Reduced memory meet-in-the-middle attack against the NTRU private key

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Abstract

NTRU is a public-key cryptosystem introduced at ANTS-III. The two most used techniques in attacking the NTRU private key are meet-in-the-middle attacks and lattice-basis reduction attacks. Howgrave-Graham combined both techniques in 2007 and pointed out that the largest obstacle to attacks is the memory capacity that is required for the meet-in-the-middle phase. In the present paper an algorithm is presented that applies low-memory techniques to find ‘golden’ collisions to Odlyzko’s meet-in-the-middle attack against the NTRU private key. Several aspects of NTRU secret keys and the algorithm are analysed. The running time of the algorithm with a maximum storage capacity of \(w\) is estimated and experimentally verified. Experiments indicate that decreasing the storage capacity \(w\) by a factor \(1 < c < \sqrt{w}\) increases the running time by a factor \(\sqrt{c}\).

1. Introduction

NTRU \cite{12} is a ring-based public-key cryptosystem that so far has survived quantum attacks (e.g. \cite{9, 30}). Therefore it is one of the proposals to become standard in a post-quantum-computer world. There has been much research into which parameters to choose for NTRU \cite{10, 11, 16} to keep up with new attacks and new proposals for the structure of the private key.

Odlyzko’s meet-in-the-middle attack is one of those attacks which attempts to recover the NTRU private key from the public key. In a CRYPTO 2007 paper \cite{14} by Howgrave-Graham it is said regarding this attack:

Odlyzko’s attack on the ees251ep6 parameter set will require too many operations \((2^{95.8}\) modular additions) and/or too much storage \((2^{94}\) bits) to be feasible, and hence the parameter set is more than adequate for a \(k = 80\) security level. Of these two constraints the storage requirement is by far the larger obstacle in today’s hardware.

The requirement of \(2^{94}\) bits of storage would mean that at a conservative estimate of a $25 1 TB hard drive, this attack would cost over $50 quadrillion.

The main contribution of this paper is to show that it is possible to apply low-memory search techniques introduced by \cite{28} to the meet-in-the-middle attacks against the NTRU private key. This extension of Odlyzko’s attack introduces the possibility of using a variable amount of memory for Odlyzko’s attack against the NTRU private key, an option that does...
not seem to have been considered before. We show that we can achieve a factor $1 < c < \sqrt{w}$ reduction in the required space $w$ of this attack at the cost of a factor of approximately $\sqrt{c}$ in the running time. Additionally this paper offers some new analysis of NTRU private keys and extensive experiments to support the results of the analysis.

1.1. Roadmap

In §2 we describe the theory of NTRU and the corresponding key recovery problem. We then explain how a meet-in-the-middle attack can be mounted to find the NTRU private key. We end the section with an explanation of how a low-memory search for collisions works in meet-in-the-middle as was described in [27].

In §3 we describe how we can use this collision search in the meet-in-the-middle attack against the NTRU private key to achieve the reduction in the required memory.

In §4 we analyse some mathematical aspects of the attack. We look at the number of rotations of a key that have the correct form to be found by the attack and the probability that one of those rotations cannot be detected by the collision search. This section ends with a heuristic for the estimated running time of the attack.

In §5 we verify the analysis by running simulated attacks implemented in the computer algebra system Sage [22].

Finally, in §6, we discuss other considerations about this work such as optimizations of the attack and applicability to other key forms and the hybrid attack of [14]. It also contains possible implications for the NTRU parameters currently advised in the EEES #1 standard.

2. Preliminaries

2.1. The NTRU cryptosystem

NTRU [12] is a ring-based public-key cryptosystem and a proposed alternative to RSA and ECC. The original proposal, which we will work with for the remainder of this paper, works over the ring $\mathcal{R} = \mathbb{Z}_q[X]/(X^n - 1)$ in which elements can be represented by vectors in $\mathbb{Z}_q^n$ and multiplications with powers of $X$ are cheap rotations. The integers $q$ and $n$ are system parameters.

Keys in this system are generated as follows. The user chooses two random small polynomials $f, g \in \mathcal{R}$. If we suppose that elements $\mathcal{R}$ are represented by coefficients in the interval $[-q/2, q/2)$, then ‘small’ in this context means that the coefficients of $f$ and $g$ are much smaller than the average $q/4$. The cryptosystem has additional parameters $d_f$ and $d_g$ determining how small they are by setting the private key $f$ to have $d_f$ coefficients equal to 1 and the rest equal to 0 and $g$ to have $d_g$ coefficients equal to 1 and the rest equal to 0. In this paper we will assume that $d = d_f = d_g$, although they can be chosen differently.

The user then computes the public key as follows:

$$h \equiv f^{-1}g.$$ 

Note that since $f, g \in \mathcal{R}$, the result is also in $\mathcal{R}$ and, as is common for NTRU, we will assume the coefficients to be centralized to fall in the interval $[-q/2, q/2)$. In the remainder of the paper we will assume the same for operations involving elements of $\mathcal{R}$. We will denote ring elements by $f$ and the corresponding vectors in $(\mathbb{Z}_q)^n$ by $\mathbf{f}$. For $q$ we choose a prime smaller than the odd prime parameter $n$. Again, many variations are possible in $n$, $q$, the form of $f$ and $g$ and the ring that is chosen, and we conjecture that the attack will work on all variants. Some of these variations will be discussed in §6, but in the next sections we will assume the cryptosystem as is described above.
2.2. Odlyzko’s meet-in-the-middle attack

The original description of a meet-in-the-middle attack on NTRU is attributed to Odlyzko and first described in [15]. We will follow the description in [14] and use its notation for the remainder of this paper.

