# Optimizing BIKE for the Intel Haswell and ARM Cortex-M4 

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#### Abstract

BIKE is a key encapsulation mechanism that entered the third round of the NIST post-quantum cryptography standardization process. This paper presents two constant-time implementations for BIKE, one tailored for the Intel Haswell and one tailored for the ARM Cortex-M4. Our Haswell implementation is much faster than the avx2 implementation written by the BIKE team: for bikel1, the level-1 parameter set, we achieve a 1.39 x speedup for decapsulation (which is the slowest operation) and a 1.33 x speedup for the sum of all operations. For bikel3, the level-3 parameter set, we achieve a 1.5 x speedup for decapsulation and a 1.46 x speedup for the sum of all operations. Our M4 implementation is more than two times faster than the non-constant-time implementation portable written by the BIKE team. The speedups are achieved by both algorithm-level and instruction-level optimizations.


Keywords: constant-time implementations • NIST PQC standardization • CortexM4

## 1 Introduction

BIKE $\left[\mathrm{ABB}^{+} 17\right]$ is a key encapsulation mechanism (KEM) involved in the NIST postquantum cryptography standardization process. It can be viewed as a variant of the code-based cryptosystem introduced in [MTSB13], which makes use of so-called "quasicyclic moderate-density parity-check" (QC-MDPC) codes. In July 2020, NIST announced that BIKE has been selected as one of 15 schemes that advanced to the third round of the process, making BIKE one of the candidates considered for standardization.

During the standardization process, BIKE holds a nice security record: unlike many other candidates, the claimed security levels of BIKE haven never been challenged. BIKE also features small public keys: the public key sizes are around $1.5 \mathrm{~KB}, 3 \mathrm{~KB}$, and 5 KB for the level-1, level-3, and level-5 parameter sets, respectively. These features make BIKE a competitive candidate in the process. In fact, NIST also commented that
"NIST views BIKE as one of the most promising code-based candidates."
in the status report for the second round candidates $\left[\mathrm{AASA}^{+} 20\right]$.
Despite the advantages, there are two factors that arguably make BIKE look weaker than some other candidates. The first factor is that, although the BIKE team has identified the conditions on the decoder that would make the scheme CCA-secure, they are not able to prove that the decoder actually satisfies the conditions. On the other hand, BIKE can still be used in applications which require only CPA security. The other factor is the relatively slow speeds of the operations. The speed issue is what this paper aims to deal with.

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### 1.1 Previous Works

There have been many previous papers on improving the speed of the cryptosystem introduced in [MTSB13] and its variants. In particular, the QcBits paper [Cho16] was the first one to present a fully constant-time software implementation. Later on, many papers tried to improve the speed of constant-time software implementations. For the purpose of this paper, instead of reviewing all related papers, we list some important optimization techniques adopted by the BIKE team. All the optimization techniques are related to $\mathcal{R}_{z}=\mathbb{Z}[y] /\left(y^{r}-1\right)$ or $\mathcal{R}=\mathbb{F}_{2}[x] /\left(x^{r}-1\right)$.

- As pointed out in [Cho16], there are many operations that can be viewed as multiplications in $\mathcal{R}_{z}$ in the decoding algorithm. These multiplications are of the form $g(y) f(y)$, where each coefficient $g_{i}$ and $f_{i}$ is either 0 or 1 , and $g(y)$ is a sparse polynomial. [Cho16] proposed to consider $g(y)$ as $\sum_{s} y^{-s}$ and compute the sum of all $y^{-s} f(y)$ to obtain $g(y) f(y)$. Each $y^{-s} f(y)$ is computed using a circular shift on the coefficients of $f$ by $s$ positions. In order to make sure that the circular shift does not leak information through timing, [Cho16] proposed to make use of the concept of a Barrel shifter to build a "Barrel rotator".
- This idea of using a "Barrel rotator" would be very efficient if $r$ was a multiple of the register size. However, as $r$ is always an odd prime in BIKE, performing the circular shift directly will introduce some overhead. In order to reduce the overhead, [DG19] proposed to represent $f$ in a "duplicated form" such that $y^{-s} f(y)$ can be obtained by performing a logical shift on the duplicated form. The logical shift is again implemented using the concept of a Barrel shifter.
- To compute the sum $\sum_{s}\left(y^{-s} f(y)\right)$, [Cho16] proposed to perform the computation in a bitsliced fashion to make use of the fact that each coefficient of $y^{-s} f(y)$ is either 0 or 1 .
- There are many multiplications in $\mathcal{R}$ in key generation, encapsulation, and decapsulation. In order to compute the product of two elements $f=\sum_{i} f_{i} x^{i}$ and $g=\sum_{j} g_{j} x^{j}$ in $\mathcal{R}$, the clmul implementation of [Cho16] represents $f$ and $g$ as $\sum_{i} F_{i} z^{i}$ and $\sum_{j} G_{j} z^{j}$, where $z=x^{64}$, and uses the pclmulqdq instruction to compute every $F_{i} G_{j}$. [DG19] proposed to use Karatsuba recursively, where the bottom level of recursions is handled by pclmulqdq.
- The public key is of the form $h_{1} \cdot h_{0}^{-1}$ where each $h_{i} \in \mathcal{R}$. Computation of $h_{0}^{-1} \in \mathcal{R}$ can be computed using a multiplication chain as shown in [Cho16]. Inside the multiplication chain, it is often required to compute the $2^{k}$ th power of an element $f \in \mathcal{R}$ for some constant $k$. [DGK20a] proposed to compute such an exponentiation by permuting the coefficients in $f$. When $k$ is large, this is much faster than carrying out $k$ squarings sequentially.

Unfortunately, even with so many optimizations, the latest software speed of BIKE is still not very satisfying: the eBACs website [Be] reports that, on titano, a machine with an Intel Haswell CPU, the level-1 parameter set bikel1 takes more than 138000 cycles and 2650000 cycles to perform encapsulation and decapsulation, respectively. For comparison, mceliece348864, one of the level-1 parameter sets of a third-round code-based candidate Classic McEliece $\left[\mathrm{ABC}^{+} 17\right]$, takes only 44652 and 132360 cycles, respectively, on the same machine. Note that this means that the situation of code-based candidates is quite different from the situation of lattice-based candidates, as structured lattice-based schemes are typically much faster than non-structured ones.

The BIKE team has some optimized implementations for x86 machines, whose speeds rely heavily on the usage of instructions for carryless multiplications such as pclmulqdq.

This raises the question whether BIKE can still run with reasonable speed on embedded systems where such instructions are not available. To our knowledge, there has not been any BIKE implementation tailored for embedded systems yet.

### 1.2 Our Contribution

This paper presents two implementations of BIKE, one optimized for the Intel Haswell and one optimized for the Cortex-M4. Compared to the avx2 implementation written by the BIKE team, for bikel1, we achieve a 1.39 x speedup for decapsulation and a $1.33 x$ speedup for the sum of all operations on Haswell. For bikel3, we achieve a 1.5 x speedup for decapsulation and a 1.46 x speedup for the sum of all operations on Haswell. Compared to the non-constant-time portable implementation written by the BIKE team, our implementation is more than two times faster for all the three operations on M4. Both of our implementations are constant-time: there is no memory access using secret indices, no branching with secret conditions, and no usage of variable-time instructions.

We believe that our implementations will make BIKE a more interesting option for standardization. Even though we did not implement the largest parameter set bikel5, which was introduced in the 3rd-round specification, we expect that our optimization techniques (see below) will be useful for optimizing bikel5 as well. We also expect the implementation techniques used for Haswell will be useful for optimizing BIKE for newer microarchitectures, e.g., those supporting AVX512. In addition, our optimization techniques for polynomial multiplications in $\mathcal{R}$ can be useful for optimizing $\mathrm{HQC}\left[\mathrm{AMAB}^{+} 17\right]$, another third-round code-based candidate.

### 1.3 Optimization Techniques

The speed of our implementations is achieved by using the following optimization techniques.
a) When performing a logical shift on the duplicated form of $f \in \mathcal{R}_{z}$, we make use of matrix transpositions to change the representation of $f$, so that a Barrel shifter is no longer required. As far as we can tell, this idea is completely original. This optimization is only used in our Haswell implementation.
b) We implement the Barrel shifter for shifting the duplicated form of $f \in \mathcal{R}_{z}$ using conditional execution and powerful multiply-and-accumulate instructions supported by the M4. We consider this as a purely instruction-level optimization.
c) The sum of all $y^{-s} f$ is computed using a sequence of half adders in the BIKE team's implementation. We follow an algorithm explained by Boyar and Peralta [BP05] to use full adders whenever possible. The algorithm is more complex and requires more

Table 1: Optimization techniques and where they are used: • means that the technique is used for the corresponding operation on the corresponding platform, and o means that the technique is not used.

|  | Haswell implementation |  | M4 implementation |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| technique | keygen | encap. | decap. | keygen | encap. | decap. | section |
| a) | $\circ$ | $\circ$ | $\bullet$ | $\circ$ | $\circ$ | $\circ$ | 3 |
| b) | $\circ$ | $\circ$ | $\circ$ | $\circ$ | $\circ$ | $\bullet$ | 4 |
| c) | $\circ$ | $\circ$ | $\bullet$ | $\circ$ | $\circ$ | $\bullet$ | 5 |
| d) | $\bullet$ | $\bullet$ | $\bullet$ | $\circ$ | $\circ$ | $\circ$ | 6 |
| e) | $\circ$ | $\circ$ | $\circ$ | $\bullet$ | $\bullet$ | $\bullet$ | 7 |

memory, but it reduces the number of logical operations required to compute the sum by a factor of more than 2 .
d) The BIKE team's implementations use the Karatsuba algorithm to optimize multiplications in $\mathcal{R}$. The Karatsuba algorithm is used recursively. Instead, we use a five-way recursive algorithm proposed by Bernstein in [Ber09a] for the top level of recursion. This optimization is only applied in our Haswell implementation.
e) We use a "Frobenius additive FFT" (FAFFT) to optimize multiplications in $\mathcal{R}$ on the M4. We found that FAFFTs fit nicely with bitslicing, such that we are able to carry out multiplications in $\mathcal{R}$ efficiently even without instructions for carryless multiplications.

To summarize, a), b), and c) are used for multiplications in $\mathcal{R}_{z}$, while d) and e) are used for multiplications in $\mathcal{R}$. Table 1 summarizes where each of these techniques is used in our implementations.

### 1.4 Availability of Source Code

Our implementations are adapted from the BIKE team's implementation and hence will be distributed under the same Apache 2.0 license of the orignal implementations. We plan to submit our Haswell implementation to the eBACS project [Be] so that the source code can be included in SUPERCOP. The source code of our M4 implementation has been included in the pqm4 project [KRSS] under the directories crypto_kem/bikel1/m4f and crypto_kem/bikel3/m4f. We also plan to submit both of our implementations as the artifact of our paper for CHES 2021.

### 1.5 Organization of the paper

Section 2 reviews the latest specification of BIKE and shows why we need to perform multiplications in $\mathcal{R}_{z}$ and multiplications in $\mathcal{R}$. Each of the remaining sections introduces one of our five optimization techniques; See Table 1. Section 8 shows our experimental results for multiplications in $\mathcal{R}_{z}$, multiplications in $\mathcal{R}$, and the three KEM operations.

## 2 BIKE

This section reviews the third-round specification of BIKE $\left[\mathrm{ABB}^{+} 17\right]$. For the sake of completeness, we repeat some contents from the specification in this section.

