# COL872 Problem Set 1

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#### Question 1: Lossy Encryption

**Question.** Show that if  $\mathcal{E} = (\text{Setup}, \text{Setup-Lossy}, \text{Enc}, \text{Dec})$  is a lossy encryption scheme satisfying all the above properties, then  $\mathcal{E}' = (\text{Setup'}, \text{Enc'}, \text{Dec'})$  is also a correct and semantically secure public key encryption scheme.

*Proof.* To Prove:  $\mathcal{E}'$  is a correct and semantically secure PKE given  $\mathcal{E}$  as described in the problem statement.

#### Correctness

For correctness, Dec'(Enc'(m, pk), sk) = m

By the definition of  $\mathcal{E}'$ , Setup' = Setup, Dec' = Dec, Enc' = Enc. In  $\mathcal{E}'$ , all the keys are a (pk, sk) pair generated by Setup, which is non-lossy hence the encrypted text can always be decrypted (correctness of  $\mathcal{E}$ ).

 $\mathsf{Dec}'(\mathsf{Enc}'(m,pk),sk) = \mathsf{Dec}(\mathsf{Enc}(m,pk)) = m$ 

Therefore the public key encryption scheme  $\mathcal{E}'$  has the correctness property.

#### Semantic Security of PKE

Proof for semantic security of PKE  $\mathcal{E}'$  by hybrid world model.

- World 0:  $m_0$  is encrypted using Setup
  - 1. Challenger creates a public key pk and a secret key sk using Setup
  - 2. Challenger sends pk to the adversary
  - 3. The adversary sends two messages  $m_0, m_1$  to the challenger
  - 4. The challenger encrypts  $m_0$  using pk and sends it to the adversary
- **Hybrid 0**:  $m_0$  is encrypted using Setup-Lossy
  - 1. Challenger creates a public key pk using Setup-Lossy
  - 2. Challenger sends pk to the adversary
  - 3. The adversary sends two messages  $m_0, m_1$  to the challenger
  - 4. The challenger encrypts  $m_0$  using pk and sends it to the adversary
- **Hybrid 1**:  $m_1$  is encrypted using Setup-Lossy
  - 1. Challenger creates a public key pk using Setup-Lossy
  - 2. Challenger sends pk to the adversary
  - 3. The adversary sends two messages  $m_0, m_1$  to the challenger
  - 4. The challenger encrypts  $m_1$  using pk and sends it to the adversary
- World 1:  $m_1$  is encrypted using Setup
  - 1. Challenger creates a public key pk and a secret key sk using Setup
  - 2. Challenger sends pk to the adversary
  - 3. The adversary sends two messages  $m_0, m_1$  to the challenger

4. The challenger encrypts  $m_1$  using pk and sends it to the adversary

World 0 and Hybrid 0: From indistinguishability of modes, we know that the public keys are sampled from computationally indistinguishable distributions. Therefore, if World 0 and Hybrid 0 were far apart, we would be able to come up with an adversary that breaks the indistinguishability of modes. Hence, these two worlds are close.

**Hybrid 0 and Hybrid 1**: We know that the distributions for any two messages are statistically indistinguishable in lossy mode. Therefore, no (unbounded) adversary can differentiate between these two hybrids.

**Hybrid 1 and World 1**: Similar argument follows as for closeness of World 0 and Hybrid 0.

Therefore, we have shown that if  $\mathcal{E}$  is a secure lossy encryption scheme, then  $\mathcal{E}'$  is a correct and semantically secure encryption scheme.

#### Question 2 $\mathbf{2}$

#### Question 2: A Lossy Encryption Scheme based on LWE

Question. In this problem, you will have to construct a lossy encryption mode for Regev encryption. The algorithms Setup, Enc, Dec are defined as in class (see Lecture Notes). You must define the Setup-Lossy algorithm, and then show that it is a secure lossy encryption scheme.

