# A Group Signature Scheme from Lattice Assumptions

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Abstract. Group signature schemes allow users to sign messages on behalf of a group while (1) maintaining anonymity (within that group) with respect to an outside observer, yet (2) ensuring traceability of a signer (by the group manager) when needed. In this work we give the first construction of a group signature scheme based on lattices (more precisely, the learning with errors assumption), in the random oracle model. Towards our goal, we construct a new algorithm for sampling a basis for an orthogonal lattice, together with a trapdoor, that may be of independent interest.

#### 1 Introduction

Group signature schemes [16] allow users to sign messages on behalf of a group administered by some manager. The group is initialized by having the group manager generate master public and secret keys; upon admission to the group, a user is given a personal secret key that is derived from the master secret key by the manager. A member of the group can sign a message using their personal secret key, enabling anyone who knows the master public key to verify that some group member signed the message. Roughly, group signatures are required to satisfy two seemingly contradictory requirements: given some legitimate group signature  $\sigma$ , the group manager should be able to determine which member of the group issued  $\sigma$  (traceability), but no one other than the group manager should be able to determine any information about the signer (anonymity). Group signatures have proven to be a popular primitive, and since their introduction several constructions have been proposed both with random oracles [5, 6, 13, 10, 14, 22] and without [8, 9, 4, 11, 12, 21].

While there exist constructions of group signature schemes based on trapdoor permutations [8,9], such schemes serve only as proofs of feasibility and are far from practical. On the other hand, practical schemes are based on a relatively small set of assumptions: namely, the strong RSA assumption [5,6,13,22] and

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various assumptions related to groups having an associated bilinear map [10, 14, 4, 11, 12, 21].

In this work we show the first construction of a group signature scheme from assumptions related to *lattices*. The use of lattice-based assumptions in cryptography has seen a flurry of activity in recent years. In part, this is due to a general desire to expand the set of assumptions on which cryptosystems can be based (i.e., beyond the standard set of assumptions related to the hardness of factoring and solving the discrete logarithm problem). Relying on lattice-based assumptions offers several concrete advantages as well: such assumptions are appealing because of the known worst-case/average-case connections between lattice problems, and also because lattice problems are currently immune to quantum attacks. Even restricting to classical attacks, the best-known algorithms for solving several lattice problems require exponential time (in contrast to the sub-exponential algorithms known, e.g., for factoring). Finally, relying on lattices can potentially yield efficient constructions because the basic lattice operations manipulate relatively small numbers and are inherently parallelizable.

While our resulting construction is less efficient than existing schemes based on number-theoretic assumptions, our construction is significantly more efficient than the generic approaches of [8,9] that rely on NIZK proofs based on a Karp reduction to some NP-complete language. (Peikert and Vaikuntanathan [26] construct NIZK proofs for specific lattice problems, however their results are not directly applicable to our work.)

#### 1.1 Our Techniques

Our construction combines ideas from several different works, tying these together using a new technical tool described below. At a high level, our group signature scheme follows a template similar (but not identical) to that of Bellare et al. [8]. The master public key in our scheme includes a public key  $pk_E$  for a public-key encryption scheme, along with n signature verification keys  $pk_1, \ldots, pk_N$ . The personal secret key given to the ith group member is  $sk_i$ , the signing key corresponding to  $pk_i$ . To sign a message M, the group member (1) signs M using  $sk_i$ ; (2) encrypts the resulting signature using  $pk_E$ ; and then (3) provides a NIZK proof of well-formedness (namely, that the given ciphertext encrypts a signature on M relative to one of the  $pk_i$ ). This implies anonymity (since no one other than the group manager knows the decryption key  $sk_E$  corresponding to  $pk_E$ ), yet ensures traceability because the group manager can decrypt the ciphertext that is included as part of any valid group signature.

To instantiate this approach using lattice-based assumptions, we need to identify candidate signature and encryption schemes along with an appropriate NIZK proof system. While constructions of the former based on lattices are known, we do not currently have constructions of NIZK for all of NP from lattice-based assumptions and we will therefore have to tailor our scheme so that it can rely on (efficient) NIZK proofs for some *specific* language. This is explained in more detail in what follows.

For the underlying signature scheme we use the GPV signature scheme [19] that works roughly as follows. The public key is a basis  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$  for a random lattice. To sign a message M, the signer uses a trapdoor  $\mathbf{T}$  to find a "short" vector  $\mathbf{e} \in \mathbb{Z}^m$  with  $\mathbf{A}\mathbf{e} = H(M)$  (where H is a hash function modeled as a random oracle). Under suitable assumptions, finding such a short vector  $\mathbf{e}$  without the trapdoor is hard.

We encrypt the resulting signature using what can be viewed as a non-standard variant of the Regev encryption scheme [27]. Given a matrix  $\mathbf{B} \in \mathbb{Z}_q^{n \times m}$  (viewed as a public key), we encrypt  $\mathbf{e} \in \mathbb{Z}^m$  by choosing a random vector  $\mathbf{s} \in \mathbb{Z}_q^n$  and outputting the ciphertext  $\mathbf{z} = \mathbf{B}^T \mathbf{s} + \mathbf{e}$ . Effectively,  $\mathbf{e}$  here is being used as the noise in an instance of the "learning with errors" (LWE) problem [27]. Before going further, we stress that this "encryption scheme" is *not* semantically secure. However, it turns out that we need something much weaker than semantic security in order to prove anonymity of our scheme; roughly, all we need is that the encryption of a uniformly random  $\mathbf{e} \in \mathbb{Z}_q^m$  is computationally indistinguishable from the encryption of a vector  $\mathbf{e}$  chosen from a certain discrete Gaussian distribution. We defer further discussion to Section 3.

As described thus far, our group signature scheme would have a master public key consisting of verification keys  $\mathbf{A}_1, \dots, \mathbf{A}_N$  along with an encryption key  $\mathbf{B}$ ; a signature would include  $\mathbf{z} = \mathbf{B}^T \mathbf{s} + \mathbf{e}$ , where  $\mathbf{e}$  is such that  $\mathbf{A}_i \mathbf{e} = H(M)$  for some i, along with a proof of well-formedness of the ciphertext z. Constructing the proof of well-formedness turns out to be the most difficult aspect of our work, and it will be useful to modify our scheme a bit in order to help construct this proof. (In doing so, we also rely on specific properties of the GPV signature scheme.) We change our scheme as follows: Now, the master public key contains N verification keys  $\mathbf{A}_1, \dots, \mathbf{A}_N$  (as before) and also N encryption keys  $\mathbf{B}_1, \dots, \mathbf{B}_N$ . To sign a message M, user i computes a real signature  $\mathbf{e}_i$  (using the trapdoor associated with  $\mathbf{A}_i$ ) and "pseudo-signatures"  $\mathbf{e}_j$  for all  $j \neq i$ . Each "pseudo-signature"  $\mathbf{e}_j$  has the property that  $\mathbf{A}_j \mathbf{e}_j = H(M)$ , however  $\mathbf{e}_j$  is not short (and thus not a valid signature). All the  $\{\mathbf{e}_j\}_{j=1}^N$  are then encrypted as before, with each  $\mathbf{e}_j$  being encrypted using  $\mathbf{B}_j$  to give a ciphertext  $\mathbf{z}_j$ . We then have the signer provide a proof that (1) each  $\mathbf{z}_i$  encrypts a correct pseudosignature with respect to  $A_i$ , and (2) at least one of these pseudo-signatures is in fact short (and hence a valid signature). Further details are given next.

To provide a way for the signer to prove that every ciphertext  $\mathbf{z}_j$  encrypts a pseudo-signature, we develop a new technical tool that we believe to be of independent interest: a way to sample a basis for an *orthogonal lattice* with its associated trapdoor.<sup>1</sup> Specifically, we show a technique that, given a matrix  $\mathbf{B}$ , generates  $(\mathbf{A}, \mathbf{T})$  such that  $\mathbf{A}\mathbf{B}^T = 0 \pmod{q}$  and  $\mathbf{T}$  is still a "good trapdoor" (in the sense required for GPV signatures) for  $\mathbf{A}$ . If we use matrices  $\{\mathbf{A}_i\}$  generated in this way as verification keys in the group signature scheme described earlier, then it is possible to verify that a given ciphertext  $\mathbf{z}_j$  encrypts a pseudo-signature with respect to  $\mathbf{A}_j$  by checking whether  $\mathbf{A}_j\mathbf{z}_j \stackrel{?}{=} H(M)$ . This works

<sup>&</sup>lt;sup>1</sup> For our definition of an orthogonal lattice, see Section 2.

because

$$\mathbf{A}_{j}\mathbf{z}_{j} = \mathbf{A}_{j} \cdot \left(\mathbf{B}_{j}^{T}\mathbf{s}_{j} + \mathbf{e}_{j}\right) = \mathbf{A}_{j}\mathbf{e}_{j} = H(M)$$

by construction.

