

Lattice

2021年4月19日 13:31

Recap:

Lattice. Given k independent vectors $b_1, \dots, b_k \in \mathbb{R}^n$.

$$L(b_1, \dots, b_n) = \left\{ \sum_{i=1}^n x_i b_i \mid x_i \in \mathbb{Z} \right\}$$

Problems.

1. SVP: Given B , find the Shortest Vector $v \neq 0^n$, s.t. $\|v\|_2$ is the shortest in $L(B)$
2. t -Approximate SVP. Find $v \neq 0^n$, $\|v\|_2 \leq t \cdot \|v^*\|_2$.
3. $\lambda_1(L(B)) = \|v^*\|_2$
 $\lambda_2(L(B)) =$ SV length linearly independent with v^*
 \dots
 $\lambda_n(L(B)) = \dots$

LLL Algorithm $\Rightarrow 2^{2/n}$ approximation of SVP.

Ajtai's One-Way function

n, q, m, β . $q = \text{poly}(n)$, $m = \Omega(n \log q)$, $\beta = O(\sqrt{m})$.
"q = n^2" $\rightarrow 2^m > q^n$, so range > image.

Matrix A : $n \times m$ $\leftarrow \sum_{q=1}^m A^{n \times m}$

$$f_A(x) = Ax \pmod{q}$$

$(x \in \{0, 1\}^m \text{ s.t. } x \neq 0^m)$.

OWness: \forall ppt Adv. \exists negl ϵ , s.t.

$$\Pr_{A, x} [\text{Adv}(A, y = Ax \pmod{q}) \rightarrow x', Ax' \pmod{q} = y \text{ and } \|x'\|_2 < \beta \ \& \ x' \neq 0] < \epsilon$$

Short-Integer Solution (SIS)

$A \leftarrow \sum_{q=1}^{n \times m}$, find x s.t. $Ax = 0 \pmod{q}$, $x \neq 0^n$ and $\|x\|_2 < \beta$

Theorem. $q = 2^{2n}$, $m = n^2$, $\beta = \sqrt{m}$

Then if \exists algo solves SIS, then \exists algo that solves t -approx-SBP for $t \in \text{poly}(n)$

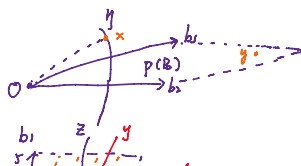
Shortest-basis
 $v_1, \dots, v_n, \|v_i\|_2 \leq \lambda_n \cdot t$

Worst-Case to Average-Case Reduction.

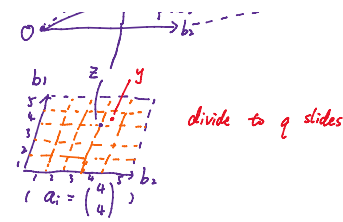
Pick η as the radius of a ball

$$\eta \in \{2\lambda_n, \dots, 2^n \lambda_n\}$$

$$x_i \leftarrow \text{Ball}(\eta)$$



$y \in \{2^n, \dots, 2^{n+1}\}$
 $x_i \leftarrow \text{Ball}(q)$
 $y_i = x_i \bmod P(B)$
 $a_i = [qB^{-1}y_i] \quad z_i = \frac{B a_i}{y}$



Let $A = [a_1, \dots, a_m]$
 Suppose Algo CA $\rightarrow w \in \mathbb{Z}^m, \|w\| \leq \beta, Aw = 0 \pmod q$.

$\sum_{i=1}^m w_i (x_i - y_i + z_i)$ is an approx. SVP.

- Proof:
- a_i distributes evenly over \mathbb{Z}_q^n (big q)
 - $v \in L(B)$.
 - v is short (small y)
 - v is not zero. (big q)
- $\Rightarrow 2. \quad v = \sum_{i=1}^m w_i (x_i - y_i + z_i)$
 $\sum_{i=1}^m w_i z_i = \sum_{i=1}^m w_i \frac{B a_i}{q} = B \sum_{i=1}^m \frac{w_i a_i}{q} \in L(B)$ (mod $q = 0$)
 $\Rightarrow 3. \quad w = \sum_{i=1}^m w_i (x_i - y_i + z_i)$
 $= \sum_{i=1}^m w_i x_i + \sum_{i=1}^m w_i (z_i - y_i)$
 $\|w\| \leq \beta$
 intuitively bounded by $\frac{1}{q}$ diameter $(P(B))$
 $\circledast y \sim \log n \cdot \lambda_n$
 (compromise to 1.8.4.) $q \sim 2^n$

