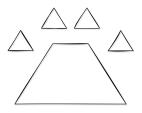
The Hitchhiker's Guide to EC-Adder HW Acceleration

Yuval Domb yuval@ingonyama.com



Mission

- Build a fully-pipelined BLS12-381 ECADDER
- Use Xilinx Virtex Ultrascale+
- Minimize DSP48 blocks

One Slice	Two Slices
26Ux19U	34Ux28U
27U×18U	35Ux27U
28U×17U	36Ux26U

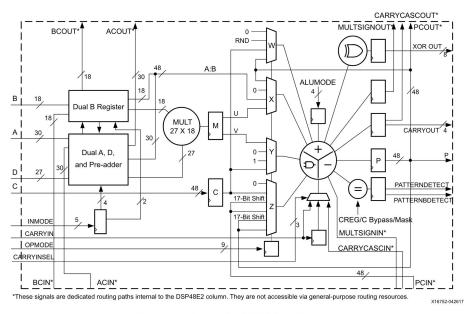


Figure 2-1: Detailed DSP48E2 Functionality

"UltraScale Architecture DSP Slice User Guide (UG579)", https://docs.amd.com/v/u/en-US/ug579-ultrascale-dsp



Elliptic Curve

EC Addition Formulas

Modular Multiplier Optimization

ECADDER Optimization

Toom-Cook Optimization

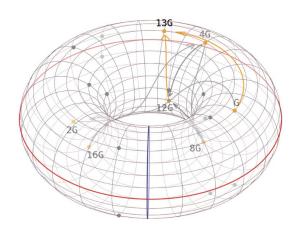
What is an Elliptic Curve?

- A smooth, projective, algebraic curve of genus 1
- Short Weierstrass Form

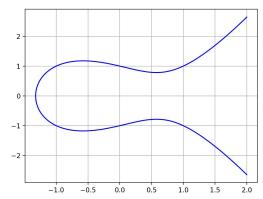
$$y^{2} = x^{3} + ax + b$$

$$4a^{3} + 27b^{2} \neq 0$$

$$a, b, x, y \in \mathbb{F}_{p^{m}}, \forall p > 3$$







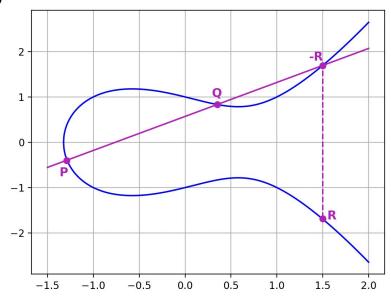


EC Additive Group $E(\mathbb{F}_p)$

- All points on curve + Point-at-Infinity O
- Define addition geometrically
- Cyclic with Generator G
- Discrete-Log hardness

$$n, G \rightarrow nG$$

$$n \not\leftarrow G, nG$$







- Distinct prime-sized subgroups, extensions, embeddings, twists...
- Pairing function a bilinear map

$$e: G_1 \times G_2 \to G_T$$

$$e(P, Q + R) = e(P, Q) \cdot e(P, R)$$

$$e(P + Q, R) = e(P, R) \cdot e(Q, R)$$

$$e(aP, bQ) = e(abP, Q) = e(P, abQ) = e(bP, aQ) = e(P, Q)^{ab}$$



Example: BLS12-381 Elliptic Curve

Barreto, Lynn, Scott

p = 0x1a0111ea397fe69a4b1ba7b6434bacd764774b84f38512bf6730d2a0f6b0f6241eabfffeb153ffffb9fefffffffaaab r = 0x73eda753299d7d483339d80809a1d80553bda402fffe5bfeffffff00000001

$$G_1 \subset E(\mathbb{F}_p), \quad E: y^2 = x^3 + 4$$

$$G_2 \subset E'(\mathbb{F}_{p^2}), \quad E': y^2 = x^3 + 4(1+i)$$

$$G_T \subset \mathbb{F}_{p^{12}}$$



Example: BLS Digital Signatures

Boneh, Lynn, Shacham

- Secret key: $s \in \mathbb{F}_r$
- Public key: $[s]_1 \in G_1$
- Message: $m \in \mathbb{F}_r$
- Signature: $[sm]_2 \in G_2$
- *Verification:* $e([s]_1, [m]_2) = e([1]_1, [sm]_2)$





Elliptic Curve

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Affine Addition (Geometric)