The only thing that remains is to provide a proof that at least one of the  $\mathbf{z}_j$  encrypts a vector  $\mathbf{e}_j$  that is also *short*. This translates to proving that at least one of the vectors  $\mathbf{z}_j = \mathbf{B}_j^T \mathbf{s}_j + \mathbf{e}_j$  is "close to" the lattice generated by the columns of  $\mathbf{B}_j^T$ . This can be done using the (statistical) zero-knowledge protocol demonstrated by Micciancio and Vadhan [23], coupled with standard techniques [17, 18] for making the proof witness indistinguishable and noninteractive in the random oracle model.

In essence, we obtain our efficiency gain by coupling together the encryption and the signature components so that the NIZK proof system we need to use is for a very simple language.

#### 1.2 Outline of the Paper

We introduce some notation and review the necessary background on lattices in Section 2. For the reader who is already familiar with lattices, we highlight the following aspects of our treatment that are new to this work:

- In Section 2.2 (cf. Lemma 1) and in the rest of the paper, we consider the LWE problem under a non-standard error distribution. Fortunately, a recent result of Peikert [25] demonstrates that the hardness of the LWE problem under this distribution is implied by standard hardness results.
- In Section 2.4 we describe a technique for sampling a basis for an orthogonal lattice and its associated trapdoor.

We turn to group signatures in Section 3. We review the standard definitions of security for group signature schemes in Section 3.1, describe our construction in Section 3.2, and prove anonymity and traceability in Sections 3.3 and 3.4.

#### 2 Preliminaries on Lattices

Throughout, we use n for the security parameter; other parameters are taken to be functions of n. When we say "statistically close" we mean "within statistical difference negligible in n."

We review some basic properties of lattices as used in prior work. This section is included mainly to fix notation and ideas, and we refer to the original papers (cited below) for further exposition.

We use bold lower-case letters (e.g.,  $\mathbf{x}$ ) to denote vectors, and bold uppercase letters (e.g.,  $\mathbf{B}$ ) to denote matrices. (Our vectors are always column vectors.) We let  $||\mathbf{x}||$  denote the Euclidean (i.e.,  $\ell_2$ ) norm of the vector  $\mathbf{x}$ , and let  $||\mathbf{B}||$ denote the maximum of the Euclidean norms of the columns of  $\mathbf{B}$ ; i.e., if  $\mathbf{B} = (\mathbf{b}_1|\cdots|\mathbf{b}_n)$  then  $||\mathbf{B}|| \stackrel{\text{def}}{=} \max_i ||\mathbf{b}_i||$ . If  $x \in \mathbb{R}$ , then  $\lfloor x \rfloor$  denotes the rounding of x to the nearest integer. For q an integer,  $\mathbb{Z}_q$  denotes the standard group of integers modulo q. We will extend modular arithmetic to the reals in the obvious way: for example, for  $q \in \mathbb{Z}^+$  and  $x \in \mathbb{R}$  we use  $x \mod q$  to represent the unique real number  $y \in [0,q)$  such that x-y is an integer multiple of q. Finally, we define a notion of distance between elements in  $\mathbb{Z}_q$  in the natural way: given  $x,y \in \mathbb{Z}_q$ , their distance is defined by mapping  $(x-y) \mod q$  to the set of integers  $\{-\lfloor q/2 \rfloor, \ldots, \lfloor q/2 \rfloor\}$  and then taking the absolute value of the result.

Fixing q and given a matrix  $\mathbf{B} \in \mathbb{Z}_q^{n \times m}$ , we define the m-dimensional lattice  $\mathcal{L}(\mathbf{B}^T)$  as  $\mathcal{L}(\mathbf{B}^T) \stackrel{\text{def}}{=} \{ y \in \mathbb{Z}^m \mid y \equiv \mathbf{B}^T \mathbf{s} \bmod q \text{ for some } \mathbf{s} \in \mathbb{Z}^n \}$ . We define the orthogonal lattice  $\Lambda^{\perp}(\mathbf{B})$  as  $\Lambda^{\perp}(\mathbf{B}) \stackrel{\text{def}}{=} \{ \mathbf{w} \in \mathbb{Z}^m \mid \mathbf{B} \cdot \mathbf{w} = 0 \bmod q \}$ . (Note that the notion of an orthogonal lattice is defined differently in some previous work.) Finally, for a vector  $\mathbf{z} \in \mathbb{Z}_q^m$  we define

$$\mathsf{dist}(\mathcal{L}(\mathbf{B}^T), \mathbf{z}) \stackrel{\mathrm{def}}{=} \min_{\mathbf{s} \in \mathbb{Z}_q^n} ||\mathbf{B}^T \mathbf{s} - \mathbf{z}||.$$

In other words,  $dist(\mathcal{L}(\mathbf{B}^T), \mathbf{z})$  is the distance of  $\mathbf{z}$  from the lattice spanned by the columns of  $\mathbf{B}^T$ .

#### 2.1 Gaussian Error Distributions

The one-dimensional (continuous) Gaussian distribution over  $\mathbb{R}$ , parameterized by  $s \in \mathbb{R}^+$ , is defined by the density function

$$\forall x \in \mathbb{R}: \quad D_s(x) = 1/s \cdot \exp(-\pi(x/s)^2).$$

In this work we always use a truncated Gaussian, i.e., the Gaussian distribution  $D_s$  whose support is restricted to numbers  $x \in \mathbb{R}$  such that  $|x| < s \cdot \omega(\sqrt{\log n})$ . The truncated and non-truncated distributions are statistically close, and we drop the word "truncated" from now on. The m-dimensional continuous Gaussian distribution is defined in a similar way, by the density function  $D_s(\mathbf{x}) = 1/s^m \cdot \exp(-\pi(||\mathbf{x}||/s)^2)$ . Finally, we denote by  $D_{s,\mathbf{c}}$  the m-dimensional continuous Gaussian distribution centered at the point  $\mathbf{c} \in \mathbb{R}^m$ . i.e.,  $D_{s,\mathbf{c}}(\mathbf{x}) = 1/s^m \cdot \exp(-\pi(||\mathbf{x}-\mathbf{c}||/s)^2)$ .

Let  $\Lambda \subseteq \mathbb{Z}^m$  be a lattice. The discrete Gaussian distribution  $D_{\Lambda,s,\mathbf{c}}$  is the m-dimensional Gaussian distribution centered at  $\mathbf{c}$ , but with support restricted to the lattice  $\Lambda$ . (We write  $D_{\Lambda,s}$  as shorthand for  $D_{\Lambda,s,\mathbf{0}}$ .) Formally, the density function of the discrete Gaussian distribution is defined as

$$\forall \mathbf{x} \in \Lambda : \qquad D_{\Lambda, s, \mathbf{c}}(\mathbf{x}) = \frac{D_{s, \mathbf{c}}(\mathbf{x})}{\sum_{\mathbf{y} \in \Lambda} D_{s, \mathbf{c}}(\mathbf{y})}.$$

Gentry et al. [19] show that given a basis **B** for  $\Lambda$ , this distribution can be sampled efficiently (to within negligible statistical distance) for  $s \ge ||\mathbf{B}|| \cdot \omega(\sqrt{\log n})$ .

#### 2.2 The Learning with Errors Problem

The "learning with errors" (LWE) problem was introduced by Regev [27] as a generalization of the "learning parity with noise" problem. We describe the problem in a form suitable for our applications in this paper.

Fix a positive integer n, integers  $m \geq n$  and  $q \geq 2$ , a vector  $\mathbf{s} \in \mathbb{Z}_q^n$ , and a probability distribution  $\chi$  on the interval  $[0,q)^m$ . Define the following two distributions over  $\mathbb{Z}_q^{n \times m} \times [0,q)^m$ :

- LWE<sub> $m,q,\chi$ </sub>(**s**) is the distribution obtained by choosing uniform  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$ , sampling  $\mathbf{e} \leftarrow \chi$ , and outputting  $(\mathbf{A}, \mathbf{A}^T \mathbf{s} + \mathbf{e} \bmod q)$ .
- $U_{m,q}$  is the distribution obtained by choosing uniform  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$  and uniform  $\mathbf{y} \in [0,q)^m$ , and outputting  $(\mathbf{A},\mathbf{y})$ .

