# A Probability Inequality with Application to Lattice Theory 

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#### Abstract

Here we mainly provide a probability inequality about GGH public-key encryption scheme. Given a constant $\sigma$, we first choose a lattice vector $v \in \mathbb{Z}^{n}$, and a small error vector $e$ is generated satisfying $|e| \leqslant \sigma$. The ciphertext result $c$ could be computed by the function $f_{B, \sigma}(v, e)=B v+e$ with a public basis $B$. To extract the message $v$, the function $f_{B, \sigma}^{-1}(c)=B^{-1}[c]_{R}$ will be used based on the private basis $R$. In this work we produce a bound for the error probability of $v \neq B^{-1}[c]_{R}$. We also illustrate the way choosing $\sigma$ such that the error probability is arbitrarily small.


Keywords Probability inequality • Encryption scheme • Lattice

## 1 Introduction

Given a full-rank lattice $L \subset \mathbb{Z}^{n}$, we denote the public basis of $L$ by $B$ and private basis of $L$ by $R$. Both $B$ and $R$ are $n \times n$ invertible matrices. In the GGH public-key encryption scheme, for a plaintext vector $v \in \mathbb{Z}^{n}$, the random error vector $e$ is chosen by setting the absolute value of each entry no more than a constant $\sigma$, where $\sigma$ is a positive real number. The ciphertext $c$ is computed by $c=f_{B, \sigma}(v, e)=B v+e \in \mathbb{R}^{n}$. Using the results of BaBai and some other ones (Ajtai, 1996; Ajtai \& Dwork, 1997; Babai, 1986; Coppersmith \& Shamir, 1997; Goldreich et al., 1997; Micciancio, 2001; Hoffstein et al., 2017, 1998), we can decipher the plaintext $v=B^{-1}[c]_{R}$ given $B$, $R$ and ciphertext $c$. Here the lattice point $[c]_{R}$ is obtained by representing $c$ as a linear combination on the columns of $R$ and rounding the coefficients in this linear combination to the nearest integers. The problem is that how $\sigma$ should be chosen so that we can get a right plaintext $v$ or guarantee a low error probability. We show three theorems to solve this problem. A probability inequality is given to estimate the bound of inversion error probability.

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## 2 Main Results

Theorem $1 B$ is the public basis and $R$ is the private basis of lattice $L . v \in \mathbb{Z}^{n}$, e is the random error vector, $|e|_{\infty} \leqslant \sigma, c=f_{B, \sigma}(v, e)=B v+e$. Then $B^{-1}[c]_{R}=v$ if and only if $\left[R^{-1} e\right]=0$, here $\left[R^{-1} e\right]$ denotes the vector in $\mathbb{Z}^{n}$ which is obtained by rounding each entry in $R^{-1} e$ to the nearest integer.

Proof Let $T=B^{-1} R$, then

$$
B^{-1}[c]_{R}=B^{-1}[B v+e]_{R}=B^{-1} R\left[R^{-1}(B v+e)\right]=T\left[T^{-1} v+R^{-1} e\right]
$$

since $T=B^{-1} R$ is a unimodular matrix, $T^{-1}$ is also a unimodular matrix. $v \in \mathbb{Z}^{n}$, so $T^{-1} v \in \mathbb{Z}^{n}$.

$$
B^{-1}[c]_{R}=T\left[T^{-1} v+R^{-1} e\right]=v+T\left[R^{-1} e\right]
$$

Thus $B^{-1}[c]_{R}=v$ is equivalent to $T\left[R^{-1} e\right]=0$, and this equality holds if and only if $\left[R^{-1} e\right]=0$.

Remark 1 This theorem gives an equivalent condition to check whether the decryption result is accurate.

Theorem 2 Let $R$ be the private basis of lattice L. e is the random error vector such that $|e|_{\infty} \leqslant \sigma$. Suppose the maximum $L_{1}$ norm of the rows in $R^{-1}$ is $\rho$. Then if $\sigma<\frac{1}{2 \rho},\left[R^{-1} e\right]=0$ holds.

Proof Let $R^{-1}=\left(c_{i j}\right)_{n \times n}, R^{-1} e=\left(a_{1}, a_{2}, \ldots, a_{n}\right)^{T}$, i.e., $a_{i}=\sum_{j=1}^{n} c_{i j} e_{j}, 1 \leqslant i \leqslant$ $n$.

$$
\left|a_{i}\right|=\left|\sum_{j=1}^{n} c_{i j} e_{j}\right| \leqslant\left|e_{j}\right|\left|\sum_{j=1}^{n} c_{i j}\right| \leqslant \sigma \rho<\frac{1}{2}
$$

This means that $\left[R^{-1} e\right]=0$.
Remark 2 Theorem 2 shows how $\sigma$ can be chosen so that no inversion error occurs.