### 2.1 System Parameters.

BIKE uses system parameters $r, w, t$, and $\ell$. The following table shows the values of the system parameters and the security level for each of the three parameter sets of BIKE, which we denote as bikel1, bikel3, and bikel5.

|  | $r$ | $w$ | $t$ | $\ell$ | level |
| :---: | ---: | ---: | ---: | :---: | :---: |
| bikel1 | 12323 | 142 | 134 | 256 | 1 |
| bikel3 | 24659 | 206 | 199 | 256 | 3 |
| bikel5 | 40973 | 274 | 264 | 256 | 5 |


| KeyGen: ()$\mapsto\left(h_{0}, h_{1}, \sigma\right), h$ | Encaps: $h \mapsto K, c$ |
| :--- | :--- |
| Output: $\left(h_{0}, h_{1}, \sigma\right) \in \mathcal{H}_{w} \times \mathcal{M}, h \in \mathcal{R}$ | Input: $h \in \mathcal{R}$ |
| $1: h_{0}, h_{1} \stackrel{\$}{\leftarrow} \mathcal{H}_{w}$ | Output: $K \in \mathcal{K}, c \in \mathcal{R} \times \mathcal{M}$ |
| $2: h=h_{1} \cdot h_{0}^{-1}$ | $1: m \leftarrow \mathcal{M}$ |
| $3: \sigma \leftarrow \mathcal{M}$ | $2: e_{0}, e_{1} \leftarrow \mathbf{H}(m)$ |
|  | $3: c=\left(e_{0}+e_{1} \cdot h, m \oplus \mathbf{L}\left(e_{0}, e_{1}\right)\right)$ |
|  | $4: K \leftarrow \mathbf{K}(m, c)$ |
| Decaps: $\left(h_{0}, h_{1}, \sigma\right), c \mapsto K$ |  |
| Input: $\left(\left(h_{0}, h_{1}\right), \sigma\right) \in \mathcal{H}_{w} \times \mathcal{M}, c=\left(c_{0}, c_{1}\right) \in \mathcal{R} \times \mathcal{M}$ |  |
| Output: $K \in \mathcal{K}$ | $\triangleright$ with the convention $\perp=(0,0)$ |
| $1: e^{\prime} \leftarrow \operatorname{decoder}\left(c_{0} h_{0}, h_{0}, h_{1}\right)$ |  |
| 2: $m^{\prime}=c_{1} \oplus \mathbf{L}\left(e^{\prime}\right)$ |  |
| 3: if $e^{\prime}=\mathbf{H}\left(m^{\prime}\right)$ then $K \leftarrow \mathbf{K}\left(m^{\prime}, c\right)$, else $K \leftarrow \mathbf{K}(\sigma, c)$ |  |

Figure 1: BIKE's key generation, encapsulation, and decapsulation.

### 2.2 Hash Functions.

The BIKE team defines the hash functions $\mathbf{H}, \mathbf{K}, \mathbf{L}$ to have the following domains and ranges.

- $\mathbf{H}: \mathcal{M} \rightarrow \mathcal{E}_{t}$,
- $\mathbf{K}: \mathcal{M} \times \mathcal{R} \times \mathcal{M} \rightarrow \mathcal{K}$,
- $\mathbf{L}: \mathcal{R}^{2} \rightarrow \mathcal{M}$,
where the spaces $\mathcal{M}, \mathcal{K}$, and $\mathcal{E}_{t}$ are defined as follows.
- $\mathcal{M}=\{0,1\}^{\ell}$.
- $\mathcal{K}=\{0,1\}^{\ell}$.
- $\mathcal{E}_{t}=\left\{\left(e_{0}, e_{1}\right) \in \mathcal{R}^{2}| | e_{0}\left|+\left|e_{1}\right|=t\right\}\right.$, where $\left|e_{i}\right|$ means the number of nonzero coefficients in $e_{i}$.

The concrete instantiation of $\mathbf{H}, \mathbf{K}, \mathbf{L}$ can be found in the specification and is beyond the scope of this paper.

### 2.3 Key Generation, Encapsulation, and Decapsulation.

BIKE's key generation, encapsulation, and decapsulation algorithms are depicted in Figure 1. As shown in the figure, the key generation algorithm generates a secret key $\left(h_{0}, h_{1}, \sigma\right)$ and a public key $h .\left(h_{0}, h_{1}\right)$ is in $\mathcal{H}_{w}$, which is defined as

$$
\left\{\left(h_{0}, h_{1}\right) \in \mathcal{R}^{2}| | h_{0}\left|=\left|h_{1}\right|=w / 2\right\} .\right.
$$

The encapsulation algorithm, on input a public key $h$, generates a session key $K$ and a ciphertext $c$ encapsulating the session key. The decapsulation algorithm, on input a secret key $\left(h_{0}, h_{1}, \sigma\right)$ and a ciphertext $c$, outputs a session key $K$ or $\perp$. There are many multiplications in $\mathcal{R}$ as one can see from the KEM operations and the decoding algorithm decoder defined in the next subsection,

### 2.4 The Black-Gray Decoder

The decoding algorithm specified in the third-round specification is the black-gray decoder, which was introduced in [DGK20b]. The algorithm is shown in Algorithm 1 of the
specification. The BIKE team described it as an algorithm that can be used for any code with a low-weight parity-check matrix. However, as shown in Figure 1, BIKE is a scheme with a ring structure. To facilitate our discussion in the following sections, we rephrased the algorithm so that operations in rings are explicitly shown. We emphasize that "rephrased" means that we do not create a new algorithm but merely write the original algorithm in a different way. Our version of the black-gray decoder uses the following notations.

- Lift $(\cdot)$, which lifts an element from $\mathcal{R}$ to $\mathcal{R}_{z}$.
- $\operatorname{Tr}(\cdot)$, which converts an element $\sum_{i=0}^{r-1} f_{i} x^{i} \in \mathcal{R}$, into $\operatorname{Lift}\left(f_{0}+\sum_{i=1}^{r-1} f_{r-i} x^{i}\right)$.

The black-gray decoder is summarized in Algorithm 1. It should be easy to see where the operations in $\mathcal{R}$ are from with respect to the specification. Note that the multiplications of the form $\operatorname{Tr}\left(h_{i}\right) \cdot \operatorname{Lift}(s)$ are in $\mathcal{R}_{z}$, and $\left|\operatorname{Tr}\left(h_{i}\right)\right|$ is always $w / 2$. These multiplications are used to compute the numbers of unsatified parity-check equations. It has been explained in [Cho16] why computation of the numbers of unsatified parity-check equations can be viewed as multiplications in $\mathcal{R}_{z}$.

For completeness, we note that the black-gray decoder uses (in addition to $r, w, t$ ) two integers $\mathrm{NbIter}, \tau$ and a function threshold as its parameters. The third-round specification defines NbIter $=5$ for all parameter sets. How $\tau$ and threshold are defined is irrelevant to the purpose of this paper.

## 3 Circular Shifts with Matrix Transposition

As mentioned in the Introduction, in order to compute $g \cdot f$ where $g, f \in \mathcal{R}_{z}, g_{i}, f_{i} \in\{0,1\}$ for all $i$, we may write $g$ as $\sum_{s} y^{-s}$ with $0 \leq s<r$ and compute $g f$ as $\sum_{s}\left(y^{-s} f\right)$. For each $s, y^{-s} f$ can be computed by carrying out a circular shift on $\left(f_{0}, \ldots, f_{r-1}\right)$ by $s$ positions due to the structure of $\mathcal{R}_{z}$. Below we review how such a circular shift can be implemented using a Barrel shifter and propose a way to avoid the Barrel shifter.

### 3.1 The Duplicated Form and the Barrel Shifter

In order to compute $y^{-s} f$, the BIKE team follows [Cho16] to use a Barrel shifter and follows [DG19] to use a duplicated form for $f$. The idea is to

1. lift $f$ to $\mathbb{Z}[y]$ and represent it as $f^{\prime}=f+\left(f \bmod y^{r-1}\right) y^{r 1}$ and
2. compute $y^{-s} f$ as $\left(\left(f^{\prime}-\left(f^{\prime} \bmod y^{s}\right)\right) / y^{s}\right) \bmod y^{r}$.

The second step might look complicated, but computationally it simply means performing a logical shift (toward the direction of $f_{0}^{\prime}$ ) of $s$ positions on the vector of $2 r-1$ coefficients of $f^{\prime}$ and taking the first $r$ coefficients. Note that making $f^{\prime}$ a longer polynomial will lead to the same result, as long as $f_{i}^{\prime}=f_{(i \bmod r)}$ for $i \in\{0, \ldots, 2 r-2\}$.

Let $b$ be a power of 2 . To implement this idea, $f^{\prime}$ is stored as an array of $b$-bit words, where the values of the words are $\sum_{i=0}^{b-1} f_{i}^{\prime} 2^{i}, \sum_{i=0}^{b-1} f_{i+b^{\prime}}^{\prime} 2^{i}$, and so on. Then, $f^{\prime}$ is first shifted by $s^{(1)}=s-(s \bmod b)$ positions and then by $s^{(0)}=s \bmod b$ positions. As $s^{(1)}$ is a multiple of $b$, the shift by $s^{(1)}=\sum_{j} s_{j}^{(1)} 2^{j}$ positions is carried out as a sequence of conditional moves from word $i+2^{j} / b$ to word $i$ for each $j$. The shift by $s^{(0)}$ positions can be carried out by a sequence of logical shifts and ORs on the words, assuming that corresponding instructions exist.

The avx2 implementation follows the strategy above, where $b$ is set to 256 to fit the size of YMM registers. Note that there is no general shift instruction for YMM registers, so the avx2 implementation follows [GAB19] to use AVX intrinsics to carry out the shift by $s^{(0)}$ positions.

[^0]```
Algorithm 1 The Black-Gray Decoder (rephrased)
Parameters: \(r, w, t, d=w / 2\), NbIter, \(\tau\), threshold
    procedure \(\operatorname{DECODER}\left(s, h_{0}, h_{1}\right)\)
        \(\left(e_{0}, e_{1}\right) \leftarrow(0,0) \in \mathcal{R}^{2}\)
        for \(i=1, \ldots\), NbIter do
            \(T \leftarrow \operatorname{threshold}\left(\left|s+e_{0} h_{0}+e_{1} h_{1}\right|, i\right)\)
            \(e_{0}, e_{1}, b_{0}, b_{1}, g_{0}, g_{1} \leftarrow \operatorname{BFIter}\left(s+e_{0} h_{0}+e_{1} h_{1}, e_{0}, e_{1}, T, h_{0}, h_{1}\right)\)
            if \(i=1\) then
                \(e_{0}, e_{1} \leftarrow \operatorname{BFMask} \operatorname{LIter}\left(s+e_{0} h_{0}+e_{1} h_{1}, e_{0}, e_{1}, b_{0}, b_{1},(d+1) / 2+1, h_{0}, h_{1}\right)\)
                    \(e_{0}, e_{1} \leftarrow \operatorname{BFMasked} \operatorname{Iter}\left(s+e_{0} h_{0}+e_{1} h_{1}, e_{0}, e_{1}, g_{0}, g_{1},(d+1) / 2+1, h_{0}, h_{1}\right)\)
            end if
        end for
        if \(s=e_{0} h_{0}+e_{1} h_{1}\) then
            return \(\left(e_{0}, e_{1}\right)\)
        else
            return \(\perp\)
        end if
    end procedure
    procedure \(\operatorname{BFITER}\left(s, e, T, h_{0}, h_{1}\right)\)
        \(\left(c_{0}, c_{1}\right) \leftarrow\left(\operatorname{Tr}\left(h_{0}\right) \cdot \operatorname{Lift}(s), \operatorname{Tr}\left(h_{1}\right) \cdot \operatorname{Lift}(s)\right)\)
        \(\left(b_{0}, b_{1}\right) \leftarrow(0,0) \in \mathcal{R}^{2}\)
        \(\left(g_{0}, g_{1}\right) \leftarrow(0,0) \in \mathcal{R}^{2}\)
        for \(i=0,1\) do
            for \(j=0, \ldots, r-1\) do
                if \(c_{i j} \geq T\) then
                    \(e_{i j} \leftarrow e_{i j}+1\)
                    \(b_{i j} \leftarrow 1\)
            else if \(c_{i j} \geq \tau\) then
                    \(g_{i j} \leftarrow 1\)
            end if
            end for
        end for
        return \(e_{0}, e_{1}, b_{0}, b_{1}, g_{0}, g_{1}\)
    end procedure
    procedure \(\operatorname{BFMASKEDITER}\left(s, e, m_{0}, m_{1}, T, h_{0}, h_{1}\right)\)
        \(\left(c_{0}, c_{1}\right) \leftarrow\left(\operatorname{Tr}\left(h_{0}\right) \cdot \operatorname{Lift}(s), \operatorname{Tr}\left(h_{1}\right) \cdot \operatorname{Lift}(s)\right)\)
        for \(i=0,1\) do
            for \(j=0, \ldots, r-1\) do
                if \(c_{i j} \geq T\) then
                    \(e_{i j} \leftarrow e_{i j}+m_{i j}\)
            end if
            end for
        end for
        return \(e_{0}, e_{1}\)
    end procedure
```


### 3.2 Avoiding the Barrel Shifter with Matrix Transposition

We do better than the avx implementation by making use of a simple observation. In short, we observed that shifting $f^{\prime}$ by $s^{(1)}$ positions can be carried out by shifting a set of