The decisional variant of the LWE problem (relative to the distribution  $\chi$ ) can be stated informally as the problem of distinguishing between  $U_{m,q}$  and  $\mathsf{LWE}_{m,q,\chi}(\mathbf{s})$  for a uniform  $\mathbf{s}$ . Formally, for m,q, and  $\chi$  that may depend on n (viewed now as a security parameter) we say the  $\mathsf{LWE}_{m,q,\chi}$  problem is hard if the following is negligible for any probabilistic polynomial-time algorithm D:

$$\begin{aligned} \left| \Pr[\mathbf{s} \leftarrow \mathbb{Z}_q^n; (\mathbf{A}, \mathbf{y}) \leftarrow \mathsf{LWE}_{m, q, \chi}(\mathbf{s}) : D(\mathbf{A}, \mathbf{y}) = 1] \right. \\ &- \left. \Pr[(\mathbf{A}, \mathbf{y}) \leftarrow U_{m, q} : D(\mathbf{A}, \mathbf{y}) = 1] \right|. \end{aligned}$$

A standard setting for the LWE problem considers the error distribution  $\Psi_{\alpha}^{m}$  over  $[0,q)^{m}$  defined as follows: Sample m numbers  $\eta_{1},\ldots,\eta_{m}\leftarrow D_{\alpha}$ , let  $e_{i}:=q\cdot\eta_{i}\pmod{q}$ , and output  $\mathbf{e}:=(e_{1},\ldots,e_{m})^{T}$ . We write  $\mathsf{LWE}_{m,q,\alpha}(\mathbf{s})$  as an abbreviation for  $\mathsf{LWE}_{m,q,\Psi_{\alpha}^{m}}(\mathbf{s})$ .

Evidence for the hardness of the  $\mathsf{LWE}_{m,q,\alpha}$  problem comes from a result of Regev [27], who gave a quantum reduction from approximating certain lattice problems to within a factor of  $\widetilde{O}(n/\alpha)$  on n-dimensional lattices in the worst case to solving  $\mathsf{LWE}_{m,q,\alpha}$ , subject to the condition that  $\alpha \cdot q > 2\sqrt{n}$ . Recently, Peikert [24] gave a classical reduction for similar parameters. For our purposes, we note that the  $\mathsf{LWE}_{m,q,\alpha}$  problem is believed to be hard — given the state-of-the-art in lattice algorithms — for any  $m,q = \mathsf{poly}(n)$  and  $\alpha = 1/\mathsf{poly}(n)$  (subject to the above condition).

A second error distribution for the LWE problem<sup>2</sup> — and one that we will use in this paper — is the discrete Gaussian distribution  $D_{\mathbb{Z}^m,s}$  (mod q). Although this distribution may seem similar to a discretized (rounded) version of  $\Psi^m_{\alpha}$ , these distributions are statistically far from each other and thus we cannot immediately conclude anything about the hardness of the LWE problem with respect to one distribution from hardness of the LWE problem with respect to the other. Fortunately, a recent result of Peikert [25] can be used to show that hardness of the LWE problem with respect to error distribution  $D_{\mathbb{Z}^m,\alpha\cdot q\cdot \sqrt{2}}$  is

<sup>&</sup>lt;sup>2</sup> When using a discrete error distribution  $\chi$  over  $\mathbb{Z}_q^m$  (rather than a continuous distribution over  $[0,q)^m$ ), the LWE problem is to distinguish LWE<sub> $m,q,\chi$ </sub> from the uniform distribution over  $\mathbb{Z}_q^{n\times m}\times\mathbb{Z}_q^m$  (rather than  $\mathbb{Z}_q^{n\times m}\times[0,q)^m$ ).

implied by hardness of the LWE problem with respect to error distribution  $\Psi^m_{\alpha}$ . We write  $\widehat{\mathsf{LWE}}_{m,q,\alpha q\sqrt{2}}$  as an abbreviation for  $\mathsf{LWE}_{m,q,D_{\mathbb{Z}^m,\alpha q\sqrt{2}}}$ 

**Lemma 1.** For any  $\alpha$ , hardness of the LWE<sub>m,q,\alpha</sub> problem implies hardness of the LWE<sub>m,q,\alpha\sqrt{2}</sub> problem.

*Proof.* We show an efficient transformation T that takes as input  $(\mathbf{A}, \mathbf{y}) \in \mathbb{Z}_q^{n \times m} \times [0, q)^m$  and has the following properties:

- If  $(\mathbf{A}, \mathbf{y})$  is uniform over  $\mathbb{Z}_q^{n \times m} \times [0, q)^m$  then the output  $T(\mathbf{A}, \mathbf{y})$  is uniform over  $\mathbb{Z}_q^{n \times m} \times \mathbb{Z}_q^m$ .
- If  $(\mathbf{A}, \mathbf{y})$  is distributed according to  $\mathsf{LWE}_{m,q,\alpha}(\mathbf{s})$  then  $T(\mathbf{A}, \mathbf{y})$  is distributed according to  $\widehat{\mathsf{LWE}}_{m,q,\alpha q}\sqrt{2}(\mathbf{s})$ .

The lemma follows immediately from these two properties.

The transformation T works as follows. Given  $(\mathbf{A}, \mathbf{y})$ , it samples a vector  $\mathbf{w} \leftarrow D_{\mathbb{Z}^m - \mathbf{v}, \alpha q}$  and outputs the pair  $(\mathbf{A}, \mathbf{y} + \mathbf{w} \pmod{q})$ .

First, say  $(\mathbf{A}, \mathbf{y})$  is distributed uniformly over  $\mathbb{Z}_q^{n \times m} \times [0, q)^m$ . Note that  $\mathbf{y} + \mathbf{w}$  is always an integer, and the distribution  $\mathbf{w} \leftarrow D_{\mathbb{Z}^m - \mathbf{y}, \alpha q}$  depends only on the fractional part of each entry in  $\mathbf{y}$ . It follows that  $(\mathbf{A}, \mathbf{y} + \mathbf{w} \pmod{q})$  is distributed uniformly over  $\mathbb{Z}_q^{n \times m} \times \mathbb{Z}_q^m$ .

On the other hand, say  $\mathbf{y} = \mathbf{A}^T \mathbf{s} + \mathbf{e} \pmod{q}$  where  $\mathbf{e} \sim \Psi_{\alpha}^m$ . Since we have  $\mathbf{A}^T \mathbf{s} \in \mathbb{Z}^m$ , sampling  $\mathbf{w} \sim D_{\mathbb{Z}^m - \mathbf{y}, \alpha q} \pmod{q}$  is equivalent to sampling  $\mathbf{w} \sim D_{\mathbb{Z}^m - \mathbf{e}, \alpha q} \pmod{q}$ . A recent theorem of Peikert [25, Theorem 3.1] shows that the following two processes produce statistically close distributions:

- Sampling  $\mathbf{e} \sim \Psi_{\alpha}^{m}$  and then setting  $\mathbf{e}' = \mathbf{e} + D_{\mathbb{Z}^{m} \mathbf{e}, \alpha q} \pmod{q}$ ;
- Sampling  $\mathbf{e}' \sim D_{\mathbb{Z}^m \cap qq\sqrt{2}} \pmod{q}$ .

We conclude that the output  $T(\mathbf{A}, \mathbf{y}) = (\mathbf{A}, \mathbf{A}^T \mathbf{s} + (\mathbf{e} + \mathbf{w}) \pmod{q})$  is distributed according to  $\widehat{\mathsf{LWE}}_{m,q,\alpha q\sqrt{2}}(\mathbf{s})$ .

### 2.3 Trapdoor Functions and the GPV Signature Scheme

Ajtai [2] and Alwen and Peikert [3] show algorithms that generate an almost uniform matrix  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$  together with a "trapdoor" matrix  $\mathbf{T} \in \mathbb{Z}^{m \times m}$  satisfying the following conditions:

**Lemma 2 ([3]).** There is a probabilistic polynomial-time algorithm TrapSamp that, on input  $1^n, 1^m, q$  with  $q \ge 2$  and  $m \ge 8n \log q$ , outputs matrices  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$  and  $\mathbf{T} \in \mathbb{Z}^{m \times m}$  such that:

- The distribution on **A** as output by TrapSamp is statistically close to uniform over  $\mathbb{Z}_q^{n \times m}$ ,
- the columns of **T** form a basis of the lattice  $\Lambda^{\perp}(\mathbf{A})$ , implying in particular  $\mathbf{A} \cdot \mathbf{T} = 0 \pmod{q}$ ,
- $\|\mathbf{T}\| \le 5\sqrt{n\log q}.$

Given an "LWE instance"  $(\mathbf{A}, \mathbf{y} = \mathbf{A}^T \mathbf{s} + \mathbf{e})$  for a "short" vector  $\mathbf{e}$ , knowledge of  $\mathbf{T}$  can be used to recover  $\mathbf{s}$ . Specifically, if  $||\mathbf{T}|| < L = 5\sqrt{n \log q}$  (as guaranteed by Lemma 2) and  $\mathbf{e}$  is drawn from  $\Psi_{\alpha}^m$  for  $\alpha \leq 1/(L \cdot \omega(\sqrt{\log n}))$ , then  $\mathbf{s}$  can be easily recovered. This is done by first computing

$$\mathbf{T}^T \mathbf{y} \pmod{q} = \mathbf{T}^T (\mathbf{A}^T \mathbf{s} + \mathbf{e}) \pmod{q} = (\mathbf{A}\mathbf{T})^T \mathbf{s} + \mathbf{T}^T \mathbf{e} \pmod{q}$$
  
=  $\mathbf{T}^T \mathbf{e} \pmod{q}$ .