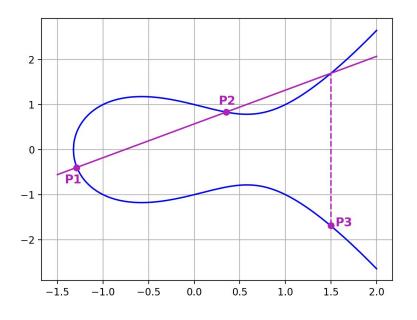
- Formula:

$$- \lambda = \frac{1}{x_2 - x_1} (y_2 - y_1)$$

-
$$x_3 = \lambda^2 - x_1 - x_2$$

-
$$y_3 = \lambda(x_1 - x_3) - y_1$$

- Cost:
 - 3 multiplication
 - 1 inversion



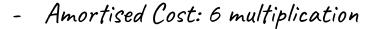
Batched Affine Addition

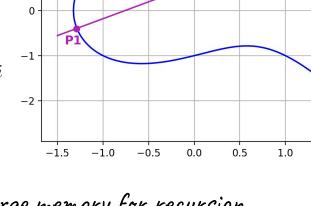


P3

1.5

- Batch many independent inversions
 - Calc iteratively: $v_0, v_0v_1, ..., \prod_0^m v_i$
 - Invert: $w=(\prod_0^m v_i)^{-1}$
 - Calc recursively: $orall j \in [m:0]$ $v_m^{-1} = w \prod_0^{m-1} v_i$ $w \leftarrow w v_m$





- Requires: many independent additions, large memory for recursion

Projective Addition



- Redundant representation: $(x,y,z) o (rac{x}{z},rac{y}{z})$
- Complete formula (Bosma and Lenstra, 1995)

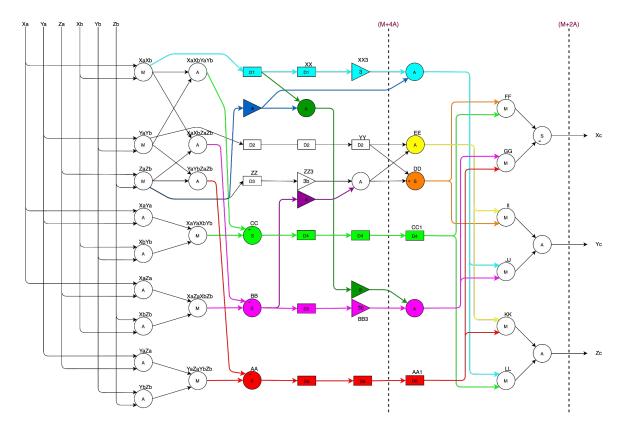
$$\begin{split} X_3 &= (X_1Y_2 + X_2Y_1)(Y_1Y_2 - a(X_1Z_2 + X_2Z_1) - 3bZ_1Z_2) \\ &- (Y_1Z_2 + Y_2Z_1)(aX_1X_2 + 3b(X_1Z_2 + X_2Z_1) - a^2Z_1Z_2), \\ Y_3 &= (3X_1X_2 + aZ_1Z_2)(aX_1X_2 + 3b(X_1Z_2 + X_2Z_1) - a^2Z_1Z_2) \\ &+ (Y_1Y_2 + a(X_1Z_2 + X_2Z_1) + 3bZ_1Z_2)(Y_1Y_2 - a(X_1Z_2 + X_2Z_1) - 3bZ_1Z_2), \\ Z_3 &= (Y_1Z_2 + Y_2Z_1)(Y_1Y_2 + a(X_1Z_2 + X_2Z_1) + 3bZ_1Z_2) \\ &+ (X_1Y_2 + X_2Y_1)(3X_1X_2 + aZ_1Z_2). \end{split}$$

- Cost: 12 field multiplications
- Highly parallelizable

@ Karthik Inbasekar



Projective Addition





Comparison of EC Addition Formulas

Туре	Mults	Sqrs	Total
Batched Affine	5	1	6
Jacobian + Affine	7	4	11
Ext. Jacobian + Affine	8	2	10
Jacobian	11	5	16
Ext. Jacobian	12	2	14
Projective	12	0	12



