Modular Multiplication and Squaring on ARM-NEON Revisited

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Abstract. Modular multiplication constitutes the "arithmetic foundation" of modern public-key cryptography with applications ranging from RSA, DSA and Diffie-Hellman over elliptic curve schemes to pairing-based cryptosystems. The increased prevalence of SIMD-type instructions in commodity processors (e.g. Intel SSE, ARM NEON) has initiated a massive body of research on vector-parallel implementations of Montgomery modular multiplication. In this paper, we introduce the Cascade Operand Scanning (COS) method to speed up multi-precision multiplication on SIMD architectures. We developed the COS technique with the goal of reducing Read-After-Write (RAW) dependencies in the propagation of carries, which also reduces the number of pipeline stalls (i.e. bubbles). The COS method operates on 32-bit words in a row-wise fashion (similar to the operand-scanning method) and does not require a "non-canonical" representation of operands with a reduced radix. We show that two COS computations can be "coarsely" integrated into an efficient vectorized variant of Montgomery multiplication, which we call Coarsely Integrated Cascade Operand Scanning (CICOS) method. Due to our sophisticated instruction scheduling, the CICOS method reaches record-setting execution times for Montgomery modular multiplication on ARM-NEON platforms. Detailed benchmarking results obtained on an ARM Cortex-A9 and Cortex-A15 processors show that the proposed CICOS method outperforms Bos et al's implementation from SAC 2013 by up to 57% (A9) and 40% (A15), respectively. Furthermore, we provide a novel squaring named Double Operand Scanning (DOS) method The method boosts squaring operation by replacing the duplicated multiplications into doubling the operands, which enhances COS method by 23.8% (A9) and 19.1% (A15), respectively. Finally, we suggest a novel fast reduction method for SIMD architecture. This operation conducts reduction in 2-way fashion. The methods result in 43% and 16.5% (APQ8060) performance enhancements in finite field multiplication and scalar multiplication than HPEC2013.

Keywords: Public-key cryptography, modular arithmetic, SIMD-level parallelism, vector instructions, ARM NEON

1 Introduction

Despite more than three decades of research efforts, public-key cryptography (PKC) is still considered computation-intensive, especially when executed on embedded processors. This is mainly because the underlying arithmetic operations (e.g. exponentiation, scalar multiplication) are performed on operands of a size of several hundreds or even thousands of bits. Multi-precision modular arithmetic is a performance-critical building block of both traditional public-key algorithms (e.g. RSA) and elliptic curve cryptosystems. This is in particular the case for the modular multiplication, which demands careful optimization to achieve acceptable performance, especially on resource-constrained embedded processors. In order to reduce the execution time of modular multiplication, cryptographers have developed several efficient reduction algorithms, while software engineers made efforts to implement them in an optimal way. One of the most important modular reduction techniques is Montgomery's algorithm, which was originally introduced in 1985 [15] and has been widely deployed in real-world applications. Some other examples for reduction algorithms are the methods of Barrett [2] and Quisquater [18, 19].

In recent years, an increasing number of embedded microprocessors started to provide Single Instruction Multiple Data (SIMD) instructions to better support multimedia workloads. In order to exploit the parallel computing power of SIMD instructions, traditional algorithms need to be redesigned and software needs to be rewritten into a vectorized form. There exist a few papers related to the implementation of cryptographic algorithms; for example, the authors of [3, 8, 21, 7] propose ways to speed up cryptography using the NEON instruction set extensions, which is a relatively new SIMD (i.e. vector) architecture for mobile devices developed by ARM. In particular, to achieve fast public-key cryptography, it is important to develop optimized SIMD implementations of multi-precision modular multiplication. In [4], an efficient 128-by-128-bit integer multiplication using Freescale's SIMD extension is introduced. Various implementations, including [10], adopt a reduced-radix representation with 29 bits per word for a better handling of the carry propagation. In [5], vector instructions on the Cell microprocessor are used to perform multiplication on operands represented with a radix of 2^{16} . More recently, Gueron et al [9] described an implementation for the new AVX2 SIMD platform (Intel Haswell architecture) that uses 256-bit wide vector instructions and a reduced-radix representation for faster accumulation of partial products. At HPEC 2013, a novel modular reduction method was introduced for the NIST primes P192 and P224 [16], which is also based on a reduced-radix representation for the operands.

However, a reduced-radix representation (sometimes also called redundant representation) requires to compute more partial products and, thus, execute more multiply instructions compared to a canonical (i.e. non-redundant) representation. For example, if we use a radix-2²⁴ representation (i.e. 24 bits per word) for 192-bit operands, the total number of partial products is $8 \times 8 = 64$. On the other hand, a conventional non-redundant representation based on a radix of 2^{32} reduces the number of partial products to only $6 \times 6 = 36$. At SAC 2013. Bos et al introduced a 2-way Montgomery multiplication for SIMD processors including ARM NEON [6]. Their implementation computes the multiplication and reduction operation simultaneously using a non-redundant representation, which allowed them to exploit the SIMD-level parallelism provided by the NEON engine. However, the performance of their implementation suffers from Read-After-Write (RAW) dependencies in the instruction flow. Such dependencies cause pipeline stalls since the instruction to be executed has to wait until the operands from the source registers are available to be read. For example, the VMULL instruction takes two clock cycles to issue the operation, but the result is only available after (at least) seven clock cycles, which means VMULL has a fairly long latency¹. If a data conflict occurs, the pipeline is halted for seven clock cycles rather than just two clock cycles.

In this paper, we describe optimizations to further push the performance of multi-precision multiplication and Montgomery multiplication on ARM-NEON processors. We present a non-redundant Cascade Operand Scanning (COS) method for multiplication, which achieves record-setting execution times on ARM Cortex-A9 and Cortex-A15 processors. The COS method processes the partial products in a non-conventional order to reduce the number of data-dependencies in the carry propagation from less to more significant words, which also reduces the number of pipeline stalls. The same strategy can be applied for a two-way NEON-optimized Montgomery multiplication method, called Coarsely Integrated Cascade Operand Scanning (CICOS) method, which essentially consists of two COS computations, whereby one contributes to the multiplication and the second to the Montgomery reduction. Our experimental results show that a Cortex-A15 processor is able to execute a CICOS Montgomery multiplication with 1024-bit operands in only 5600 clock cycles, which is almost 40% faster than the NEON implementation of Bos et al (8527 cycles according to [6, Table 3]²).

 $^{^{1}}$ A brief description of some important NEON instructions along with the instruction timings can be found in Appendix D.

² Note that the timings in the proceedings version of Bos et al's paper differ from the version in the IACR eprint archive at https://eprint.iacr.org/2013/519. We used the faster timings from the eprint version for comparison with our work.

Furthermore, we introduce a novel Double Operand Scanning (DOS) method. The method doubles the operand rather than intermediate results. This way of computation can handle carry propagation properly, because except doubling the operand, whole procedures are same with COS multiplication, which handles carry propagation in order of partial products. The 512-bit squaring implementation on A9 and A15 takes 800 and 520 clock cycles respectively. In case of reduction process, we suggest a novel fast reduction method on SIMD architecture. Unlike SISD reduction process, our method carefully designed to compute the reduction in 2-way parallel fashion. The results show that our method ourperforms HPEC'13 results by 43% on APQ8060 snapdragon board.

The remainder of this paper is organized as follows. In Section 2, we recap the previous best known results for multiplication and Montgomery multiplication on 32-bit SIMD-based architectures. In Section 3, we present novel methods for multi-precision multiplication, squaring and Montgomery multiplication and fast reduction on SIMD-based processors, especially ARM-NEON. Thereafter, we will summarize our experimental results in Section 4. Finally, in Section 5, we conclude the paper.

2 Previous Work

Long integer arithmetic is not straightforward to implement on SIMD-based architectures, mainly due to the propagation of carries from one word to the next, which has to be carried out in addition, multiplication, and other operations. In order to deal with this problem, many recent SIMD implementations adopt a redundant representation with a reduced number of active bits per register with the goal of keeping the final result within remaining capacity of a register so that no carry propagations is needed. In [9], by exploiting the AVX2 instruction set extension with redundant representation, the authors showed a performance enhancement of 51% over the OpenSSL 1.0.1 implementation. In [16], Pabbuleti et al. implemented the NISTrecommended prime-field curve including P192 and P224 on the Snapdragon APQ8060 within 404, 405 clock cycles via applying multiplicand reduction method into SIMD-based machine. Recently, in SAC'13, a different approach to split the Montgomery multiplication into two parts, being computed in parallel, was introduced [6]. They flip the sign of the precomputed Montgomery constant and accumulate the result in two separate intermediate values that are computed concurrently while avoiding a redundant representation. This method is to compute the multiplication and reduction step simultaneously using 2-way SIMD instructions at the cost of some overhead and shows a performance increase of a factor of 1.5 or more than sequential implementation on the Atom platform for 2048-bit modulo. In this paper, we take a different approach computing the multiplication using 2-way SIMD instructions first and subsequently the reduction using 2-way SIMD. The approach uses non-redundant representation and computes the carry propagation using 2-way SIMD instructions

3 Proposed Method

Throughout the paper, we will use the following notations. Let A and B be two operands with a length of m-bit that are represented by multiple-word arrays. Each operand is written as follows: A = (A[n-1], ..., A[2], A[1], A[0]) and B = (B[n-1], ..., B[2], B[1], B[0]), whereby $n = \lceil m/w \rceil$, and w is the word size. The result of multiplication $C = A \cdot B$ is twice length of A, and represented by C = (C[2n-1], ..., C[2], C[1], C[0]). For clarity, we describe the method using a multiplication structure and rhombus form. The multiplication structure describes order of partial products from top to bottom and each point in rhombus form represents a multiplication $A[i] \times B[j]$. The rightmost corner of the rhombus represents the lowest indices (i, j = 0), whereas the leftmost represents corner the highest indices (i, j = n - 1). The lowermost side represents result indices C[k], which ranges from the rightmost corner (k = 0) to the leftmost corner (k = 2n - 1). Particularly, SIMD architecture computes two 32-bit

partial products with single instruction, so we use two multiplication structures to describe SIMD operations. These block structures placed in the row represent two partial products with single instruction.

