathbf{m} + 2\mathbf{s}$	$6\mathbf{m} + 4\mathbf{s}$

Remark 2. In [GW20], the authors suggest to use affine coordinates for batch addition. That is, they only compute the numerators in the affine addition, accumulate the denominators and then batch-invert them using the Montgomery trick [Mon87]. An affine addition costs $3\mathbf{m} + 1\mathbf{i}$ ($\mathbf{i}$ being a field inversion). For a single addition this is not worth it as $1\mathbf{i} > 7\mathbf{m}$ ($= 10\mathbf{m} - 3\mathbf{m}$). If we accumulate L points and batch-add them with cost $3L\mathbf{m} + L\mathbf{i} = 6L\mathbf{m} + 1\mathbf{i}$ (the Montgomery trick costing $L\mathbf{i} = 3L\mathbf{m} + 1\mathbf{i}$), this might be worth it. Assuming I=C**m**, there might be an improvement if we accumulate a number of points $L > C/4$. However, we did not observe a significant improvement in our implementation in compared to the extended Jacobian approach. This is mainly because C is large due the optimized finite field arithmetic in `bigint` library we use. This means L should be large requiring more memory.

We work over fields of large prime characteristic ($\neq 2, 3$), so the elliptic curves in question have always a short Weierstrass (SW) form $y^2 = x^3 + ax + b$. Over this form, the fastest mixed addition is achieved using extended Jacobian coordinates. However, there are other forms that enable even faster mixed additions (cf. Table 3).

Table 3: Cost of mixed addition in different elliptic curve forms and coordinate systems assuming $1\mathbf{m} = 1\mathbf{s}$. Formulas and references from [BL22].

Form	Coordinates system	Equation	Mixed addition cost
short Weierstrass	extended Jacobian	$y^2 = x^3 + ax + b$	$10\mathbf{m}$
Jacobi quartics	$XXYZZ$, doubling-oriented $XXYZZ$, $XXYZZR$, doubling-oriented $XXYZZR$	$y^2 = x^4 + 2ax^2 + 1$	$9\mathbf{m}$
Edwards	projective, inverted	$x^2 + y^2 = c^2(1 + dx^2y^2)$	$9\mathbf{m}$
twisted Edwards	extended ($XYZT$) $x = X/Z, y = Y/Z, x \cdot y = T/Z$	$ax^2 + y^2 = 1 + dx^2y^2$	$8\mathbf{m}$ (dedicated) $9\mathbf{m}$ (unified)
twisted Edwards	extended ($XYZT$) $x = X/Z, y = Y/Z, x \cdot y = T/Z$	$-x^2 + y^2 = 1 + dx^2y^2$ ($a = -1$)	$7\mathbf{m}$ (dedicated) $8\mathbf{m}$ (unified)

It appears that a twisted Edwards (tEd) form is appealing for the bucket method since it has the lowest cost for the mixed addition in extended coordinates. Furthermore, the arithmetic on this form is *complete*, i.e. the addition formulas are defined for all inputs. This improves the run time by eliminating the need of branching in case of adding the neutral element or doubling compared to a SW form. We showed in Lemma 1 that all inner BLS curves admit a tEd form.

For the arithmetic, we use the formulas in [HWCD08] alongside some optimizations. We take the example of BLS12-377 for which $a = -1$:

- To combine the c-bit MSMs into a b-bit MSM we use unified additions [HWCD08, Sec. 3.1] ($9\mathbf{m}$) and dedicated doublings [HWCD08, Sec. 3.3] ($4\mathbf{m} + 4\mathbf{s}$).

- To combine the bucket sums we use unified additions (9**m**) to keep track of the running sum and unified re-additions (8**m**) to keep track of the total sum. We save 1**m** by caching the multiplication by $2d'$ from the running sum.
- To accumulate the G_i in the c-bit MSM we use unified re-additions with some precomputations. Instead of storing G_i in affine coordinates we store them in a custom coordinates system (X, Y, T) where $y - x = X$, $y + x = Y$ and $2d' \cdot x \cdot y = T$. This saves 1**m** and 2**a** (additions) at each accumulation of G_i.