*Proof.* Following are the steps to generate the desired lossy encryption scheme-

#### 1. Definition: Lossy Setup Algorithm

Setup-Lossy(1<sup>n</sup>): pk = (A, b) where  $A \leftarrow \mathbb{Z}_q^{n \times m}$  and  $b \leftarrow \mathbb{Z}_q^m$ pk is the lossy public key.

#### 2. Indistinguishability of the modes

**To Prove:**  $pk \leftarrow \mathsf{Setup}(1^n)$  is computationally indistinguishable from  $pk' \leftarrow \mathsf{Setup}\mathsf{-Lossy}(1^n)$ pk'(A,b) where  $A \leftarrow \mathbb{Z}_q^{n \times m}$  and  $b \leftarrow \mathbb{Z}_q^m$ , therefore pk' is completely random. pk = (A,b) where  $A \leftarrow \mathbb{Z}_q^{n \times m}$  and  $b^T = s^T \cdot A + e^T$  where  $s \leftarrow \mathbb{Z}_q^n$  is a secret and  $e \leftarrow \chi^m$  is

random error as per Regev's PKE Scheme.

Following LWE, the distribution of pk and pk' should be computationally indistinguishable as  $b^T = s^T \cdot A + e^T$  and  $b \leftarrow \mathbb{Z}_q^m$  are computationally indistinguishable due to LWE.

#### 3. Statistical indistinguishability in the lossy mode

To Prove: given  $m_0, m_1$ ;

 $\{pk, \mathsf{Enc}(pk, m_0) : pk \leftarrow \mathsf{Setup}\mathsf{-}\mathsf{Lossy}(1^n)\} = \{pk, \mathsf{Enc}(pk, m_1) : pk \leftarrow \mathsf{Setup}\mathsf{-}\mathsf{Lossy}(1^n)\}$ Let's take  $m_0$ ,  $\text{Enc}(m_0, pk_{lossy}) = (A \cdot r, b^T \cdot r + m_0 \times \frac{q}{2})$  where  $r \leftarrow \{0, 1\}^m$ 

Using Leftover Hash Lemma,  $A \cdot r$  is same as a random vector (statistically). Therefore r can not be recovered from  $A \cdot r$  and thereby  $b^T \cdot r$  is random.  $m_0 \times \frac{q}{2}$  + random is still random. Similarly, from  $m_1$ , above steps are repeated and we again arrive at a random vector.

Therefore having sent encryption of either  $m_0$  or  $m_1$  in lossy encryption mode, it is statistically impossible to figure out which message was encrypted.

#### **3.1** Question **3.1**

#### Question 3.1: Small Secrets LWE - Matrix Version

**Question.** In this problem, you have to prove the indistinguishability of the matrix version of the ss-LWE assuming that the normal version of ss-LWE is computationally hard.

*Proof.* We have the following Distributions:

$$calD_1 = \{(A, B) : B = S.A + E, A \leftarrow \mathbb{Z}_q^{n \times m}, S \leftarrow \chi^{n \times n}, E \leftarrow \chi^{n \times m}\}$$
 and  $\mathcal{D}_2 = \{(A, B) : A, B \leftarrow \mathbb{Z}_q^{n \times m}\}$ 

We have to show that the distributions  $\mathcal{D}_1$  and  $\mathcal{D}_2$  are computationally indistinguishable.

Let's create n-1 hybrid distributions  $\mathcal{H}_1, \mathcal{H}_2, \cdots, \mathcal{H}_{n-1}$  where

$$\mathcal{H}_i = \{(A, B) : B = [b_j] \text{ where } b_j = \begin{cases} j^{th} \text{ row of } S.A + E, & \text{if } j \leq i \\ \mathbb{Z}_q^{1 \times m} & i < j \leq n - 1 \end{cases}$$

Claim 3.1.  $\mathcal{D}_1$  and  $\mathcal{H}_1$  are computationally indistinguishable.

*Proof.* The distributions  $\mathcal{D}_1$  and  $\mathcal{H}_1$  differ only at the first row of the second matrix in an instance of the distribution. If an adversary is able to distinguish between the two distributions, this means it is able to distinguish between the rows of two distributions. This can be used to achieve a reduction that breaks ss-LWE<sub> $n,m,q,\chi$ </sub> but this contradicts the fact that ss-LWE<sub> $n,m,q,\chi$ </sub> is a hard computational problem.