```
_INLINE_ void
rotate256_big(OUT syndrome_t *out, IN __m256i *in, IN size_t ymm_num)
{
    int i;
    for (i = 0; i < 64; i++)
        STORE(&out->qw[4*i], SRLI_I64(in[i], ymm_num) |
                                SLLI_I64(in[i+64], 64-ymm_num));
    transpose_64x256_sp_asm(out->qw);
}
```

Figure 2: The source code for computing $M^{(1)}$ from $N^{(0)}$ for bikel1.
$b$ vectors by $s^{(1)} / b$ positions, if $f^{\prime}$ is represented in a different format. The observation is explained via the following example.

Let $r=b^{2} / 2$. For a concrete number, one can imagine that $b=64$. We choose $f^{\prime}$ to have a length of $2 r=b^{2}$ instead of $2 r-1$. Consider $f^{\prime}$ as a $b \times b$ matrix $M^{(0)}$, where the entry at row $i$ and column $j$ contains the bit $f_{b \cdot i+j}^{\prime}$. The matrix is illustrated below.

$$
M^{(0)}=\left(\begin{array}{cccc}
f_{0}^{\prime} & f_{1}^{\prime} & \ldots & f_{b-1}^{\prime} \\
f_{b}^{\prime} & f_{b+1}^{\prime} & \ldots & f_{2 b-1}^{\prime} \\
\vdots & \vdots & \vdots & \vdots
\end{array}\right)
$$

Our main observation is that shifting $f^{\prime}$ by $s^{(1)}$ positions, in the matrix view, simply means updating row $i$ by row $i+s^{(1)} / b$ (and if row $i+s^{(1)} / b$ is undefined, row $i$ is set to 0 ). If we can somehow make each column of $M^{(0)}$ lie in one $b$-bit word, this can be carried out easily via $b$ shift operations by $s^{(1)} / b$ positions on the columns.

In order to make use of this observation, given $f$ and $s$, we perform the following steps to compute $y^{-s} f$.

1. Compute $f^{\prime}$, which can be viewed as the matrix $M^{(0)}$, from $f$.
2. Compute $N^{(0)}=\left(M^{(0)}\right)^{T}$. How to perform such a matrix transposition efficiently given $b=64$ has been explained in, say, [Cho17].
3. Perform a shift of $s^{(1)} / b$ positions on each row of $N^{(0)}$ to obtain $N^{(1)}$.
4. Compute $M^{(1)}=\left(N^{(1)}\right)^{T}$. Note that we only need $\lceil(r+b-1) / b\rceil$ rows of $M^{(1)}$ for the next step.
5. Shift the vector of length $r+b-1$ represented by the first $\lceil(r+b-1) / b\rceil$ rows of $M^{(1)}$, by $s^{(0)}$ positions. Then, the first $r$ entries will be $y^{-s} f$.

Recall that what we actually want to compute is $\sum_{s} y^{-s} f$. In fact, the first two steps above do not depend on $s$, so $N^{(0)}$ is the same for any $s$. Therefore, one can save operations by computing $N^{(0)}$ with the first two steps, and then for each $s$ perform the last three steps.

### 3.3 The Haswell Implementation for bikel1

For ease of discussion, we define the following operators that can be applied to any matrix.

- $\Leftarrow_{j}(\cdot)$, meaning replacing column $i$ by column $i+j$ for all $i$.
- $\Rightarrow_{j}(\cdot)$, meaning replacing column $i$ by column $i-j$ for all $i$.
- $\Uparrow_{j}(\cdot)$, meaning replacing row $i$ by row $i+j$ for all $i$.
- $\Downarrow_{j}(\cdot)$, meaning replacing row $i$ by row $i-j$ for all $i$.

For each operator, the destination row or column is set to 0 if the source is not well-defined. The parentheses will be omitted whenever it is convenient.

Our Haswell implementation for bikel1 uses $b=256$ because it mainly works on YMM registers. bikel1 has $r=12323$, so $f^{\prime}$ consists of at least $\lceil(2 r-1) / 256\rceil=97$ words and $s^{(1)} / b \leq 48$. For ease of implementation, we let $f^{\prime}$ be of length 32768 , so $M^{(0)}$ is a matrix with 128 rows:

$$
M^{(0)}=\left(\begin{array}{llll}
M_{00} & M_{01} & M_{02} & M_{03} \\
M_{10} & M_{11} & M_{12} & M_{13}
\end{array}\right) \in \mathbb{F}_{2}^{128 \times 256}
$$

where each $M_{i j}$ is a $64 \times 64$ matrix. Following the discussion in Section 3.2, we only need the first $\lceil(12323+255) / 256\rceil=50$ rows of $\Uparrow_{s^{(1)} / b}\left(M^{(0)}\right)$ to perform the final shift of $s^{(0)}$ positions. Let $s^{\prime}=s^{(1)} / b$, for ease of implementation, we compute the first 64 rows of $\Uparrow s^{\prime}\left(M^{(0)}\right)$, which is simply

$$
M^{(1)}=\left(\begin{array}{cccc}
\Uparrow s^{\prime} & M_{00} & \Uparrow s^{\prime} & M_{01} \\
+ & + & s_{s^{\prime}} M_{02} & \Uparrow s^{\prime} \\
+ & + & M_{03} \\
\Downarrow_{64-s^{\prime}} M_{10} & \Downarrow_{64-s^{\prime}} M_{11} & \Downarrow_{64-s^{\prime}} M_{12} & \Downarrow_{64-s^{\prime}} M_{13}
\end{array}\right) \in \mathbb{F}_{2}^{64 \times 256} .
$$

Let $N_{i j}=M_{i j}^{T}$. Our implementation stores $f^{\prime}$ in a way that the corresponding matrix is

$$
N^{(0)}=\left(\begin{array}{llll}
N_{00} & N_{01} & N_{02} & N_{03} \\
N_{10} & N_{11} & N_{12} & N_{13}
\end{array}\right) \in \mathbb{F}_{2}^{128 \times 256}
$$

instead of $M^{(0)}$. And then for each $s$, we compute

$$
N^{(1)}=\left(\begin{array}{cccc}
\Leftarrow_{s^{\prime}} N_{00} & \Leftarrow_{s^{\prime}} N_{01} & \Leftarrow_{s^{\prime}} N_{02} & \Leftarrow_{s^{\prime}} N_{03} \\
+ & + & + & + \\
\Rightarrow_{64-s^{\prime}} N_{10} & \Rightarrow_{64-s^{\prime}} N_{11} & \Rightarrow_{64-s^{\prime}} N_{12} & \Rightarrow_{64-s^{\prime}} N_{13}
\end{array}\right) \in \mathbb{F}_{2}^{64 \times 256}
$$

and transpose each of the $464 \times 64$ submatrices in $N^{(1)}$ to obtain $M^{(1)}$.
We chose the implementation strategy above, because it can be implemented easily using AVX2 intrinsics/instructions. Our source code for computing $M^{(1)}$ from $N^{(0)}$ is given in Figure 2. The variable in is a pointer to $N^{(0)}$. The variable ymm_num is set to $s^{\prime}$. SRLI_I64 is defined (by the BIKE team) as _mm256_srli_epi64, and we use it to perform $\Leftarrow_{s^{\prime}}$. SLLI_I64 is defined (by the BIKE team) as _mm256_slli_epi64, and we use it to perform $\Rightarrow_{64-s^{\prime}}$. The discussion above suggests that we need to compute XOR of the results of SRLI_I64 and SLLI_I64, but the source code uses OR because it has the same effect. The function transpose_ $64 \times 256$ _sp_asm is an assembly function that computes the " $64 \times 64$-block-wise" matrix transposition for obtaining $M^{(1)}$ from $N^{(1)}$. The assembly function performs a vectorized version of the $64 \times 64$ matrix transposition mentioned in Section 3.2. We note that the assembly function is included in the source code of Classic McEliece. The source code of Classic McEliece, including the function, is in the public domain. In fact, we also compute $N^{(0)}$ from $M^{(0)}$ by calling the assembly function twice.

### 3.4 The Haswell Implementation for bikel3

bikel3 has $r=24659$, so $f^{\prime}$ consists of at least $\lceil(2 r-1) / 256\rceil=193$ words and $s^{\prime}=$ $s^{(1)} / b \leq 96$. For ease of implementation, we let $f^{\prime}$ be of lenght 65536 , which means $M^{(0)}$ is a matrix with 256 rows:

$$
M^{(0)}=\left(\begin{array}{llll}
M_{00} & M_{01} & M_{02} & M_{03} \\
M_{10} & M_{11} & M_{12} & M_{13} \\
M_{20} & M_{21} & M_{22} & M_{23} \\
M_{30} & M_{31} & M_{32} & M_{33}
\end{array}\right) \in \mathbb{F}_{2}^{256 \times 256}
$$

where each $M_{i j}$ is a $64 \times 64$ matrix. Following the dicussion in Section 3.2, we only need the first $\lceil(24659+255) / 256\rceil=98$ rows of $\Uparrow_{s^{\prime}}\left(M^{(0)}\right)$ to perform the final shift of $s^{(0)}$ positions. For ease of implementation, we instead compute 128 rows of $\Uparrow s^{\prime}\left(M^{(0)}\right)$. Let $M^{(0 k)} \in \mathbb{F}_{2}^{128 \times 256}$ be the first 128 rows of $\Uparrow_{64 k} M^{(0)}$. We observed that there are two cases for the first 128 rows of $\Uparrow s^{\prime}\left(M^{(0)}\right)$, which we denote as $M^{(1)}$ :

$$
M^{(1)}= \begin{cases}\Uparrow_{s^{\prime}}\left(M^{(00)}\right)+\Downarrow_{64-s^{\prime}}\left(M^{(01)}\right) & \text { if } 0 \leq s^{\prime}<64 \\ \Uparrow_{s^{\prime}-64}\left(M^{(01)}\right)+\Downarrow_{128-s^{\prime}}\left(M^{(02)}\right) & \text { if } 64 \leq s^{\prime}<128\end{cases}
$$

Let $N_{i j}=M_{i j}^{T}$. Our implementation stores $f^{\prime}$ in a way that the corresponding matrix is

$$
N^{(0)}=\left(\begin{array}{llll}
N_{00} & N_{01} & N_{02} & N_{03} \\
N_{10} & N_{11} & N_{12} & N_{13} \\
N_{20} & N_{21} & N_{22} & N_{23} \\
N_{30} & N_{31} & N_{32} & N_{33}
\end{array}\right) \in \mathbb{F}_{2}^{256 \times 256}
$$

If $0 \leq s^{\prime}<64$, we compute
$N^{(1)}=\left(\begin{array}{cccc}\Leftarrow_{s^{\prime}}\binom{N_{00}}{N_{10}} & \Leftarrow_{s^{\prime}}\binom{N_{01}}{N_{11}} & \Leftarrow_{s^{\prime}}\binom{N_{02}}{N_{12}} & \Leftarrow_{s^{\prime}}\binom{N_{03}}{N_{13}} \\ + & + & + & + \\ \Rightarrow_{64-s^{\prime}}\binom{N_{10}}{N_{20}} & \Rightarrow_{64-s^{\prime}}\binom{N_{11}}{N_{21}} & \Rightarrow_{64-s^{\prime}}\binom{N_{12}}{N_{22}} & \Rightarrow_{64-s^{\prime}}\binom{N_{13}}{N_{23}}\end{array}\right) \in \mathbb{F}_{2}^{128 \times 256}$
Otherwise, we compute
$N^{(1)}=\left(\begin{array}{rrrr}\Leftarrow_{s^{\prime}-64}\binom{N_{10}}{N_{20}} & \Leftarrow_{s^{\prime}-64}\binom{N_{11}}{N_{21}} & \Leftarrow_{s^{\prime}-64}\binom{N_{12}}{N_{22}} & \Leftarrow_{s^{\prime}-64}\binom{N_{13}}{N_{23}} \\ + & + & + \\ \Rightarrow_{128-s^{\prime}}\binom{N_{20}}{N_{30}} & \Rightarrow_{128-s^{\prime}}\binom{N_{21}}{N_{31}} & \Rightarrow_{128-s^{\prime}}\binom{N_{22}}{N_{32}} & \Rightarrow_{128-s^{\prime}}\binom{N_{23}}{N_{33}}\end{array}\right) \in \mathbb{F}_{2}^{128 \times 256}$
In either case, each of the $864 \times 64$ submatrices in $N^{(1)}$ is transposed to obtain $M^{(1)}$.
It might seem that the implementation strategy above will result in branching on $s^{\prime}$, and branching on $s^{\prime}$ is not allowed for constant-time implementations. However, this can be avoided with the ability of AVX intrinsics. Our source code for computing $M^{(1)}$ from $N^{(0)}$ is given in Figure 3. The variable in, again, is a pointer for $N^{(0)}$. The variable ymm_num, again, is set to $s^{\prime}$. When $0 \leq s^{\prime}<64$, the SRLI_I64 and SLLI_I64 in line 12 and 13 will give all-zero results because their shift amounts are out-of-range. Similarly, when $64 \leq s^{\prime}<128$, the SRLI_I64 and SLLI_I64 in line 10 and 11 will give all-zero results because their shift amounts are out-of-range. In this way, we are able to avoid branching on $s^{\prime}$. Finally, we compute $M^{(1)}$ from $N^{(1)}$. by calling the assembly function transpose_64x256_sp_asm twice. We also compute $N^{(0)}$ from $M^{(0)}$ by calling the assembly function four times.

## 4 Circular Shifts with Conditional Moves

Following the discussion in Section 3.1, our M4 implementation uses a Barrel shifter to shift the duplicated form of $f$ by $s$ bits. In the setting of M4, we have $s=s^{(1)}+s^{(0)}$ where $s^{(0)}=s \bmod 32$, and the duplicated form of $f$ is stored as an array of 32 -bit words. Our M4 implementation first performs the shift by $s^{(1)}$ bits and then the shift by $s^{(0)}$ bits.