The goal of a general meet-in-the-middle attack is to find specific elements \( x, x' \) in a search space of which it is known that \( F_1(x) = F_2(x') \) for some functions \( F_1, F_2 \). There may be more solutions to the equation, also known as collisions, but usually a specific solution is sought. This solution is called the golden collision [28]. In NTRU this idea can be applied by splitting the search space of the possible private keys into two parts and searching through them separately to find the full private key. Recall that the private key \( f \) is a polynomial of degree \( n-1 \). Searching for the key \( f \) is split up by looking for polynomials \( f_1 \) in the polynomials of degree at most \( n/2 - 1 \) (\( \mathcal{F}_1 \)) and polynomials \( f_2 \) which only contain terms of degree between \( n/2 \) and \( n-1 \) (\( \mathcal{F}_2 \)). For \( f = f_1 + f_2 \), we have that

\[
\begin{align*}
  h &= (f_1 + f_2)^{-1}g, \\
  f_1 \cdot h &= g - f_2 \cdot h.
\end{align*}
\]

This implies that the correct pair of \( f_1 \) and \( f_2 \) will only differ by a binary vector after multiplication with \( h \). All rotations of the keys \( f \) and \( g \) will be a solution to this equation. Therefore the mean value theorem implies the search can be restricted to \( f_1, f_2 \) with (almost) equal number of ones. Without the \( g \) term the search would boil down to a simple collision search of drawing elements \( f_1, f_2 \) and computing \( f_1 \cdot h, f_2 \cdot h \) until we find a match. We are not searching for a collision, however, but a near-collision.

To search for near-collisions an auxiliary function \( a(x) \) is needed. This function takes a vector of length \( n \) and in each coordinate \( x_i \) returns \( 1 (x_i > 0) \). If \( g \) does not cause the coordinates of \( -f_2 \cdot h \) to change sign, that is, \( a(-f_2 \cdot h) \neq a(-f_2 \cdot h + g) \), we have that \( a(f_1 \cdot h) = a(-f_2 \cdot h) \).

The attack then works as follows. We generate elements \( f_i \) from the sets \( \mathcal{F}_1 \) and \( \mathcal{F}_2 \) and store them in a hash table according to their auxiliary function values, or addresses, \( a_i \) until a collision is found. For colliding values \( f_i, f_j \) we compute if \( (f_i + f_j) \cdot h \) is binary. If it is we have almost surely found the private key \( f \).

The expected number of loops before a collision is found [15] is a constant multiple of

\[
L = \frac{1}{\sqrt{n}} \binom{n/2}{d/2}.
\]

If we assume that \( nd/2 \) operations each are required for the computation of \( f_1 \cdot h \) and \( f_2 \cdot h \) and that each \( f_1, f_2 \) takes at least \( \log_2(2(\binom{n/2}{d/2})) \) space to store, then this attack requires expected time \( ndL/2 \) and expected space \( \log_2(2(\binom{n/2}{d/2})L) \). Note that the expected time is a lower bound, since it does not take the computation of \( (f_1 + f_2) \cdot h \) after a collision into account.

There is a probability that \( g \) does change the sign of \( -f_2 h \) (see more on this in §4.1). This can be taken into account by storing it in each address that could be achieved by all possible values of \( g \). The described complexity does not take this probability into account. If we do do this the space complexity needs to be multiplied by the average number of addresses that \( f_1 \cdot h \) can have after adding a sparse, binary \( g \).

2.3. Parallel collision search

Parallel collision search [26] is a method to search for colliding values \( x, x' \) in the function values \( F(x), F(x') \) for a given function \( F : S \rightarrow S \). Its techniques are based on the Pollard rho [21] and try to find the collisions by creating deterministic sequences.
Let \( D \) be a nonempty subset of \( S \) and let \( \theta = |D|/|S| \). We call this set of points the distinguished points. We then create sequences in \( S \), also called trails, by picking a random point \( x_0 \in S \) and computing \( x_i = F(x_{i-1}) \) until a point \( x_i \in D \) is detected. Note that the expected value of \( t \) is \( 1/\theta \). For each trail we store the triples \( (x_0, x_i, t) \). Whenever we find two triples \( (x_0, x_i, t), (x'_0, x'_i, t') \) with \( x_i = x'_i \) and \( x_0 \neq x'_0 \), we have found a collision. These trails can be rerun from their starting values to find the steps \( x_i \neq x_j \) for which \( F(x_i) = F(x_j) \). It can then be checked if this is the golden collision we were looking for.

This method can fail in one of two ways. It can happen that a sequence gets stuck in a cycle of which none of the points are distinguished. This risk can be managed by setting a maximal sequence length after which it is highly likely that a distinguished point has been found, for instance \( 20/\theta \). The second type of hazard is called a Robin Hood. This is whenever one trail hits the starting point of another trail. Such an occurrence does not lead to a collision, but can be easily detected. The occurrence is also highly unlikely.

