

# CS 258: Quantum Cryptography (Fall 2025)

## Homework 3 (100 points)

### 1 Problem 1 (30 points)

In class, for a matrix  $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$ ,  $m > n$ , we defined the lattices

$$\begin{aligned}\Lambda_q^\perp(\mathbf{A}) &= \{\mathbf{x} \in \mathbb{Z}^m : \mathbf{A} \cdot \mathbf{x} \bmod q = 0\} \\ \Lambda_q(\mathbf{A}) &= \{\mathbf{x} \in \mathbb{Z}^n : \exists \mathbf{r} \in \mathbb{Z}^n : \mathbf{x} = \mathbf{A}^T \cdot \mathbf{r} \bmod q\}\end{aligned}$$

We said that  $\Lambda_q^\perp(\mathbf{A})$  is spanned by the integer vectors in the kernel of  $\mathbf{A}$  as well as  $q\mathbf{I}$ . However, this is not a basis. Likewise,  $\Lambda_q(\mathbf{A})$  is spanned by the rows of  $\mathbf{A}$  and  $q\mathbf{I}$ , but this is not a basis. Here, you will find explicit bases for these lattices.

Assume for simplicity that the first  $n$  columns of  $\mathbf{A}$  are linearly independent mod  $q$ . In general, we will always work in a regime where the some set of  $n$  columns are linearly independent with overwhelming probability. If they are not the first  $n$  columns, we can adjust the derivation below, but you are not required to do so.

**Part (a). 10 points.** Write  $\mathbf{A} = (\mathbf{A}_0 | \mathbf{A}_1)$  where  $\mathbf{A}_0 \in \mathbb{Z}_q^{n \times n}$  is full rank mod  $q$ . Define  $\mathbf{A}' = \mathbf{A}_0^{-1} \mathbf{A} \bmod q = (\mathbf{I}, \mathbf{A}'_1)$  where  $\mathbf{A}'_1 = \mathbf{A}_0^{-1} \mathbf{A}_1 \bmod q$ . Here, the inverse is taking mod  $q$ , so that  $\mathbf{A}_0^{-1} \in \mathbb{Z}_q^{n \times n}$ ; this inverse exists by assumption that  $\mathbf{A}_0$  is full rank mod  $q$ . Show that  $\Lambda_q(\mathbf{A}') = \Lambda_q(\mathbf{A})$  and  $\Lambda_q^\perp(\mathbf{A}') = \Lambda_q^\perp(\mathbf{A})$ .

**Part (b). 10 points.** Show that the columns of

$$\begin{pmatrix} q\mathbf{I}_n & -\mathbf{A}'_1 \\ 0 & \mathbf{I}_{m-n} \end{pmatrix}$$

form a basis for  $\Lambda_q^\perp(\mathbf{A})$ , where  $\mathbf{I}_k$  is the  $k \times k$  identity matrix. To do so, show that the columns of this matrix are each in  $\Lambda_q^\perp(\mathbf{A})$ , and that any vector  $\mathbf{x} \in \Lambda_q^\perp(\mathbf{A})$  can be written as an integer linear combination of the columns.

**Part (c). 10 points.** Show that the columns of

$$\begin{pmatrix} \mathbf{I}_n & 0 \\ (\mathbf{A}'_1)^T & q\mathbf{I}_{m-n} \end{pmatrix}$$

form a basis for  $\Lambda_q(\mathbf{A})$ .

### 2 Problem 2 (30 points)

Let  $\mathcal{L} \subseteq \mathbb{R}^n$  be a lattice, which we will assume to be full rank. The dual lattice, denoted  $\mathcal{L}^*$ , is defined as:

$$\mathcal{L}^* = \{\mathbf{x} \in \mathbb{R}^n : \forall \mathbf{y} \in \mathcal{L}, \mathbf{x} \cdot \mathbf{y} \in \mathbb{Z}\}$$

That is, the inner product of any vector  $\mathbf{x} \in \mathcal{L}^*$  with any vector  $\mathbf{y} \in \mathcal{L}$  is an integer.

**Part (a). 10 points** Suppose  $\mathcal{L} = \mathcal{L}(\mathbf{B})$  for some basis  $\mathbf{B} \in \mathbb{R}^{n \times n}$  ( $\mathbf{B}$  is square since  $\mathcal{L}$  is assumed to be full rank). Show that  $\mathcal{L}^* = \mathcal{L}(\mathbf{B}^{-T})$ , where  $\mathbf{B}^{-T} = (\mathbf{B}^{-1})^T = (\mathbf{B}^T)^{-1}$ . [Hint: write any vector  $\mathbf{x} \in \mathbb{R}^n$  as  $\mathbf{x} = \mathbf{B}^{-T} \cdot \mathbf{r}$  for a unique  $\mathbf{r}$ , which is possible since  $\mathbf{B}$  and hence  $\mathbf{B}^{-T}$  is full rank. If  $\mathbf{r} \in \mathbb{Z}^n$  (meaning that  $\mathbf{x} \in \mathcal{L}(\mathbf{B}^{-T})$ ), what is the inner product of  $\mathbf{x}$  with the elements of  $\mathcal{L}$ ? If  $\mathbf{r} \notin \mathbb{Z}^n$ , what is the inner product of  $\mathbf{x}$  with the elements of  $\mathcal{L}$ ?]