So,  $\mathcal{D}_1$  and  $\mathcal{H}_1$  are computationally indistinguishable Distributions.

Claim 3.2.  $\mathcal{H}_{i-1}$  and  $\mathcal{H}_i$  are computationally indistinguishable.

*Proof.* The distributions  $\mathcal{H}_{i-1}$  and  $\mathcal{H}_i$  differ only at the  $i^{th}$  row of the second matrix in an instance of the distribution. If an adversary is able to distinguish between the two distributions, this means it is able to distinguish between the rows of two distributions. This can be used to achieve a reduction that breaks ss-LWE<sub> $n,m,q,\chi$ </sub> but this contradicts the fact that ss-LWE<sub> $n,m,q,\chi$ </sub> is a hard computational problem.

So,  $\mathcal{H}_{i-1}$  and  $\mathcal{H}_i$  are computationally indistinguishable Distributions.

Claim 3.3.  $\mathcal{H}_{n-1}$  and  $\mathcal{D}_2$  are computationally indistinguishable.

*Proof.* The distributions  $\mathcal{H}_{n-1}$  and  $\mathcal{D}_2$  differ only at the last row of the second matrix in an instance of the distribution. If an adversary is able to distinguish between the two distributions, this means it is able to distinguish between the rows of two distributions. This can be used to achieve a reduction that breaks ss-LWE<sub> $n,m,q,\chi$ </sub> but this contradicts the fact that ss-LWE<sub> $n,m,q,\chi$ </sub> is a hard computational problem.

So,  $\mathcal{H}_{n-1}$  and  $\mathcal{D}_2$  are computationally indistinguishable Distributions.

Using the claims proven above, we can say that the distributions  $\mathcal{D}_1$  and  $\mathcal{D}_2$  are computationally indistinguishable.

3.2 Question 3.2

### Question 3.2: Small Secrets LWE - Matrix Version

**Question.** Let  $q = 2^{\sqrt{n}}$  and  $\chi \equiv \mathsf{Unif}_{\left[-\sqrt{q},\sqrt{q}\right]}$ . Would this problem remain hard if even A is chosen from  $\chi^{n \times m}$  in both the distributions?

*Proof.* Nothing can be determined for the current range of  $\chi$  since the range of  $\mathbf{S} \cdot \mathbf{A} + \mathbf{E}$  spans  $\mathbb{Z}_q^{n \times m}$ 

But if  $\chi \equiv \mathsf{Unif}_{\left[-q^{\frac{1}{4}},q^{\frac{1}{4}}\right]}$ , then it can be said with certainty that these two distributions are **not** 

**equivalent** as the max value in the second matrix of distribution  $\mathcal{D}_1$  will be  $n \times q^{\frac{1}{2}} + q^{\frac{1}{4}}$  which is less than q/2 and thus It will not contain a lot of values present in the uniform distribution  $\mathbb{Z}_q$  and thus will be distinguishable from second distribution drawn over  $\mathbb{Z}_q^{n \times m}$ .

3.3 Question 3.3

Question 3.3: Small Secrets LWE - Matrix Version

**Question.** Recall, in the Learning-with-errors problem, if m = n, then the two distributions are statistically indistinguishable. Does the same hold true in the small- secrets setting?

Proof. When we calculate the randomness of both the distributions, for  $\mathcal{D}_2$ , randomness is  $(nm+nm)\log q$ , and for  $\mathcal{D}_1$ , the randomness comes out to be  $nm\log q+(n^2+nm)\log B$  where B depends on the range of  $\chi$ . If the range of  $\chi$  is less than  $\sqrt{q}$ , then both the distributions have different randomness and thus are statistically far-apart. So, in general, nothing conclusive can be said about statistical indistinguishability of Small Secrets LWE-Matrix Version as the indistinguishability can change depending upon the range of  $\chi$ .