Elliptic Curve

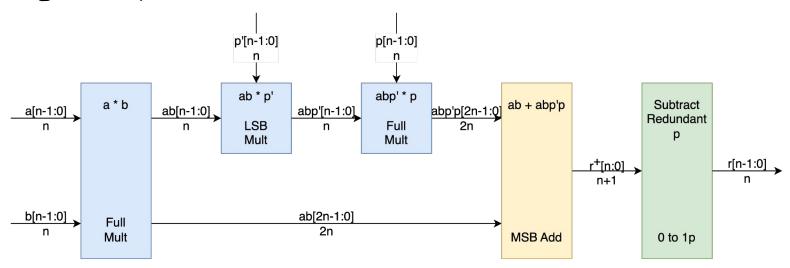
EC Addition Formulas

Modular Multiplier Optimization

ECADDER Optimization

Toom-Cook Optimization

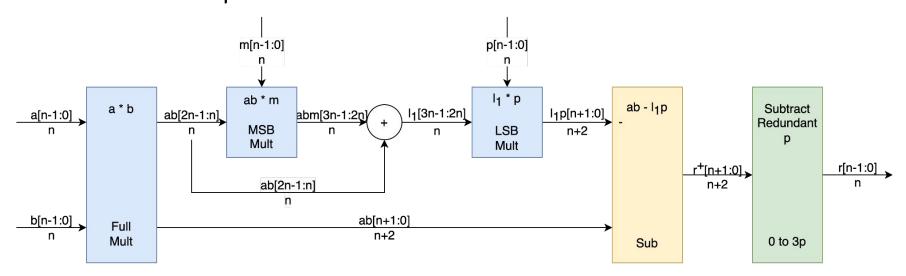
Montgomery Multiplier



- Requires special transformation to Montgomery representation
- Good for multi-precision systems such as CPUs

Barrett Multiplier

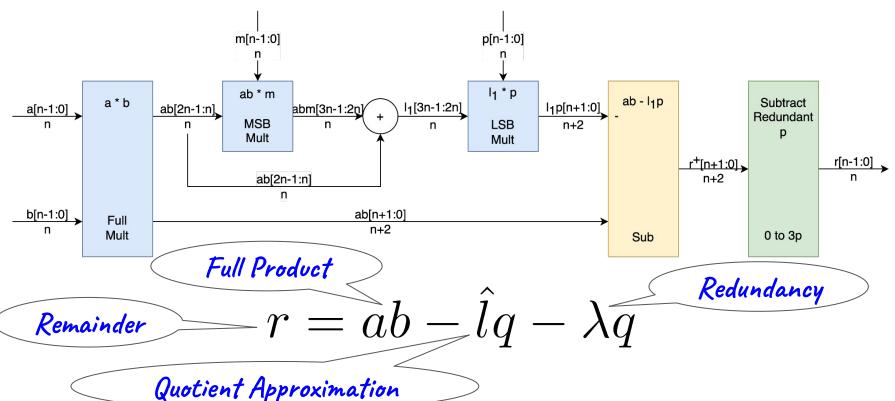




- Works for any modulo reduction
- Uses native number representation
- Provides many opportunities for general-purpose hardware optimization

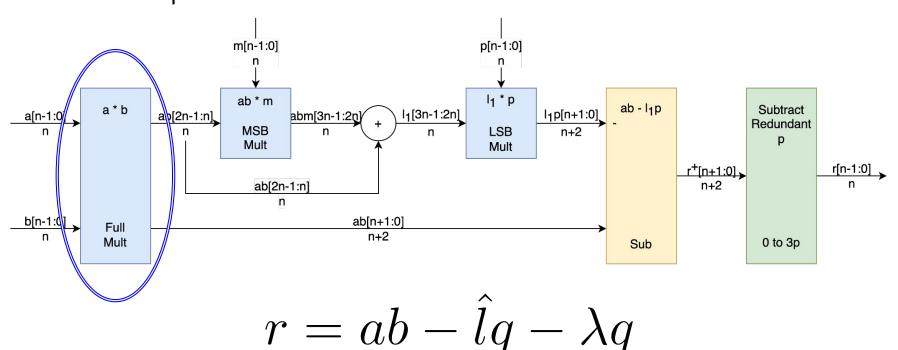








Barrett Optimization 1 - Karatsuba





Barrett Optimization 1 - Karatsuba

- Apply initially to ab full-multiplier
- General concept:

$$ab = (a_h 2^{\frac{n}{2}} + a_l)(b_h 2^{\frac{n}{2}} + b_l)$$

= $a_h b_h 2^n + ((a_h + a_l)(b_h + b_l) - a_h b_h - a_l b_l)2^{\frac{n}{2}} + a_l b_l$