In [27] and subsequently [28] the application of collision search to, among others, meet-in-the-middle attacks was shown. We deviate slightly from the description in [28], but the idea and heuristic running times remain the same. We define the two functions mentioned before in the description of the meet-in-the-middle attack as \( F_1 : S_1 \to S \) and \( F_2 : S_2 \to S \) and we wish to find \( x,x' \) such that \( F_1(x) = F_2(x') \). Assume without loss of generality that \( |S_1| \geq |S_2| \).

To apply parallel collision search to meet-in-the-middle attacks, we construct a single function \( F \) which has identical domain and codomain to do the collision search on. This function is constructed from the functions \( F_1 \) and \( F_2 \) as follows. Let \( I \) be a large enough interval of integers \( \{0, \ldots, I\} \) \((|I| \geq |S_1|)\) and let \( h_1 : I \to S_1 \) and \( h_2 : I \to S_2 \) functions which map elements of the interval onto elements of respectively \( S_1 \) and \( S_2 \). There are many options for these functions but they should map to each element in the codomain with (almost) uniform probability.

Now let \( G : S \to I \times \{0,1\} \) be a mapping that maps elements from the codomain of the \( F_i \) to elements of \( I \) and a bit selector. A hash function in which the most significant bit is split off is a good example for \( G \). The function \( F(x,i) = G(F_{i+1}(h_{i+1}(x))) \) is now a function of equal domain and codomain in which collision search can be performed. Note that even though \(|I|\) can be as large as we want, only \(|S| = |S_1| + |S_2|\) different values of the interval will be used.

The (almost) uniform codomain stipulation, as well the requirement for a large interval \( I \), makes finding a surjective \( G \) improbable and it can occur that none of the paths lead to the golden collision. A solution for this is to vary \( F \). For instance, different hash functions for \( G \) could be used, or the \( h_i \) could be randomized. Assuming there is one golden collision, the following (heuristic) running time for the attack is given.

**Heuristic 1 (From [28]).** Let \( F : S \to S \) and \( w \geq 2^{10} \) be the number of triples \( (x_0, x_i, t) \) that can be stored. Then the conjectured optimum proportion of distinguished points is \( \theta \approx 2.25 \sqrt{w/|S|} \), and one should generate about \( 10w \) trails per version of \( F \). The expected number of iterations of \( F \) required to complete a meet-in-the-middle attack using these parameters is \( (2.5|S|^{3/2}/w^{1/2})r \), and the expected number of memory accesses is \( 4.5|S| \).

By ‘a version of \( F \)’ we mean that the function needs to be varied to increase the probability of success and \( r \) is the time required for a function iteration. This attack time can be linearly reduced with the number of processors used.

In [28] it is noted that with a small variation of the described attack, it can also be run in \( (7|S_1| \sqrt{|S_2|/w})t \). Although this is an improvement when \( S_2 \) is much smaller than \( S_1 \), we will focus on cases where \( 1 < |S_1|/|S_2| < 2 \).
3. The reduced memory meet-in-the-middle algorithm

In this section we describe how to apply low-memory search for golden collisions to NTRU.

Recall that the secret keys are denoted by $f$ and $g$, the public key $h \equiv f^{-1}g \mod q$ and $f = f_1 + f_2$. To apply collision techniques we first need a function $F$, of equal domain and codomain, in which $f_1$ and $f_2$ collide.

Let $H$ be a hash function. This can be any function with codomain larger than the keyspace of the NTRU private key that is being attacked. We will denote the codomain of this hash function by $\mathcal{I} = I \times \{0, 1\}$. If the space $\mathcal{F}_1$ is the space of possibilities for $f_1$ and $\mathcal{F}_2$ the space of possibilities for $f_2$, then $|I| > \max(|\mathcal{F}_1|, |\mathcal{F}_2|)$. There is an easy surjective function that maps elements from an interval $I$ to elements in either of the spaces $\mathcal{F}_1$ and $\mathcal{F}_2$:

$$ I_i : I \rightarrow \mathcal{F}_i $$

$$ x \rightarrow \text{detcomb}(x \mod |\mathcal{F}_i|), $$

where $i = 1, 2$ and detcomb is a function that deterministically assigns the integer $x$ to a vector $\mathbf{f}_i$ in $\{0, 1\}^n$ which corresponds to a ring element $f_i \in \mathcal{F}_i$. The problem of assigning deterministic indices to binary sequences of length $n$ and weight $d$ is well known in the combinatorial literature (e.g. [18]). An algorithm for the application of data compression can be found in [6, 7, 20, 23].

Note that if $|\mathcal{F}_1| = |\mathcal{F}_2|$, then elements in both sets have a one-to-one correspondence by division by $x^{n/2}$. This can be useful for efficient implementation.