Since both **T** and **e** contain only "small" entries, each entry of the vector  $\mathbf{T}^T \mathbf{e}$  is smaller than q and thus  $\mathbf{T}^T \mathbf{e}$  (mod q) is equal to  $\mathbf{T}^T \mathbf{e}$  (over the integers). Multiplying by  $(\mathbf{T}^T)^{-1}$  thus gives  $\mathbf{e}$ , after which it is easy to recover  $\mathbf{s}$ .

The GPV signature scheme. Gentry, Peikert, and Vaikuntanathan [19] showed how to use the trapdoor sampling procedure described above to construct a one-way preimage-sampleable function. This can then be turned into a digital signature scheme using an "FDH-like" construction [7]. (See [19] for a formal definition of preimage-sampleable functions and the construction of the signature scheme.) Here, we describe how the preimage-sampleable function works.

Take  $q = \mathsf{poly}(n)$ ,  $m \ge 5n \log q$ , and  $s \ge 5\sqrt{n \log q} \cdot \omega(\sqrt{\log n})$ . The one-way preimage-sampleable function is defined by the following algorithms:

- GPVGen(1<sup>n</sup>) runs TrapSamp(1<sup>n</sup>, 1<sup>m</sup>, q) to obtain (**A**, **T**). The matrix **A** (and q) defines the function  $f_{\mathbf{A}}(\mathbf{e}) = \mathbf{A}\mathbf{e} \pmod{q}$ , with domain  $\{\mathbf{e} \in \mathbb{Z}^m : ||\mathbf{e}|| \le s\sqrt{m}\}$  and range  $\mathbb{Z}_q^n$ . Hardness of inversion is with respect to the distribution  $D_{\mathbb{Z}^m,s}$  over the domain.
- The trapdoor inversion algorithm GPVInvert( $\mathbf{A}, \mathbf{T}, s, \mathbf{u}$ ) samples from  $f_{\mathbf{A}}^{-1}(\mathbf{u})$  as follows: first, it computes (using standard linear algebra) an arbitrary  $\mathbf{t} \in \mathbb{Z}^m$  such that  $\mathbf{A}\mathbf{t} = \mathbf{u} \pmod{q}$  (except for a negligible fraction of  $\mathbf{A}$ , such a  $\mathbf{t}$  always exists). Then it samples  $\mathbf{v} \leftarrow D_{\Lambda^{\perp}(\mathbf{A}),s,-\mathbf{t}}$  and outputs  $\mathbf{e} = \mathbf{t} + \mathbf{v}$ .

Gentry et al. show that the above is one-way if  $\mathsf{GapSVP}_{\gamma}$  is hard for polynomial approximation factor  $\gamma$ .

#### 2.4 Sampling a Basis for an Orthogonal Lattice with Trapdoor

We show a variant of the trapdoor sampling algorithm described in Lemma 2. In our variant, the algorithm is additionally given a matrix  $\mathbf{B} \in \mathbb{Z}_q^{n \times m}$  and (informally) should output a matrix  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$  with an associated trapdoor  $\mathbf{T} \in \mathbb{Z}^{m \times m}$  with the additional requirement that the rows of  $\mathbf{A}$  are orthogonal (over  $\mathbb{Z}_q$ ) to the rows of  $\mathbf{B}$ . In other words, we require that  $\mathbf{A}\mathbf{B}^T = \mathbf{0} \pmod{q}$ .

Overview of the construction. The basic idea is as follows. Write  ${\bf B}$  as

$$\mathbf{B}^T = \begin{pmatrix} \mathbf{B}_1 \\ \mathbf{B}_2 \end{pmatrix},$$

with  $\mathbf{B}_2$  a square, invertible matrix of dimension  $n \times n$ . We then generate an orthogonal matrix  $\mathbf{A} = [\mathbf{A}_1 \mid \mathbf{A}_2]$  in two steps. We generate the first component  $\mathbf{A}_1$  using the TrapSamp protocol. Recall, this returns a matrix that is statistically close to uniform, along with an associated trapdoor  $\mathbf{T}_1$ . Once we have chosen  $\mathbf{A}_1$  the second component  $\mathbf{A}_2$  is constrained to a fixed value by the requirement that  $\mathbf{A}\mathbf{B}^T = \mathbf{0} \pmod{q}$ ; we thus generate  $\mathbf{A}_2$  by solving the linear equations that define this constraint.

All that remains is to find a trapdoor  $\mathbf{T}$  such that the columns of  $\mathbf{T}$  are short and  $\mathbf{A} \cdot \mathbf{T} = \mathbf{0}$ . Here we rely on the recent basis delegation techniques of Cash et al. [15], which allows us to "extend" the basis  $\mathbf{T}_1$  into a larger basis  $\mathbf{T}$  for  $\Lambda^{\perp}(\mathbf{A})$  as desired. The details follow.

**Lemma 3.** There is a probabilistic polynomial-time algorithm OrthoSamp that on input  $1^n, 1^m, q$  (with  $q \ge 2$  and  $m \ge 5n \log q$ ) and a matrix  $\mathbf{B} \in \mathbb{Z}_q^{n \times m}$  whose columns span  $\mathbb{Z}_q^n$ , outputs matrices  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$  and  $\mathbf{T} \in \mathbb{Z}^{m \times m}$  such that:

- $-\mathbf{A}\mathbf{B}^T = \mathbf{0} \pmod{q}$ . Moreover, the distribution on  $\mathbf{A}$  is statistically close to uniform over  $\mathbb{Z}_q^{n \times m}$ , subject to this condition,
- the columns of  $\mathbf{T}$  form a basis of the lattice  $\Lambda^{\perp}(\mathbf{A})$ , implying in particular  $\mathbf{A} \cdot \mathbf{T} = 0 \pmod{q}$ ,
- $\|\mathbf{T}\| \le 5\sqrt{n\log q}.$

*Proof.* Let  $m_1 = 3n \log q$  and  $m_2 = 2n \log q$ . Write

$$\mathbf{B}^T = \begin{pmatrix} \mathbf{B}_1 \\ \mathbf{B}_2 \end{pmatrix},$$

where  $\mathbf{B}_1 \in \mathbb{Z}_q^{m_1 \times n}$  and  $\mathbf{B}_2 \in \mathbb{Z}_q^{m_2 \times n}$ . The algorithm OrthoSamp works as follows:

- 1. If the rows of  $\mathbf{B}_2$  do not span  $\mathbb{Z}_q^n$ , output  $\perp$ . This occurs with only negligible probability [1,27]
- 2. Compute  $(\mathbf{A}_1, \mathbf{T}_1) \leftarrow \mathsf{TrapSamp}(1^n, 1^{m_1}, q)$ . Let  $\mathbf{A}_2 \in \mathbb{Z}_q^{n \times m_2}$  be a uniformly random matrix satisfying

$$\mathbf{A}_2\mathbf{B}_2 = -\mathbf{A}_1\mathbf{B}_1 \pmod{q}$$

(such a matrix exists by the assumption that the rows of  $\mathbf{B}_2$  span  $\mathbb{Z}_q^n$ ). If the columns of  $\mathbf{A}_1$  do not span  $\mathbb{Z}_q^n$ , output  $\perp$ . This occurs only with negligible probability.