We note that although the dedicated addition (resp. the dedicated mixed addition) in [HWCD08, Sec. 3.2] saves the multiplication by $2d'$, it costs 4**m** (resp. 2**m**) to check the operands equality: $X_1Z_2 = X_2Z1$ and $Y_1Z_2 = Y_2Z1$ (resp. $X_1 = X_2Z1$ and $Y_1 = Y_2Z1$). This cost offset makes both the dedicated (mixed) addition and the dedicated doubling slower than the unified (mixed) addition in the MSM case. We also note that the conversion of all the G_i points given on a *SW* curve with affine coordinates to points on a *tEd* curve (also with $a = -1$) with the custom coordinates (X, Y, T) is a one-time computation dominated by a single inverse using the Montgomery batch trick. In SNARKs, since the G_i are points from the proving key, this computation can be part of the *Setup* algorithm and do not impact the *Prove* algorithm. If the *Setup* ceremony is yet to be conducted, it can be performed directly with points in the twisted Edwards form.

Our implementation shows that an MSM instance of size 2^{16} on the BLS12-377 curve is 30% faster when the $\mathbb{G}_i$ points are given on a *tEd* curve with the custom coordinates compared to the Jacobian-extended-based version which takes points in affine coordinates on a *SW* curve.

5 Implementation

Submissions to the ZPrize "Accelerating MSM on Mobile" division must run on Android 12 (API level 32) and are tested on the Samsung Galaxy A13 5G (Model SM-A136ULGDXAA with SoC MediaTek Dimensity 700 (MT6833)). The MSM must be an instance of 2^{16} $\mathbb{G}_1$-points on the BLS12-377 curve. The baseline is the `arkworks` [aC22] MSM implementation in Rust (the bucket-list method), a widely used library in SNARK projects. Submissions must beat this baseline by at least 10% in order to be eligible for the prize. We implemented our algorithm in Go language using the `gnark-crypto bigint` library [BPH+]. The ZPrize judges have chosen this mobile device as a representative Android device, with specifications similar to older higher-end devices and new budget devices, and as a result represents the kind of hardware that is common today in wealthy markets, and will become common over the next 3-5 years in middle income markets. It is also widely available and relatively inexpensive. We achieved a speedup of 78% (cf. Table 4). The source code is available under MIT or Apache-2 licenses at:

`https://github.com/gbotrel/zprize-mobile-harness`

The speedup against the baseline/`arkworks` comes from the algorithmic optimizations discussed in this paper and the `bigint` arithmetic optimizations in `gnark-crypto` aimed at the `arm64` target. We use a Montgomery CIOS variant to handle the field multiplication (Details of the algorithms and proofs are in Appendix A). On `x86` architectures, `gnark-crypto` leverages the `ADX` and `BMI2` instructions to efficiently handle the interleaved carry chains in the algorithm. For `arm64` architecture, we "untangled" the carry propagation in the pure Go code to ensure the carry chains were uninterrupted. Moreover, the large number of registers available (in practice 28 for `arm64` against 14 for `x86`) allowed for an efficient implementation of the squaring function – the 64 word-word multiplications are performed at the beginning of each iteration, and the results are stored in registers. This

Table 4: Comparison of the ZPrize baseline and the submission MSM instances of 2^{16} $\mathbb{G}_1$-points on the BLS12-377 curve.

Implementation	Timing	Curve form and coordinates system	Parallelism?	Precomputation?	2-NAF buckets?
Baseline	2309 ms	SW Jacobian (X, Y, Z)	✓	✗	✗
Submission	509 ms	tEd $(a = -1)$ Custom (X, Y, T)	✓	✗	✓

allows to have uninterrupted carry chains when doubling the intermediate product. The impact of these optimizations is $\sim 17\%$ for $\mathbb{F}_p$ multiplication and $\sim 25\%$ for the squaring. For an ext-Jac MSM instance of size 2^{16}, the timing was 821ms before these `arm64` field arithmetic optimizations and 620ms after. For the *tEd-custom* version the speedup is only related to the $\mathbb{F}_p$-multiplication since there are no squaring in the mixed addition. For this same version, we stored $(y - x, y + x)$ in the coordinates system instead of (x, y) and added ~ 40 lines of `arm64` assembly for a small function in $\mathbb{F}_p$ (`Butterfly(a, b)` $\rightarrow$ `a = a + b; b = a - b`). The butterfly performance impact was $\sim 5\%$, as it speeds up the unified (mixed) addition in the *tEd* form.