Theorem 3 Let an $n \times n$ matrix $R$ be the private basis used in the inversion of $f_{B, \sigma}$, and denote the maximum $L_{\infty}$ norm of the rows in $R^{-1}$ by $\frac{r}{\sqrt{n}}$. Then the probability of inversion errors is bounded by

$$
P\left\{\left[R^{-1} e\right] \neq 0\right\} \leqslant 2 n \cdot \exp \left(-\frac{1}{8 \sigma^{2} r^{2}}\right),
$$

here $e=\left(e_{1}, e_{2}, \ldots, e_{n}\right)^{T}$ and $e_{1}, e_{2}, \ldots, e_{n}$ are $n$ independent random variables such that $\left|e_{i}\right| \leqslant \sigma$ and $E\left(e_{i}\right)=0$ for $1 \leqslant i \leqslant n$.

Lemma 1 For any non-negative random variable $X$ with finite expectation $E(X)$ and any positive real number $\mu$, we have

$$
P\{X \geqslant \mu\} \leqslant \frac{E(X)}{\mu} .
$$

Proof Here we treat $X$ as a random variable of continuous type. For the other situations, the proof is similar. Let $f(x)$ be the probability density function of $X$. Since $E(X)=\int_{0}^{+\infty} x f(x) \mathrm{d} x \geqslant \int_{\mu}^{+\infty} x f(x) \mathrm{d} x \geqslant \int_{\mu}^{+\infty} \mu f(x) \mathrm{d} x=\mu P\{X \geqslant \mu\}$, then we have $P\{X \geqslant \mu\} \leqslant \frac{E(X)}{\mu}$.

Lemma 2 Given random variable $X$ satisfying $-a \leqslant X \leqslant a$ with $E(X)=0$, here $a>0$. For any real number $\lambda$, we have

$$
E\left(e^{\lambda X}\right) \leqslant \exp \left(\frac{\lambda^{2} a^{2}}{2}\right)
$$

Proof For any real number $\lambda, f(x)=e^{\lambda x}$ is a convex function. Notice that

$$
x=\frac{x+a}{2 a} \cdot a+\frac{a-x}{2 a} \cdot(-a), \quad-a \leqslant x \leqslant a
$$

then

$$
\begin{gathered}
f(x) \leqslant \frac{x+a}{2 a} f(a)+\frac{a-x}{2 a} f(-a) \\
e^{\lambda x} \leqslant \frac{x+a}{2 a} e^{\lambda a}+\frac{a-x}{2 a} e^{-\lambda a} \\
E\left(e^{\lambda X}\right) \leqslant E\left(\frac{X+a}{2 a} e^{\lambda a}+\frac{a-X}{2 a} e^{-\lambda a}\right)=\frac{1}{2}\left(e^{\lambda a}+e^{-\lambda a}\right)
\end{gathered}
$$

Let $t=\lambda a$, next we prove that $\frac{1}{2}\left(e^{t}+e^{-t}\right) \leqslant \exp \left(\frac{t^{2}}{2}\right)$. This inequality is equivalent to

$$
\ln \frac{e^{t}+e^{-t}}{2} \leqslant \frac{t^{2}}{2}
$$

Let $g(t)=\frac{t^{2}}{2}-\ln \frac{e^{t}+e^{-t}}{2}$, then $g^{\prime}(t)=t-\frac{e^{t}-e^{-t}}{e^{t}+e^{-t}}$ and $g^{\prime}(0)=0$. Since $g^{\prime \prime}(t) \geqslant 0$, we get $g^{\prime}(t) \leqslant 0$ if $t \leqslant 0$ and $g^{\prime}(t) \geqslant 0$ if $t \geqslant 0$. Then $g(t) \geqslant g(0)=0$ and we complete the proof.

Lemma 3 Suppose $X_{1}, X_{2}, \ldots, X_{n}$ are $n$ independent random variables. For $1 \leqslant$ $i \leqslant n$, we have $-a \leqslant X_{i} \leqslant a$ and $E\left(X_{i}\right)=0$, here $a>0$. Let $S_{n}=\sum_{i=1}^{n} X_{i}, \varepsilon>0$, then

$$
P\left\{\left|S_{n}\right| \geqslant \varepsilon\right\} \leqslant 2 \exp \left(-\frac{\varepsilon^{2}}{2 n a^{2}}\right)
$$

Proof For any $\lambda>0$, based on Lemma 1, we can get

$$
P\left\{S_{n} \geqslant \varepsilon\right\}=P\left\{e^{\lambda S_{n}} \geqslant e^{\lambda \varepsilon}\right\} \leqslant \frac{E\left(e^{\lambda S_{n}}\right)}{e^{\lambda \varepsilon}}
$$

Since $X_{1}, X_{2}, \ldots, X_{n}$ are independent random variables, combine with Lemma 2,

$$
\begin{gathered}
E\left(e^{\lambda S_{n}}\right)=\prod_{i=1}^{n} E\left(e^{\lambda X_{i}}\right) \leqslant \prod_{i=1}^{n} e^{\frac{\lambda^{2} a^{2}}{2}}=e^{\frac{n \lambda^{2} a^{2}}{2}} \\
P\left\{S_{n} \geqslant \varepsilon\right\} \leqslant \frac{E\left(e^{\lambda S_{n}}\right)}{e^{\lambda \varepsilon}} \leqslant e^{-\lambda \varepsilon+\frac{n \lambda^{2} a^{2}}{2}}
\end{gathered}
$$