3.1 Cascade Operand Scanning Multiplication for SIMD

SIMD architecture is able to compute multiple data with single instruction. However, SIMD instruction does not provide carry handling registers and therefore, results in imposing huge overheads on SIMD machine to manage carry propagation. In order to alleviate this problem, many of the previous work adopted the so-called redundant representation which absorbs carry propagation into remaining bits in the destination registers, but this architecture also has performance degradation because redundant representation increases number of partial products. To address both drawbacks, we choose non-redundant representation, suggesting efficient carry handling with simple operand realignments.

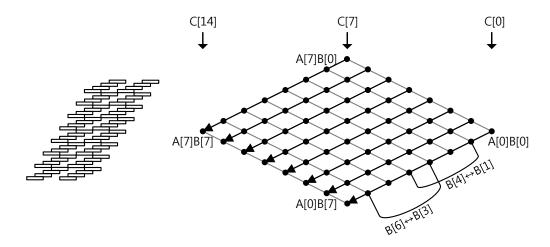


Fig. 1. Cascade operand scanning multiplication for 256-bit operand

In Figure 1, we designed multi-precision multiplication for SIMD architecture. Taking the 32-bit word with 256-bit multiplication as an example, our method works as follows³. Firstly, we re-organized operands by conducting transpose operation, which can efficiently shuffle inner vector by 32-bit wise. Instead of a normal order ((B[0], B[1]), (B[2], B[3]), (B[4], B[5]), (B[6], B[7])), we actually classify the operand as groups ((B[0], B[4]), (B[2], B[6]), (B[1], B[5]), (B[3], B[7])) for computing multiplication where each operand ranges from 0 to 0xfffff_ffff. Secondly, multiplication A[0] with ((B[0], B[4]), (B[2], B[6]), (B[1], B[5]), (B[3], B[7])) is computed, generating partial product pairs including ((C[0], C[4]), (C[2], C[6]), (C[1], C[5]), (C[3], C[7])) where the results are located from 0 to 0xfffff_fffe_0000_0001. Third, partial products are separated into higher bits $(64 \sim 33)$ and lower bits $(32 \sim 1)$ by using transpose operation with 64-bit initialized registers $(0x0000_0000_0000_0000_0000)$, which outputs 32-bit results ranging from 0 to 0xfffff_ffff. After then the higher bits are added to lower bits of upper intermediate results. For example, higher bits of ((C[0], C[4]), (C[1], C[5]), (C[2], C[6]), (C[3], C[7]), (C[4])). By referring Figure 2, our method establishes SIMD friendly addition process and the carry

³ Operands $A[0 \sim 7]$ and $B[0 \sim 7]$ are stored in 32-bit registers. Intermediate results $C[0 \sim 15]$ are stored in 64-bit registers. The 32-bit register means that we use half capacity of 64-bit register.

values are propagated in grouped cascade way. The proposed method finely rearranges each intermediate result and order of carry propagation to conduct carry handling with only four additions in non-redundant representation and no pipeline stalls. After addition, the least significant word $(C[0], lower bits of B[0] \times A[0])$ is placed within 32-bit in range of [0, 0xffff_ffff] so this is directly saved into temporal registers or memory. On the other hand, remaining intermediate results from C[1] to C[7] are placed within $[0, 0x1_ffff_fffe]^4$, which exceed range of 32-bit in certain cases. Fortunately, the addition of intermediate results $(C[1 \sim 7])$ and 32bit by 32-bit multiplication in next step are placed into 64-bit registers without overflowing, because addition of maximum multiplication result (0xffff_fffe_0000_0001) and intermediate result (0x1_ffff_fffe) outputs the final results within 64-bit (0xffff_ffff_ffff)⁵. For this reason, we don't need to propagate 33th carry bit of intermediate results ($C[1 \sim 7]$) in each round but we delay the carry propagation to very end of round and conduct whole carry propagations at once. Before move to next round, lower bits of C[4] and higher bits of C[7] are re-grouped into (C[4], C[8]) and then intermediate result pairs are re-shaped in ((C[1], C[5]), (C[2], C[6]), (C[3], C[7]), (C[4], C[8])). This process is repeated with remaining operands $(A[1 \sim 7])$ by seven times more to complete the multiplication⁶. After eight rounds of multiplication, the results from C[0] to C[7] are perfectly fit into 32-bit, because the least significant word is outputted in 32-bit in every round. However remaining intermediate results $(C[8] \sim C[15])$ are not placed within 32-bit so we should process a chain of carry propagation over 32-bit by conducting final alignment. The final alignment executes carry propagation results from C[8] to C[15] with sequential addition and transpose instructions. This process causes pipeline stalls by 8 times, because higher bits of former results are directly added to next intermediate results. Therefore, proposed COS incurs pipeline stalls by the number of nfor final alignment. In case of 512-, 1024- and 2048-bit COS multiplications, we should conduct 4, 16 and 64 times of 256-bit COS multiplications because the 256-bit COS multiplication is maximum operand size on NEON processor due to limited number of registers⁷. Unlike 256-bit version, intermediate result should be stored and re-loaded, so we assigned temporal memory and stack storages to retain intermediate results. Finally, several load, store, push and pop instructions are used to establish 512-, 1024- and 2048-bit implementations.

3.2 Coarsely Integrated Cascade Operand Scanning Multiplication for SIMD

In [6], Bos et al. introduced a 2-way Montgomery multiplication for SIMD architecture. However, the proposed 2-way Montgomery multiplication has high data interdependency because they used ordinary operand-scanning method for multiplication and reduction procedures which compute partial products in incremental order and previous partial product results are directly used in next step. Finally this resource access conflicts results in pipeline stalls. In order to resolve this problem, we rescheduled the order of operations to achieve a dependency-free design. We implemented the Coarsely Integrated Cascade Operand Scanning (CICOS) algorithm by using COS method since separated version needs to load and store intermediate results twice more than integrated version, while the finely integrated mode leads

⁴ In the first round, the range is within $[0, 0x1_ffff_fffd]$, because higher bits and lower bits of intermediate results ($C[0 \sim 7]$) are located in range of $[0, 0xffff_fffe]$ and $[0, 0xffff_ffff]$, respectively. From second round, the addition of higher and lower bits are located within $[0, 0x1_ffff_fffe]$, because both higher and lower bits are located in range of $[0, 0xffff_fffff]$.

⁵ In the first round, intermediate results ($C[0 \sim 7]$) are in range of [0, 0x1_ffff_fffd] so multiplication and accumulation results are in range of [0, 0xffff_fffffffff]. From second round, the intermediate results are located in [0, 0x1_ffff_fffe] so multiplication and accumulation results are in range of [0, 0xffff_fffffffffffff].

⁶ Pseudo and program code of cascade operand scanning is available in Appendix. A and C, the rhombus form in 512-bit is depicted in Figure 12.

⁷ NEON engine supports sixteen 128-bit registers. We assigned four registers for operands (A, B), four for intermediate results (C) and four for temporal storages.

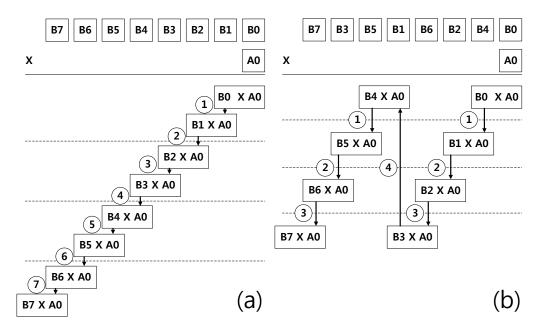


Fig. 2. Carry propagation in non-redundant representation, (a) ordinary operand scanning method, (b) proposed method(operand B is transposed before computations)

Algorithm 1 Calculation of the Montgomery reduction

Require: An odd *m*-bit modulus M, Montgomery radix $R = 2^m$, an operand T where $T = A \cdot B$ in the range [0, 2M - 1], and pre-computed constant $M' = -M^{-1} \mod R$

Ensure: Montgomery product $Z = \text{MonRed}(T, R) = T \cdot R^{-1} \mod M$

- 1: $Q \leftarrow T \cdot M' \mod R$
- 2: $Z \leftarrow (T + Q \cdot M)/R$
- 3: if $Z \geq M$ then $Z \leftarrow Z M$ end if
- 4: return Z

to high interdependency between each intermediate result. To describe Montgomery multiplication properly, we grouped two rhombus forms in Figure 3. Upper rhombus represents multi-precision multiplication and under rhombus represents Montgomery reduction. In order to distinguish both computations, we painted structure forms of multiplication process in white and reduction process in yellow.