#### Question 4: Full Rank Matrices vs Noisy Low Rank Matrices

**Question.** Let  $\mathcal{D} = \{(A, A.B + E) : A \leftarrow \mathbb{Z}_q^{2n \times n}, B \leftarrow \mathbb{Z}_q^{n \times 2n}, E \leftarrow \chi^{2n \times 2n}\}.$  Show that  $\mathcal{D}$  is computationally indistinguishable from the uniform distribution over  $\mathbb{Z}_q^{2n \times n} \times \mathbb{Z}_q^{2n \times 2n}$ .

Proof. Let us start by defining 3 Distributions over  $\mathbb{Z}_q^{n\times 2n}\times \mathbb{Z}_q^{n\times 2n}$ :  $\mathcal{D}_0 = \{(A^T, B^T.A^T + E^T): A^T \leftarrow \mathbb{Z}_q^{n\times 2n}, B^T \leftarrow \mathbb{Z}_q^{2n\times n}, E^T \leftarrow \chi^{2n\times 2n}\},$ 

$$\mathcal{D}_1 = \{ (A^T, p) \text{ where } p = [p_{ij}] \text{ and } p_{ij} = \begin{cases} e_{ij} + \sum_{k=0}^n b_{ik} \times a_{kj}, & \text{if } i \leq n \\ q, q \leftarrow \mathbb{Z}_q, & \text{if } i > n \end{cases}$$

where  $b_{ik} \leftarrow B^T, a_{kj} \leftarrow A^T, e_{ij} \leftarrow E^T$ 

$$\mathcal{D}_2 = \{ (A^T, U^T) : A^T \leftarrow \mathbb{Z}_q^{n \times 2n}, U^T \leftarrow \mathbb{Z}_q^{2n \times 2n} \}.$$

Claim 4.1. Distributions  $\mathcal{D}_0$  and  $\mathcal{D}_1$  are computationally indistinguishable.

*Proof.* The Distributions  $\mathcal{D}_0$  and  $\mathcal{D}_1$  differ only at the last n rows of the second matrix which forms a submatrix of dimensions  $n \times 2n$ .

We can divide the Distribution  $\mathcal{D}_0$  into two Distributions  $\mathcal{D}_{01}$  and  $\mathcal{D}_{02}$  over  $\mathbb{Z}_q^{n \times 2n} \times \mathbb{Z}_q^{n \times 2n}$  where  $\mathcal{D}_{01}$  consists of first n rows of Second matrix in  $\mathcal{D}_0$  and  $\mathcal{D}_{02}$  consists of last n rows of Second Matrix.

Similarly we can create Distributions  $\mathcal{D}_{11}$  and  $\mathcal{D}_{12}$  from  $\mathcal{D}_1$ .

Now We can see that the Distributions  $\mathcal{D}_{01}$  and  $\mathcal{D}_{11}$  are identical.

So to show  $\mathcal{D}_0$  and  $\mathcal{D}_1$  are computationally indistinguishable, it suffices to show that  $\mathcal{D}_{02}$  and  $\mathcal{D}_{12}$  are computationally indistinguishable.

now, 
$$\mathcal{D}_{02} = \{(A^T, F) : F = B'.A^T + E', A^T \leftarrow \mathbb{Z}_q^{n \times 2n}, B' \leftarrow \mathbb{Z}_q^{n \times n}, E' \leftarrow \mathbb{Z}_q^{n \times 2n}\}$$
  
 $B'$  is bottom n rows of  $B^T$ ,  $E'$  is bottom n rows of  $E^T$ , and
$$\mathcal{D}_{12} = \{(A^T, F) : A^T, F \leftarrow \mathbb{Z}_q^{n \times 2n}\}$$

Form Questions 3, we know about the matrix version of Small Secrets LWE. As LWE and Small Secrets LWE are equally hard, we can get the matrix version of LWE.

Using the matrix Version of LWE,  $\mathcal{D}_{02}$  and  $\mathcal{D}_{12}$  are computationally indistinguishable.

Therefore, Distributions  $\mathcal{D}_0$  and  $\mathcal{D}_1$  are computationally indistinguishable.