- Uses 3 instead of 4 half-length multipliers
- Can be applied recursively

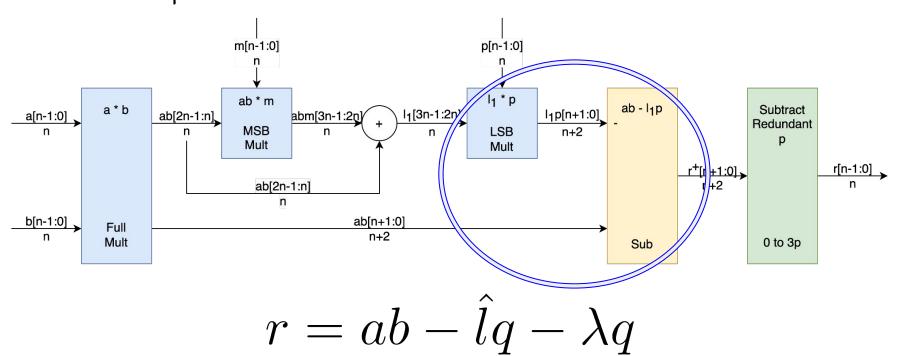


BLS12-381: Karatsuba Full Mult

Base Mult	Strategy	Calculation	Result
19U×19U	Schoolbook	ceil(381/19)^2	441 slices
27Ux27U	Schoolbook	2 * ceil(381/27)^2	450 slices
24Ux24U	Karatsuba	u = ceil(381/24) = 16	162 slices
		2 * 3^log2(u)	



Barrett Optimization 2 - LSB Mult



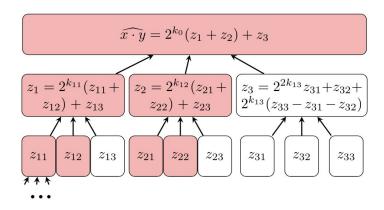
Barrett Optimization 2 - LSB Mult

- Calculate the remainder: $r=ab \mod q=ab-lq$
- By approximating $l-\lambda \leq \hat{l} \leq l$ we get: $r+\lambda q = ab-\hat{l}q$
- Using long subtraction for the case $\lambda=0$:

- <u>Conclusion</u>: Only need to calculate approximately $\,n + log_2(\lambda)\,$ LSBs of $\,\hat{l}q$



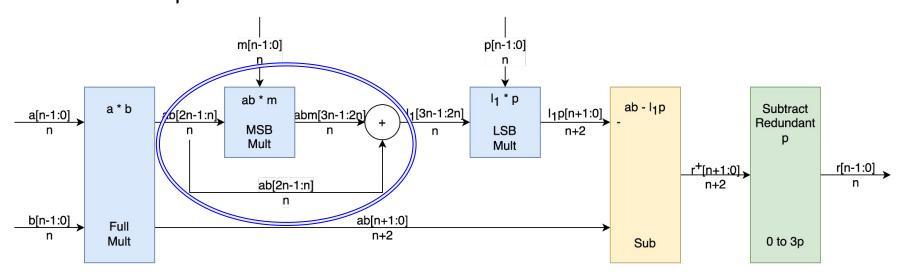
BLS12-381: Karatsuba LSB Mult with Pruning



- Using 24Ux24U multipliers with recursive Karatsuba: <120 slices



Barrett Optimization 3 - MSB Mult



$$r = ab - \hat{l}q - \lambda q$$



Barrett Optimization 3 - MSB Mult

- Since $mpprox rac{1}{a}$ cannot be represented with fixed precision:

$$\hat{l}_0 = \left\lfloor \frac{abm}{2^{2n}} \right\rfloor$$
 $e(\hat{l}_0) < 1$

By taking just the MSBs of the result:
$$\hat{l}_1 = \left\lfloor \left(\frac{ab}{2^n}\right) \cdot \frac{m}{2^n} \right\rfloor$$
 $e(\hat{l}_1) < 3$

Since:
$$\frac{abm}{2^{2n}} = \frac{ab[2n-1:n]\cdot m}{2^n} + \frac{ab[n-1:0]\cdot m}{2^{2n}} < \frac{ab[2n-1:n]\cdot m}{2^n} + 2$$