However, the large gap cannot be justified by these facts only. The target device SoC can run 32-bit and 64-bit instruction sets. However, the stock firmware runs a 32-bit ARM architecture (armv7) on which the baseline implementation is benchmarked by the ZPrize judges. For the sake of the competition, we performed a static build targeting a 64-bit ARM architecture (arm64), which allowed us without a complicated build process to run the 64-bit code on the target device.

For the sake of this paper, and for a fair comparison, we perform the same architecture hack on the baseline implementation. We report in Figure 3 a comparison of our code to the baseline. We report timings of several MSM instances of different sizes and with different curve parameterizations (SW in extended Jacobians vs. tEd $(a = -1)$ in custom/extended coordinates).

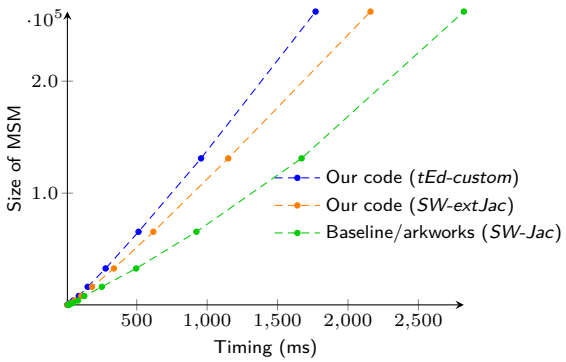


Figure 3: Comparison of our MSM code and the `arkworks` one for different instances on the BLS12-377 $\mathbb{G}_1$ group.

For the ZPrize MSM instance of size 2^{16} the speed up is 45% with the *tEd* version and 33% with the more generic *SW-extJac* version. For different sizes ranging from 2^8 to 2^{18} the speed up is 40-47% with the *tEd* version and 20-35% with *SW-extJac*.

6 Conclusion

Multi-scalar-multiplication dominates the proving cost in most elliptic-curve-based SNARKs. Inner curves such as the BLS12-377 are optimized elliptic curves suitable for both proving generic-purpose statements and in particular for proving composition and recursive statements. Hence, it is critical to aggressively optimize the computation of MSM instances on these curves. We showed that our work yield a very fast implementation both when the points are given on a short Weierstrass curve and even more when the points are given on a twisted Edwards curve. We showed that this is always the case for inner curves such as BLS12-377 and that the conversion cost is a one-time computation that can be performed in the *Setup* phase. We note that, more generally, these tricks apply to any elliptic curve that admits a twisted Edwards form — particularly SNARK-friendly 2-cycles of elliptic curves. We suggest that this should be taken into account at the design level of SNARK-friendly curves.

Open question: for Groth16, the same scalars a_i are used for both $\mathbb{G}_1$ and $\mathbb{G}_2$ MSMs. Is it possible to mutualize a maximum of computations between these two instances? It seems that moving to a type-2 pairing would allow to deduce the $\mathbb{G}_1$ instance from the $\mathbb{G}_2$ one using an efficient homomorphism (the trace map) over the resulting single point. However, $\mathbb{G}_2$ computations would be done on the much slower full extension $\mathbb{F}_{p^k}$. The pairing, needed for proof verification, would also be moderately slower because of the anti-trace map.

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A Faster `bigint` modular multiplication for most moduli

We discovered an optimization that reduces the number of operations needed to compute the modular multiplication of two big integers for most (but not quite all) choices of modulus. To the best of our knowledge, we are not aware of any prior art describing this optimization, though it is possible that we missed something in the literature.

A.1 The Montgomery multiplication: theory

The modular multiplication problem. Given integers a, b and p the modular multiplication problem is to compute the remainder of the product

$$ab \mod p\ .$$

On computers a division operation is much slower than other operations such as multiplication. Thus, a naive implementation of $ab \mod p$ using a division operation is prohibitively slow. In 1985, Montgomery introduced a method to avoid costly divisions [Mon85]. This method, now called the Montgomery multiplication, is among the fastest solutions to the problem and it continues to enjoy widespread use in modern cryptography.