In the following, we will take a 256-bit case (M' has a length of 32-bit) as an examples to further explain our implementation of Montgomery reduction for Algorithm 1. As shown in Figure 3, part of the operand Q (i.e. $Q[0 \sim 7]$) are computed by multiplying intermediate results with M'. All partial products with Q[0] and $M[0 \sim 7]$ are executed throughout the row. After then $T+Q\cdot M$ (i.e. Step 2 of Algorithm 1) can be computed using COS method. For Montgomery multiplication on SIMD architecture, we integrated two COS methods⁸. Firstly, we re-organized multiplicands (B) by conducting transpose operation. This process takes ((B[0], B[1]), (B[2], B[3]), (B[4], B[5]), (B[6], B[7])) as inputs and generates the new operand groups as ((B[0], B[4]), (B[2], B[6]), (B[1], B[5]), (B[3], B[7])) where each operand ranges from 0 to 0xfffff_fff. Secondly, multiply A[0] with ((B[0], B[4]), (B[2], B[6]), (B[1], B[5]), (B[3], B[7])), and generating partial product pairs including ((C[0], C[4]), (C[2], C[6]), (C[1], C[5]), (C[3], C[7])) where multiplication results range from 0 to 0xffff_fffe_0000_0001. Third, partial products are separated into higher bits and lower bits

⁸ Operands $A[0 \sim 7]$, $B[0 \sim 7]$, $M[0 \sim 7]$, $Q[0 \sim 7]$ and M' are stored in 32-bit registers. Intermediate results $C[0 \sim 15]$ are stored in 64-bit registers.

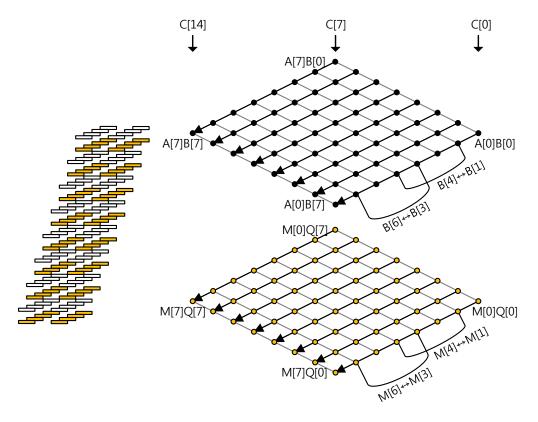


Fig. 3. Coarsely integrated cascade operand scanning for Montgomery multiplication in 256-bit

by using transpose operation with initialized registers, which outputs results in range of [0, Oxffff_ffff]. After then the higher bits are added to lower bits of upper intermediate results. For example, higher bits of (C[0], C[4]), (C[2], C[6]), (C[1], C[5]), (C[3]) are added to lower bits of (C[1], C[5]), (C[3], C[7]), (C[2], C[6]), (C[4]). These intermediate results are placed between 0 and $0x1_ffff_fffe^9$. The remaining higher bits of (C[7]) are copied to (C[8]). After then lower bits of C[0] in range of [0, 0xffff_ffff] is multiplied with operand M' to generate Q[0] in range of $[0, 0xffff_ffff]$. In case of operand (M), the variables are stored in ordered way like this ((M[0], M[4]), (M[2], M[6]), (M[1], M[5]), (M[3], M[7])). Fourthly, multiply Q[0]with ((M[0], M[4]), (M[2], M[6]), (M[1], M[5]), (M[3], M[7])) are executed with intermediate results pairs including ((C[0], C[4]), (C[2], C[6]), (C[1], C[5]), (C[3], C[7])) where the results range from 0 to Oxffff_ffff_ffff. Lastly, partial products are separated into higher bits and lower bits. After then the higher bits of ((C[0], C[4]), (C[1], C[5]), (C[2], C[6]), and the lower bits of (C[4]) are added to lower bits of ((C[1], C[5]), (C[2], C[6]), (C[3], C[7]),and the higher bits of (C[3]), which output the accumulated results ranging from 0 to $0x1_ffff_fffe$. Particularly, higher bits of (C[7]) are added to lower bits of (C[8]) and then higher bits of (C[3]) and lower bits of (C[8]) are re-grouped in (C[4], C[8]). Finally, the result C[0] is set to zero which is discarded and intermediate results are re-constructed like this ((C[1], C[5]), (C[2], C[6]), (C[3], C[7]), (C[4], C[8])). The multiplication and reduction process is repeated with remaining operands $(A[1 \sim 7])$ by seven times more to complete the Mont-

⁹ In the first round, the range is within $[0, 0x1_ffff_fffd]$, because higher bits and lower bits of intermediate results $(C[0 \sim 7])$ are located in range of $[0, 0xffff_fffe]$ and $[0, 0xffff_ffff]$, respectively. From second round, the addition of higher and lower bits are located within $[0, 0x1_ffff_fffe]$, because both higher and lower bits are located in range of $[0, 0xffff_fffff]$.

gomery multiplication¹⁰. After then, final alignment and final subtraction follow¹¹. As like COS, maximum operand size of CICOS method is 256-bit. For longer integers such as 512-, 1024- and 2048-bit, we should conduct 4, 16 and 64 times of 256-bit CICOS Montgomery multiplication.

3.3 Final Subtraction without Conditional Statements

As shown in Algorithm 1, the calculation of the Montgomery multiplication may require a final subtraction of the modulus M to get a fully reduced result in range of [0, M). In order to get the results, the final subtraction is conducted. The operation is computable with conditional branch by checking the carry bit. However, this method has two drawbacks. First two operands should be compared byte by byte via the compare function and the attacker can catch the leakage information because conditional statements consumes different clock cycles [23]. To resolve this, in [13], author suggested without conditional branch method for Montgomery multiplication. We adopted this method into SIMD architecture. Based on the concept of incomplete modular arithmetic, we don't compare exact value between Z and M, but we use most significant bit (z_m) of Z. If z_m is set, modulus remains. If not, modulus M is set to zero. In order to compute modulus, we perform SIMD multiplication M by z_n , because the M is saved in 32-bit packed way (M[0], M[4]) like this (0x?????.????.????.????). This representation is not favorable to conduct 32-bit wise subtraction because it cannot handle borrow bit in non-redundant subtraction. By conducting multiplication by z_n , each modulus can get masked 64-bit operand with higher 32-bit in zero padding and lower 32-bit in values ((0, M[0]), (0, M[4])) like this $(0 \times 0000 - 0000 - ???? - ????)$. After then we perform the subtraction on intermediate results (Z) by modulus (M). Final result may not be the at least non-negative residue but this is always in the range $[0, 2^m)$. This incomplete reduction does not introduce any problems in practice because incomplete representation can still be used as operand in a subsequent Montgomery multiplication [24]. In the following subsections, we further explore the algorithm in instruction level to analyze the impacts of pipelined operation order.

3.4 Requirements for Pipeline

By referring to Appendix. D. Table 6, NEON engine consists of four timing models. Among them, we should carefully concern on both issue and result clock cycles. Issue clock is minimum clock cycles to trigger instruction. After the issue clock cycle, we can issue next instructions in pipelined way. In a contrast, result clock is time to get final results. If we access to result registers before result clock, we should wait until the data is ready during remaining result clock cycles. For fully pipelined computation, we should issue several instructions and then access to certain results which satisfy the result clock cycle. We analyze proposed method by referring Figure 4.

Part 1 This part starts with operand alignments. Operands B are transposed by conducting the single VTRN instruction 12 on eight 32-bit packed variables. The operation needs 1 clock cycle to issue the instruction and 2 clock cycles take to get final results.

Part 2 The second part consists of multiplication and carry propagation. The VMULL operation needs 2 clock cycles to issue the instruction and 7 clock cycle for result. By conducting four

 $^{^{10}}$ The rhombus form in 512-bit is described in Figure 13

 $^{^{11}}$ Pseudo and program code of coarsely integrated cascade operand scanning is available in Appendix. B and C

 $^{^{12}}$ Detailed descriptions of instructions are found in Table 6

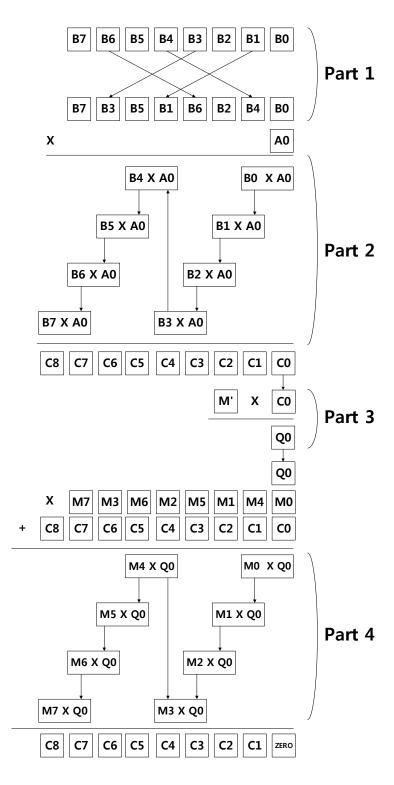


Fig. 4. Structure of coarsely integrated cascade operand scanning Montgomery multiplication

VMULL instructions for $((B[0] \times A[0]), (B[4] \times A[0]), ((B[1] \times A[0]), (B[5] \times A[0])), ((B[2] \times A[0]), (B[6] \times A[0]))$ and $((B[3] \times A[0]), (B[7] \times A[0]))$, we can get final results within 13 clock cycles, because 8 (2×4) clock cycles for triggering four VMULL instructions and 5 additional clock cycles to get last results are needed. Furthermore this computation order satisfies result clock cycle because we conduct four VMULL operations taking 8 clock cycles to issue so results of first instruction are ready. In a contrast, if there is interdependency between former destination and latter source registers, the costs rise to 28 clock cycles, because we should wait for four result clock cycles. After then four VTRN instructions for intermediate results on previous partial products are conducted. This takes 5 clock cycles for 4 and 1 clock cycles to trigger the instruction and wait for final results. Lastly, the five carry propagations are conducted with five VADD instructions. This takes 5 and 2 clock cycles to issue and get results, respectively. If this is not pipelined, the overheads are 15 clock cycles.