Barrett Optimization 3 - Karatsuba MSB Mult

- By approximating the MSB multiplier using Karatsuba pruning

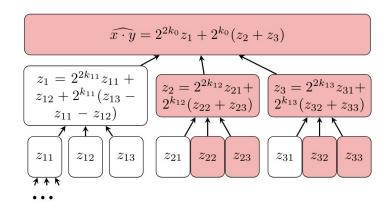
$$e(\hat{l_1}) < 3 + \left\lceil \frac{\Delta^{\{k_{ij}\}}}{2^n} \right\rceil$$

Since:
$$0 \le \Delta(x \cdot y) < 2^{2k_0} + 2^{k_0}(\Delta(z_1) + \Delta(z_2)) < 2^{2k_0} + 2^{k_0}(2^{2k_{12}} + 2^{2k_{13}} + 2^{k_{12}}(\Delta(z_{22}) + \Delta(z_{23})) + 2^{k_{13}}(\Delta(z_{32}) + \Delta(z_{33})) < \dots = \Delta^{\{k_{ij}\}}$$

@ Dmytro (Dima) Tymokhanov



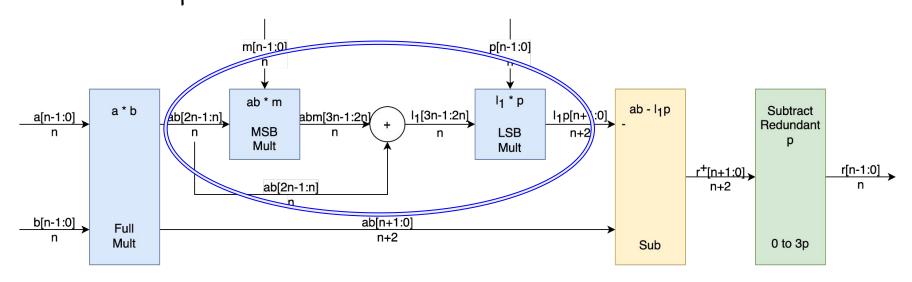
BLS12-381: Karatsuba MSB Mult with Pruning



- Using 24Ux24U multipliers with recursive Karatsuba: <120 slices



Barrett Optimization 4 - Constant Mult and CSD



$$r = ab - \hat{l}q - \lambda q$$



Barrett Optimization 4 - Constant Mult and CSD

- Redundant Signed Digit (RSD) representation uses {-1,0,1} "bits"
- Canonical Signed Digit (CSD)
 - RSD with minimal number of non-zero terms
 - Canonical form
- Example: 7 = [0,1,1,1] = [1,0,0,-1]
- A CSD form of a typical n-bits number has n/3 non-zeros



Barrett Optimization 4 - Constant Mult and CSD

- A constant multiplier is equivalent to k additions where k is the number of non-zeros terms in the constant
- For CSD the average k is n/3 but sometimes it's much less, especially for small n
- Strategy:
 - Trade the "CSD-light-weight" Karatsuba leaf multipliers with additions
 - When constant part has many leading zeros use one-slice multipliers



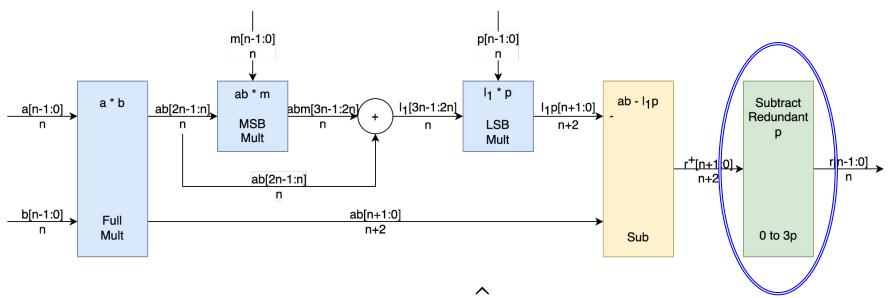
BLS12-381: Constant Mult and CSD

- Replace multipliers by additions when "CSD-weight" does no exceed 4
- Use one-slice multipliers when possible (sufficiently small constants)
- And extensive "under-the-hood" optimization...
- MSB constant mult: 62 slices
- LSB constant mult: 30 slices

@ Hadar Sackstein



Barrett Optimization 5 - Conditional Subtraction



$$r = ab - \hat{l}q - \lambda q$$

@ Tony Wu



Barrett Optimization 5 - Conditional Subtraction

- Inspired by Niall Emmart's work on MSM (see slide 26)
- Observation: The conditional subtraction of redundant copies of q depends only on the inaccuracy of the quotient estimate \hat{l} and is bounded, regardless of the size of the product ab (Reduction accuracy is independent of full product size).
- Conclusion:
 - Plan the reduction for a larger product, say 384x384.
 - Design the reduction such that $r+\lambda q$ does not exceed 384 bits.