Let $\lambda=\frac{\varepsilon}{n a^{2}}$, therefore, the above inequality becomes to

$$
P\left\{S_{n} \geqslant \varepsilon\right\} \leqslant \exp \left(-\frac{\varepsilon^{2}}{2 n a^{2}}\right)
$$

In the same way, we can prove that

$$
P\left\{S_{n} \leqslant-\varepsilon\right\} \leqslant \exp \left(-\frac{\varepsilon^{2}}{2 n a^{2}}\right)
$$

Thus

$$
P\left\{\left|S_{n}\right| \geqslant \varepsilon\right\} \leqslant 2 \exp \left(-\frac{\varepsilon^{2}}{2 n a^{2}}\right)
$$

Proof of Theorem 3. Now we can prove Theorem 3 given at first according to Lemma 3.

Let $R^{-1}=\left(c_{i j}\right)_{n \times n}, e=\left(e_{1}, e_{2}, \ldots, e_{n}\right)^{T}$, here $e_{1}, e_{2}, \ldots, e_{n}$ are $n$ independent random variables such that $\left|e_{i}\right| \leqslant \sigma$ and $E\left(e_{i}\right)=0$ for $1 \leqslant i \leqslant n$.
We denote $R^{-1} e=\left(a_{1}, a_{2}, \ldots, a_{n}\right)^{T}$, i.e., $a_{i}=\sum_{j=1}^{n} c_{i j} e_{j}, 1 \leqslant i \leqslant n$.
Since $\left|c_{i j}\right| \leqslant \frac{r}{\sqrt{n}}$ and $\left|e_{j}\right| \leqslant \sigma$, then the random variable $c_{i j} e_{j}$ is limited to the interval $\left[-\frac{r \sigma}{\sqrt{n}}, \frac{r \sigma}{\sqrt{n}}\right]$. Based on Lemma 3,

$$
\begin{gathered}
P\left\{\left|a_{i}\right| \geqslant \frac{1}{2}\right\}=P\left\{\left|\sum_{j=1}^{n} c_{i j} e_{j}\right| \geqslant \frac{1}{2}\right\} \leqslant 2 \exp \left(-\frac{\left(\frac{1}{2}\right)^{2}}{2 n\left(\frac{r \sigma}{\sqrt{n}}\right)^{2}}\right)=2 \exp \left(-\frac{1}{8 \sigma^{2} r^{2}}\right) \\
P\left\{\left[R^{-1} e\right] \neq 0\right\} \leqslant \sum_{i=1}^{n} P\left\{\left|a_{i}\right|>\frac{1}{2}\right\} \leqslant \sum_{i=1}^{n} P\left\{\left|a_{i}\right| \geqslant \frac{1}{2}\right\} \leqslant 2 n \cdot \exp \left(-\frac{1}{8 \sigma^{2} r^{2}}\right)
\end{gathered}
$$

Thus the inequality in Theorem 3 holds.

Corollary $1 \quad P\left\{\left[R^{-1} e\right] \neq 0\right\}<\varepsilon$ if $\sigma<\left(2 r \sqrt{2 \ln \frac{2 n}{\varepsilon}}\right)^{-1}$.
Proof $\sigma<\left(2 r \sqrt{2 \ln \frac{2 n}{\varepsilon}}\right)^{-1} \Leftrightarrow 2 n \cdot \exp \left(-\frac{1}{8 \sigma^{2} r^{2}}\right)<\varepsilon$, from Theorem 3,

$$
P\left\{\left[R^{-1} e\right] \neq 0\right\} \leqslant 2 n \cdot \exp \left(-\frac{1}{8 \sigma^{2} r^{2}}\right)<\varepsilon
$$

Remark 3 Theorem 3 provides a way to estimate the bound of inversion error probability, and Corollary 1 gives a detailed bound for $\sigma$ based on Theorem 3 to get the error probability no more than a constant $\varepsilon$.

## 3 Conclusions

In this work we mainly present a probability inequality about GGH public-key encryption scheme. In this scheme, we first take a lattice vector $v \in \mathbb{Z}^{n}$ and generate a small error vector $e$ such that $|e| \leqslant \sigma$. Given a public basis $B$, the function $f_{B, \sigma}(v, e)=B v+e$ computes the ciphertext result $c$. To decrypt, the private basis $R$ and the function $f_{B, \sigma}^{-1}(c)=B^{-1}[c]_{R}$ will be used to extract the message $v$. We give a bound for the error probability of $v \neq B^{-1}[c]_{R}$ and explain how to choose $\sigma$ in order to obtain the error probability no more than a given constant $\varepsilon$.

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