```
_INLINE_ void
rotate256_big(OUT syndrome_t *out, IN __m256i *in, IN size_t ymm_num)
{
    int j;
    __m256i tmp;
    for (j = 0; j < 128; j++)
    {
        tmp = SRLI_I64(in[j], ymm_num)
            | SLLI_I64(in[j+64], 64 - ymm_num)
            | SRLI_I64(in[j+64], ymm_num - 64)
            | SLLI_I64(in[j+128], 128 - ymm_num);
        STORE(&out->qw[4*j], tmp);
    }
    transpose_64x256_sp_asm(&out->qw[0]);
    transpose_64x256_sp_asm(&out->qw [64*4]);
}
```

Figure 3: The function for computing $M^{(1)}$ from $N^{(0)}$ for bikel3.

### 4.1 The Shift by $s^{(1)}$ Bits

In the first phase of the Barrel shifter, each word $i$ is conditionally replaced by word $i+2^{j} / 32$ for several $j$ 's with $j \geq 5$ to carry out the shift by $s^{(1)}$ bits. Note that $j \in\{13,12, \ldots, 5\}$ for bikel1 and $j \in\{14,13, \ldots, 5\}$ for bikel3. To complete the computation for each $j$, we can first create a 32 -bit mask which is set to 0xFFFFFFFF if the shift of $2^{j} / 32$ words needs to be carried out or $0 x 0$ otherwise. Then, for $i=0,1,2, \ldots$, the portable implementation does

```
w[i] = (w[i] & ~mask) | (w[i+idx] & mask),
```

where idx holds the value of $2^{j} / 32$. This is essentially the code of the portable implementation, except that it actually works on 64 -bit words.

In our implementation, for each $k$ in the first phase, the words are partitioned into $2^{j} / 32$ sets: each word $i$ with $k=i \bmod 2^{j} / 32$ is in the $k$ th set. For each set, we use two general-purpose registers Rx and Ry , where Rx is initialized with word $k$. Then, we set $i$ to $k$ and perform the following loop.

1. Load word $i+2^{j} / 32$ to Ry.
2. Conditionally move Ry to Rx.
3. Store Rx to word $i$.
4. If $i<\min \left(\left\lceil\left(r+2^{j}-1\right) / 32\right\rceil\right.$ and $i+2^{j} / 32<\lceil(2 r-1) / 32\rceil$, increase $i$ by $2^{j} / 32$. Otherwise, break the loop.
5. Load word $i+2^{j} / 32$ to Rx .
6. Conditionally move Rx to Ry.
7. Store Ry to word $i$.
8. If $i<\min \left(\left\lceil\left(r+2^{j}-1\right) / 32\right\rceil\right.$ and $i+2^{j} / 32<\lceil(2 r-1) / 32\rceil$, increase $i$ by $2^{j} / 32$ and go back to Step 1. Otherwise, break the loop.

In Step 4 and 8 , we check if $i<\min \left(\left\lceil\left(r+2^{j}-1\right) / 32\right\rceil\right.$ because we only need

$$
\left\lceil\left(r+\sum_{i=0}^{j-1} 2^{i}\right) / 32\right\rceil=\left\lceil\left(r+2^{j}-1\right) / 32\right\rceil
$$

words for the remaining conditional shifts of $2^{j-1}, 2^{j-2}, \ldots, 32, s^{(0)}$ bits. We check if $i+2^{j} / 32<\lceil(2 r-1) / 32\rceil$ because there are only $\lceil(2 r-1) / 32\rceil$ words. In our implementation, the if statements are simplified to "If $i<i_{\max }, \ldots$ ", where $i_{\max }$ is a precomputed value that depends on $k$ and $j$.

To carry out the conditional moves, our M4 implementation uses the instruction SEL. As suggested by the name, SEL can move one of its two source registers into its destination register. Which source register is chosen is determined by the "GE" flags ${ }^{2}$. We note that the SEL instruction always takes 1 cycle.

In order to reduce the average cost of each conditional move, our implementation actually processes $n=\min \left(2^{j} / 32,4\right)$ words in one iteration of the loop: we first load $n$ words to Ry $1, \ldots$, Ryn, use the SEL instruction $n$ times to conditionally move Ry1, ..., Ryn to the corresponding Rx1, ..., Rxn, store Ry1, ..., Ryn to the memory of $n$ words. We also partially unroll the loop so that Step 4 is omitted. A piece of code for a loop of $j=12$ is given in Appendix A.

### 4.2 The Shift by $s^{(0)}$ Bits

In the second phase, the Barret shifter performs the shift of $f$ by $s^{(0)}<32$ bits. Suppose $f$ is stored as an array of 32 -bits words $\mathrm{w}[\mathrm{n}]$. In the portable implementation of BIKE team, the shift is carried out by doing

$$
w[i]=(w[i] \geqslant s 0) \mid(w[i+1] \ll(32-s 0)) ;
$$

for $i=1,2, \ldots$, and so on. This approach takes 2 logic shits and 1 logic OR operations for each 32 -bits word. We note that portable actually works on 64 -bit words, but the idea is the same. In fact, we can also deal with the 32 -bit words in the reversed order:

```
v = 0;
for (i = n-1; i >= 0; i--)
{
    tmp = w[i] << (32-s0);
    w[i] = (w[i] >> s0) | v;
    v = tmp;
}
```

Note that the logical OR can be replaced by an addition.
Now consider the following C function.

```
void F(uint32* w, uint32* v, s0)
{
    uint32_t tmp = *w << (32-s0);
    *W = (*W >> sO) + v;
    *v = tmp;
}
```

The shift by $s^{(0)}$ bits can then be carried out by doing simply

[^1]```
v = 0;
for (i = n-1; i >= 0; i--)
    F(w[i], &v, s0);
```

The interesting thing is that F can be replaced by a single instruction UMLAL. The instruction operates as

UMLAL Rdlo, Rdhi, Rm, Rn
where Rdlo and Rdhi are the 2 destination registers and Rm and Rn are the 2 source registers. The instruction multiplies Rm and Rn and adds the 64 -bit result into the 64 -bit integer represented by Rdhi (the top 32 bits) and Rdlo (the bottom 32 bits). To carry out F (w[i], \&v, s0) with the UMLAL, the 4 registers Rdlo, Rdhi, Rm, and Rn are set to $\mathrm{w}[\mathrm{i}], \mathrm{v}$, (1《s0)-1, and w[i] respectively. Note that we set Rdlo to w[i] and Rm to (1《 (32-s0))-1 instead of setting Rdlo to 0 and Rm to 1 < ( $32-\mathrm{s} 0$ ) to avoid the overflow of Rm when $s^{(0)}=0$. In this way, we are able to replace the 2 logical shifts and 1 logical OR with only 1 UMAAL.

A previous version of our M4 implementation actually used IT blocks instead of SEL to carry out conditional moves and use the same way as the portable implementation to perform the shift by $s^{(0)}$ bits. The ideas of using SEL and replacing shift instructions by multiplication instructions was suggested by an anonymous reviewer of this paper. In the current version of our M4 implementation, which uses the ideas, one circular shift is about $5 \%$ faster than in the previous version.

## 5 Hamming Weight Computation with Full Adders

The implementations of the BIKE team follows [Cho16] to compute the sum $\sum_{s}\left(y^{-s} f\right)$ in a bitsliced fashion to obtain $g f$. As the result, a sequence of bitwise logical operations are performed to obtain the sum. Let $n$ be the cardinality for the set of $s$, and let $b_{1}$ be the constant term of the first $y^{-s} f, b_{2}$ be the constant term of the second $y^{-s} f$, and so on. It is easy to see that the problem of computing $\sum_{s}\left(y^{-s} f\right)$ boils down to computing the Hamming weight of $\left(b_{1}, b_{2}, \ldots, b_{n}\right) \in\{0,1\}^{n}$ using a sequence of gates. This sections shows how previous implementations and our implementation compute the Hamming weight.