3. Extend the basis  $\mathbf{T}_1$  into basis  $\mathbf{T} \in \mathbb{Z}_q^{m \times m}$  for  $\Lambda^{\perp}(\mathbf{A})$  using the technique of Cash et al. [15, Lemma 3.2]. We present their technique for completeness. Let  $\mathbf{T}$  be of the form

$$\mathbf{T} = \left(egin{array}{cc} \mathbf{T}_1 \ \mathbf{W} \ \mathbf{0} & \mathbf{I} \end{array}
ight)$$

where  $\mathbf{W} \in \mathbb{Z}_q^{m_1 \times m_2}$  is an arbitrary matrix satisfying  $\mathbf{A}_1 \mathbf{W} = -\mathbf{A}_2$ , and  $\mathbf{I} \in \mathbb{Z}_q^{m_2 \times m_2}$  is the identity matrix. (Note that  $\mathbf{W}$  exists by the assumption that the columns of  $\mathbf{A}_1$  span  $\mathbb{Z}_q^n$ .) Output  $\mathbf{A} = [\mathbf{A}_1 \mid \mathbf{A}_2]$  and  $\mathbf{T}$ .

We now verify that this algorithm satisfies the required properties. First observe that

$$\mathbf{A}\mathbf{B}^T = \mathbf{A}_1\mathbf{B}_1 + \mathbf{A}_2\mathbf{B}_2 = \mathbf{A}_1\mathbf{B}_1 - \mathbf{A}_1\mathbf{B}_1 = \mathbf{0} \pmod{q}.$$

The claim regarding the distribution of  ${\bf A}$  follows directly from the construction. We also have

$$\mathbf{A} \cdot \mathbf{T} = [\mathbf{A}_1 | \mathbf{A}_2] \cdot \begin{pmatrix} \mathbf{T}_1 & \mathbf{W} \\ \mathbf{0} & \mathbf{I} \end{pmatrix}$$
$$= [\mathbf{A}_1 \mathbf{T}_1 + \mathbf{A}_2 \mathbf{0} | \mathbf{A}_1 \mathbf{W} + \mathbf{A}_2]$$
$$= \mathbf{0} \pmod{q},$$

where the final equality holds because  $\mathbf{A}_1\mathbf{T}_1 = \mathbf{0}$  by the properties of TrapSamp, and  $\mathbf{A}_1\mathbf{W} = -\mathbf{A}_2$  by construction. Finally, we refer the reader to the work of Cash et al. [15] for a proof that  $||T|| \leq 5\sqrt{n\log q}$ 

The following corollary follows from the above construction, and will be used in the security proof of our signature scheme.

Corollary 1. The distributions

$$\left\{\mathbf{B} \leftarrow \mathbb{Z}_q^{n \times m}; \mathbf{A} \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{B}) : (\mathbf{A}, \mathbf{B})\right\}$$

and

$$\left\{\mathbf{A} \leftarrow \mathbb{Z}_q^{n \times m}; \mathbf{B} \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{A}) : (\mathbf{A}, \mathbf{B})\right\}$$

are statistically close.

#### 2.5 Efficient NIWI Proofs for Lattice Problems

Let  $\mathbf{B}_1, \dots, \mathbf{B}_N \in \mathbb{Z}_q^{n \times m}$  and  $\mathbf{z}_1, \dots, \mathbf{z}_N \in \mathbb{Z}_q^m$ . In this section we briefly describe how it is possible to construct a noninteractive witness-indistinguishable (NIWI) proof (in the random oracle model) for the gap language  $L_{s,\gamma} = (L_{YES}, L_{NO})$  defined by:

$$L_{YES} = \left\{ \begin{pmatrix} \mathbf{B}_1, \dots, \mathbf{B}_N \\ \mathbf{z}_1, \dots, \mathbf{z}_N \end{pmatrix} \middle| \exists \mathbf{s} \in \mathbb{Z}_q^n \text{ and } i \in [N] : ||\mathbf{z}_i - \mathbf{B}_i^T \mathbf{s}|| \le s\sqrt{m} \right\}$$
$$L_{NO} = \left\{ \begin{pmatrix} \mathbf{B}_1, \dots, \mathbf{B}_N \\ \mathbf{z}_1, \dots, \mathbf{z}_N \end{pmatrix} \middle| \forall \mathbf{s} \in \mathbb{Z}_q^n \text{ and } i \in [N] : ||\mathbf{z}_i - \mathbf{B}_i^T \mathbf{s}|| > \gamma \cdot s\sqrt{m} \right\}.$$

Here,  $L_{YES}$  is a collection of N points at least one of which is close to the corresponding lattice, and  $L_{NO}$  is a collection of N points all of which are far from the corresponding lattices.

Our starting point is an (interactive) witness-indistinguishable (WI) proof system for the gap version of the closest vector problem, i.e., for the language  $L'_{\gamma} = \{L'_{YES}, L'_{NO}\}$  [20, 23]:

$$L'_{YES} = \{ (\mathbf{B}, \mathbf{z}, t) \mid \exists \mathbf{s} : ||\mathbf{z} - \mathbf{B}^T \mathbf{s}|| \le t \}.$$

$$L'_{NO} = \{ (\mathbf{B}, \mathbf{z}, t) \mid \forall \mathbf{s} : ||\mathbf{z} - \mathbf{B}^T \mathbf{s}|| > \gamma \cdot t \}.$$

Our language  $L_{s,\gamma}$  can be described as the OR of several instance of  $L'_{\gamma}$ ; that is,

$$\begin{pmatrix} \mathbf{B}_1, \dots, \mathbf{B}_N \\ \mathbf{z}_1, \dots, \mathbf{z}_N \end{pmatrix} \in L_{YES} \Leftrightarrow \bigvee_i \big( (\mathbf{B}_i, \mathbf{z}_i, s\sqrt{m}) \in L_{YES}' \big).$$

$$\begin{pmatrix} \mathbf{B}_1, \dots, \mathbf{B}_N \\ \mathbf{z}_1, \dots, \mathbf{z}_N \end{pmatrix} \in L_{NO} \Leftrightarrow \bigwedge_i \left( (\mathbf{B}_i, \mathbf{z}_i, s\sqrt{m}) \in L'_{NO} \right).$$

We can thus use the techniques of Cramer, Damgård, and Schoenmakers [17] to obtain an interactive WI proof for  $L_{s,\gamma}$  with negligible soundness error. Using the Fiat-Shamir transformation [18], the resulting protocol can be made non-interactive in the random oracle model.

We remark that for our application we only require soundness (and do not require the proof system to be a proof of knowledge) and witness indistinguishability (rather than zero knowledge). The observations in this section are summarized in the following lemma.

**Lemma 4.** Let  $\gamma \geq O(\sqrt{m/\log m})$ . Then there is a noninteractive witness-indistinguishable proof system for the language  $L_{s,\gamma}$  in the random oracle model, where the length of the proof is  $O(mnN\log q)$  bits.

### 3 A Group Signature Scheme Based on Lattices

#### 3.1 Definitions

We adopt the definition of group signature schemes from the work of Bellare, Micciancio, and Warinschi [8], with the relaxation suggested by Boneh, Boyen, and Shacham [10] (and considered also in, e.g., [11]). Formally, a group signature scheme  $\mathcal{GS} = (G.KeyGen, G.Sign, G.Vrfy, G.Open)$  is a collection of four polynomial-time algorithms defined as follows.

- The group key-generation algorithm  $\mathsf{G.KeyGen}(1^n,1^N)$  is a randomized algorithm that takes a security parameter  $1^n$  and the group size  $1^N$  as input, and outputs  $(\mathsf{PK},\mathsf{TK},\mathsf{gsk})$ , where  $\mathsf{PK}$  is the group public key,  $\mathsf{TK}$  is the group manager's tracing key, and  $\mathsf{gsk}$  is a vector of N signing keys with  $\mathsf{gsk}[i]$  being the signing key given to the  $i^{th}$  group member.
- The group signature algorithm  $\mathsf{G.Sign}(\mathsf{gsk}[i], M)$  is a randomized algorithm that takes as input a secret signing key  $\mathsf{gsk}[i]$  and a message M, and outputs a signature  $\sigma$ .
- The group signature verification algorithm  $\mathsf{G.Vrfy}(\mathsf{PK}, M, \sigma)$  is a deterministic algorithm that takes as input the group public key  $\mathsf{PK}$ , a message M, and a signature  $\sigma$ , and outputs either 1 or 0 (signifying accept or reject, respectively).
- The opening algorithm  $G.Open(TK, M, \sigma)$  is a deterministic algorithm that takes as input the tracing key TK, a message M, and a signature  $\sigma$ , and outputs an identity  $i \in [N]$ .