Overview of the solution: the Montgomery multiplication. There are many good expositions of the Montgomery multiplication (e.g. [BM17]). As such, we do not go into detail on the mathematics of the Montgomery multiplication. Instead, this section is intended to establish notation that is used throughout this appendix.

The Montgomery multiplication algorithm does not directly compute $ab \mod p$. Instead it computes $abR^{-1} \mod p$ for some carefully chosen number R called the Montgomery radix. Typically, R is set to the smallest power of two exceeding p that falls on a computer word boundary. For example, if p is 381 bits then $R = 2^{6\times 64} = 2^{384}$ on a 64-bit architecture.

In order to make use of the Montgomery multiplication the numbers a and b must be encoded into the Montgomery form: instead of storing (a, b), we store the numbers $(\tilde{a}, \tilde{b})$ given by $\tilde{a} = aR \mod p$ and $\tilde{b} = bR \mod p$. A simple calculation shows that the Montgomery multiplication produces the product $ab \mod p$, encoded in the Montgomery form: $(aR)(bR)R^{-1} = abR \mod p$. The idea is that numbers are always stored in the Montgomery form so as to avoid costly conversions to and from the Montgomery form.

Other arithmetic operations such as addition, subtraction are unaffected by the Montgomery form encoding. But the modular inverse computation $a^{-1} \mod p$ must be adapted to account for the Montgomery form. We do not discuss modular inversion in this appendix.

A.2 The Montgomery multiplication: implementation

For security purposes, cryptographic protocols use large moduli — a ,b and p are stored on multiple machine words (multi-precision). In this appendix, we let D denote the basis in which integers are represented. (For example, $D = 2^{64}$ if a word is 64 bits). A large number n can be represented by its base-D digits $n_0, \ldots, n_m$ stored in machine words (`uint`): $\sum_{i=0}^{m} n_i D^i$.

There are several variations of multi-precision Montgomery multiplication. A popular choice is the Coarsely Integrated Operand Scanning (CIOS) variant [Aca98]. In some settings, factors such as modulus size, CPU cache management, optimization techniques, architecture and available instruction set might favor other variants.

How fast is the CIOS method? Let N denote the number of machine words needed to store the modulus p. For example, if p is a 381-bit prime and the hardware has 64-bit word size then $N = 6$. The CIOS method solves modular multiplication using $4N^2 + 4N + 2$ unsigned integer additions and $2N^2 + N$ unsigned integer multiplications.

Our optimization reduces the number of additions needed in the CIOS Montgomery multiplication to only $4N^2 - N$, a saving of $5N + 2$ additions. This optimization can be used whenever the highest bit of the modulus is zero (and not all of the remaining bits are set — see below for details).

The core of the state-of-the-art CIOS Montgomery multiplication is reproduced below. This listing is adapted from Section 2.3.2 of Tolga Acar's thesis [Aca98]. The symbols in this listing have the following meanings:

- N is the number of machine words needed to store the modulus p.
- D is the word size. For example, on a 64-bit architecture D is 2^{64}.
- $a[i], b[i], p[i]$ are the i-th words of the integers a, b and p.
- $p'[0]$ is the lowest word of the number $-p^{-1} \mod R$. This quantity is precomputed, as it does not depend on the inputs a and b.
- t is a temporary array of size $N + 2$.
- C, S are machine words. A pair (C, S) refers to (high-bits, low-bits) of a two-word number. For short we denote them (`hi`, `lo`).

Algorithm 3: The CIOS Montgomery multiplication

```
1  for i = 0 to N − 1 do
2  |  C := 0;
3  |  for j = 0 to N − 1 do
4  |  |  (C, t[j]) := t[j] + a[j] · b[i] + C
5  |  (t[N + 1], t[N]) := t[N] + C;
6  |  C := 0;
7  |  m := t[0] · p′[0]  mod D;
8  |  (C, _) := t[0] + m · p[0];
9  |  for j = 1 to N − 1 do
10 |  |  (C, t[j − 1]) := t[j] + m · p[j] + C
11 |  (C, t[N − 1]) := t[N] + C;
12 |  t[N] := t[N + 1] + C;
```

Next, we show that we can save the additions in lines 5 and 12 of Alg. 3 when the highest word of the modulus p is at most $(D-1)/2-1$. This condition holds if and only if the highest bit of the modulus is zero and not all of the remaining bits are set.