### 5.1 Simple Algorithms for Computing the Hamming Weight

The avx2 implementation of BIKE, following [Cho16], first prepares a counter of $\left\lfloor\log _{2} n\right\rfloor+1$ zero bits. The counter is used to store the Hamming weight. Then, for each $i \in\{1, \ldots, n\}$, $b_{i}$ is added to the counter. For each addition, if $n$ is a power of 2 , each of the least significant $\left\lfloor\log _{2} n\right\rfloor$ bits can be updated using one half adder, and the most significant bit can be set to the carry-out bit of the last half adder. If $n$ is not a power of 2 , each of the least significant $\left\lfloor\log _{2} n\right\rfloor$ bits can be updated using one half adder, and the most significant bit can be updated using 1 XOR. Let $T(n)$ be the number of bit operations it takes to compute the Hamming weight of $n$ bits. We have

$$
T(n)= \begin{cases}\left(2\left\lfloor\log _{2} n\right\rfloor+1\right) n, & \text { if } n \text { is not a power of } 2 \\ 2\left\lfloor\log _{2} n\right\rfloor n, & \text { if } n \text { is a power of } 2\end{cases}
$$

In [GAB19], the authors suggested a simple way to improve the approach above. The idea is that, since adding $b_{i}$ only affects the first $\left\lfloor\log _{2} i\right\rfloor+1$ bits of the counter, there is no need to update the remaining bits. In fact, adding $b_{i}$ takes $2\left\lfloor\log _{2} i\right\rfloor+1$ bit operations if $i$ is not a power of 2 , and adding $b_{i}$ takes $2\left\lfloor\log _{2} i\right\rfloor$ bit operations if $i$ is a power of 2 . Therefore, we have

$$
T(n)=\sum_{i=1}^{n} C(i), \quad C(i)= \begin{cases}\left(2\left\lfloor\log _{2} i\right\rfloor+1\right), & \text { if } i \text { is not a power of } 2 . \\ 2\left\lfloor\log _{2} i\right\rfloor, & \text { if } i \text { is a power of } 2 .\end{cases}
$$

### 5.2 The Boyar-Peralta Algorithm

Our implementation uses a much faster algorithm which makes use of full adders. The algorithm was explained by Boyar and Peralta in [BP05]. We note that the technique is probably much older: [FS71] presents an algorithm that seems equivalent. However, as we are not able to tell who was the first to propose the algorithm, we simply call the algorithm the Boyar-Peralta Algorithm. Boyar and Peralta focused on proving that the algorithm gives the minimal number of ANDs required to compute the Hamming weight. In the following discussion, we consider the number of XORs as well as the space requirement of the algorithm.

To explain the idea, consider the case when $n=2^{k}-1$. Boyar and Peralta described the following recursive algorithm: apply the algorithm recursively to compute the Hamming weight $x_{1}$ of the first $2^{k-1}-1$ bits, apply the algorithm recursively to compute the Hamming weight $x_{2}$ of the next $2^{k-1}-1$ bits, and use a standard adder to compute the sum $x_{1}+x_{2}+b_{n}$, where $b_{n}$ serves as the carry-in bit. Then the sum $x_{1}+x_{2}+b_{n-1}$ can be computed using $k-1$ full adders, resulting in $5(k-1)$ bit operations. Therefore, we have

$$
T\left(2^{k}-1\right)=5(k-1)+2 T\left(2^{k-1}-1\right)
$$

for $k>1$, and obviously $T(0)=T(1)=0$.
In order to handle general $n$, the Boyar-Peralta algorithm computes $k, n_{1}, n_{2}$ such that $n=2^{k-1}+n_{2}=n_{1}+n_{2}+1$, where $0 \leq n_{2}<2^{k-1}$. Then, the algorithm is applied recursively to compute the Hamming weight $x_{1}$ of the first $n_{1}$ bits, and the Hamming weight $x_{2}$ of the next $n_{2}$ bits, and finally a standard adder is used to compute the sum $x_{1}+x_{2}+b_{n-1}$. Let $k^{\prime}=\left\lfloor\log _{2} n_{2}\right\rfloor+1$ if $n_{2}>0$, i.e., $k^{\prime}$ is the number of bits required to represent $n_{2}$. $k^{\prime}$ is set to 0 if $n_{2}=0$. Then, the sum $x_{1}+x_{2}+b_{n-1}$ can be computed using $k^{\prime}$ full adders and $k-1-k^{\prime}$ half adders, resulting in $5 k^{\prime}+2\left(k-1-k^{\prime}\right)$ bit operations. Therefore, we have

$$
T(n)=5 k^{\prime}+2\left(k-1-k^{\prime}\right)+T\left(2^{k-1}-1\right)+T\left(n_{2}\right)
$$

for $n \geq 2$, and $T(0)=T(1)=0$.
The reason why we need to perform multiplications of the form $g f=\sum_{s}\left(y^{-s} f\right)$ in the first place is because there are multiplications between $\operatorname{Tr}\left(h_{i}\right)$ and $\operatorname{Lift}(s)$ in the decoder (see Section 2.4). As each $\operatorname{Tr}\left(h_{i}\right)$ is always of weight $w / 2$, our goal is actually to compute the Hamming weight of $w / 2$ bits. Now we can easily compute $T(w / 2)$ for the Boyar-Peralta algorithm and the simple algorithms in the previous subsection. The numbers are summarized in Table 2.

### 5.3 Code Generation for the Boyar-Peralta Algorithm

To implement the Boyar-Peralta algorithm, we wrote a code generator to generate the sequence of additions in the Boyar-Peralta algorithm. The code generator is included in Appendix B. It assumes that there is a buffer consisting of infinitely many bits that can be used for storing the results of the additions and the Hamming weight.

Our code generator is written as a Python function named bp. One can simply call $\mathrm{bp}(\mathrm{n}, 0)$ to to generate operations for $n$ input bits. Its the second argument, the variable off in its signature, indicates the index of the first bit in the buffer of the computation. The function, on input $n>1$, computes $k, n_{1}, n_{2}$ such that $n=2^{k}+n_{2}=n_{1}+n_{2}+1$. Then, the function first recursively calls

```
bp(n1, off);
bp(n2, off + nbits(n1));
```

to generate operations for computing $x_{1}$ and $x_{2}$, where nbits (n1) means the number of bits required to represent $n_{1}$. Note $x_{1}$ will be stored in the first nbits(n1) bits starting

Table 2: The numbers of gates required to compute the Hamming weight for the three methods.

|  | $n=w / 2$ | $[$ Cho16] | [GAB19] | $[$ BP05 $]$ |
| :---: | :---: | ---: | ---: | ---: |
| bikel1 | 71 | 923 | 676 | 326 |
| bikel3 | 103 | 1339 | 1092 | 484 |
| bikel5 | 137 | 2055 | 1553 | 655 |

from the buffer bit of index off after the operations generated by the first recursive call are carried out, and $x_{2}$ will be stored in the next nbits(n2) bits after the operations generated by the second recursive call are arried out.

After the recursive calls, the function first outputs

```
"rotate_right(bn, f, s[{0}]);".format(count)
```

which means a new $s$ is used to compute $y^{-s} f$, so that the bit bn is generated. Here count is a global variable which is always increased by 1 after printing the line of rotate_right. Then, the function outputs

```
"adder_size_eq(buf, bn, {0}, {1});".format(off, nbits(n1))
```

if nbits(n1) is equal to nbits(n2) or
"adder_size_neq(buf, bn, \{0\}, \{1\}, \{2\});".format(off, nbits(n1), nbits(n2))
if they are different. In either case, this means that a standard adder is used to compute $x_{1}+x_{2}+b_{n}$. The variable buf is a pointer to the buffer, and the offset of $x_{1}$ is indicated by off. The difference between the two cases is that adder_size_eq only requires one argument to indicate the bit length of $x_{1}$ or $x_{2}$, while adder_size_neq requires two arguments. If $n=1$, the function outputs the line of rotate_right, increases count by 1 , and outputs

```
"adder_size_eq(buf, bn, {0}, 0);".format(off)
```

where the last argument 0 indicates that $b_{n}$ is simply copied into the buffer bit of index off. If $n=0$, the function returns immediately. The only remaining task is to implement the C functions adder_size_eq and adder_size_neq.

By analysing the output of the function, we found that a buffer of 16,17 , and 22 bits are required for $n=71, n=103$, and $n=137$. Note that the simple algorithms requires a buffer of only 7,7 , and 8 bits respectively. In other words, using the BoyarPeralta algorithm to compute the sum $\sum_{s}\left(y^{-s} f\right)$ in a bitsliced fashion takes at least $\lceil 16 \cdot 12323 / 8\rceil=24646,\lceil 17 \cdot 24659 / 8\rceil=52401$, and $\lceil 22 \cdot 40973 / 8\rceil=112676$ bytes for bikel1, bikel3, and bikel5, respectively.

## 6 Multiplications in $\mathbb{F}_{2}[x]$ with Bernstein's 5 -way Recursive Algorithm

The implementations of the BIKE team use Karatsuba for multiplying polynomials in $\mathbb{F}_{2}[x]$ : given two polynomials $F=F_{0}+F_{1} x^{n}$ and $G=G_{0}+G_{1} x^{n}$ where $F_{0}, F_{1}, G_{0}, G_{1}$ are of degree smaller than $n$, the algorithm computes $F G$ as

$$
F_{0} G_{0}+F_{1} G_{1} x^{2 n}+\left(\left(F_{0}+F_{1}\right)\left(G_{0}+G_{1}\right)-F_{0} G_{0}-F_{1} G_{1}\right) x^{n}
$$

This means that Karatsuba is a 3-way recursive algorithm: it reduces the multiplication into 3 smaller ones, where the smaller multiplications can be handled recursively. The complexity of Karatsuba is $O\left(n^{\log _{2} 3}\right)$.

Recall that bikel1 has $r=12323$ and bikel3 has $r=24659$. The BIKE team always picks $n$ to be the smallest possible power of 2 , which means at the top level of recursion, we have $n=8192$ for bikel1 and $n=16384$ for bikel3. They then apply Karatsuba recursively. At the bottom level of recursion, the avx2 implementation uses pclmulqdq, and the portable implementation uses a non-constant-time function to handle multiplications of polynomials of degree smaller than 64 .

As $r$ seems to be large enough to make use of algorithms with lower complexities, our Haswell implementation uses a 5 -way recursive algorithm by Bernstein.