Elliptic Curve

EC Addition Formulas

5292

Modular Multiplier Optimization

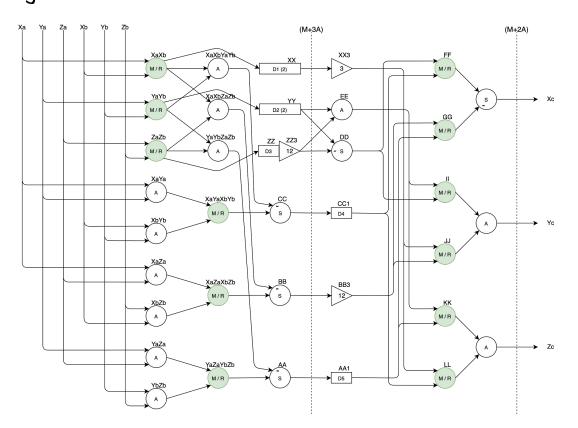
3048

ECADDER Optimization

Toom-Cook Optimization



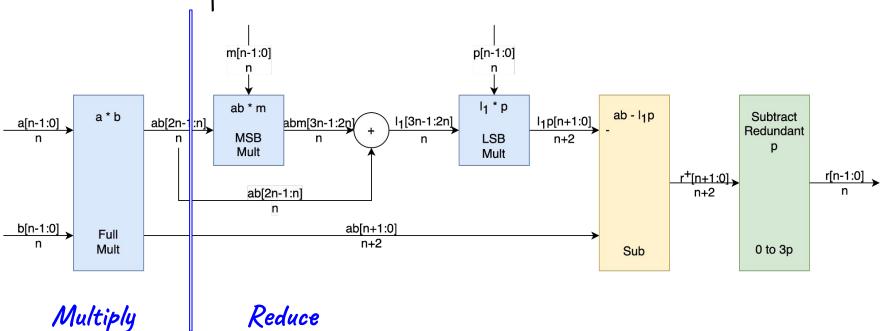
Projective ECADDER for BLS12-381 (a=0, b=4)



- 12 Multiply (162)
- 12 Reduce (92)
- 3048 slices

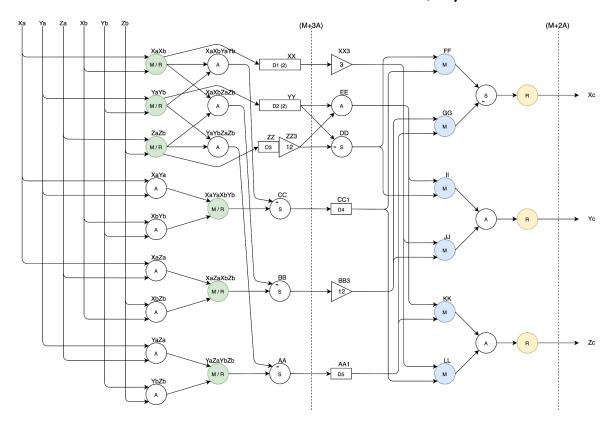


Barrett Multiplier





Projective ECADDER - Multiply/Reduce Separation



- Separate Multiply and Reduce
- Reduce input increased by 1 bit
- 12 Multiply (162)
- 9 Reduce (92)
- 2772 slices



Sidenote: G2 Field Multiplication

- Reminder: $G_2 \subset E'(\mathbb{F}_{p^2}), \quad E': y^2 = x^3 + 4(1+i)$
- General concept (Karatsuba):