We define

$$ F_1 : [0, I) \rightarrow \{0, 1\}^n $$

$$ x \rightarrow a(I_1(x) \cdot h), $$

$$ F_2 : [0, I) \rightarrow \{0, 1\}^n $$

$$ x \rightarrow a(-I_2(x) \cdot h), $$

where, as before, $a$ is an address function, and

$$ G : \{0, 1\}^n \rightarrow \mathcal{I} $$

$$ x \rightarrow (H(x) \mod |I|) \times \text{MSB}(H(x)), $$

where MSB denotes the most significant bit. Note that the address function $a$, which is independent of $i$, could also be moved to the $G$ function. Also note that in theory we want to map uniformly onto $I$. In practice, taking $|I| = \max(|\mathcal{F}_1|, |\mathcal{F}_2|)$, there is a large risk of $H$ mapping to the same values after reduction modulo $|I|$. To avoid these ‘fake’ collisions we took $I = |\mathcal{F}_1| \cdot |\mathcal{F}_2|$, which seems to solve this problem. The function we will use for the collision search is now defined as

$$ F : \mathcal{I} \rightarrow \mathcal{I} $$

$$ (x, i) \rightarrow G(F_{i+1}(x)). $$

The function $F$ can now be used for a collision attack as follows. We define a distinguishing property $\mathcal{D}$ on $I$ and create trails starting from $x_0 \in \mathcal{I}$ chosen at random. We run trails until they reach a distinguished point in $x_t \in \mathcal{D}$ and then store the triple $(x_0, x_t, t)$ in a hash list, where $t$ is the number of function iterations until the distinguished point was reached. When the memory is full we do not append new values to the hash list, but instead replace a random triple. Upon a collision $(x_0, x_t, t)$, $(x'_0, x'_t, t')$ the trails are rerun to find the collision point (this can be done efficiently by storing $t$ and $t'$). It is then checked if it is a golden collision, and if it is, the key $f = f_1 + f_2$ is returned. If a collision is non-golden, we replace $(x_0, x_t, t)$ with $(x'_0, x'_t, t')$ in the hash list. After producing $10w$ points we delete the hash list and start from scratch after randomizing $H$ by appending a seed to the addresses that are hashed.
4. Analysis

This section will present analysis of the new attack, of which aspects are also applicable to Odlyzko’s meet-in-the-middle attack. We will end with an estimated running time for our new reduced-memory meet-in-the-middle attack. To ease the readability of the analysis, in this section we will assume that \( n \) and \( d \) are even, such that \( |\mathcal{F}_1| = |\mathcal{F}_2| = \binom{n/2}{d/2} = |S|/2 \). All statements can be generalized to odd cases by adding \( \lfloor \rfloor \) and \( \lceil \rceil \) in the appropriate places.

4.1. Success probability

As mentioned before, the meet-in-the-middle attack will work if \( g \) does not cause the addresses of the parts to change sign. To solve this problem we could store each sample \( f_1 \) in all boxes it could fall into if \( g \) were to change the address. In the attack we use, however, this will not be practical. To this end we analyse the probability that \( g \) changes sign.

**Lemma 1.** Suppose \( f \) and \( g \) are randomly chosen of degree \( n - 1 \) with \( d \) coefficients set to 1. Under the assumption that the public key \( h \) is uniformly distributed over \( \mathcal{R} \), the probability that \( g \) will change the address of \(-f_2h\) in equation (2.1) is \((1 - d/nq)^n \approx e^{-d/q}\).

**Proof.** If \( g \) is randomly chosen with \( d \) ones, then the probability that an entry of \( g \) can change a sign is \( d/n \). If we assume that the public key \( h \) is uniformly distributed over \( \mathcal{R} \), then each entry of \( h \) is uniformly distributed over \([0,q]\). Multiplying \( f_2 \) and \( h \) can be seen as summing over \( d/2 \) rotations of \( h \). Since the rotations are uniform we get a sum of uniform distributions modulo \( q \), which is again uniform. Therefore we assume that \(-f_2 \cdot h\), where \( f_2 \) has \( d/2 \) ones, is uniformly distributed over \( \mathcal{R} \) and so we have a probability of \( d/nq \) of one specific entry changing sign and thus a probability of

\[
(1 - \frac{d}{nq})^n \approx e^{-d/q}
\]

of having no sign change. \(\square\)

If \( q \approx 2/3n \) and \( d \approx n/3 \) then this boils down to a probability of \( \exp^{-1/2} \approx 0.61 \).

Because of the mean value theorem we always have a rotation of \( f \) that has \( d/2 \) ones in its first \( n/2 \) coordinates. To get a better estimate of our running times we first present a heuristic which estimates how many rotations have this form.

**Heuristic 2.** Let \( f \) be a polynomial of degree \( n - 1 \) with binary coefficient vector \( f \). Define a correct form of \( f \) as one of its \( n \) rotations \( x^i f \) that has \( d/2 \) ones in the first \( n/2 \) coordinates of \( f \). Then the average number of correct forms for \( f \) can be approximated by

\[
\frac{n\sqrt{2}}{\sqrt{\pi d}}
\]  

(4.1)

**Proof.** Let \( f \) be of degree \( n - 1 \) with \( d \) coordinates equal to 1. Then the probability that \( f \) has \( d/2 \) ones in its first \( n/2 \) coordinates is

\[
\Pr[X = d/2] = \frac{\binom{n/2}{d/2}^2}{\binom{d}{d/2}}.
\]  

(4.2)