Part 3 In third part, the partial product of M' and C[0] is executed. The result Q[0] is directly used for Montgomery reduction in Part 4. If we conduct this instruction in Part 3, we should wait 7 clock cycles for results to be ready for next step. However, we can avoid the cost by conducting the product during Part 2. After obtaining C[0] and before transpose operation, we conduct VMULL on M' and C[0]. This only takes 2 clock cycles to issue so we can save 5 clock cycles.

Part 4 The last part calculates the reduction process. In contrast to Part 2, VMLAL instruction is exploited to compute multiplication and addition simultaneously. The four partial products of Q[0] and $M[0\sim7]$ are conducted, accumulating the intermediate results $C[0\sim7]$. The computation takes only 8 and 5 clock cycles to issue and get results. If we conduct the addition and multiplication computations separately in non-pipelined way, the overheads are up-to 40 clock cycles consisting of 12 and 28 for four VADD and four VMULL instructions, respectively. After then, four VTRN and five VADD instructions are conducted to manage carry propagation and accumulate intermediate results, which take 5 and 7 clock cycles, respectively.

Final Alignment and Subtraction/Addition The final results from C[8] to C[15] should be aligned to propagate carry bits from least significant word (C[8]) to most (C[15]). Firstly, higher bits of C[8] are added to C[9] and this is iterated to most significant intermediate result (C[15]). Unlike multiplication and reduction process, the final alignment occurs pipeline stalls due to interdependency between former and next variables. After final alignment, final subtraction follows and this also requires to conduct carry propagation incurring pipeline stalls. Finally, our Montgomery method includes the overheads of pipeline stalls by 2n times (n for final alignment and n for final subtraction). In case of pipelined instruction, we conduct n^2 times of carry propagations during multiplication and reduction process. On the other hand, Bos's 2-way Montgomery multiplication has $n^2 + n$ times of pipeline stalls because former multiplication results are directly used in next operations by following ordinary operand scanning. Particularly, Bos's method incurs pipeline stalls during multiplication and reduction process by n^2 times and final subtraction/addition is required to conduct sequential subtraction by n times. The comparison results are drawn in Table 1.

Table 1. Comparison of pipeline stall for Montgomery multiplication

Our CICOS MM		Bos's 2-way MM [6]	
Pipelined	Pipeline Stall	Pipelined	Pipeline Stall
n^2	2n	-	$n^2 + n$

3.5 Karatsuba Multiplication for SIMD

General complexity of m-bit integer multiplication is $\mathcal{O}(m^2)$. However, in 1960, Karatsuba presented a multiplication algorithm with asymptotic complexity $\mathcal{O}(m^{\log_2 3})$. This method was published in 1962 and thousands of multiplication implementations adopted Karatsuba method [11]. For this reason, we wonder that the impacts and feasibilities of Karatsuba method on SIMD machine. In this section, we use following notations to describe the method. Each operand has even m bits and k = m/2. The operands is divided into higher bits [k, m-1]and lower bits [0, k-1]. Let operands A and B $A_{LOW}+A_{HIGH}$ and $B_{LOW}+B_{HIGH}$. The Karatsuba multiplication can be established in two different ways. First approach is additive Karatsuba and this is executed as follows. The partial products (L, H and M) are computed with $A_{LOW} \cdot B_{LOW}$, $A_{HIGH} \cdot B_{HIGH}$ and $(A_{LOW} + A_{HIGH})(B_{LOW} + B_{HIGH})$. Finally, the results are obtained in $A \cdot B = L + 2^k(M - L - H) + 2^m H$. There are two drawbacks to use additive Karatsuba method on SIMD architecture. First problem is the addition of lower and higher operands $(A_{LOW} + A_{HIGH})$ and $(B_{LOW} + B_{HIGH})$ produce carry bits. This imposes more number of partial products. Second problem is that the subtraction on non-reduced form is not efficient enough, because non-reduced form does not have signed bit so sequential 2's complement should be conducted and this causes huge performance degradation. The alternative subtractive Karatsuba is given as follows. The partial products (L, H and M) are computed with $A_{LOW} \cdot B_{LOW}$, $A_{HIGH} \cdot B_{HIGH}$ and $|A_{LOW} - A_{HIGH}| \cdot |B_{LOW} - B_{HIGH}|$. If M is $(A_{LOW} - A_{HIGH}) \cdot (B_{LOW} - B_{HIGH})$, t is 0. Otherwise t is 1 where $M^* = (-1)^t M$. Finally, result is obtained in $A \cdot B = L + 2^k(L + H - M^*) + 2^mH$. Unlike additive Karatsuba method, the method can be established with only addition operations when t is 1. Otherwise one subtraction is included. In order to avoid subtraction, we present conditional subtractive Karatsuba multiplication in Algorithm 3. If partial product M is minus, we use subtractive Karatsuba multiplication, otherwise we conduct COS multiplication. By reodering partial product computations, we can efficiently accumulate intermediate results in product scanning way. As described in Figure 5 (b), addition of partial products (L and H) is used twice in first and last steps and this order of computation efficiently avoids carry propagation on accumulation process in every round. Compared with COS multiplication, Karatsuba multiplication has a major disadvantage that the chains of carry propagations for additions of M and (L+ H). This imposes 2n times of pipeline stalls so complete Karatsuba multiplication is not significant at all on SIMD machine.

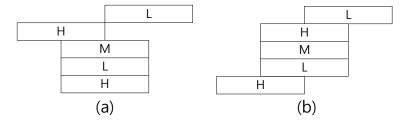


Fig. 5. Karatsuba multiplication order

3.6 Double Operand Scanning Squaring for SIMD

A typical software implementation of squaring method can be realized either using one of the above mentioned multiplication techniques or the specialized squaring method. Implementation using a specialized squaring method may have two advantages than simply using multiplication method for squaring. Firstly, only one operand (A) is used for squaring computation, thus, the number of operand load is reduced to half times of multiplication. Second,

Algorithm 2 Conditional subtractive Karatsuba multiplication

```
Require: An even m-bit operands A and B and partial products L = (A_{LOW} \cdot B_{LOW}), H = (A_{HIGH} \cdot B_{HIGH}) and M = (A_{LOW} - A_{HIGH}) \cdot (B_{LOW} - B_{HIGH})

Ensure: 2m-bit result C

1: if (A_{LOW} - A_{HIGH}) \cdot (B_{LOW} - B_{HIGH}) < 0

2: return L + 2^k(L + H + M) + 2^mH

3: else

4: return A \cdot B(COS multiplication)
```

there are duplicated partial products exist. For this reason, these parts are multiplied once and added twice to intermediate results. This computation generates same results, we expected. There are several optimal squaring methods developed. Efficient doubling method

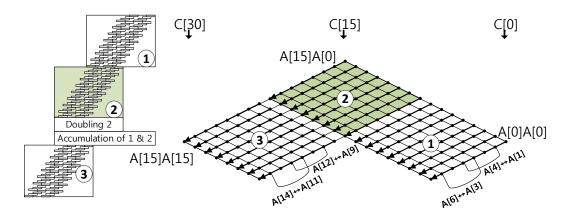


Fig. 6. Lazy Doubling squaring in SIMD

named lazy-doubling was introduced in [12]. The strong feature is doubling process which is delayed to end of inner structure and then computed. This method reduces number of arithmetic operations by conducting doubling computations on accumulated intermediate results. This technique significantly reduces the number of doubling process to one doubling process. In Figure 6, COS based LD method is drawn. Squaring method on SIMD is available from 512-bit word because 256-bit word size is optimal size for multiplication and 512-bit can be readily divided into duplicated and non-duplicated parts in 256-bit wise. The 512-bit LD squaring is established with three COS multiplications, i.e. single operand doubling and accumulation operations. Firstly part 1 and part 2 are computed in sequential way. While computing part 2, former intermediate results should be placed into memory or stack because the results of part 2 should not be mixed with part 1. After then part 2 is doubled and the results are added to results of part 1. This causes chains of carry propagations, because the least significant bit of part2 should be propagated. And then part 3 is conducted with COS.