The basic consistency requirements of a group signature scheme are that an honest signature generated by a group member should be accepted as correct, and must be traceable to the group member who issued it. That is, for any  $(\mathsf{PK},\mathsf{TK},\mathsf{gsk})$  output by  $\mathsf{G.KeyGen}(1^n,1^N)$ , any M, and any  $i\in[N]$ , if  $\sigma\leftarrow\mathsf{G.Sign}(\mathsf{gsk}[i],M)$  then

$$G.Vrfy(PK, M, \sigma) = 1$$
 and  $G.Open(TK, M, \sigma) = i$ ,

except with negligible probability over the entire experiment.

Group signature schemes are also required to satisfy two basic security properties: anonymity and traceability. Anonymity means that without the tracing key it should be infeasible to determine which group member issued a particular signature (even given all the signing keys). Bellare et al. [8] defined a "CCAversion" of this notion, where the adversary is given access to a tracing oracle. Following [10] we use a "CPA-version" of anonymity where such oracle access is not given.

**Definition 1.** A group signature scheme  $\mathcal{GS} = (G.KeyGen, G.Sign, G.Vrfy, G.Open)$  is anonymous if for all polynomials  $N(\cdot)$  and all probabilistic polynomial-time adversaries  $\mathcal{A}$ , the advantage of  $\mathcal{A}$  in the following experiment is negligible in n:

- 1. Compute  $(PK, TK, gsk) \leftarrow G.KeyGen(1^n, 1^N)$  and give (PK, gsk) to A.
- 2. A outputs two identities  $i_0, i_1 \in [N]$ , along with a message M. A random bit b is chosen, and A is given  $G.Sign(gsk[i_b], M)$ . Finally, A outputs a bit b'.

 $\mathcal{A}$  succeeds (denoted Succ) if b' = b, and the advantage of  $\mathcal{A}$  is  $|\Pr[\mathsf{Succ}] - \frac{1}{2}|$ .

Traceability means that it should be infeasible for an adversary who corrupts some set of users  $\mathcal{C}$  to output a valid signature that cannot be traced to some member of  $\mathcal{C}$ .

**Definition 2.** A group signature scheme  $\mathcal{GS} = (G.KeyGen, G.Sign, G.Vrfy, G.Open)$  is traceable if for all polynomials  $N(\cdot)$  and all probabilistic polynomial-time adversaries  $\mathcal{A}$ , the success probability of  $\mathcal{A}$  in the following experiment is negligible in n:

- 1. Compute  $(PK, TK, gsk) \leftarrow G.KeyGen(1^n, 1^N)$  and give PK and TK to A.
- 2. A may query the following oracles adaptively and in any order:
  - A Corrupt oracle that on input  $i \in [N]$  returns gsk[i].
  - A Sign oracle that on input i, M outputs G.Sign(gsk[i], M).

Let C be the set of identities queried to Corrupt.

3. At some point, A outputs a message M and a signature  $\sigma$ .

 $\mathcal{A}$  succeeds if (1) G.Vrfy(PK,  $M, \sigma$ ) = 1 and (2) Sign(i, M) was never queried for  $i \notin \mathcal{C}$ , yet (3) G.Open(TK,  $M, \sigma$ )  $\notin \mathcal{C}$ .

#### 3.2 Our Construction

We let n be the security parameter,  $q = \mathsf{poly}(n), m \ge 5n \log q$  and  $s \ge 5\sqrt{n \log q} \cdot \omega(\sqrt{\log m})$  be parameters of the system. We let  $H : \{0,1\}^* \to \mathbb{Z}_q^n$  be a hash function, to be modeled as a random oracle.

- G.KeyGen $(1^n, 1^N)$ : First compute  $(\mathbf{B}_1, \mathbf{S}_1), \dots, (\mathbf{B}_N, \mathbf{S}_N) \leftarrow \mathsf{TrapSamp}(1^n, 1^m, q)$  and then, for  $1 \leq i \leq N$ , compute  $(\mathbf{A}_i, \mathbf{T}_i) \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{B}_i)$ . Output  $\mathsf{PK} = \left( (\mathbf{A}_i, \mathbf{B}_i)_{i=1}^N \right)$  as the public key,  $\mathsf{TK} = \left( \mathbf{S}_i \right)_{i=1}^N$  as the tracing key, and  $\mathsf{gsk} = (\mathbf{T}_i)_{i=1}^N$  as the users' signing keys.
- G.Sign(gsk[j], M): To sign message M using secret key gsk[j] =  $\mathbf{T}_j$ , choose random  $r \leftarrow \{0,1\}^n$ , set  $\bar{M} = M \| r$ , and then compute  $\mathbf{h}_i = H(\bar{M} \| i)$  for  $1 \leq i \leq N$ . Then:
  - Compute  $\mathbf{e}_j \leftarrow \mathsf{GPVInvert}(\mathbf{A}_j, \mathbf{T}_j, s, \mathbf{h}_j)$ .
  - For  $i \neq j$ , choose  $\mathbf{e}_i \in \mathbb{Z}_q^m$  uniformly subject to the condition that  $\mathbf{A}_i \mathbf{e}_i = \mathbf{h}_i \pmod{q}$ .

For all i, sample  $\mathbf{s}_i \leftarrow \mathbb{Z}_q^n$  and compute  $\mathbf{z}_i = \mathbf{B}_i^T \mathbf{s}_i + \mathbf{e}_i \pmod{q} \in \mathbb{Z}_q^m$ . Finally, construct an NIWI proof  $\pi$  for the gap language  $L_{s,\gamma}$  as discussed in Section 2.5 (and using the witness  $(\mathbf{s}_i, i)$ ). Output the signature  $(r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$ .

- G.Vrfy(PK, M,  $\sigma$ ): Parse the signature as  $(r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$  and set  $\bar{M} = M \| r$ . Output 1 iff the proof  $\pi$  is correct, and  $\mathbf{A}_i \mathbf{z}_i = H(\bar{M} \| i) \pmod{q}$  for all i.
- G.Open(TK, M,  $\sigma$ ): Parse the signature as  $(r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$ . Using the  $\{\mathbf{S}_i\}$ , output the smallest index i for which<sup>3</sup>  $\operatorname{dist}(\mathcal{L}(\mathbf{B}_i^T), \mathbf{z}_i) \leq s\sqrt{m}$ .

We first check correctness. Let  $(r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$  be a signature produced by an honest signer. It is clear that  $\pi$  is a valid proof. Moreover, for any i we have

$$\mathbf{A}_i \mathbf{z}_i = \mathbf{A}_i (\mathbf{B}_i^T \mathbf{s}_i + \mathbf{e}_i) = \mathbf{A}_i \mathbf{e}_i = H(\bar{M} | i) \pmod{q},$$

and so verification succeeds. Correctness of the opening algorithm follows easily.

**Theorem 1.** Let m, q, and s be as described above. If  $\mathsf{LWE}_{m,q,\alpha}$  is hard for  $\alpha = s/(q\sqrt{2})$ , and the proof system used is witness indistinguishable, then the group signature scheme described above is anonymous. If  $\mathsf{GapSVP}_{\gamma}$  is hard for  $\gamma = O(n\log^4 n)$ , then the group signature scheme described above is traceable.

We note that for values of s as described above, the hardness of  $\mathsf{LWE}_{m,q,\alpha}$  is implied by the difficulty of finding a quantum algorithm for approximating  $\mathsf{GapSVP}_{\hat{\gamma}}$  for  $\hat{\gamma} = \widetilde{O}(n/\alpha)$  [27], so our entire scheme can be based on the difficulty of finding a quantum algorithm for  $\mathsf{GapSVP}$ .

We prove anonymity in Section 3.3 and traceability in Section 3.4.

<sup>&</sup>lt;sup>3</sup> Soundness of the proof system ensures that if  $\sigma$  is valid, then some such i exists except with negligible probability.

#### 3.3 Anonymity

Fix  $N = \mathsf{poly}(n)$  and let  $\mathcal{A}$  be a PPT adversary attacking the group signature scheme in the sense of Definition 1. Let  $\mathsf{G}_0$  denote the experiment of Definition 1 with b = 0, and let  $\mathsf{G}_1$  be the same experiment with b = 1. We consider a sequence of experiments  $\mathsf{G}_0, \mathsf{G}'_0, \mathsf{G}'_1, \mathsf{G}_1$  and show that each experiment is indistinguishable from the one preceding it. This implies anonymity.