Then because each \( f \) has \( n \) rotations, if we assume independence, we have an approximate number of \( n\Pr[X = d/2] \) rotations. This is not an equality, because we neglect the cases in
which $f$ has overlapping rotations. We can approximate this quantity as follows:

$$n \Pr[X = d/2] \approx \frac{n \binom{n/2}{d/2}^2}{\binom{n}{d}} \approx n \cdot \frac{\left(\frac{(n/2)/(d/2) - 1/2}{d/2}\right)^2 \cdot \sqrt{2\pi d}}{(n/d - 1/2)^{d/2}} \cdot \frac{\sqrt{2\pi d}}{(n/d - 1/2)^{d/2}}$$

$$= \frac{n\sqrt{2\pi d}}{\pi d}$$

$$= \frac{n\sqrt{2}}{\sqrt{\pi d}}.$$

Here the first approximation applies (4.2) and the second approximation is due to Stirling’s approximation.

As [15] discovered experimentally, this is more than $\sqrt{n}$ rotations for common parameters.

### 4.2. Performance

We can now estimate how many collisions we need to find before we find a golden collision.

**Lemma 2.** Under the assumption that Heuristic 2 is correct, when mounting a meet-in-the-middle attack as described in §2.2 we expect to find a golden collision after $|S|/2 \cdot (e^{d/q} \cdot (\sqrt{\pi d}/\sqrt{2n}))^{1/2}$ collisions.

**Proof.** The number of unordered pairs of elements in $S$ is close to $|S|^2/2$. For each of these pairs the probability that they map to the same value in $S$ by $F$ is $1/|S|$. This means that the total number of collisions is approximately equal to $|S|^2/2$. By Heuristic 2 and Lemma 1 on average $e^{−(d/q) \cdot (n\sqrt{2}/\sqrt{\pi d})}$ of them are golden. By combining this with the birthday paradox, the lemma follows.

The next lemma shows, as [27] argues intuitively, that analysis using a full memory does not suffice.

**Lemma 3.** Let $\theta$ be the fraction of distinguished points and assume they are distributed uniformly over $S$. Furthermore, assume each one has equal probability of being reached. Then if paths are run until the memory of $w$ points is full, an expected

$$1 - e^{w/\theta^2|S|}$$

$$1 - e^{1/\theta^2|S|}$$

collisions have been found.

**Proof.** Suppose there are $i$ distinguished points in the memory. The average walk length is $1/\theta$, so the expected number of points covered by these $i$ distinguishing points is $i/\theta$. Let $X_i$ be a new path. The probability that this path does not hit any of the paths leading up to the distinguished points is then

$$\left(1 - \frac{i}{\theta|S|}\right)^{1/\theta} \approx e^{-i/\theta^2|S|}.$$

This gives an expected $e^{i/\theta^2|S|}$ collisions at the point where $i$ memory elements are filled. Therefore when the allotted memory is full the expected number of collisions that have
Let $D$ be a set of distinguished points with $|D|/|S| = \alpha \sqrt{w/|S|} = 2.25 \sqrt{w/|S|}$. Then the expected running time to find a golden collision is slightly overestimated as $2^{\gamma} \cdot (\alpha \sqrt{|S|/w} + 2/\alpha \sqrt{|S|/w})$ function iterations if the memory is full.

There are multiple problems with this heuristic time. If the memory is not yet full, the running time per collision is longer. In Lemma 3 we saw that a number of collisions are already found during the filling of memory. On top of this, not all distinguished points have the same probability of being hit.

Therefore we use the experimentally found formula of [27] that for $\theta = 2.25 \sqrt{w/|S|}$ the expected running time to find a golden collision is slightly overestimated as $2.5 r \sqrt{|S|^3/\theta}$. This is, however, an estimate for the case where there is only one golden collision. In Lemmas 2 and 1 it was shown that on average $\left(\frac{n(n/2)^2}{\binom{n}{d/2}}\right) e^{-d/q}$ golden collisions can be found. We get an estimated number of function evaluations of

$$2.5 \left(\frac{e^{-d/q} n\binom{n/2}{d/2}^2}{\binom{n}{d}}\right)^{-1/2} \frac{\left(2\binom{n/2}{d/2}\right)^3}{w} = 5 \left(\frac{2\binom{n}{d}(n/2)^2}{nw e^{-d/q}}\right)^{-1/2},$$

which concludes our proof.

The cost for each function iteration of $F$ will be slightly higher than the cost of $F'$ in each loop of the classic meet-in-the-middle attack. We argue that it will not make a large difference:

- Both $F$ and $F'$ compute a product of $nd/2$ operations.
- Both $F$ and $F'$ compute an address by checking $n$ bits in $n$ operations.
- $F$ first computes $\text{detcomb}$ of an integer. The function performs at most $n/2$ steps and at each step there is a comparison, a multiplication and a division. We therefore require a computational cost of $1.5n$. Since the random sampling of an element in $F'$ will also incur at least $n$ operations, it only makes a small difference.
- In $F$ a hash function $H$ is evaluated, which is not used in $F'$. The number of operations this requires depends on the hash function that is used and can be optimized based on the sample space.