### 6.1 Bernstein's 5-way Recursive Algorithm

In [Ber09a], Bernstein proposed a 5 -way recursive algorithm for multiplying two polynomials in $\mathbb{F}_{2}[x]$ with five multiplications. The complexity of Bernstein's algorithm is $O\left(n^{\log _{3} 5}\right)$. Given two polynomials $F_{0}+F_{1} z+F_{2} z^{2}$ and $G_{0}+G_{1} z+G_{2} z^{2}$ in $\mathbb{F}_{2}[x]$, where $F_{i}$ 's and $G_{i}$ 's are polynomials of degree smaller than $n$ and $z=x^{n}$. The algorithm reduces the multiplication into five polynomial multiplications of polynomials of length about $n$ :

$$
\begin{aligned}
H(0) & =F_{0} \cdot G_{0}, \\
H(1) & =\left(F_{0}+F_{1}+F_{2}\right) \cdot\left(G_{0}+G_{1}+G 2\right), \\
H(x) & =\left(F_{0}+F_{1} x+F_{2} x^{2}\right) \cdot\left(G_{0}+G_{1} x+G_{2} x^{2}\right), \\
H(x+1) & =\left(\left(F_{0}+F_{1}+F_{2}\right)+F_{1} x+F_{2} x^{2}\right) \cdot\left(\left(G_{0}+G_{1}+G_{2}\right)+G_{1} x+G_{2} x^{2}\right), \\
H(\infty) & =F_{2} \cdot G_{2} .
\end{aligned}
$$

Bernstein showed that $H=F \cdot G$ can be constructed by the following formula

$$
\begin{equation*}
H=U+H(\infty)\left(z^{4}+z\right)+\frac{U+V+H(\infty)\left(x^{4}+x\right)}{x^{2}+x}\left(z^{2}+z\right) \tag{1}
\end{equation*}
$$

where

$$
U=H(0)+(H(0)+H(1)) z \text { and } V=H(x)+(H(x)+H(x+1))(x+z) .
$$

### 6.2 Implementing Bernstein's algorithm

Our Haswell implementation uses Bernstein's algorithm at the top level of the recursion and Karatsuba for other levels. At the top level, we pick $n=4096$ for bikel1 and $n=8192$ for bikel3. Note that $12323-4096 \cdot 3=35$ and $24659-8192 \cdot 3=83$. These "leftover bits" are handle separately, and the costs of handling them is very small. In order to make the operands of the 5 multiplications fit into $n=4096$ or $n=8192$ bits, we also handle the top 2 bits of $F_{2}$ and $G_{2}$ and the top bit of $F_{1}$ and $G_{1}$ separately, together with the "leftover bits". At the bottom level of recursion, we also use pclmulqdq to handle multiplications for polynomials of degree smaller than 64 .

Let $h(x)=\sum_{i=0}^{\ell-1} h_{i} x^{i}=\left(U+V+H(\infty)\left(x^{4}+x\right)\right)$. While most of the operations in Bernstein's algorithm are either multiplications or additions, there is one division between $h(x)$ and $\left(x^{2}+x\right)$ in Eq. 1. We carry out the division by computing

$$
h_{\ell-1} x^{\ell-3}+\left(h_{\ell-1}+h_{\ell-2}\right) x^{\ell-4}+\ldots+\left(\Sigma_{i=2}^{\ell-1} h_{i}\right) .
$$

Note that instead of computing the coefficients sequentially, the computation can be easily parallelized: let $h^{\prime}$ be $\left(h-\left(h \bmod x^{2}\right)\right) / x^{2}$, we update $h^{\prime}$ to $h^{\prime}+\left(h^{\prime}-\left(h^{\prime} \bmod x\right)\right) / x$, update $h^{\prime}$ to $h^{\prime}+\left(h^{\prime}-\left(h^{\prime} \bmod x^{2}\right)\right) / x^{2}$, update $h^{\prime}$ to $h^{\prime}+\left(h^{\prime}-\left(h^{\prime} \bmod x^{4}\right)\right) / x^{4}$, and so on. This results in a much faster implementation for the division.

## 7 Polynomial Multiplications in $\mathbb{F}_{2}[x]$ with Frobenius Additive FFTs

This section presents how we carry out the polynomial multiplications on the Cortex-M4. Section 7.1 reviews how the Gao-Mateer additive FFT works. Section 7.2 reviews how to perform multiplications in $\mathbb{F}_{2}[x]$ with the additive FFT in $\left[\mathrm{LCK}^{+} 18, \mathrm{CCK}^{+}\right]$. Section 7.3 shows how we implement the algorithm for the Cortex-M4.

### 7.1 The Gao-Mateer Additive FFT

Below we introduce how the Gao-Mateer additive FFT [GM10], which we denote as AFFT, can be used for polynomials over $\mathbb{F}_{2^{m}}$, where $m$ is a power of 2 .

Cantor Basis An $\mathbb{F}_{2^{2}}$-linear basis $\left\{v_{0}, v_{1}, \ldots, v_{m-1}\right\}$ of $\mathbb{F}_{2^{m}}$, where $m$ is a power of 2 , is called a Cantor basis [Can89] if $v_{0}=1$ and $v_{i-1}=v_{i}^{2}+v_{i}$ for all $i>0$. We define $W_{\ell} \subseteq \mathbb{F}_{2^{m}}$ as the span of $\left\{v_{0}, \ldots, v_{\ell-1}\right\}$ for any $1 \leq \ell \leq m$ and $W_{0}$ as $\{0\}$. When $\ell$ is a power of $2, W_{\ell}$ is the subfield of size $2^{\ell}$. In [Can89], Cantor defined a series of polynomials that are closely related to $v_{i}$ 's and $W_{\ell}$ 's:

$$
s_{0}(x)=x, \text { and } s_{i}(x)=\left(s_{i-1}(x)\right)^{2}+s_{i-1}(x) \text { for } i \geq 1 .
$$

In fact, each $s_{i}$ satisfies the following properties.

- $s_{i}(x)$ is linear in the sense that $s_{i}(\alpha+\beta)=s_{i}(\alpha)+s_{i}(\beta)$.
- $s_{i}\left(v_{k+i}\right)=v_{k}$ for all $i \geq 0$.

Divide and Conquer Given $f(x) \in \mathbb{F}_{2^{m}}[x]$ of degree smaller than $n$, where $n=2^{\ell}$, $v \in \mathbb{F}_{2^{m}}$, and $\ell \leq m$, the AFFT computes $S=\left\{f(\alpha) \mid \alpha \in v+W_{\ell}\right\}$. When $\ell=0$, the algorithm simply returns $f(x)$ itself. When $\ell>0$, the algorithm first writes $f(x)$ as

$$
f^{(0)}\left(s_{1}(x)\right)+x \cdot f^{(1)}\left(s_{1}(x)\right) .
$$

This is called "radix conversion" in [BC14]: the radix is changed from $x$ to $s_{1}(x)=x^{2}+x$. Note that the computation $f^{(0)}$ and $f^{(1)}$ from $f$ requires only field additions, which are much cheaper than field multiplications, and each $f^{(i)}$ is of degree smaller than $2^{\ell-1}$. If we have $S^{(0)}=\left\{f^{(0)}\left(s_{1}(\alpha)\right) \mid \alpha \in v+W_{\ell}\right\}$ and $S^{(1)}=\left\{f^{(1)}\left(s_{1}(\alpha)\right) \mid \alpha \in v+W_{\ell}\right\}$, then every $f(\alpha)$ can be computed as $f(\alpha)=f^{(0)}(\alpha)+\alpha f^{(1)}(\alpha)$.

Observe that $s_{1}\left(v+W_{\ell}\right)=s_{1}(v)+W_{\ell-1}$, which is of size $2^{\ell-1}$. Gao and Mateer make use of this observation to compute $S^{(0)}$ and $S^{(1)}$ by recursively evaluating $f^{(0)}$ and $f^{(1)}$ at all elements of $s_{1}(v)+W_{\ell-1}$. In addition, as

$$
v+W_{\ell}=\left(v+\operatorname{span}\left\{v_{\ell-1}, \ldots, v_{1}\right\}\right) \cup\left(v+\operatorname{span}\left\{v_{\ell-1}, \ldots, v_{1}\right\}+1\right)
$$

Gao and Mateer suggest to compute $S$ by computing

$$
f(\beta)=f^{(0)}\left(s_{1}(\beta)\right)+\beta f^{(1)}\left(s_{1}(\beta)\right), \quad f(\beta+1)=f(\beta)+f^{(1)}\left(s_{1}(\beta)\right),
$$

for all $\beta \in v+\operatorname{span}\left\{v_{\ell}-1, \ldots, v_{1}\right\}$. Computation of $f(\beta)$ and $f(\beta+1)$ is depicted by the following figure.



Figure 4: The three stages of butterflies in a size-8 AFFT. Each arrow going upwards is associated with a constant which is not shown, so it actually means "adding the product of the source and the constant to the destination" instead of "adding the source to the destination".

This is one "butterfly" in an AFFT.
To have a clearer picture regarding the pattern of computation in an AFFT, one can "unroll" the recursion. In this way, it can be seen that an AFFT always consists of two phases. The first phase consists of a bunch of field additions from radix conversions in all levels of recursion. The second phase consists of several stages of butterflies, where the last stage consists of the butterflies in level 0 (i.e., the top level) of the recursion, the second last stage consists of the butterflies in level 1 of the recursion, and so on. Figure 4 shows the three stages in the second phase of a size-8 $(n=8)$ AFFT.

### 7.2 Multiplication in $\mathbb{F}_{2}[x]$ with FAFFT

The AFFT allows us to perform multiplications in $\mathbb{F}_{2}[x]$ : given polynomials $f, g \in \mathbb{F}_{2}[x]$ such that $f g$ is of degree smaller than $n=2^{\ell}$, we can lift $f$ and $g$ to $\mathbb{F}_{2^{m}}[x]$ such that $m$ is a power of 2 and $\ell \leq m$, perform an AFFT for each of $f$ and $g$ to evaluate them at $W_{\ell}$, perform the pointwise multiplication, and perform the inverse AFFT to obtain $f g$. Such an AFFT-based polynomial multiplication always works, but one can reduce the cost of the AFFT by carefully choosing the evaluation points for $f$ and $g$.

Reducing the Number of Evaluation Points In $\left[\mathrm{LCK}^{+} 18\right]$, Li et al. suggested to pick the set of evaluation points to be

$$
\Sigma=W_{\ell^{\prime}}+v_{\ell^{\prime}+m / 2}
$$

with

$$
\begin{equation*}
\ell^{\prime}+m / 2<m, \quad n^{\prime}=|\Sigma|=\frac{n}{m}, \quad \ell^{\prime}=\log _{2}^{n^{\prime}}<\frac{m}{2} . \tag{2}
\end{equation*}
$$

In this way, the function that maps $f$ to $f(\Sigma)$ will be invertible, which means we can use the AFFT and the inverse AFFT to carry out multiplications in $\mathbb{F}_{2}[x]$. As explained by Li et al., it is the property that $\Sigma$ can reach $n$ elements in $\mathbb{F}_{2^{m}}$ via the Frobenius map which guarantees that the function is invertible. This is why Li et al. call such an AFFT a Frobenius AFFT (FAFFT). With the FAFFT, the number of evaluation points is reduced by a factor of $m$.