We review  $G_0$  as applied to our group signature scheme. First, the keygeneration algorithm  $G.\mathsf{KeyGen}(1^n,1^N)$  is run and  $\mathcal{A}$  is given the public key  $\mathsf{PK} = \left( (\mathbf{A}_i, \mathbf{B}_i)_{i=1}^N \right)$  and the users' secret keys  $\mathsf{gsk} = (\mathbf{T}_i)_{i=1}^N$ , where each  $\mathbf{B}_i$ is statistically close to uniform and  $(\mathbf{A}_i, \mathbf{T}_i) \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{B}_i)$ . (The tracing key TK is irrelevant in the CPA-version of the anonymity experiment that we are considering.) Next,  $\mathcal{A}$  outputs  $i_0, i_1, M$ , and is given a signature of user  $i_0$ on M, computed as follows. Let  $\mathbf{h}_i = H(M||r||i)$ , for a random  $r \in \{0,1\}^n$ . Then  $\mathbf{e}_{i_0}$  is computed as  $\mathbf{e}_{i_0} \leftarrow \mathsf{GPVInvert}(\mathbf{A}_{i_0}, \mathbf{T}_{i_0}, s, \mathbf{h}_{i_0})$ , whereas  $\mathbf{e}_i$  (for  $i \neq i_0$ ) is chosen uniformly subject to the condition that  $\mathbf{A}_i \mathbf{e}_i = \mathbf{h}_i \pmod{q}$ . Then, for all  $i \in [N]$ , choose random  $\mathbf{s}_i \leftarrow \mathbb{Z}_q^n$  and compute  $\mathbf{z}_i = \mathbf{B}_i^T \mathbf{s}_i + \mathbf{e}_i$ . Finally, a proof  $\pi$  is generated and  $\mathcal{A}$  is given the signature  $(r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$ .

In  $G_0'$  we introduce the following modification with respect to  $G_0$ : when generating the signature, we now compute  $\mathbf{e}_{i_0} \leftarrow \mathsf{GPVInvert}(\mathbf{A}_{i_0}, \mathbf{T}_{i_0}, s, \mathbf{h}_{i_0})$  and  $\mathbf{e}_{i_1} \leftarrow \mathsf{GPVInvert}(\mathbf{A}_{i_1}, \mathbf{T}_{i_1}, s, \mathbf{h}_{i_1})$ . (For  $j \notin \{i_0, i_1\}$ , the value  $\mathbf{e}_j$  is computed as before.)

Claim. If the  $\mathsf{LWE}_{m,q,\alpha}$  problem is hard, then  $\mathsf{G}_0$  and  $\mathsf{G}_0'$  are computationally indistinguishable.

*Proof.* Recall (cf. Lemma 1) that hardness of the  $\mathsf{LWE}_{m,q,\alpha}$  problem implies hardness of the  $\widehat{\mathsf{LWE}}_{m,q,\alpha q\sqrt{2}}$  problem. We use  $\mathcal{A}$  to construct a PPT algorithm  $\mathcal{D}$  for the  $\widehat{\mathsf{LWE}}_{m,q,\alpha q\sqrt{2}}$  problem.  $\mathcal{D}$  is given as input  $(\mathbf{B},\mathbf{y}) \in \mathbb{Z}_q^{n \times m} \times \mathbb{Z}_q^m$ , where  $\mathbf{B}$  is uniform and  $\mathbf{y}$  is either uniform or equal to  $\mathbf{B}^T\mathbf{s} + \mathbf{e}$  for  $\mathbf{e} \sim \mathcal{D}_{\mathbb{Z}^m,\alpha q\sqrt{2}}$ .

 $\mathcal{D}$  first chooses a random index  $i^* \leftarrow [N]$  and sets  $\mathbf{B}_{i^*} = \mathbf{B}$ . For all  $i \neq i^*$ , it chooses  $\mathbf{B}_i$  uniformly at random. Then, for  $1 \leq i \leq N$  algorithm  $\mathcal{D}$  computes  $(\mathbf{A}_i, \mathbf{T}_i) \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{B}_i)$ . It gives  $\mathsf{PK} = \left( (\mathbf{A}_i, \mathbf{B}_i)_{i=1}^N \right)$  and  $\mathsf{gsk} = (\mathbf{T}_i)_{i=1}^N$  to  $\mathcal{A}$ . All H-queries of  $\mathcal{A}$  are answered with random elements from the appropriate domain.

Eventually  $\mathcal{A}$  outputs two identities  $i_0, i_1 \in [N]$  along with a message M. If  $i^* \neq i_1$  then  $\mathcal{D}$  outputs a random bit and aborts. Otherwise,  $\mathcal{D}$  creates a signature by choosing random  $r \in \{0,1\}^n$  and fixing  $\mathbf{h}_{i_1} \stackrel{\text{def}}{=} H(M||r||i_1) = A_{i_1}\mathbf{y}$ . (The value  $\mathbf{h}_i = H(M||r||i)$  for  $i \neq i_1$  is chosen uniformly.) Then  $\mathcal{D}$  computes  $\mathbf{e}_{i_0} \leftarrow \mathsf{GPVInvert}(\mathbf{A}_{i_0}, \mathbf{T}_{i_0}, s, \mathbf{h}_{i_0})$  and, for  $i \notin \{i_0, i_1\}$ , chooses  $\mathbf{e}_i$  uniformly subject to the condition that  $\mathbf{A}_i \mathbf{e}_i = \mathbf{h}_i \pmod{q}$ . ( $\mathcal{D}$  does not explicitly compute any value  $\mathbf{e}_{i_1}$ .) For  $i \neq i_1$ , the ciphertext  $\mathbf{z}_i$  is computed as in  $\mathsf{G}_0$  and  $\mathsf{G}_0'$ . However,  $\mathcal{D}$  sets  $\mathbf{z}_{i_1} = \mathbf{y}$ .

<sup>&</sup>lt;sup>4</sup> Note that, except with negligible probability,  $H(M||r||i_1)$  has not been queried thus far.

Let  $\mathcal{D}_{rand}$  denote the above experiment when  $\mathcal{D}$ 's input  $\mathbf{y}$  is uniformly distributed. We claim that  $\mathcal{A}$ 's view in  $\mathcal{D}_{\mathrm{rand}}$  is statistically close to its view in  $\mathsf{G}_0$ .

- In  $\mathsf{G}_0$  we have  $\mathbf{h}_{i_1}$  chosen uniformly in  $\mathbb{Z}_{q}^n$ ; then  $\mathbf{e}_{i_1}$  is chosen uniformly subject to  $\mathbf{A}_{i_1}\mathbf{e}_{i_1} = \mathbf{h}_{i_1}$ ; and finally  $\mathbf{z}_{i_1} = \mathbf{B}_{i_1}^T\mathbf{s}_{i_1} + \mathbf{e}_{i_1}$ .
- In  $\mathcal{D}_{\text{rand}}$  we have  $\mathbf{z}_{i_1} = \mathbf{y} = \mathbf{B}_{i_1}^T \mathbf{s}_{i_1} + \mathbf{e}_{i_1}$  for  $\mathbf{e}_{i_1}$  chosen uniformly in  $\mathbb{Z}_q^m$ ; then  $\mathbf{h}_{i_1} = \mathbf{A}_{i_1} \mathbf{e}_{i_1}$ .

To see that these are statistically close, we demonstrate that the choice of  $e_{i_1}$ in  $\mathsf{G}_0$  is statistically close to uniform over  $\mathbb{Z}_q^m$ . We view **A** as a function from  $\mathbb{Z}_q^m \to \mathbb{Z}_q^n$ , and note that this function is regular. Furthermore, since the columns of **A** generate all of  $\mathbb{Z}_q^n$  with all but negligible probability (over the choice of **A**), our randomly chosen h is uniform over the image of A. For a regular function, choosing a uniform element from the image, followed by a uniform element from its pre-image, is equivalent to choosing a uniform element from the domain, as is done in  $\mathcal{D}_{rand}$ .

On the other hand, let  $\mathcal{D}_{LWE}$  denote the above experiment when  $\mathcal{D}$ 's input  $\mathbf{y}$  is distributed according to  $\mathbf{y} = \mathbf{B}^T \mathbf{s} + \mathbf{e}$  for  $\mathbf{e} \sim D_{\mathbb{Z}^m, \alpha q \sqrt{2}}$ . We claim that  $\mathcal{A}$ 's view in  $\mathcal{D}_{LWE}$  is statistically close to its view in  $\mathsf{G}_0'$ . Indeed:

- In experiment  $\mathsf{G}_0'$  we have  $\mathbf{h}_{i_1}$  chosen uniformly in  $\mathbb{Z}_q^n$ . Next, we compute
- $\mathbf{e}_{i_1} \leftarrow \mathsf{GPVInvert}(\mathbf{A}_{i_1}, \mathbf{T}_{i_1}, s, \mathbf{h}_{i_1}); \text{ and finally } \mathbf{z}_{i_1} = \mathbf{B}_{i_1}^T \mathbf{s}_{i_1} + \mathbf{e}_{i_1}.$   $\text{ In } \mathcal{D}_{LWE} \text{ we have } \mathbf{z}_{i_1} = \mathbf{y} = \mathbf{B}_{i_1}^T \mathbf{s}_{i_1} + \mathbf{e}_{i_1} \text{ for } \mathbf{e}_{i_1} \sim \mathcal{D}_{\mathbb{Z}^m, \alpha \cdot q \cdot \sqrt{2}}; \text{ then } \mathbf{h}_{i_1} = \mathbf{e}_{i_1} = \mathbf{e}_{i_1} + \mathbf{e}_{i_2} = \mathbf{e}_{i_2} + \mathbf{e}_{i_2} = \mathbf{e}_{i_1} + \mathbf{e}_{i_2} = \mathbf{e}_{i_2} + \mathbf{e}_{i_2} = \mathbf{e}_{i_2} + \mathbf$  $A_{i_1}e_{i_1}$ .