Claim 4.2. Distributions  $\mathcal{D}_1$  and  $\mathcal{D}_2$  are computationally indistinguishable.

*Proof.* The Distributions  $\mathcal{D}_1$  and  $\mathcal{D}_2$  differ only at the first n rows of the second matrix which forms a submatrix of dimensions  $n \times 2n$ .

We can divide the Distribution  $\mathcal{D}_1$  into two Distributions  $\mathcal{D}_{11}$  and  $\mathcal{D}_{12}$  over  $\mathbb{Z}_q^{n \times 2n} \times \mathbb{Z}_q^{n \times 2n}$  where  $\mathcal{D}_{11}$  consists of first n rows of Second matrix in  $\mathcal{D}_1$  and  $\mathcal{D}_{12}$  consists of last n rows of Second Matrix.

Similarly we can create Distributions  $\mathcal{D}_{21}$  and  $\mathcal{D}_{22}$  from  $\mathcal{D}_{2}$ .

Now We can see that the Distributions  $\mathcal{D}_{12}$  and  $\mathcal{D}_{22}$  are identical.

So to show  $\mathcal{D}_1$  and  $\mathcal{D}_2$  are computationally indistinguishable, it suffices to show that  $\mathcal{D}_{11}$ and  $\mathcal{D}_{21}$  are computationally indistinguishable.

now, 
$$\mathcal{D}_{11} = \{(A^T, F) : F = B'.A^T + E', A^T \leftarrow \mathbb{Z}_q^{n \times 2n}, B' \leftarrow \mathbb{Z}_q^{n \times n}, E' \leftarrow \mathbb{Z}_q^{n \times 2n}\}$$
  
 $B'$  is top n rows of  $B^T$ ,  $E'$  is top n rows of  $E^T$ , and
$$\mathcal{D}_{21} = \{(A^T, F) : A^T, F \leftarrow \mathbb{Z}_q^{n \times 2n}\}$$

Using the matrix Version of LWE,  $\mathcal{D}_{11}$  and  $\mathcal{D}_{21}$  are computationally indistinguishable. Therefore, Distributions  $\mathcal{D}_1$  and  $\mathcal{D}_2$  are computationally indistinguishable. 

From the above two claims, we can say that the Distributions  $\mathcal{D}_0$  and  $\mathcal{D}_2$  are computationally indistinguishable.

As Transposing a matrix is just changing the positions of the elements in the matrix, it cannot change the matrix computationally. Using this, we can define two new distributions by just taking the transpose of the matrices in the computationally indistinguishable Matrices which will also be computationally indistinguishable.

Taking Transpose,

$$\mathcal{D}'_0 = \{ (A, A.B + E) : A \leftarrow \mathbb{Z}_q^{2n \times n}, B \leftarrow \mathbb{Z}_q^{n \times 2n}, E \leftarrow \chi^{2n \times 2n} \}.$$
 and

$$\mathcal{D}'_2 = \{ (A, U) : A \leftarrow \mathbb{Z}_q^{2n \times n}, U \leftarrow \mathbb{Z}_q^{2n \times 2n} \}.$$

 $\mathcal{D}'_2 = \{(A, U) : A \leftarrow \mathbb{Z}_q^{2n \times n}, U \leftarrow \mathbb{Z}_q^{2n \times 2n}\}.$ here  $\mathcal{D}'_0 = \mathcal{D}$  and  $\mathcal{D}'_2$  is a uniform distribution over  $\mathbb{Z}_q^{2n \times n} \times \mathbb{Z}_q^{2n \times 2n}$ .

Thus  $\mathcal{D}$  is computationally indistinguishable from a Uniform Distribution over  $\mathbb{Z}_q^{2n\times n}$  ×  $\mathbb{Z}_q^{2n\times 2n}$ .

#### Question 5: Random-looking t matrices with a special structure

**Question.** Define distribution  $\mathcal{D}$  that is indistinguishable from  $(\mathbb{Z}_q^{n \times 2n})^t$  and no subset sum of an element sampled from this distribution has only small entries.