FAFFT by Truncating A Size-n AFFT Consider $\hat{\Sigma}=W_{\ell^{\prime}+\log _{2} m}+v_{\ell^{\prime}+m / 2}$, such that $\Sigma \subset \hat{\Sigma}$ and $|\hat{\Sigma}|=n$. The way Li et al. evaluate $f$ as $\Sigma$ is to perform a size- $n$ AFFT to evaluate $f$ at $\hat{\Sigma}$ but skip all the redundant operations. In other words, the approach of Li et al. is to carry out a truncated AFFT. The truncated AFFT starts with a usual

```
Algorithm 2 FAFFT-based Polynomial Multiplication
Parameters: \(n\) : maximum length of output polynomial
    procedure \(\operatorname{FAFFT}\left(f(x) \in \mathbb{F}_{2}[x]_{<n}, \Sigma\right)\)
        \(\left(f_{0}, f_{1}, \ldots, f_{n-1}\right) \in \mathbb{F}_{2}^{n} \leftarrow \operatorname{RadixConversions}(f(x))\)
        \(\left(f_{0}^{\prime}, \ldots, f_{n^{\prime}-1}^{\prime}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}} \leftarrow \operatorname{Encode}\left(\left(f_{0}, f_{1}, \ldots, f_{n-1}\right), \Sigma\right)\)
        \(\left(\hat{f}_{0}, \ldots, \hat{f}_{n^{\prime}-1}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}} \leftarrow \operatorname{Butterflies}\left(\left(f_{0}^{\prime}, \ldots, f_{n^{\prime}-1}^{\prime}\right), \Sigma\right)\)
        return \(\left(\hat{f}_{0}, \ldots, \hat{f}_{n^{\prime}-1}\right)\)
    end procedure
    procedure \(\operatorname{FAFFT}^{-1}\left(\left(\hat{f}_{0}, \ldots, \hat{f}_{n^{\prime}-1}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}}, \Sigma\right)\)
        \(\left(f_{0}^{\prime}, \ldots, f_{n^{\prime}-1}^{\prime}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}} \leftarrow\) Butterflies \({ }^{-1}\left(\left(\hat{f}_{0}, \ldots, \hat{f}_{n^{\prime}-1}\right), \Sigma\right)\)
        \(\left(f_{0}, \ldots, f_{n^{\prime}-1}\right) \in \mathbb{F}_{2}^{n} \leftarrow \operatorname{Encode}^{-1}\left(\left(f_{0}^{\prime}, f_{1}^{\prime}, \ldots, f_{n-1}^{\prime}\right), \Sigma\right)\)
        \(f(x) \in \mathbb{F}_{2}[x]_{<n} \leftarrow\) RadixConversions \(^{-1}\left(f_{0}, f_{1}, \ldots, f_{n-1}\right)\)
        return \(f(x)\)
    end procedure
    procedure \(\operatorname{BitPolyMul}\left(a(x) \in \mathbb{F}_{2}[x]_{<\frac{n}{2}}, b(x) \in \mathbb{F}_{2}[x]_{<\frac{n}{2}}\right)\)
        \(\left(\hat{a}_{0}, \ldots, \hat{a}_{n^{\prime}-1}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}} \leftarrow \operatorname{FAFFT}(a(x), \Sigma)\)
        \(\left(\hat{b}_{0}, \ldots, \hat{b}_{n^{\prime}-1}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}} \leftarrow \operatorname{FAFFT}(b(x), \Sigma)\)
        \(\left(\hat{c}_{0}, \ldots, \hat{c}_{n^{\prime}-1}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}} \leftarrow\left(\hat{a}_{0} \cdot \hat{b}_{0}, \ldots, \hat{a}_{n^{\prime}-1} \cdot \hat{b}_{n^{\prime}-1}\right) \quad \triangleright\) pointwise multiplication
        \(c(x) \in \mathbb{F}_{2}[x]_{<n} \leftarrow \operatorname{FAFFT}^{-1}\left(\left(\hat{c}_{0}, \ldots, \hat{c}_{n^{\prime}-1}\right), \Sigma\right)\)
        return \(c(x)\)
    end procedure
```

radix conversion phase for a size- $n$ AFFT. The second phase consists of the first $\log _{2} m$ levels of butterflies in the truncated AFFT. This phase is called encoding, in which the operations is defined below. The third phase consists of the last $\ell^{\prime}$ stages of butterflies in the truncated AFFT. The last stage is identical to the butterfly phase in a usual size- $n^{\prime}$ AFFT for $\Sigma$.

As each of the three phase is invertible, one can carry out the inverse FAFFT by performing the inverses of the three phases in the reverse order. The FAFFT-based polynomial multiplication algorithm is shown in Algorithm 2.

Encoding Encoding performs the first $\log _{2} m$ levels of butterflies in the truncated AFFT. With respect to $\hat{\Sigma}$, it is an invertible linear map which maps radix converted coefficients of $f(x)$, e.g., $\left(f_{0}, f_{1}, \ldots, f_{n-1}\right) \in \mathbb{F}_{2}^{n}$ to $\left(f_{0}^{\prime}, \ldots, f_{n^{\prime}-1}^{\prime}\right) \in \mathbb{F}_{2^{m}}^{n^{\prime}}$ where the $f_{i}^{\prime}$ is defined as

$$
\begin{equation*}
f_{i}^{\prime}=\sum_{j=0}^{m-1} r_{j} \cdot f_{j \cdot n_{p}+i}, \quad r_{j}=\prod_{k=0}^{\log _{2} m-1}\left(v_{m / 2-k}\right)^{j_{k}}, \tag{3}
\end{equation*}
$$

where $\left(j_{0}, j_{1}, \ldots, j_{\log _{2} m-1}\right) \in\{0,1\}^{\log _{2} m}$ is the binary representation of the integer $j$, i.e., $j=\sum_{k} j_{k} \cdot 2^{k}$.

### 7.3 Implementing the FAFFT-based Polynomial Multiplication

Below we show how we implemented the FAFFT-based multiplication for the M4.

Parameter Selection The following table shows our parameters for the algorithm 2.
We pick the parameters to minimize $m$ satisfying Eq. (2). The field $\mathbb{F}_{2^{32}}$ is constructed as

|  | $r$ | $n$ | $m$ | $n^{\prime}$ | $\ell^{\prime}$ | $\Sigma$ |
| :---: | ---: | :---: | :---: | :---: | :---: | :---: |
| BIKE-1 | 12323 | 32768 | 32 | 1024 | 10 | $v_{26}+W_{10}$ |
| BIKE-3 | 24659 | 65536 | 32 | 2048 | 11 | $v_{27}+W_{11}$ |

a tower of extension fields of $\mathbb{F}_{2}$ as in $\left[\mathrm{BC}^{2} 4, \mathrm{LCK}^{+} 18\right]$ :

$$
\begin{aligned}
& \mathbb{F}_{2^{2}}=\mathbb{F}_{2}\left[x_{1}\right] /\left(x_{1}^{2}+x_{1}+1\right), \\
& \mathbb{F}_{2^{4}}=\mathbb{F}_{2^{2}}\left[x_{2}\right] /\left(x_{2}^{2}+x_{2}+x_{1}\right), \\
& \mathbb{F}_{2^{8}}=\mathbb{F}_{2^{4}}\left[x_{4}\right] /\left(x_{4}^{2}+x_{4}+x_{2} x_{1}\right), \\
& \mathbb{F}_{2^{16}}=\mathbb{F}_{2^{8}}\left[x_{8}\right] /\left(x_{8}^{2}+x_{8}+x_{4} x_{2} x_{1}\right), \\
& \mathbb{F}_{2^{32}}=\mathbb{F}_{2^{16}}\left[x_{16}\right] /\left(x_{16}^{2}+x_{16}+x_{8} x_{4} x_{2} x_{1}\right) .
\end{aligned}
$$

For completeness, we note that our Cantor basis is defined by

$$
\begin{aligned}
v_{31}= & 1+x_{1} x_{2}+x_{1} x_{2} x_{4}+x_{8}+x_{1} x_{2} x_{8}+x_{2} x_{4} x_{8}+x_{1} x_{2} x_{4} x_{8} \\
& +x_{16}+x_{1} x_{2} x_{16}+x_{4} x_{16}+x_{1} x_{4} x_{16}+x_{2} x_{4} x_{16}+x_{1} x_{2} x_{4} x_{16} \\
& +x_{1} x_{8} x_{16}+x_{4} x_{8} x_{16}+x_{1} x_{2} x_{4} x_{8} x_{16} .
\end{aligned}
$$

The Pointwise Multiplication The pointwise multiplication consists of 1024 or 2048 independent multiplications in $\mathbb{F}_{2^{32}}$. To carry out these multiplications, we follow [BC14, Cho17] to represent field elements in a bitsliced format. As the M4 is a 32-bit platform, bitslicing means that we always perform 32 multiplications in parallel.

To multiply $a_{0}+a_{1} x_{16} \in \mathbb{F}_{2^{32}}$ and $b_{0}+b_{1} x_{16} \in \mathbb{F}_{2^{32}}$ (in parallel with 31 other multiplications), we use the Karatsuba algorithm to reduce the task into 3 multiplications $a_{0} b_{0}, a_{1} b_{1},\left(a_{0}+b_{0}\right)\left(a_{1}+b_{1}\right)$ in $\mathbb{F}_{2^{16}}$ and apply the Karatsuba recursively as [BC14]. We note that the amount of data for multiplication in $\mathbb{F}_{2^{4}}$ fits into the general-purpose registers, so there is no register spilling in the case. For larger fields, we use floating-point registers as a pool for register spilling.

The Butterflies Section 7.1 shows that, at level $i$ of recursion (level 0 represents the top level), multiplications in the butterflies are always of the form $\beta \cdot f^{(1)}\left(s_{1}(\beta)\right)$, where

Table 3: Cost of multiplying an element in $\mathbb{F}_{2^{32}}$ by various field elements. Note that the cycle counts are for carrying out 32 such multiplications. For the environment of benchmarking, see Section 8 .

| mult. by | ANDs | XORs | cycles |
| :--- | ---: | ---: | ---: |
| $\mathbb{F}_{2^{4}}$ | $12 \cdot 8$ | $18 \cdot 8$ | 465 |
| $\mathbb{F}_{2^{8}}$ | $36 \cdot 4$ | $74 \cdot 4$ | 871 |
| $\mathbb{F}_{2^{16}}$ | $108 \cdot 2$ | $270 \cdot 2$ | 1652 |
| $\mathbb{F}_{2^{32}}$ | 324 | 930 | 2849 |
| $v_{17}$ | 0 | 210 | 760 |
| $v_{18}$ | 0 | 211 | 742 |
| $v_{19}$ | 0 | 231 | 800 |
| $v_{20}$ | 0 | 226 | 798 |
| $v_{21}$ | 0 | 231 | 795 |
| $v_{22}$ | 0 | 222 | 795 |
| $v_{23}$ | 0 | 239 | 836 |
| $v_{24}$ | 0 | 249 | 854 |
| $v_{25}$ | 0 | 251 | 881 |
| $v_{26}$ | 0 | 248 | 861 |
| $v_{27}$ | 0 | 243 | 847 |

$\beta \in v_{\ell^{\prime}-i+m / 2}+W_{\ell^{\prime}-i}$. Following [BC14, Cho17], we pre-compute all $\beta^{\prime}$ 's and store them (in the bitsliced format) in a static table. The multiplications can of course be carried out by our function for 32 multiplications in $\mathbb{F}_{2^{32}}$. However, our implementation does better by exploiting the fact that $\beta=v+w$, where $v \in v_{\ell-i+m / 2}$ and $w \in W_{\ell^{\prime}-i}$. To carry out a multiplication between $\beta$ and $\gamma$, our implementation multiplies $\gamma$ by $v$, multiplies $\gamma$ by $w$, and adds the two results to obtain $\beta \cdot \gamma$.

Observe that $w$ is always in a proper subfield of size $2^{2^{\left[\log _{2}\left(\ell^{\prime}-i\right)\right]} \text {. Due to the tower field }}$ structure, we carry out the multiplication by $w$ as $32 / 2^{\left\lceil\log _{2}\left(\ell^{\prime}-i\right)\right\rceil}$ multiplications in the subfield. To carry out the multiplication by $v$, we consider the operation as an invertible linear map and use the circuit generator introduced in the ePrint version of [BC14] to generate a sequence of XORs for the multiplication. We note that $m=32$ is actually too large for the circuit generator to handle: it stops making progress usually after several tens of iterations. When the circuit generator stops to make progress, we generate another sequence of XORs reaching the state in a naive way. Then, we combine the 2 sequences of XORs to complete the whole invertible linear operation. We've considered Paar's circuit generator [Paa97], but it hard to fit everything into available registers since it does not generate in-place code. We also tried Bernstein's circuit generator [Ber09b], but the results (in terms of numbers of XORs) are worse than the one from [BC14].