The above are easily seen to be statistically close for our choice of parameters, again using the results of [19]. Since the probability that  $\mathcal{D}$  does not abort is 1/N, and its decision to abort is independent of A's success, the proof is complete.

The rest of the proof of anonymity is straightforward, and so we merely provide a sketch. Experiment  $G'_1$  is identical to  $G'_0$  with the exception that the proof  $\pi$  is now computed using the witness  $(\mathbf{s}_{i_1}, i_1)$  rather than  $(\mathbf{s}_{i_0}, i_0)$ . Witness indistinguishability of the proof system implies that  $G'_1$  and  $G'_0$  are computationally indistinguishable.

Computational indistinguishability of  $G'_1$  and  $G_1$  (the experiment from Definition 1 with b=1) can be proved exactly as in the proof of the previous claim.

#### 3.4 Traceability

Fix N = poly(n) and let  $\mathcal{A}$  be a PPT adversary attacking the group signature scheme in the sense of Definition 2. We construct a PPT forger  $\mathcal{F}$  for the GPV signature scheme [19] (in the random oracle model) whose success probability is polynomially related to that of A. Since the GPV signature scheme is secure assuming hardness of the  $\mathsf{GapSVP}_{\gamma}$  problem, this completes the proof.

We first observe that we may, without loss of generality, assume that A never corrupts all users in [N] because A can succeed with only negligible probability in this case. (Given a valid signature  $(r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$ , soundness of the proof system implies that G.Open outputs some  $i \in [N]$  except with negligible probability.) We will assume this in what follows.

 $\mathcal{F}$  is given a public key **A** for the GPV signature scheme, and begins by choosing a random index  $i^* \in [N]$  and setting  $\mathbf{A}_{i^*} = \mathbf{A}$ . Next, it computes the values  $(\mathbf{B}_{i^*}, \mathbf{S}_{i^*}) \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{A}_{i^*})$ . For all the remaining indices  $i \neq i^*$ , the forger computes the values  $(\mathbf{B}_i, \mathbf{S}_i) \leftarrow \mathsf{TrapSamp}(1^n, 1^m, q)$  and  $(\mathbf{A}_i, \mathbf{T}_i) \leftarrow \mathsf{OrthoSamp}(1^n, 1^m, q, \mathbf{B}_i)$  exactly as in the legitimate key-generation algorithm. After this,  $\mathcal{F}$  gives  $\mathsf{PK} = (\mathbf{A}_i, \mathbf{B}_i)_{i=1}^N$  and  $\mathsf{TK} = (\mathbf{S}_i)_{i=1}^N$  to  $\mathcal{A}$ . We note that by Corollary 1, the distribution of these keys is statistically close to the distribution that is expected by the adversary.

 $\mathcal{F}$  answers random oracle queries of  $\mathcal{A}$  by simply passing these queries to its own random oracle.  $\mathcal{F}$  responds to the other queries of  $\mathcal{A}$  as follows:

- Corrupt(i): if  $i \neq i^*$  then  $\mathcal{F}$  gives  $\mathbf{T}_i$  to  $\mathcal{A}$ , while if  $i = i^*$  then  $\mathcal{F}$  aborts.
- Sign(i, M): If  $i \neq i^*$  then  $\mathcal{F}$  computes the signature using  $\mathbf{T}_i$  and the honest signing algorithm. If  $i = i^*$ , then:
  - 1.  $\mathcal{F}$  chooses random  $r \in \{0,1\}^n$  and queries its own signing oracle on the message  $M||r||i^*$ . It receives in return a signature  $\mathbf{e}_{i^*}$ .
  - 2. The remainder of the signature is computed using the honest signing algorithm. (Note that computation of  $\mathbf{e}_{i^*}$  the only aspect of signing that relies on the secret key of user  $i^*$ .)

Let  $\mathcal{C}$  denote the set of identities that  $\mathcal{A}$  has queried to Corrupt. (Recall that if  $\mathcal{F}$  has not aborted, then  $i^* \notin \mathcal{C}$ .) At some point  $\mathcal{A}$  outputs a message Mand signature  $\sigma = (r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$ . Assume  $\mathsf{G.Vrfy}(\mathsf{PK}, M, \sigma) = 1$ , and that Sign(i, M) was never queried for  $i \notin \mathcal{C}$ . Since  $\mathcal{F}$  has the tracing key TK, it can compute  $j \leftarrow \mathsf{G.Open}(\mathsf{TK}, M, \sigma)$ . If  $j \neq i^*$  then  $\mathcal{F}$  aborts. Otherwise,  $\mathcal{F}$  does:

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1. Use S_{i*} to recover e_{i*} such that
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- $||\mathbf{e}_{i^*}||_{\infty} \leq s\sqrt{m}, \text{ and }$   $\mathbf{z}_{i^*} \mathbf{e}_{i^*} \in \mathcal{L}(\mathbf{B}_{i^*}^T).$
- 2. Output the forgery  $(M||r||i^*, \mathbf{e}_{i^*})$ .

Let  $\epsilon$  denote the probability with which  $\mathcal{A}$  succeeds in the experiment of Definition 2. It is easy to see that  $\mathcal{F}$  aborts with probability at most<sup>5</sup> (N-1)/Nand, conditioned on not aborting, the view of A when run as a sub-routine by  $\mathcal{F}$  is statistically close to its view in the experiment described in Definition 2. Thus, with probability at least  $\epsilon/N$  it holds that  $\mathcal{A}$  outputs  $(M, \sigma)$  with  $\mathsf{G.Vrfy}(\mathsf{PK}, M, \sigma) = 1 \text{ and } \mathsf{G.Open}(\mathsf{TK}, M, \sigma) = i^*, \text{ and where } \mathcal{A} \text{ never queried}$  $Sign(i^*, M)$ . We show that whenever this occurs, then  $\mathcal{F}$  outputs a valid forgery (except with negligible probability).

Fix  $(M, \sigma)$  such that the above hold, and let  $\sigma = (r, \mathbf{z}_1, \dots, \mathbf{z}_N, \pi)$ . Since  $\mathsf{G.Open}(\mathsf{TK}, M, \sigma) = i^*$ , this implies that  $\mathcal{F}$  will indeed be able to recover  $\mathbf{e}_{i^*}$ 

<sup>&</sup>lt;sup>5</sup> Actually,  $\mathcal{F}$  aborts with probability at most  $(N-1)/N + \mathsf{negl}(n)$ , where the negligible term arises from the possibility that A violates soundness of the proof system. We ignore this for simplicity.

such that (1)  $||\mathbf{e}_{i^*}||_{\infty} \leq s\sqrt{m}$  and (2)  $\mathbf{z}_{i^*} - \mathbf{e}_{i^*} \in \mathcal{L}(\mathbf{B}_{i^*}^T)$ . Moreover, since  $\mathsf{G.Vrfy}(\mathsf{PK}, M, \sigma) = 1$  we have  $\mathbf{A}_{i^*}\mathbf{z}_{i^*} = H(M\|r\|i^*)$ ; since  $\mathbf{A}_{i^*}(\mathbf{z}_{i^*} - \mathbf{e}_{i^*}) = \mathbf{0}$  this means  $\mathbf{A}_{i^*}\mathbf{e}_{i^*} = H(M\|r\|i^*)$ . Thus  $\mathbf{e}_{i^*}$  is a valid GPV signature on the message  $M\|r\|i^*$ . Since  $\mathcal{A}$  never queried  $\mathsf{Sign}(i^*, M)$ , we know that  $\mathcal{F}$  never queried its own signing oracle for a signature on  $M\|r\|i^*$ . It follows that the output of  $\mathcal{F}$  is indeed a valid forgery.

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