*Proof.* We define the distribution as follows:

$$\mathcal{D} = (\mathbf{B}_{0}, \mathbf{B}_{1}, \dots, \mathbf{B}_{t})$$

$$\mathbf{B}_{i} = [\mathbf{A}_{i} | \mathbf{C}_{i}], \forall i \in [t]$$

$$\mathbf{C}_{i} = \mathbf{S} \cdot \mathbf{A}_{i} + \mathbf{E}_{i} + \mathbf{D}$$

$$\mathbf{A}_{i} \leftarrow \mathbb{Z}_{q}^{n \times n}, \mathbf{S} \leftarrow \chi^{n \times n}, \mathbf{E}_{i} \leftarrow \chi^{n \times n}, \mathbf{D} \leftarrow d^{n \times n}$$

$$(1)$$

We now define  $\chi$  and d appropriately so that the required properties are satisfied. Consider any matrix  $\mathbf{B}_p$  from the tuple sampled from  $\mathcal{D}$ . This is made up of  $\mathbf{A}_p$  and  $\mathbf{C}_p$ . We will now choose d in such a way that exactly one of these two matrices can have *only small entries*. Now, if  $\mathbf{A}_p$  does not have *only small entries*, any value of d will work. Therefore, we consider the case when  $\mathbf{A}_p$  is made up of *only small entries*. Also, consider  $\chi$  to be of the form  $\mathsf{Unif}_{[-m,m]}$  where we have to determine the value of m (which has to be some power of q for LWE to remain a hard computational problem, say  $\alpha$ ). Also, WLOG, we assume d to be positive and it lies between 0 to q/2. We now consider the worst-case scenario such that d has to be the largest possible value:

$$\sum_{k \in [n]} (s_{ik} \cdot a_{kj}) + e_{ij} + d > q^{0.75}$$

$$\implies d > q^{0.75} - e_{ij} - \sum_{k \in [n]} (s_{ik} \cdot a_{kj})$$

Taking  $e_{ij}$ ,  $a_{kj}$ ,  $s_{ik}$  to be the largest, with appropriate signs

$$> q^{0.75} + m + n \times (q^{0.75} \cdot m)$$

$$> q^{0.75} + q^{\alpha} + n \cdot q^{\alpha + 0.75}$$

$$\implies d \ge 1 + q^{0.75} \times (1 + n \cdot q^{\alpha}) + q^{\alpha}$$

$$\ge q^{0.75} \times (3 + n \cdot q^{\alpha})$$

$$\ge (n+3) \cdot q^{\alpha + 0.75}$$

$$> q^{\epsilon + \alpha + 0.75} = q^{\alpha' + 0.75}$$

We replace (n+3) by  $q^{\epsilon}$  since we are dealing with large numbers and  $O(\log^2(q)) < O(q^{\gamma})$ . Now, since we assumed that d has to be  $\leq q/2 = O(q)$ , therefore,  $\alpha < \alpha' < 0.25$ . Also, let  $\alpha' + 0.75 = \beta$ . We cannot have  $d = \Theta(q)$  since this can lead to the sum ending up in the range of *only small entries* which is something that we do not want. Therefore, we conclude that d = o(q).

This construction also satisfies the requirement for a subset sum since for any subset  $T \subseteq [t]$ , we can consider  $\mathbf{A}' = \sum_{i \in T} \mathbf{A}_i$  and the noise  $\sum_{i \in T} \mathbf{E}_i$  only increases by a factor of  $t = \mathsf{poly}(n)$  which is much lesser than  $d = \Theta(q^\beta)$ .