Our optimization. Observe that lines 4 and 10 have the form $(\texttt{hi}, \texttt{lo}) := m_1 + m_2 \cdot B + m_3$, where $\texttt{hi}$, $\texttt{lo}$, m_1, m_2, m_3 and B are machine-words where each is at most $D-1$. If $B \leq (D-1)/2-1$ then a simple calculation shows that

$$\begin{aligned} m_1 + m_2 \cdot B + m_3 &\leq (D-1) + (D+1)(\tfrac{D-1}{2} - 1) + (D-1) \\ &\leq D \underbrace{(\tfrac{D-1}{2})}_{\texttt{hi}} + \underbrace{(\tfrac{D+1}{2} - 1)}_{\texttt{lo}} \end{aligned}$$

From which we derive the following Lemma:

Lemma 2. *If $B \leq (D-1)/2-1$, then $\texttt{hi} \leq (D-1)/2$.*

We use Lemma 2 to prove the following Proposition:

Proposition 2. *If the highest word of p is at most $(D-1)/2-1$, then the variables $t[N]$ and $t[N+1]$ always store the value 0 at the beginning of each iteration of the outer i-loop.*

Proof. We prove this proposition by induction. The base case $i = 0$ is trivial, since the t array is initialized to 0. For the inductive step at the iteration i, we suppose that $t[N] = t[N+1] = 0$ and trace the execution through the iteration. Begin at the final iteration of the first inner loop ($j = N-1$) on line 4. Because $a < p$ and because the highest word of p is smaller than $(D-1)/2$, we may use Lemma 2 to see that the carry C is at most $(D-1)/2$. Then line 5 sets

$$\begin{aligned} t[N] &= C \\ t[N+1] &= 0 \ . \end{aligned}$$

A similar observation holds at the end of the second inner loop ($j = N-1$) on line 10: Lemma 2 implies that the carry C is at most $(D-1)/2$. We previously observed that $t[N]$ is also at most $(D-1)/2$, so $t[N] + C$ is at most

$$\tfrac{D-1}{2} + \tfrac{D-1}{2} = D - 1$$

which fits entirely into a single word. Then line 11 sets C to 0 and line 12 sets $t[N]$ to 0. The proof by induction is now complete. □

With this proposition, we no longer need the addition at line 5. We store C in a variable A and at line 11 we compute instead

$$(\_, t[N-1]) := A + C$$

The final algorithm. For $N < 12$, we merge the two inner loops

Algorithm 4: Our optimized CIOS Montgomery multiplication

```
1 for i = 0 to N − 1 do
2 |   (A, t[0]) := t[0] + a[0] · b[i];
3 |   m := t[0] · p′[0]  mod W;
4 |   C,: = t[0] + m · p[0];
5 |   for j = 1 to N − 1 do
6 |   |   (A, t[j]) := t[j] + a[j] · b[i] + A;
7 |   |   (C, t[j − 1]) := t[j] + m · p[j] + C;
8 |   t[N − 1] := C + A;
```

Our optimized Montgomery squaring. The condition on the modulus differs, here

$$p[N-1] \leq \tfrac{D-1}{4} - 1 \ .$$

However, the reasoning is similar and we end up with

Algorithm 5: Our optimized CIOS Montgomery squaring

```
1 for i = 0 to N − 1 do
2 |   C, t[i] := a[i] · a[i] + t[i];
3 |   p := 0;
4 |   for j = i + 1 to N − 1 do
5 |   |   p, C, t[j] := 2a[j] · a[i] + t[j] + (p, C)
6 |   A := C;
7 |   m := t[0] · p′[0]; C,: = t[0] + p[0] · m for j = 1 to N − 1 do
8 |   |   C, t[j − 1] := p[j] · m + t[j] + C
9 |   t[N − 1] := C + A;
```

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