HEURISTIC 3. Let $w$ be the number of triples $(x_0, x_t, t)$ for which there is available memory. Let $D$ be a set of distinguished points with $|D|/|S| = \alpha \sqrt{w/|S|} = 2.25 \sqrt{w/|S|}$. Then the algorithm is expected to run in

$$L^* = 5r \sqrt{\frac{2\binom{n}{d/2}^3}{nw}},$$

(4.3)

operations, where $r$ is the number of operations needed for a function evaluation of $F$.

Proof. As was noted in [27] a simple analysis leads to a cost per detected collision of $\alpha \sqrt{|S|/w} + 2/\alpha \sqrt{|S|/w}$. In Lemma 2 we proved that we need on average $\gamma = |S|/2 \cdot (e^{d/q} - (\sqrt{n}/\sqrt{2n}))^{1/2}$ collisions before we find a golden one, which means a runtime of $\gamma \cdot (\alpha \sqrt{|S|/w} + 2/\alpha \sqrt{|S|/w})$ function iterations if the memory is full.

Of course the above lemma does not take into account that in reality some paths are more likely to be hit than others. It should, however, serve as a good estimate. A similar analysis implies that if the memory is full with $w$ points then a new collision is found after every $\theta |S|/w$ paths. Using the above lemmas, we can now derive a heuristic running time and memory requirement for the reduced memory meet-in-the-middle attack.

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- In $F$ a hash function $H$ is evaluated, which is not used in $F'$. The number of operations this requires depends on the hash function that is used and can be optimized based on the sample space.
As for the storage costs, each triple \((x_0, x_t, t)\) has the following storage requirements:

\[
\text{total storage} = \text{storage}(x_0) + \text{storage}(x_d) + \text{storage}(t) \\
= |\mathcal{F}| + |\mathcal{F}|\theta + \left\lceil \frac{20}{\theta} \right\rceil \\
= |S| + \alpha \sqrt{w|S|} + \frac{\sqrt{|S|}}{\alpha \sqrt{w}}
\]

The total storage per triple is therefore approximately \(2 \log_2 \left( \frac{n}{d} \right)\) bits. Therefore if we have \(b\) bits of memory available, we should choose \(w \approx b / \left(2 \log_2 \left( \frac{n}{d} \right) \right)\).

5. Results

We implemented tests and the attack in the computer-algebra system Sage [22] and ran it on a single core of an AMD FX-8350 Vishera 4.0 GHz CPU of the Saber cluster [1]. The Sage code for our implementation of the attack can be found in [29]. For the hybrid attack of the next section the implementation is a small variation of the PARI/GP code of Schanck [24].

5.1. Rotations

First we verified the correctness of Heuristic 2. Counting the number of correct rotations of 100 000 random keys, we observed that the number is independent of \(q\). For the dependence on \(n\) and \(d\) we graph the results in Figure 1. We see that the experimental results of the number of rotations of the right form converge to the theoretical estimate.

5.2. Probability of failure

Next we verified the validity of Lemma 1. For \(n = 37\) we generated 100 000 random keys \(f\) and \(g\) for different values of \(q\) and \(d\). We then checked if \(g - f_2h\) had a different address from \(f_1h\). The results are displayed in Figures 2 and 3. We see that Lemma 1 predicts the probability well and that the assumption that the coefficients of \(-f_2h\) can be modelled as uniformly distributed is realistic. We also observed that for larger \(n\), the value of \(n\) no longer significantly influences the probability.
5.3. Performance of the algorithm

Although on average there are multiple rotations of a key $f$ that has $d/2$ ones in each half of its vector $f$, it can happen that a chosen key has only a very small number of correct rotations. If this is the case it can happen that our algorithm does not find the key at all, because $g$ might change all possible addresses.

In the experiments to measure the performance of the attack algorithm we picked a prime $n$, $q \approx (2/3)n$, $d \approx (1/3)n$ and varied the number of triples $w$ that could be stored. As advised in [27], we took the number of distinguished points as $D = 1/\theta$ with $\theta = 2.25 \sqrt{w/(F_1 + F_2)} = 2.25 \sqrt{w/(\binom{n/2}{d/2} + \binom{n/2}{d/2})}$ and randomized the step function $F$ after every $10w$ distinguished points. If a distinguished point was to be added to a full memory, we deleted a random entry. We ran the algorithm until at least 20 times the value predicted by Heuristic 3.

We looked at the number of evaluations of $F$ and the success probability of the algorithm. On each set of parameters we ran 1000 attacks, each time taking a fresh random $f$ and $g$. The results are given in Table 1.

We see that there is a high variance in the results. The expectation is that the first version of $F$ will find the solution; if it does not, then it takes on average $10w|D|$ steps to get another chance. If our algorithm uses approximately the same memory as required on average by the...
classic meet-in-the-middle attack, it uses more iterations. As noted before, however, $L$ does not take collision checks into account.

We also see that a decrease in memory capacity by a factor of 4 leads to an increase in the runtime by a factor of 2, which we expected from equation (4.3). Note that this increase is limited by the search space. Increasing the allotted memory above $|F_1 + F_2|$ is no longer useful. We see this by allowing the storage of $2^{16}$ triples for the $n = 37$ set and $2^{18}$ for the $n = 43$ set, which exceeds the size of the search space.