**Part (b). 10 points** Show that  $(\mathcal{L}^*)^* = \mathcal{L}$ .

**Part (c). 10 points** Suppose you have a basis  $\mathbf{B}$  for  $\mathcal{L}$  that is “short”, in the sense that each column of  $\mathbf{B}$  has norm at most  $\sigma$ . Here, you will see how to solve (approximate) CVP in  $\mathcal{L}^*$ .

Let  $\mathbf{y}$  be a vector that is “close” to  $\mathcal{L}^*$ , in the sense that there exist a short vector  $\mathbf{e}$  such that  $\mathbf{u} = \mathbf{y} - \mathbf{e} \in \mathcal{L}^*$ . In particular, assume that  $|\mathbf{e}| < 1/2n\sigma$ . Your goal is to compute  $\mathbf{u}$ , or equivalently  $\mathbf{e}$ .

To do so, compute  $\mathbf{B}^T \mathbf{y}$ , and use that  $\mathbf{y} = \mathbf{u} + \mathbf{e}$ . What happens if you round to the nearest integer? What happens if instead you remove the integer part (this is the same as reducing mod 1)? [Hint: recall that  $|\mathbf{a} \cdot \mathbf{b}| \leq |\mathbf{a}| \times |\mathbf{b}|$ ]

### 3 Problem 3 (40 points)

The  $S|LWE\rangle$  problem ( $S$  for “state”) problem asks to compute  $\mathbf{s}$  from the state

$$|\tau_{\mathbf{s}}\rangle := \sum_{\mathbf{e}} \sqrt{\Pr[\mathbf{e} \leftarrow D_{\sigma}^m]} |\mathbf{A}^T \cdot \mathbf{s} + \mathbf{e} \bmod q\rangle$$

**Part (a). 10 points.** Assume that  $\mathbf{e}$  in the support of  $D_{\sigma}^m$  has support constrained to  $\sigma m$ . Also, assume that there are no vectors of norm at most  $2\sigma m$  in  $\Lambda_q(\mathbf{A})$ . Then show that the states  $|\tau_{\mathbf{s}}\rangle$  for different  $\mathbf{s} \in \mathbb{Z}_q^n$  are orthogonal. Thus, the task of finding  $\mathbf{s}$  from  $|\tau_{\mathbf{s}}\rangle$  is at least information-theoretically possible.

**Part (b). 10 points** Show how to construct  $|\tau_{\mathbf{s}}\rangle$  from  $\mathbf{s}$  efficiently. You may assume the ability to create Gaussian-weighted superpositions  $\sum_{\mathbf{e}} \sqrt{\Pr[\mathbf{e} \leftarrow D_{\sigma}]} |\mathbf{e}\rangle$ , that  $D_{\sigma}$  has support only on integers of absolute value at most  $\sigma\sqrt{m}$ , and that  $\sigma\sqrt{m} < q/2$ .

**Part (c). 10 points.** Explain that if search LWE is easy, then  $S|LWE\rangle$  is easy, for the same parameters  $q, n, m, \sigma$ .

**Part (d). 10 points.** Show that if it is possible to solve  $S|LWE\rangle$  perfectly with parameter  $\sigma = q/2\gamma$ , then it is possible to solve SIS with parameter  $\beta = \gamma m$  (these are the same parameters we saw in class for Regev’s reduction). Concretely, you may assume that the  $S|LWE\rangle$  solver is a unitary mapping  $|\tau_{\mathbf{s}}\rangle |\mathbf{y}\rangle$  to  $|\tau_{\mathbf{s}}\rangle |\mathbf{y} + \mathbf{s} \bmod q\rangle$ .

For this problem, you may assume the ability to construct Gaussian-weighted superpositions of arbitrary parameter  $\gamma$  such that  $1 \ll \gamma \ll q$ , that the support of such distributions constrained to “small” vectors as described above, and also the ability to construct the uniform superpositions over linear subspaces.

**Remark 1.** The above shows that  $S|LWE\rangle$  is at least as hard as SIS, and that LWE is at least as hard as  $S|LWE\rangle$ . But it could be that  $S|LWE\rangle$  is actually an easier problem than ordinary LWE. We know, under some loss in the parameters  $n, m, q, \sigma$  that  $S|LWE\rangle$  and LWE are equivalent, but we do not know if they are equivalent for the exact same parameters.