We now show that this distribution is computationally indistinguishable from the uniform distribution over  $(\mathbb{Z}^{n\times 2n})^t$ . Notice that the distribution is equivalent to the following distribution:

$$\{(\mathbf{A}_i, \mathbf{S} \cdot \mathbf{A}_i + \mathbf{E}_i + \mathbf{D})\}_{i \in [t]}$$
(3)

Now, from the solution of Question 3.1, we know that the matrix-version of ss-LWE<sub> $n,m,q,\chi$ </sub> is also a hard computational problem. Our distribution can be reduced to matrix version of ss-LWE<sub> $n,tm,q,\chi$ </sub> by just appending the t matrices column-wise. The addition of  $\mathbf{D}$  does not affect the reduction since  $\mathbf{D}$  is just a constant and a constant being added to a uniformly random number is also uniformly random. Therefore,  $\mathcal{D}$  is computationally indistinguishable from the uniform distribution and it also ensures that the event with a subset sum having only small entries has a probability of 0.

Therefore, we have shown how to sample  $\mathcal{D}$  with  $\chi = \mathsf{Unif}_{[-q^{\alpha},q^{\alpha}]}$  and  $d = q^{0.75 + \alpha + \epsilon}$  such that  $0 < \alpha < \alpha + \epsilon < 0.25$ .

#### 6.1 Question 6.1

#### Question 6.1: Code Obfuscation

Question. What is the issue with the attempt involving integers?

*Proof.* The proposed **Attempt 1** does not ensure correctness since it is possible that for some other integer z',  $\Sigma_i a_{i,z'_i} = 0$  since all entries other than  $a_{t,z_t}$  are randomly sampled. For instance, consider the following case when  $a_{t,1-z_t} = -\Sigma_{i < t} a_{i,z_i} \pmod{q}$ . Then,  $\mathsf{Eval}(\mathsf{Obf}(f_z), z) = \mathsf{Eval}(\mathsf{Obf}(f_z), z \oplus 1) = 1$ , which is incorrect.

#### 6.2 Question 6.2, 6.3, 6.4

#### Question 6.1: Code Obfuscation

**Question.** Propose an attempt at obfuscation of  $f_z$  using matrices and ideas developed in Question 5. Describe Obf and Eval. Show that your scheme satisfies correctness. Prove security of your scheme, assuming ss-LWE<sub>n,m,q,\chi</sub>. Note that t must be large for this proof to work. How large must t be?

*Proof.* We use a similar sampling approach as done in Question 5. We choose  $\chi$  and d such that they form a valid solution to Question 5. For simplicity we first define the following function:

$$\mathsf{Sample-Dirac-Matrix}_{\chi}(\mathbf{A},\mathbf{S},\mathbf{D}) = [\mathbf{A}|\mathbf{S}\cdot\mathbf{A} + \mathbf{E} + \mathbf{D}], \text{ where } \mathbf{E} \leftarrow \chi^{n\times n} \tag{4}$$

We define  $\mathsf{Obf}(f_z)$  as:

$$\begin{aligned} \mathsf{Obf}(f_z) &= (\mathbf{B}_{1,0}, \mathbf{B}_{1,1}, \mathbf{B}_{2,0}, \mathbf{B}_{2,1}, \dots, \mathbf{B}_{t,0}, \mathbf{B}_{t,1}) \,, \, \text{where} \\ \mathbf{B}_{i,z_i} &= \mathsf{Sample-Dirac-Matrix}_{\chi}(\mathbf{A}_{i,z_i}, \mathbf{S}, 0), 1 \leq i < t \\ \mathbf{B}_{t,z_t} &= \mathsf{Sample-Dirac-Matrix}_{\chi}(-\sum_{i=1}^{t-1} \mathbf{A}_{i,z_i}, \mathbf{S}, 0) \\ \mathbf{B}_{i,1-z_i} &= \mathsf{Sample-Dirac-Matrix}_{\chi}(\mathbf{A}_{i,1-z_i}, \mathbf{S}, \mathbf{D}), i \in [t] \end{aligned} \tag{5}$$

$$\begin{aligned} \mathbf{B}_{i,1-z_i} &= \mathsf{Sample-Dirac-Matrix}_{\chi}(\mathbf{A}_{i,1-z_i},\mathbf{S},\mathbf{D}), i \in [t] \\ \mathbf{S} &\leftarrow \chi^{n \times n} \text{ and } \forall (i,b) \in ([t] \times \{0,1\}) \setminus \{(t,z_t)\} : \mathbf{A}_{i,b} \leftarrow \mathbb{Z}_q^{n \times n} \end{aligned}$$