We show the costs of various multiplications in Table 3 to evaluate our strategy of performing multiplication by $\beta=v+w$. As shown in the table, carrying out 32 multiplications by $v$ takes at most 881 cycles, and carrying out 32 multiplications by 32 w's takes at most 1652 cycles. Carrying out 32 multiplications in $\mathbb{F}_{2^{32}}$, however, takes $2849>881+1652$ cycles.

Figure 4 shows the stride of the butterflies decreases by a factor of 2 in each stage. For the first few stages, the strides are always greater than 32 , so 32 butterflies can be carried out in parallel with bitslicing. However, in the last 5 stages, there are stride-16, $8,4,2,1$ butterflies, which make it hard to carry out 32 butterflies in parallel. The same issue has appeared in [Cho17] regarding Beneš networks as Beneš networks have a similar structure as the butterflies. The author of [Cho17] solve the issue by carrying out matrix transpositions on the data. We make use of the same idea: one can imagine that our implementation performs a matrix transposition on the data right before the last 5 stages. The matrix transposition turns stride- $16,8,4,2,1$ butterflies into butterflies with strides greater than or equal to 32 , so that we are able to carry out 32 butterflies in parallel. At the end of the forward FAFFTs, the order of the field elements is changed because of the matrix transposition, but this does not affect the pointwise multiplication since the multiplication can be carried out in any order. Similarly, in the inverse FAFFT, one can imagine that our implementation performs a matrix transposition on the data right after the first 5 stages.

We note that instead of carrying out the whole matrix transpositions, we actually carry out only a part of the matrix transposition in each of the 5 stages. The part of the matrix transposition we carry out in a stage is just enough to makes it possible to carry out 32 butterflies in parallel in that stage. This is a different way of implementing the idea in [Cho17].

Encoding Consider the input $f^{\prime}$ as a matrix $M \in \mathbb{F}_{2}^{32 \times n^{\prime}}$ where the $M_{j, i}=c_{j \cdot n^{\prime}+i}$. One way to understand the encoding is to consider it as multiplying $M$ by a $32 \times 32$ invertible matrix (from the left), where the invertible matrix is defined by $v_{m / 2}, \ldots, v_{m / 2-4}$. As multiplying by the invertible matrix is a linear map, we again make use of the circuit generator in [Cho17] to optimize it. We also use the circuit generator to optimize the inverse of encoding, which is used in the inverse FAFFT.

Radix Conversions Our implementation follows the algorithm presented in [LANH16], which seems to be implicitly shown in [GM10, Section IV]. As the implementation is quite straight-forward, we do not describe how we implemented the algorithm in detail.

## 8 Experiment Results and Discussions

We chose the STM32F407 Discovery board to be the platform for benchmarking. It is a low-cost development board by STMicroelectronics featuring the STM32F407VG microcontroller that can be clocked with up to 168 MHz , and it has 192 KB of SRAM and 1 MB of flash memory. All implementations for M4 are benchmarked in the pqm4 [KRSS19, KRSS] benchmarking framework on the development board with the compiler arm-none-eabi-gcc-10.2.1. Note that pqm4 always sets the frequency to 24 MHz so that the CPU can be set to have zero wait states when fetching instructions from the flash memory.

All the Haswell cycle counts in this section, unless specified otherwise, are measured on one core of an Intel Xeon E3-1275 v3 CPU, with Turbo Boost and hyper-threading disabled.

### 8.1 Multiplications in $\boldsymbol{R}_{z}$

The following table shows the cycle counts for carrying out one circular shift in our Haswell and M4 implementations with the techniques in Section 3 and 4.

|  | Haswell | M4 |
| ---: | ---: | ---: |
| bikel1 | 1112 | 24856 |
| bikel3 | 2547 | 52722 |

The following table shows the numbers of cycles on the Haswell core to carry out a multiplication $\operatorname{Tr}\left(h_{i}\right) \cdot \operatorname{Lift}(s)$ in $\mathcal{R}_{z}$. Our code uses the techniques in Section 3 and 5.

|  | our code | avx2 |
| ---: | ---: | ---: |
| bikel1 | 113822 | 202176 |
| bikel3 | 369782 | 629952 |

The way the avx2 implementation carries out circular shifts is similar to the approach in Section 4, because it also makes use of conditional moves. The conditional moves are realized by using the intrinsic _mm256_blendv_epi8.

The following table shows the numbers of cycles on the Cortex-M4 to carry out a multiplication $\operatorname{Tr}\left(h_{i}\right) \cdot \operatorname{Lift}(s)$ in $\mathcal{R}_{z}$. Our code uses the techniques in Section 4 and 5 .

|  | our code | portable |
| ---: | ---: | ---: |
| bikel1 | 2195104 | 4346484 |
| bikel3 | 6746105 | 13981293 |

We actually tried to use the techniques in Section 3 for our M4 implementation but found that the approach in Section 4 is still faster. We believe this is due to the fact that the size
of YMM registers on Haswell is much larger than the size of general-purpose registers on the Cortex-M4, and that there are powerful intrinsics/instructions (as shown in [GAB19]) that can be used to manipulate YMM registers so that even a shift of $s^{(0)} \in\{0, \ldots, 255\}$ bits can be carried out efficiently.

### 8.2 Multiplications in $\mathcal{R}$

The following table shows the Haswell cycles for a multiplication between two elements in $\mathcal{R}$. Our code uses techniques in Section 6. We actually tried to implement the FAFFTbased multiplication described in Section 7. Instead of bitslicing, our FAFFT-based implementation uses pclmulqdq for field multiplications. However, as shown in the last column of the table, this leads to a much slower implementation.

|  | our code | avx2 | FAFFT |
| ---: | ---: | ---: | ---: |
| bikel1 | 22856 | 29169 | 66265 |
| bikel3 | 68912 | 91641 | 136194 |

The following table shows the M4 cycles for a multiplication between two elements in $\mathcal{R}$ with respect to methods in Section 7 and from the BIKE team.

|  | our code | portable |
| ---: | ---: | ---: |
| bikel1 | 1320940 | 2897887 |
| bikel3 | 2929293 | 9606051 |

We can further reduce the cost of FAFFT multiplication by reusing the intermediate data while multiplying same elements. As shown in Algorithm 1, there are many multiplications in $\mathcal{R}$ of the form $e_{i} h_{i}$ in the decoder. Since $h_{0}$ and $h_{1}$ do not change during decapsulation, our implementation computes two forward FAFFTs, one for $h_{0}$ and one for $h_{1}$, at the beginning of decapsulation. Then, each time we need to compute $e_{i} h_{i}$, one forward FFT can be saved. In this way, about $30 \%$ of the 1320940 cycles and 2929293 cycles can be saved for each $e_{i} h_{i}$.

It is mentioned in [Cho16, Section 5.1] that one can use circular shifts to carry out each multiplication of $e_{i} h_{i}$ : write $h_{i}$ as $\sum_{s} x^{-s}$ and compute $e_{i} h_{i}$ as $\sum_{s}\left(x^{-s} e_{i}\right)$. The numbers in the first table of Section 8.1 show that the resulting cycle counts are expected to be

- $1112 \cdot 71=78952$ cycles for bikel1 on Haswell,
- $2547 \cdot 103=262341$ cycles for bikel3 on Haswell,
- $24856 \cdot 71=1764776$ cycles for bikel1 on M4,
- $52722 \cdot 103=5430366$ cycle for bikel3 on M4.


### 8.3 Key Generation, Encapsulation, and Decapsulation

The Haswell cycles for key generation, encapsulation, and decapsulation are shown in the following table. The cycle counts of the avx implementation are from [Be], and they are measured on a machine titan0 with a Haswell CPU.

|  | key gen. | encap. | decap. | imple. |
| :---: | ---: | :---: | :---: | :---: |
| bikel1 | 833968 | 131276 | 2636108 | avx2 |
|  | 688372 | 124892 | 1900908 | our code |
| bikel3 | 2515016 | 313976 | 9040604 | avx2 |
|  | 1857256 | 283932 | 6010004 | our code |

The M4 cycles for key generation, encapsulation, and decapsulation are shown in the following table.

|  | key gen. | encap. | decap. | imple. |
| :---: | ---: | ---: | ---: | :---: |
| bikel1 | 65414337 | 4824059 | 114592442 | portable |
|  | 24935033 | 3253379 | 49911673 | our code |
| bikel3 | 212999628 | 15041356 | 374777003 | portable |
|  | 59820502 | 8376212 | 139234176 | our code |

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## A A Loop with Conditional Shift in Our M4 Implementation

Note that the selection instruction SEL depends on the GE flag, which is not affected by the TEQ instruction that is used to control the loop.

```
lsr Ry0, s, #13
and Ry0, #1
mov Rx0, #983040
mul Ry0, Rx0
msr APSR_g, Ry0
add ptr, ptr_a, #-16
add ptr_end, ptr, #1024
loop256_a:
    ldr Rx0, [ptr, #16]!
    ldr Rx1, [ptr, #4]
    ldr Rx2, [ptr, #8]
    ldr Rx3, [ptr, #12]
        ldr RyO, [ptr, #1024]
        ldr Ry1, [ptr, #1028]
        ldr Ry2, [ptr, #1032]
        ldr Ry3, [ptr, #1036]
        sel Rx0, Ry0, Rx0
        sel Rx1, Ry1, Rx1
        sel Rx2, Ry2, Rx2
        sel Rx3, Ry3, Rx3
        str Rx0, [ptr, #O]
        str Rx1, [ptr, #4]
        str Rx2, [ptr, #8]
        str Rx3, [ptr, #12]
        ldr Rx0, [ptr, #2048]
        ldr Rx1, [ptr, #2052]
        ldr Rx2, [ptr, #2056]
        ldr Rx3, [ptr, #2060]
        sel Ry0, Rx0, Ry0
        sel Ry1, Rx1, Ry1
        sel Ry2, Rx2, Ry2
        sel Ry3, Rx3, Ry3
        str Ry0, [ptr, #1024]
        str Ry1, [ptr, #1028]
        str Ry2, [ptr, #1032]
        str Ry3, [ptr, #1036]
    teq ptr_end, ptr
    bne loop256_a
```


## B Python Code for the Boyar-Peralta Algorithm

```
count = 0
index = []
def nbits(n):
    ret = 0;
    while n > 0:
        ret += 1
        n >>= 1
    return ret
```

```
def find(n):
    exp = 0
    while True:
        if 2**exp-1>n:
            break
        exp += 1
    return 2**(exp-1)-1
def bp(n, off):
    global count
    global index
    if n == 1:
        print ("rotate_right(&bn, f, s[{0}]);".format(count))
        count += 1
        print ("adder_size_eq(buf, bn, {0}, 0);".format(off))
        index.append(off + 0)
        return
n1 = find(n-1)
n2 = n-1-n1
    if n1 >= 1: bp(n1, off)
    if n2 >= 1: bp(n2, off + nbits(n1))
print ("rotate_right(&bn, f, s[{0}]);".format(count))
count += 1
if n1 == n2:
        print ("adder_size_eq(buf, bn, {0}, {1});".format(off, nbits(n1)))
        index.append(off + 2*nbits(n1) - 1);
else:
        print ("adder_size_neq(buf, bn, {0}, {1}, {2});".format(off, nbits(n1), nbits(n2)))
        index.append(off + nbits(n1) + nbits(n2) - 1);
```


[^0]:    ${ }^{1}$ [DG19] actually uses $\left(1+y^{r}\right) f$, which has one more coefficient than $f+\left(f \bmod y^{r-1}\right) y^{r}$.

[^1]:    ${ }^{2}$ Actually the selection is done on a byte per byte basis. For example, with the right values in the GE flags, one can write the lower 8 bits of the first source register to the corresponding bits of the destination register, and write the upper 24 bits of the second source register to corresponding bits of the destination register.