Although there is a bit of fluctuation in the success rate with varying $w$, this seems to be due to the sample size. The values do seem to correlate with the probability of $g$ not changing the address of a single rotation. Equation (1) gives values of 0.625, 0.591, 0.592, 0.617 for the parameters $n = 23, 31, 37, 43$, respectively. Although this is not very strong evidence, it leads us to believe that this probability will remain similar as long as $q \approx 2d$.

Table 1. Results of 1000 trials for each set of parameters. For each set the experimental success rate and $(\log_2)$ runtime (in number of loops), as well as the theoretical $(\log_2)$ runtime (4.3) and the classic meet-in-the-middle $(\log_2)$ runtime and storage capacity (2.2) are given.

<table>
<thead>
<tr>
<th>$n$</th>
<th>$q$</th>
<th>$d$</th>
<th>$w$</th>
<th>Success</th>
<th>Experimental</th>
<th>$L^\ast$ (4.3)</th>
<th>$L$ (2.2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>23</td>
<td>17</td>
<td>8</td>
<td>$2^4$</td>
<td>0.53</td>
<td>12.82</td>
<td>12.85</td>
<td>6.40</td>
</tr>
<tr>
<td>23</td>
<td>17</td>
<td>8</td>
<td>$2^6$</td>
<td>0.50</td>
<td>11.61</td>
<td>11.85</td>
<td>6.40</td>
</tr>
<tr>
<td>23</td>
<td>17</td>
<td>8</td>
<td>$2^8$</td>
<td>0.53</td>
<td>10.56</td>
<td>10.85</td>
<td>6.40</td>
</tr>
<tr>
<td>31</td>
<td>19</td>
<td>10</td>
<td>$2^6$</td>
<td>0.44</td>
<td>16.02</td>
<td>16.48</td>
<td>9.35</td>
</tr>
<tr>
<td>31</td>
<td>19</td>
<td>10</td>
<td>$2^8$</td>
<td>0.45</td>
<td>14.59</td>
<td>15.48</td>
<td>9.35</td>
</tr>
<tr>
<td>31</td>
<td>19</td>
<td>10</td>
<td>$2^{10}$</td>
<td>0.45</td>
<td>14.60</td>
<td>14.48</td>
<td>9.35</td>
</tr>
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<td>31</td>
<td>19</td>
<td>10</td>
<td>$2^{12}$</td>
<td>0.42</td>
<td>13.44</td>
<td>13.48</td>
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<tr>
<td>37</td>
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<td>12</td>
<td>$2^{10}$</td>
<td>0.50</td>
<td>18.70</td>
<td>18.38</td>
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</tr>
<tr>
<td>37</td>
<td>23</td>
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<td>0.46</td>
<td>17.59</td>
<td>17.38</td>
<td>11.85</td>
</tr>
<tr>
<td>37</td>
<td>23</td>
<td>12</td>
<td>$2^{14}$</td>
<td>0.49</td>
<td>16.13</td>
<td>16.38</td>
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</tr>
<tr>
<td>37</td>
<td>23</td>
<td>12</td>
<td>$2^{16}$</td>
<td>0.46</td>
<td>15.99</td>
<td>15.38</td>
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<tr>
<td>43</td>
<td>29</td>
<td>14</td>
<td>$2^{14}$</td>
<td>0.55</td>
<td>20.57</td>
<td>20.26</td>
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<tr>
<td>43</td>
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<td>43</td>
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<td>14</td>
<td>$2^{18}$</td>
<td>0.56</td>
<td>18.68</td>
<td>18.26</td>
<td>14.39</td>
</tr>
</tbody>
</table>

6. Other considerations

Generalization of key forms. In this paper we assumed that the system uses binary form keys. This is the most basic version of the NTRU scheme, but other forms have been proposed. For various reasons other proposals include ternary polynomials [10, 11, 16], sampling keys from a discrete Gaussian distribution [25], keys with a set number of ones, minus ones, twos and minus twos [8], and so-called product form polynomials [13].

This attack will work for these forms. Although we chose to attack the keys by splitting them in the middle, the only requirement is that the key space $F$ can be rewritten as a decomposition $F = F_1 \oplus F_2$. This can also be done for the other forms. It does have to be taken into account that for some forms there might be fewer rotations to take advantage of; in these cases the running time will increase accordingly.

Rekeying. Given that the attack has a significant probability of failing, it would be possible to adapt the key generation algorithm of NTRU such that none of the rotations of $f$ and $g$ will cause addresses to collide. However, since we can choose any $d/2$-sized set of positions for $F_1$ and split the keyspace into $F_1$ and $F \setminus F_1 = F_2$, the attack will still work.
Also the attack could be adapted to succeed with probability 1. As discussed in §2.2, we can store the keys in all possible addresses that can occur by $g$ changing the sign of $-f_2h$. In our attack, we could add a few extra bits to $I$ that would not only map it to a polynomial, but also select a deterministic option out of its possible addresses. If the expected number of addresses that can occur is $A$, then we could add $\log_2(2A)$ bits to $I$. This would increase the search space by a factor of $2A$ and therefore also increase the expected running time by $(2A)^{3/2}$, but it would remove the possibility of failure.