Now, Eval(Obf $(f_z)$ , x) is defined as:

$$\mathsf{Eval}(\mathsf{Obf}(f_z), x) = \begin{cases} 1 & \text{if } \sum_{i \in [t]} (\mathbf{B}_{i, x_i}) \text{ has only small entries} \\ 0 & \text{otherwise} \end{cases}$$
 (6)

For correctness, notice that when x = z the sum is  $[0|\sum_{i \in [t]} \mathbf{E}_{i,z_i}]$ . Since the noise is much smaller than  $q^{0.75}$ , it is guaranteed that the sum will have *only small entries*. Therefore,  $\mathsf{Eval}(\mathsf{Obf}(f_z), z) = 1$ . Now, we need to show that the evaluation results in 0 for all other points. Consider any  $x \neq z$ . Now, this number differs from the binary representation of z at

at least one index. Therefore,  $Eval(Obf(f_z), x)$  will be of the form:

$$[\mathbf{A}'|\mathbf{S}\cdot\mathbf{A}' + \sum_{i\in[t]}\mathbf{E}_{i,x_i} + c\cdot\mathbf{D}], t \ge c > 0$$
(7)

Now, using the same arguments as in Question ??, we can show that at least one entry of this matrix is not small and therefore,  $\mathsf{Eval}(\mathsf{Obf}(f_z), x) = 0$ . Therefore, the proposed scheme satisfies correctness.

We will now prove the security of our scheme by constructing computationally indistinguishable distributions until we arrive at a uniformly random distribution:

- Distribution  $\mathcal{D}_1$ : Same as  $\mathsf{Obf}(f_z)$  but  $\mathsf{Sample-Dirac-Matrix}_\chi'$  is used where  $\mathsf{Sample-Dirac-Matrix}_\chi'(\mathbf{A}, \mathbf{S}, \mathbf{M}) = \mathsf{Sample-Dirac-Matrix}_\chi(\mathbf{A}, \mathbf{S}, \mathbf{D})$  (i.e., same  $\mathbf{D}$  is used for all the samples). This is statistically indistinguishable from the distribution obtained from  $\mathsf{Obf}(f_z)$  since we are only changing the value of the constant matrix being added to the  $2^{\mathrm{nd}}$  half of the matrix. As shown in Question 5, the distribution  $\mathcal{D}$  is computationally indistinguishable from a uniformly random matrix irrespective of the constant value that is added. Therefore,  $\mathsf{Obf}(f_z)$  is statistically indistinguishable from  $\mathcal{D}_1$ .
- Distribution  $\mathcal{D}_2$ : We sample  $\mathbf{B}_{i,b}$  uniformly randomly for  $i \in [t], b \in \{0, 1\}$  except when  $(i, b) = (t, z_t)$ . We set  $\mathbf{B}_{t, z_t}$  the same way as it has been set in  $\mathcal{D}_1$ . Now, from hardness of ss-LWE<sub> $n,(2t-1)n,q\chi$ </sub>, we can claim that  $\mathcal{D}_1$  and  $\mathcal{D}_2$  are computationally indistinguishable since we can reduce both these distribution into distributions from the ss-LWE  $_{n,(2t-1)n,q\chi}$  problem.
- Distribution  $\mathcal{D}_3$ : We sample all 2t matrices uniformly randomly. Now, from **Fact** 1 given in the assignment statement, we can say that  $\mathcal{D}_2$  and  $\mathcal{D}_3$  are statistically indistinguishable if t is large enough. In this instance, m = 2n, and therefore,  $t \geq n \cdot 2n \cdot \log q + n = 2n^2 \cdot \log q + n = 2n^{2.5} + n$  (since  $q = 2^{\sqrt{n}}$ ).

Therefore, we have shown that the matrix sampled from  $\mathsf{Obf}(f_z)$  is computationally indistinguishable from a uniformly random tuple of matrices. Therefore, our scheme  $\mathsf{Obf}$  is secure as well.