In INDOCRYPT'13, a novel multi-precision squaring on embedded platforms, named "sliding block doubling" (SBD) was proposed. On one hand, SBD method computes doubling using "1-bit left shifting" operation at the end of duplicated partial product computation, which accumulates all partial product results with only consuming few arithmetic operations. On the other hand, contrary to previously known solutions, SBD method adopts product-scanning method to compute duplicated product parts. After then the intermediate results are doubled, and added into the final results [22]. In Figure 7, SBD on SIMD method is described. Unlike LD method, duplicated part 2 is firstly computed and then doubled. After then part 1 is computed and both intermediate results of parts 1 and 2 are added. While accumulation,

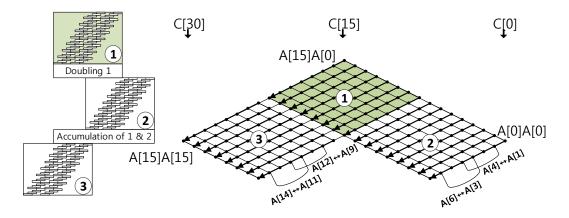


Fig. 7. Sliding Block Doubling squaring in SIMD

carry propagation is occurred and this imposes huge overheads on SIMD machine. Finally part 3 is computed with COS multiplication method. Due to inefficient accumulation process the method does not show high performance.

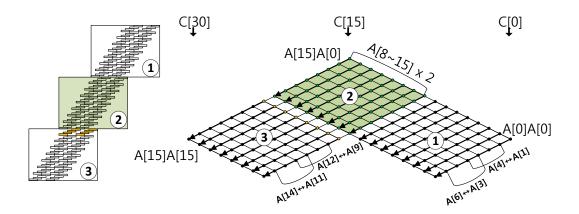


Fig. 8. Double Operand Scanning squaring in SIMD

In this paper, we suggest different approach to eliminate inefficient accumulation process. In Figure 8, a novel multi-precision squaring method named "Double Operand Scanning" (DOS) is introduced. Unlike previous methods, we don't need to conduct inefficient carry propagations occurred from intermediate results accumulations. To do this, we conduct doubling operation on operands rather than intermediate results. Before starting the DOS squaring, we doubled operand (A) ranging from m/2 to (m-1)-bit. And then we conduct parts 1, 2 and 3 in sequential way. When we move from part 2 to 3, we conduct carry-bit handling, because doubled operand has range of [m/2+1,m] bits. The highest(m-th) bit would be set in 50% possibility so it would trigger one addition of operands in range of [0,(m/2)-1]. For this reason, if the m-th bit is set, the operand A in range of [0,(m/2)-1] is added to intermediate results, which is described as yellow block and dot in Figure 8. Otherwise no additional addition is needed. The detailed algorithm is drawn in Algorithm 3.

Constant time computation for DOS squaring Doubled operand A should be added when the highest bit is set. However, this leaks such information of secret value. To ensure

Algorithm 3 Double Operand Scanning squaring

Require: An even m-bit operand A

Ensure: 2m-bit result C

- 1: $(A_{hbit}, A_{DBL}) = A_{\lceil \frac{m}{2}, m-1 \rceil} \times 2$
- 2: $C = A_{[0, \frac{m}{2} 1]} \times A_{[0, \frac{m}{2} 1]}$
- 3: $C = C + A_{[0, \frac{m}{2} 1]} \times A_{DBL}$
- 4: $C = C + (A_{hbit} \& A_{[0,\frac{m}{2}-1]}) + A_{[\frac{m}{2},m-1]} \times A_{[\frac{m}{2},m-1]}$
- 5: return C

secure against side channel attack, we conduct constant time computation whether the highest bit is set or not. In order to resolve this problem, we masked operands with highest bit. If bit is set, mask is set to <code>0xffff_ffff</code>. Otherwise the mask is set to <code>0</code>. Following is detailed process. After computing part 2 in Figure 8, we conduct <code>VMLAL</code> operation to multiply the operand and highest bit and then results are added to intermediate results. This can conduct masking operation because if the highest bit is zero, no values are added to results. Otherwise operands are added to intermediate results.

3.7 Fast Reduction for SIMD

Since specific prime fields have their own modulars, we can estimate that which words will be reduced or not. In this section, we selected several prime field for SIMD friendly curves which consist of only one arithmetic including addition or subtraction. If two operations are mixed, it is not readily computable because non-redundant form does not have signed bit and both arithmetic cannot be handled properly at once.

SGCM: $P = 2^{128} + 2^{13} + 2^{12} + 2^7 + 2^5 + 2 + 1$ The Galois/Counter Mode (GCM) is a NIST standardized authenticated encryption mode, which is used with block cipher such as AES [14]. However the method has been found to be susceptible to weak key / short cycle forgery attacks. SGCM is intended for those who want a concrete, largely technically compatible alternative to GCM [20]. Unlike GCM, SGCM uses GF(p) with $p = 2^{128} + 12451$, where $\frac{p-1}{2}$. This prime only consists of subtraction for reduction. This can be conducted over non-redundant representation. As described in Figure 9, outputs ranging from 129-bit to 256-bit $(C4 \sim C7)$ are multiplied by 12451 and then the results are subtracted to $(C0 \sim C3)$. Both multiplication and subtraction is readily established with single VMLSL instruction 13. After then carry propagation from 1-bit to 128-bit $(C0 \sim C3)$ is conducted. If there is carry/borrow bit (129-th) set, additional reduction is conducted again.

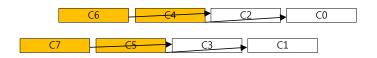


Fig. 9. Fast reduction for SGCM

secp192r1: $P = 2^{192} - 2^{64} - 1$ For elliptic curve cryptography, secp192r1 is selected, because it is divided into 32-bit word and only consists of two addition operations for reduction. Addition is much more favorable than subtraction when it comes to SIMD architecture because

¹³ Each block represents 64-bit register and from C0 to C7, these registers contain the bits from first bit to 256-th bit.

we don't need to care signed bit. The detailed process is described in Figure 10. Firstly, in (a), intermediate results from 321-bit to 384-bit are added to $193 \sim 256$ -bit. The reason to conduct $193 \sim 256$ -bit first is to accumulate same bit position and compute at once. And then the results from 193-bit to 384-bit are added to $1 \sim 192$ -bit in (b). Finally, in (c), results from 193-bit to 320-bit are added to $65 \sim 192$ -bit. If there is carry bit set, additional reduction is conducted more.

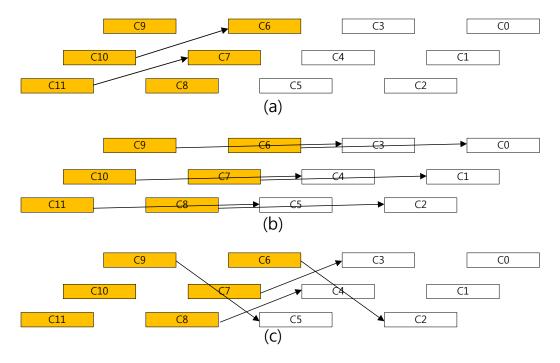


Fig. 10. Fast reduction for secp192r1

secp256k1: $P = 2^{256} - 2^{32} - 2^9 - 2^8 - 2^7 - 2^6 - 2^4 - 1$ For 256-bit case, we selected secp256k1 curve. The detailed process is described in Figure 11. In (a), we firstly conduct reduction on 2^{32} so intermediate results from 257-bit to 480-bit are added to $33 \sim 256$ -bit. The range of [481,512]-bit should be added to [257,288]-bit and this is again added to $33 \sim 64$ -bit, so we can reduce duplicated addition by directly adding the results to $33 \sim 64$ -bit. After then intermediate results from 257 to 512-bit are multiplied by 977 and then added to $1 \sim 256$ -bit in (b). As a similar approach of previous step, the range of [481,512]-bit is directly multiplied and added to bits from 1 to 32-bit. If there is carry bit set, additional reduction is conducted more.

4 Results

In this section, we report the execution time of the proposed methods on 32-bit ARM Cortex-A9 and Cortex-A15 processors and compare our results with the related works.

4.1 Target Platforms

Cortex Series The ARM Cortex series are full implementation of the ARMv7 architecture including NEON engine. Register size is expanded to double(D) and quadruple(Q) word

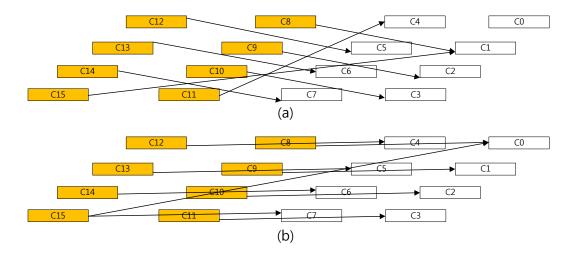


Fig. 11. Fast reduction for secp256k1

which represent 64-bit and 128-bit word registers. Each register provides short bit size computations such as 8-bit, 16-bit, 32-bit and 64-bit. This feature provides more precise operation and benefit to various word size computation. Particularly, Cortex-A9 provides the high efficiency, dual-issue super scalar, out-of-order execution and 8-11 stage pipeline. The processor is adopted in several devices including iPad 2, iPhone 4S, Galaxy S2, Galaxy S3, Galaxy Note 2, and Kindle Fire. The Cortex-A15 processor delivers twice the performance of Cortex-A9 processor. The processor provides multi-issue, out-of-order superscalar pipeline with a tightly coupled low latency. The pipeline is supported from 15 to 25 stages. The Cortex-A15 is used in Chromebook, NEXUS 10, Tegra 4 and Odroid-XU.