**Optimization.** Many optimizations are possible to increase the speed of the algorithm further. The values for $\alpha$, the number of distinguished points to try before switching functions, as well as the method to replace distinguished points are taken from [27], but could be optimized for this particular attack. To remove the constraint of good rotations, the attack can be adapted to choose random polynomials of $d/2$ ones instead of splitting the space into two disjoint parts. Also, parallelization could easily be achieved by using a central server that collects the distinguished points and thus lets multiple cores give a linear speed-up in the runtime. Furthermore, as is for instance suggested in [2], it might be better not to replace the distinguished points randomly when the memory is full, but instead replace the ‘less-used’ points. The goal of this paper, however, was not to implement an optimized attack, but to show that the amount of memory that is used can be varied and, in particular, that this does not constitute a bottleneck.

**Hybrid attack.** The hybrid lattice-basis reduction and meet-in-the-middle attack [14] mentioned in the introduction works by taking the NTRU lattice and applying lattice-basis reduction to part of the matrix and then applying a meet-in-the-middle approach to the remaining key space. Although we did not write down the details of such an attack, it should be possible to apply the reduced memory algorithm of this paper to reduce the memory requirement for the hybrid attack.

For the ees251ep6 parameter set of NTRU, [14] estimated a storage cost of $2^{65.6}$ after applying the lattice reduction phase of the hybrid attack with $2^{56.7}$ loops of the algorithm. We conjecture that if the attack had to be run with only 1 TB storage available, then an attack similar to that described in this paper could be mounted at the cost of only a factor $\sim 2^{13}$ more running time.

We leave the details of the attack for possible future research, but if we do a quick analysis and we have a cap on the memory of $w$ triples, then, using Heuristic 1, the running time of the collision search on the meet-in-the-middle phase of the hybrid attack is equal to $2.5r\sqrt{(2^{H(p)}/N)^3/w}$. Here $r$ is the time for one iteration of the step function, which we will assume to be one multiplication, and $H(p)$ is the entropy of the search space calculated by [11, equation (9)], which is conservatively reduced by a factor of $N$ to take rotations into account.

**Present-day parameters.** The ees251ep6 parameter set from [14] is now outdated. In this concluding section we look at the currently advised parameters and what the results of this paper could mean for them.

The parameters currently in the EEES #1 standard for NTRU and their security analysis can be found in [11]. We summarize the results in the first six columns of Table 2. The hybrid MITM memory req. is the memory requirement for the meet-in-the-middle phase of the hybrid attack against the NTRU secret key, while the direct MITM memory req. is the memory needed to mount a meet-in-the-middle attack on the entire key space of NTRU.

The first thing we observe is that for the larger parameter sets, the classic meet-in-the-middle outperforms the hybrid attack and in this case the results of this paper are directly applicable. Secondly, the assumption in these security estimates is that memory is equal to running time, that is, storing an element in the attack and doing one multiplication are of equal cost in deciding the security level. However, one might consider other metrics. If we take
into account the estimate of [3], we have that $2^{70}$ bytes of storage is approximately 1500 times as expensive as $2^{80}$ floating point operations. This trend is reinforced by the fact that a year of computing power on a c4.8xlarge of the Amazon EC2 Cloud gives you $2^{60}$ operations and $2^{36}$ bits of storage for under $3000 ($0.34 per hour). On top of this, memory access can be much slower than computation; for example, sorting $n$ numbers uses energy proportional to $n^{1.5}$ while their computation is slightly superlinear in $n$. If we thus conservatively assume that a bit is at least $2^{10}$ times as expensive as an operation, then using a dollar metric will change the operation/memory proportion that will minimize the cost of the attack. The memory reduction techniques of this paper can achieve this.

Another argument is that the amount of storage available might be limited, due to either cost or availability. Previous security analysis did not suggest that in this case an attack can still be mounted.

As an example, suppose we cap the memory at storing $2^{10}$ times fewer triples than the number of operations needed for the attack. For the above parameters we reran the parameter generation algorithm of [11] (implementation [24]) with the heuristic running times given in this paper (implementation [29]). The results are given in the two rightmost columns of Table 2. The memory cap indicates the amount of triples that can be stored for either the hybrid or direct meet-in-the-middle attack on the NTRU secret key. The operation count estimates how many operations are required to find the when we have such a memory cap.

**Other schemes.** Another class of secret key cryptographic schemes that have so far withstood quantum attacks are based on the learning with errors (LWE) problem (see, for example, [4, 8, 17, 19]). It has recently been shown that the hybrid attack applies to LWE with binary error as well [5]. This paper shows that the storage capacity is not as unambiguous as portrayed in [14] and reduced-memory variation of the attack should be taken into account for parameter considerations of NTRU and LWE-based schemes.

### Table 2. An overview of NTRU parameters. All security and memory values are log$_2$ values.

<table>
<thead>
<tr>
<th>Sec. goal</th>
<th>$N$</th>
<th>$q$</th>
<th>Hybrid MITM estimate</th>
<th>Direct MITM memory req.</th>
<th>Memory cap</th>
<th>Operation count</th>
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<td>401</td>
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<td>743</td>
<td>2048</td>
<td>256</td>
<td>279</td>
<td>252</td>
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### References


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