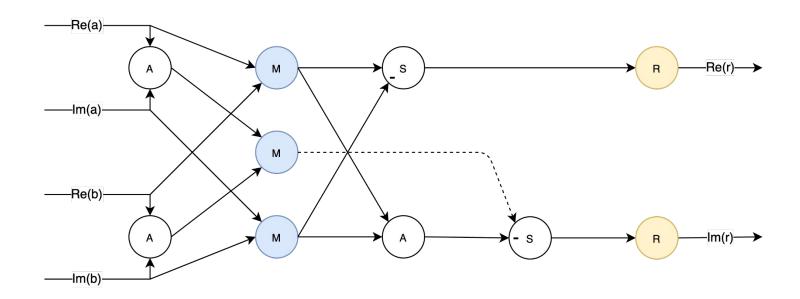
$$ab = (a_r + ia_i)(b_r + ib_i)$$

$$= a_r b_r - a_i b_i + i((a_r + a_i)(b_r + b_i) - a_r b_r - a_i b_i)$$

- Multiply: 3 base-field multipliers
- Reduce: 2 base-field reductions



Sidenote: G2 Field Multiplication





Elliptic Curve

EC Addition Formulas 5292

Modular Multiplier Optimization 3048

ECADDER Optimization 2772

Toom-Cook Optimization

Toom-Cook Multiplication



- A generalization of Karatsuba
- Represent multiplicands as polynomials sampled at $\,2^k$ where $\,n=mk$

$$a = a_{m-1}2^{(m-1)k} + \dots + a_12^k + a_0$$

$$a(x) = a_{m-1}x^{m-1} + \dots + a_1x + a_0$$

$$\Rightarrow a = a(x = 2^k)$$

- Calculate polynomial product in a suitable evaluations domain
- Transform product-polynomial back to coefficients domain and sample at $\,\,_{2k}$



Toom-Cook Multiplication (e.g. n=3k)

- Represent as polynomial product:

$$p(x) = (a_2x^2 + a_1x + a_0)(b_2x^2 + b_1x + b_0)$$

- Evaluate over domain $\{0,1,-1,-2,\infty\}$ of length 5:

$$\begin{bmatrix} p(0) \\ p(1) \\ p(-1) \\ p(-2) \\ p(\infty) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & -2 & 4 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \\ a_2 \end{bmatrix} \odot \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & -1 & 1 \\ 1 & -2 & 4 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} b_0 \\ b_1 \\ b_2 \end{bmatrix}$$

Element-wise Multiplication

[&]quot;Toom-cook multiplication", https://en.wikipedia.org/wiki/Toom-Cook multiplication

Toom-Cook Multiplication (e.g. n=3k)



- The product polynomial follows:

$$\begin{bmatrix} p(0) \\ p(1) \\ p(-1) \\ p(-2) \\ p(\infty) \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 & 1 \\ 1 & -2 & 4 & -8 & 16 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} p_0 \\ p_1 \\ p_2 \\ p_3 \\ p_4 \end{bmatrix}$$

- Inverting the above results in.

$$\begin{bmatrix} p_0 \\ p_1 \\ p_2 \\ p_3 \\ p_4 \end{bmatrix} = \underbrace{1}_{3} \begin{bmatrix} 3 & 0 & 0 & 0 & 0 & 0 \\ 3/2 & 1 & -3 & 1/2 & -6 \\ -3 & 3/2 & 3/2 & 0 & -3 \\ -3/2 & 1/2 & 3/2 & -1/2 & 6 \\ 0 & 0 & 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} p(0) \\ p(1) \\ p(-1) \\ p(-2) \\ p(\infty) \end{bmatrix}$$



Toom-Cook Multiplication Implications

- Pro: Number of multiplications equals order of resulting product-polynomial
- Pro: Like Karatsuba, can be applied recursively
- Pro: Extendable to asymmetric multi-precision schemes
- Pro: Extendable to products of more than 2 arguments
- Con: Involves many small constant multiplications
- Con: Requires division 😕





Mult	381/Width	Full Mult	MSB+LSB	Total
19U×19U	20.1 < 24 = 2*3*4	3*5*7 = 105	60	1800
24U×24U	15.9 < 16 = 2^4	2(3^4) = 162	92	2772
27Ux27U	14.1 < 15 = 3*5	2(5*9) = 90	52	1548



Elliptic Curve

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Toom-Cook Optimization	1548



Ingonyama's Aleo IP uses TC ECADDER

	Aleo IP	RTX 4090	RTX 3090	RTX 3080
Proofs / joule	2000	35.6	21.9	18.8
Proofs per sec / mm2	2000	26.3	11.1	9.6
Proofs / second	6000 x N	16000	7000	6000
Power (W)	3 x N	450	320	320

Thank You

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