Snapdragon Series The Qualcomm Snapdragon S3 features a dual CPU architecture with two NEON processors. This architecture has combined design of Cortex-A8 and Cortex-A9 CPUs. The processor is adopted in several devices including LG Lucid, Samsung Galaxy Note, Xiaomi MI-One. In case of Snapdragon 800, L0, L1 and L2 cache and quad-core clocking at 2.4 GHz and ARMv7 instruction set are realized. This architecture is used in various gadgets including Samsung Galaxy S4 LTE-A, Samsung Galaxy Note 3.

4.2 Evaluation

We prototyped our methods for ARM Cortex-A9 and A15 processors, which are equivalent to the target processors used in Bos et al's work [6]. We developed the source code using the NDK android library and measured the execution time using the system time function clock_gettime() and scaled to clock cycles based on the processor's frequency. We compared our results with best previous results from proceedings version of Bos et al's paper presented at SAC 2013 [6]. In Table 2, we categorize the timings with respect to the architecture that served as experimental platform. In the case of 1024-bit Montgomery multiplication, we achieve an execution time of 8358 clock cycles on the Cortex-A9 series, while Bos et al's SIMD implementation requires 17464 clock cycles. Furthermore, on a Cortex-A15, we compute a 1024-bit Montgomery multiplication within 5600 clock cycles rather than 8527 clock cycles as specified in [6, Table 3]. Thus, our 1024-bit implementation outperforms Bos et al's work by approximately 52% and 34% on a Cortex-A9 and Cortex-A15, respectively. The speed-ups are even more significant for 512-bit operands: 57% on the A9 and 40% on the A15. Our results do not agree with those of Bos et al, who found that using the NEON instructions yields consistently worse results than a baseline (i.e. non-vector) implementation using only

conventional ARM instructions. In fact, our results show that the NEON instructions improve the execution time by 29% and 18% (512 and 1024-bit on Cortex-A9) as well as 37% and 35% (512 and 1024-bit on Cortex-A15) over a sequential implementation that does not take advantage of the NEON engine. The case for 2048-bit also shows, our method improves performance by 48.7% and 21.5% on A9 and A15, respectively. Compared with baseline implementations, still our 2048-bit implementations have enhancements by 10.9% and 22.7%. The following is reason for the significant speed-up compared to Bos et al's NEON implementation. First, we process the operands in a special order so as to reduce pipeline stalls caused by RAW data dependencies. Second, we perform the carry propagation in an efficient fashion in the NEON engine by adding grouped intermediate results.

Our squaring method is implemented on 512-bit size. The method reduces the number of multiplication to three quarters of total overheads. The results show that DOS method outperforms COS method by 23.8% (A9) and 19.1% (A15), respectively¹⁴. This is due to efficient doubling and carry propagation processes than previous approaches. The interesting approach, Karatsuba multiplication, is not recommended for non-redundant representation. The Karatsuba multiplication replaces partial products into addition operations. The addition operation is readily executed with redundant representation. However, non-redundant representation should conduct chain of carry propagations while addition operation. As a result of this, the subtractive Karatsuba multiplication improves performance by 2.2% (A9) and 2.7% (A15) when $(A_{LOW} - A_{HIGH}) \cdot (B_{LOW} - B_{HIGH}) < 0$. However, the other case is slower than common COS method due to overhead of signed bit computations. Both case happen in 50% possibility so average clock cycles are 1085 (A9) and 661 (A15) and these are 3.2% and 2.7% slower than COS method, respectively.

Table 2. Results of multiplication, squaring and Montgomery multiplication in clock cycle on ARM Cortex-A9/A15 platforms

Bit	Cortex-A9			Cortex-A15			
	Our NEON	NEON [6]	ARM [6]	Our NEON	NEON [6]	ARM [6]	
Cascade Operand Scanning Multiplication							
256	308	n/a	n/a	219	n/a	n/a	
512	1050	n/a	n/a	643	n/a	n/a	
1024	4298	n/a	n/a	2810	n/a	n/a	
2048	17080	n/a	n/a	10672	n/a	n/a	
Karat	Karatsuba Multiplication $(t = 1/t = 0)$						
512	1026/1144	n/a	n/a	625/697	n/a	n/a	
Doubl	Double Operand Scanning Squaring						
512	800	n/a	n/a	520	n/a	n/a	
Montgomery Multiplication							
256	658	n/a	n/a	308	n/a	n/a	
512	2254	5236	3175	1485	2473	2373	
1024	8358	17464	10167	5600	8527	8681	
2048	32732	63900	36746	26232	33441	33961	

The comparison results of prime field multiplication with fast reduction on APQ8060 target device is drawn in Table 3. Previous method directly reduced multiplicand while we compute multiplication with 2-way COS multiplication and then conduct fast reduction in SIMD architecture. As a results of this, we achieved 43% performance enhancements in secp192r1 case than HPEC2013. The performance enhancement is directly impacted on scalar multiplication

¹⁴ To the best of our knowledge, there is no non-redundant representation squaring method released so we compared with our multiplication method.

operations and we achieved performance gain by 16.5%. The case for SGCM128 is par with secp192r1 because fast reduction on SGCM128 consists of subtraction operation which causes huge performance degradation. Under non-redundant representation, subtraction needs series of 2's complement and bit-wise exclusive-or instructions.

Table 3. Results of prime fie	eld multiplication on	Qualcomm Snapdragon
--------------------------------------	-----------------------	---------------------

Bit	APQ8060				
	Our NEON NEON [17] ARM [17]				
Prime Field Multiplication with Fast Reduction					
SGCM128	228	n/a	n/a		
secp192r1	228	404	574		
secp256k1	402	n/a	n/a		
Scalar Multiplication					
secp192r1	1,035K	1,243K	1,591K		

5 Conclusions

We presented optimization techniques to improve the performance of multi-precision arithmetic operations (in particular multiplication, squaring, Montgomery multiplication and fast reduction) on 2-way SIMD platforms. More specifically, we introduced the COS method for multi-precision multiplication, which processes the words of one of the operands in a nonconventional order so as to reduce pipeline stalls caused by data dependencies. Furthermore, we described the CICOS method for performing Montgomery multiplication by coarsely interleaving COS-based multiplication and Montgomery reduction steps. Thanks to these optimizations, we were able to achieve record-setting execution times for conventional multiplication as well as Montgomery multiplication on ARM NEON platforms. For example, on an ARM Cortex-A15 processor, our CICOS method performs a 1024-bit Montgomery multiplication only 5600 clock cycles, which is roughly 34% faster than the NEON implementation of Bos et al (8527 cycles). On a Cortex-A9, the performance gain is even higher, namely 52% (8358 vs. 17464 cycles, i.e. we save 9106 cycles). In case of 2048-bit operands, our methods have improved performance by 21.5% (A15) and 48.7% (A9). These gaps further increase for 512-bit operands to 40% (A15) and 57% (A9). We also outperform Bos et al's non-vectorized implementation (which uses only standard ARM instructions) on the two platforms by between 11% and 37%. Based on these results, we can draw the following two conclusions. First, our re-ordering of operands along with a "coarse" integration of multiplication and reduction is significantly faster than the conventional operand ordering and "fine" integration approach followed by Bos et al. Second, and even more important, using the NEON instructions for Montgomery multiplication yields better performance than a sequential (i.e. non-vectorized) implementation, which contradicts the findings of Bos et al from SAC 2013. Furthermore, we introduce novel squaring method named double operand scanning, which doubles operands rather than intermediate results to enhance performance. The 512-bit results on A9 and A15 show that 23.8% and 19.1% enhancements respectively. Finally, we introduce fast reduction on SIMD machine, which conducts fast reduction in 2-way fashion. This improves multiplication and scalar multiplication by 43% and 16.5% on APQ8060 than HPEC 2013.

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Appendix. A. Pseudo Code: COS Multiplication in 256-bit¹⁵

```
Input: word size 32-bit, Operands A, B \in [0, 256).
Output: C = A \cdot B.
//Register R, Memory M, Temporal registers R_{AC}, R_{LOW}, R_{HIGH} \in [0, 8]^{16}.
R_A[7,...,0] \leftarrow M_A[7,...,0].
R_B[7,...,0] \leftarrow M_B[7,...,0].
R_{AC}[8,...,0] \leftarrow 0.
transpose R_B.
for i = 0 to 7
     (R_{AC}[4], R_{AC}[0]) \leftarrow (R_{AC}[4], R_{AC}[0]) + (R_A[i], R_A[i]) \times (R_B[4], R_B[0]).
     (R_{AC}[5], R_{AC}[1]) \leftarrow (R_{AC}[5], R_{AC}[1]) + (R_A[i], R_A[i]) \times (R_B[5], R_B[1]).
     (R_{AC}[6], R_{AC}[2]) \leftarrow (R_{AC}[6], R_{AC}[2]) + (R_A[i], R_A[i]) \times (R_B[6], R_B[2]).
     (R_{AC}[7], R_{AC}[3]) \leftarrow (R_{AC}[7], R_{AC}[3]) + (R_A[i], R_A[i]) \times (R_B[7], R_B[3]).
     R_{LOW}[8,...,0], R_{HIGH}[8,...,0] \leftarrow 0.
     transpose R_{AC} into R_{HIGH} and R_{LOW}.
     (R_{LOW}[4]) \leftarrow (R_{LOW}[4]) + (R_{HIGH}[3]).
     (R_{LOW}[5], R_{LOW}[1]) \leftarrow (R_{LOW}[5], R_{LOW}[1]) + (R_{HIGH}[4], R_{HIGH}[0]).
     (R_{LOW}[6], R_{LOW}[2]) \leftarrow (R_{LOW}[6], R_{LOW}[2]) + (R_{HIGH}[5], R_{HIGH}[1]).
     (R_{LOW}[7], R_{LOW}[3]) \leftarrow (R_{LOW}[7], R_{LOW}[3]) + (R_{HIGH}[6], R_{HIGH}[2]).
     (R_{LOW}[8]) \leftarrow (R_{HIGH}[7]).
     M_C[i] \leftarrow R_{LOW}[0].
     for j = 0 to 3
         (R_{AC}[j+4], R_{AC}[j]) \leftarrow (R_{LOW}[(j+5)], R_{LOW}[(j+1)]).
     end for
end for
final alignment R_{AC}.
M_C[15, ..., 8] \leftarrow R_{AC}.
Return M_C.
```

 $^{^{15}}$ 512-, 1024- and 2048-bit cases conduct 256-bit COS multiplication by 4, 16 and 64 times, respectively. The rhombus form of 512-bit COS is available in Figure 12.

 $^{^{16}}$ 32-bit register: $R_A[0\sim7],~R[0\sim7],~64$ -bit register: $R_{AC}[0\sim8],~R_{LOW}[0\sim8],~R_{HIGH}[0\sim8]$

Appendix. B. Pseudo Code: CICOS Montgomery Multiplication in 256-bit¹⁷

```
Input: word size 32-bit, Operands A, B, Modulus M^* \in [0, 256),
Inverse of Modulus M' \in [0, 32).
Output: C = MonRed(A \cdot B, R).
 //Register R, Memory M, Temporal registers R_{AC}, R_{LOW}, R_{HIGH} \in [0, 8]^{18}.
 R_A[7,...,0] \leftarrow M_A[7,...,0].
 R_B[7,...,0] \leftarrow M_B[7,...,0].
 R_{M'} \leftarrow M_{M'}.
 R_{M^*}[7,...,0] \leftarrow M_{M^*}[7,...,0].
 R_{AC}[8,...,0] \leftarrow 0.
 transpose R_B.
 for i = 0 to 7
     (R_{AC}[4], R_{AC}[0]) \leftarrow (R_{AC}[4], R_{AC}[0]) + (R_A[i], R_A[i]) \times (R_B[4], R_B[0]).
     (R_{AC}[5], R_{AC}[1]) \leftarrow (R_{AC}[5], R_{AC}[1]) + (R_A[i], R_A[i]) \times (R_B[5], R_B[1]).
     (R_{AC}[6], R_{AC}[2]) \leftarrow (R_{AC}[6], R_{AC}[2]) + (R_A[i], R_A[i]) \times (R_B[6], R_B[2]).
     (R_{AC}[7], R_{AC}[3]) \leftarrow (R_{AC}[7], R_{AC}[3]) + (R_A[i], R_A[i]) \times (R_B[7], R_B[3]).
     R_{LOW}[8,...,0], R_{HIGH}[8,...,0] \leftarrow 0.
     transpose R_{AC} into R_{HIGH} and R_{LOW}.
     (R_{LOW}[4]) \leftarrow (R_{LOW}[4]) + (R_{HIGH}[3]).
     (R_{LOW}[5], R_{LOW}[1]) \leftarrow (R_{LOW}[5], R_{LOW}[1]) + (R_{HIGH}[4], R_{HIGH}[0]).
     (R_{LOW}[6], R_{LOW}[2]) \leftarrow (R_{LOW}[6], R_{LOW}[2]) + (R_{HIGH}[5], R_{HIGH}[1]).
     (R_{LOW}[7], R_{LOW}[3]) \leftarrow (R_{LOW}[7], R_{LOW}[3]) + (R_{HIGH}[6], R_{HIGH}[2]).
     (R_{LOW}[8]) \leftarrow (R_{HIGH}[7]).
     R_O[i] \leftarrow R_{LOW}[0] \times R_{M'}.
     (R_{AC}[4], R_{AC}[0]) \leftarrow (R_{LOW}[4], R_{LOW}[0]) + (R_Q[i], R_Q[i]) \times (R_{M^*}[4], R_{M^*}[0]).
     (R_{AC}[5], R_{AC}[1]) \leftarrow (R_{LOW}[5], R_{LOW}[1]) + (R_Q[i], R_Q[i]) \times (R_{M^*}[5], R_{M^*}[1]).
     (R_{AC}[6], R_{AC}[2]) \leftarrow (R_{LOW}[6], R_{LOW}[2]) + (R_Q[i], R_Q[i]) \times (R_{M^*}[6], R_{M^*}[2]).
     (R_{AC}[7], R_{AC}[3]) \leftarrow (R_{LOW}[7], R_{LOW}[3]) + (R_Q[i], R_Q[i]) \times (R_{M^*}[7], R_{M^*}[3]).
     R_{LOW}[8,...,0], R_{HIGH}[8,...,0] \leftarrow 0.
     transpose R_{AC} into R_{HIGH} and R_{LOW}.
     (R_{HIGH}[3]) \leftarrow (R_{HIGH}[3]) + (R_{LOW}[4]).
     (R_{LOW}[4]) \leftarrow (R_{HIGH}[3]).
     (R_{LOW}[5], R_{LOW}[1]) \leftarrow (R_{LOW}[5], R_{LOW}[1]) + (R_{HIGH}[4], R_{HIGH}[0]).
     (R_{LOW}[6], R_{LOW}[2]) \leftarrow (R_{LOW}[6], R_{LOW}[2]) + (R_{HIGH}[5], R_{HIGH}[1]).
     (R_{LOW}[7], R_{LOW}[3]) \leftarrow (R_{LOW}[7], R_{LOW}[3]) + (R_{HIGH}[6], R_{HIGH}[2]).
     (R_{LOW}[8]) \leftarrow (R_{LOW}[8]) + (R_{HIGH}[7]).
     for j = 0 to 3
         (R_{AC}[j+4], R_{AC}[j]) \leftarrow (R_{LOW}[(j+5)], R_{LOW}[(j+1)]).
     end for
 end for
 final alignment R_{AC}.
 final subtraction R_{AC}.
 M_C \leftarrow R_{AC}.
 Return M_C.
```

¹⁷ 512-, 1024- and 2048-bit cases conduct 256-bit CICOS Montgomery multiplication by 4, 16 and 64 times, respectively. The rhombus form of 512-bit CICOS is available in Figure 13.

¹⁸ 32-bit register: $R_A[0 \sim 7]$, $R[0 \sim 7]$, $R_M[0 \sim 7]$, $R_Q[0 \sim 7]$, $R_{M'}$, 64-bit register: $R_{AC}[0 \sim 8]$, $R_{LOW}[0 \sim 8]$, $R_{HIGH}[0 \sim 8]$

Appendix. C. Program Code: COS Multiplication and CICOS Montgomery Multiplication in 256-bit

Table 4. Program code: COS multiplication in 256-bit

vtrn.32 q0, q1	vtrn.32 q10, q7	veor q0,q0,q0
viiii.52 q0, q1	vtrn.32 q8 , q9	veor qo,qo,qo
//step 1	viiii.92 qo , qo	vtrn.32 d20, d10
vmull.u32 q13, d0, d4[0]	vadd.i64 q11, q11, q5	vqadd.u64 d16, d16, d10
vmull.u32 q12, d2, d4[0]	vadd.i64 q10, q10, q6	vext.8 d2, d2, d20, #4
vmull.u32 q12, d2, d4[0] vmull.u32 q11, d1, d4[0]	vadd.i64 q8 , q8 , q7	vext.8 d2, d2, d20, #4
vmull.u32 q10, d3, d4[0]	vgadd.u64 d18, d18, d25	vtrn.32 d16, d11
viiiuii.u52 q10, u5, u4[0]	vext.8 d28, d28, d24, #4	vqadd.u64 d18, d18, d11
veor q5,q5,q5	vext.o u2o, u2o, u24, #4	vext.8 d2, d2, d16, #4
veor q6,q6,q6	//skip step3-7	vext.8 d2, d2, d10, #4
	//skip steps-1	utrn 22 d18 d12
veor q7,q7,q7	//step 8	vtrn.32 d18, d12
veor q8,q8,q8		vqadd.u64 d26, d26, d12
	vmlal.u32 q11, d0, d7[1]	vext.8 d3, d3, d18, $\#4$
vtrn.32 q13, q5	vmlal.u32 q10, d2, d7[1]	-4 20 406 412
vtrn.32 q12, q6	vmlal.u32 q8 , d1, d7[1]	vtrn.32 d26, d13
vtrn.32 q11, q7	vmlal.u32 q9, d3, d7[1]	vqadd.u64 d21, d21, d13
vtrn.32 q10, q8	F F F	vext.8 d3, d3, d26, $\#4$
11:04 10 10 5	veor q5,q5,q5	. 20 101 114
vadd.i64 q12, q12, q5	veor $q6,q6,q6$	vtrn.32 d21, d14
vadd.i64 q11, q11, q6	veor q7,q7,q7	vqadd.u64 d17, d17, d14
vadd.i64 q10, q10, q7	veor $q13,q13,q13$	vext.8 d4, d4, d21, #4
vqadd.u64 d16, d16, d27		
vext.8 d28, d28, d26, #4	vtrn.32 q11, q5	vtrn.32 d17, d15
	vtrn. $32 q10, q6$	vqadd.u64 d19, d19, d15
//step 2	vtrn.32 q8, q7	vext.8 d4, d4, d17, #4
vmlal.u32 q12, d0, d4[1]	vtrn. 32 q 9 , q 13	
vmlal.u32 q11, d2, d4[1]		vtrn.32 d19, d0
vmlal.u32 q10, d1, d4[1]	vadd.i64 q10, q10, q5	vqadd.u64 d27, d27, d0
vmlal.u32 q8 , d3, d4[1]	$vadd.i64\ q8\ ,\ q8\ ,\ q6$	vext.8 d5, d5, d19, #4
	vadd.i64 q9, q9, q7	vext.8 d5, d5, d27, $\#4$
veor $q5,q5,q5$	vqadd.u64 d26, d26, d23	
veor $q6,q6,q6$	vext.8 d31, d31, d22, $\#4$	//output q14, q15, q1, q2
veor $q7,q7,q7$		
veor $q9,q9,q9$	$//{ m final~alignment}$	
	veor $q5,q5,q5$	
vtrn. $32 q12, q5$	veor $q6,q6,q6$	
vtrn.32 q11, q6	veor q7,q7,q7	

Operand A registers : q0, q1, Operand B registers : q2, q3, Inverse of M register : d8 Operand M registers : q14, q15, r0 indicates pointer of value(0x00000001ffffffff)

vtrn. 32 q 0 , q 1	//final alignment	vadd.i64 d20, d17, d20
// , 1	veor q5,q5,q5	vtrn.32 d20, d10
//step 1	veor q6,q6,q6	
vmull.u32 q13, d0, d4[0]	veor $q7,q7,q7$	vqsub.u64 d18, d18, d10
vmull.u32 q12, d2, d4[0]	veor $q0,q0,q0$	veor d18, d18, d16
vmull.u32 q11, d1, d4[0]	20 400 410	vadd.i64 d18, d18, d17
vmull.u32 q10, d3, d4[0]	vtrn.32 d20, d10	vtrn.32 d18, d11
rroom at at at	vqadd.u64 d18, d18, d10	vegub v64 d26 d26 d11
veor q5,q5,q5	t 20 J10 J11	vqsub.u64 d26, d26, d11
veor q6,q6,q6	vtrn.32 d18, d11	veor d26, d26, d16
veor q7,q7,q7	vqadd.u64 d26, d26, d11	vadd.i64 d26, d26, d17
veor q8,q8,q8	return 22 d26 d12	vtrn.32 d26, d12
rmull u22 c4 d9 d26	vtrn.32 d26, d12	wasub u64 d94 d94 d19
vmull.u32 q4, d8, d26	vqadd.u64 d24, d24, d12	vqsub.u64 d24, d24, d12
rrtum 22 a12 a5	rtm 22 d24 d12	veor d24, d24, d16
vtrn.32 q13, q5	vtrn.32 d24, d13	vadd.i64 d24, d24, d17
vtrn.32 q12, q6 vtrn.32 q11, q7	vqadd.u64 d21, d21, d13	vtrn.32 d24, d13
	vtrn.32 d21, d14	wasub u64 d21 d21 d13
vtrn. $32 q10, q8$,	vqsub.u64 d21, d21, d13
redd;64 a19 a19 a5	vqadd.u64 d19, d19, d14	veor d21, d21, d16
vadd.i64 q12, q12, q5	vtrn.32 d19, d15	vadd.i64 d21, d21, d17
vadd.i64 q11, q11, q6 vadd.i64 q10, q10, q7	vqadd.u64 d27, d27, d15	vtrn.32 d21, d14
vqadd.u64 d27, d27, d16	vqadd.u04 d21, d21, d15	vqsub.u64 d19, d19, d14
vqadd.u04 d21, d21, d10	vtrn 32 d27 d0	
woor a5 a5 a5	vtrn.32 d27, d0 vqadd.u64 d25, d25, d0	veor d19, d19, d16 vadd.i64 d19, d19, d17
veor q5,q5,q5 veor q6,q6,q6	vqadd.u04 d25, d25, d0	vadd.104 d19, d19, d17 vtrn.32 d19, d15
veor q7,q7,q7	vtrn.32 d25, d1	viiii.92 d19, d19
veor q9,q9,q9	viiii.92 d29, d1	vqsub.u64 d27, d27, d15
veor q <i>5</i> ,q <i>5</i> ,q <i>5</i>	//final subtraction	veor d27, d27, d16
vmlal.u32 q13, d28, d8[0]	vmull.u32 q1, d28, d1[0]	vadd.i64 d27, d27, d17
vmlal.u32 q12, d29, d8[0]	vmull.u32 q2, d29, d1[0]	vtrn.32 d27, d0
vmlal.u32 q12, d29, d6[0] vmlal.u32 q11, d30, d8[0]	vmull.u32 q2, d29, d1[0] vmull.u32 q3, d30, d1[0]	viiii.92 d21, d0
vmlal.u32 q10, d31, d8[0]	vmull.u32 q4, d31, d1[0]	vqsub.u64 d25, d25, d0
viiiiai.u92 q10, u91, u0[0]	vinum.uo2 q1, uo1, u1[o]	veor d25, d25, d16
vtrn.32 q13, q5	vldmia r0!, {q8}	vadd.i64 d25, d25, d17
vtrn.32 q12, q6	vidilia 10., [qo]	vadd.101 d20, d20, d11
vtrn.32 q11, q7	veor q5,q5,q5	vtrn. 32 q 10 , q 9
vtrn.32 q10, q9	veor q6,q6,q6	vtrn.32 q13, q12
viii.02 q10, q0	veor q7,q7,q7	voim.02 q10, q12
vadd.i64 q12, q12, q5	veor q0,q0,q0	//output d20, d26, d21, d27
vadd.i64 q11, q11, q6	7001 q0,q0,q0	// output u20, u20, u21, u21
vadd.i64 q10, q10, q7	vsub.i64 q10, q10, q1	
vqadd.u64 d18, d18, d27	vsub.i64 q9, q9, q2	
vqadd.u64 d19, d19, d17	vsub.i64 q13, q13, q3	
. 4844.401 410, 410, 411	vsub.i64 q12, q12, q4	
//skip step2-8	.545.101 412, 412, 41	
, , ship step2 0	veor d20, d20, d16	
	.551 420, 420, 410	

Appendix. D. Instruction Set Summary for NEON

Cycles This is the number of issue cycles the particular instruction consumes, and is the absolute minimum number of cycles per instruction if no operand interlocks are present.

Source The Source field indicates the execution cycle where the source operands must be available if the instruction is to be permitted to issue. The comma separated list matches that of the Format field, indicating the register that each value applies to. Where two lines are provided for a single format containing quad (Q) registers, this indicates that the source registers are handled independently between top and bottom half double (D) registers. The first line provides the information for the lower half of the quad register and the second line provides the same information for the upper half of the quad register.

Result The Result field indicates the execution cycle when the result of the operation is ready. At this point, the result might be ready as source operands for consumption by other instructions using forwarding paths. However, some pairs of instructions might have to wait until the value is written back to the register file.

Writeback The Writeback field indicates the execution cycle that the result is committed to the register file. From this cycle, the result of the instruction is available to all other instructions as a source operand.

Table 6. Instruction set summary for NEON [1]

Mnemonics	Description	Cycles	Source	Result	Writeback
VADD	Vector Addition	1	-,2,2	3	6
VSUB	Vector Subtraction	1	-,2,1	3	6
VEOR	Vector Exclusive-or	1	-,2,2	3	6
VMULL	Vector Multiplication	2	-,2,1	7	7
VMLAL	Vector Multiplication with Addition	2	3,2,1	7	7
VTRN	Vector Traspose	1	-,1	2	6

Appendix E. 512-bit COS Multiplication and CICOS Montgomery Multiplication in Rhombus $Form^{19}$

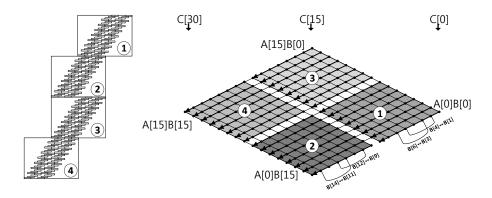
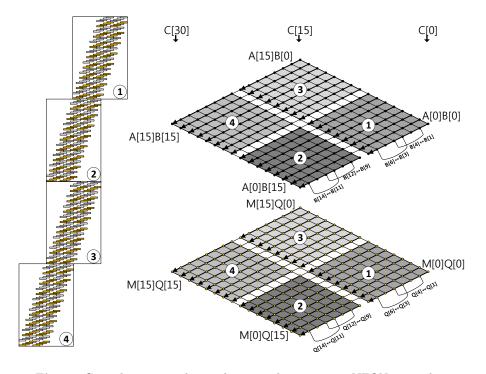


Fig. 12. Cascade operand scanning on NEON in 512-bit



 ${\bf Fig.\,13.}$ Coarsely integrated cascade operand scanning on NEON in 512-bit

¹⁹ 1024-and 2048-bit COS and CICOS methods can be established with 4 and 16 times of 512-bit COS and CICOS computations.