Learning with Errors

- Recap. Secret vector $s = (s_1, \dots, s_n) \in \mathbb{Z}_q^n$
 Oracle $\text{Oss} \Rightarrow \{ \sum_{i=1}^n a_i s_i + e \bmod q; a_i \}$, $a \leftarrow \mathbb{Z}_q^n$, $n \leq B < q$
 $e \leftarrow [-B, B]$
 Goal: Find s .



Define. Statistically close X, Y . $X \approx_s Y$ if distribution on A

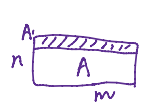
Given $n \times m$ matrix A , $y = A^T s + e \bmod q$.
 $e_i \leftarrow e^{-\frac{x_i^2}{\sigma^2}}$, $\sigma = O(\sqrt{n})$.

$\sum_{a \in A} |Pr[X=a] - Pr[Y=a]| < \text{negl.}$

Decisional LWE

Decide either $(y = A^T s + e \bmod q)$
 or $(y = U \in \mathbb{Z}_q^m)$

Search LWE = Decision LWE



1) guess $s_i = k$. Let $y' = y - A^T \cdot k + (h \cdot 0) \cdot k$, $A' = \begin{matrix} A_i + h \\ \vdots \\ A_{i+n} \end{matrix}$
 i. if correct, A' & y' be another LWE \Rightarrow not random
 ii. if wrong, $A' \cdot (k - s) = A_i(k - s) + h(k - s) = \text{random}$

Public key Enc.

sk: LWE secret $s \leftarrow U(\mathbb{Z}_q^n)$
 pk: LWE sample $A \leftarrow U(\mathbb{Z}_q^{n \times m})$, $m = \Omega(n \log q)$, $q \geq n^2$
 $y = A^T s + e \bmod q$, $e \leftarrow X_\sigma$, $\sigma = O(\sqrt{n})$.

$\text{Enc}(b \in \{0,1\}) = \underbrace{A^T r \bmod q}_z, \langle y, r \rangle + \underbrace{b \cdot L \cdot \frac{q}{2}}_v \bmod q$ ($r \in U(m)$)

$\text{Dec}(z, v) = \begin{cases} 1 & \text{if } |v - \langle s, z \rangle - \frac{q}{2}| < \frac{q}{4} \\ 0 & \text{if } |v - \langle s, z \rangle - 0| < \frac{q}{4} \end{cases}$
 $v = S^T A^T r + e^T r + b L \frac{q}{2} \bmod q$

easily derived. $e_i \leftarrow X_\sigma$
 $\langle e, r \rangle \leftarrow X_\sigma$, $\sigma' = \sqrt{n} m \ll q \Rightarrow \text{Correctness}$

Security: $\text{Enc}(0) \approx U(\mathbb{Z}_q^n \times \mathbb{Z}_q) \approx \text{Enc}(1)$.

$$\langle e, r \rangle \leftarrow X_{\sigma}^n \cdot \sigma = \sqrt{n} m \ll q$$

$$\text{Security: } \text{Enc}_{pk}^{(D)} \approx_c \mathcal{U}(\mathbb{Z}_q^h \times \mathbb{Z}_q) \approx_c \text{Enc}_{pk}^{(C)}$$

$$\begin{aligned} & A \cdot r, \langle y, r \rangle + b \lfloor \frac{q}{2} \rfloor, |A, y \\ \approx_c & A \cdot r, \langle r', r \rangle + b \lfloor \frac{q}{2} \rfloor, |A, u \leftarrow \mathcal{U}(\mathbb{Z}_q^h) \\ \approx_c & \mathcal{U}(\mathbb{Z}_q^h \times \mathbb{Z}_q) \mid \mathcal{U}(\mathbb{Z}_q^{h+m}), u \end{aligned}$$

Fully Homomorphic Encryption

Def. A (public key) encryption is called a fully homomorphic encryption if:

$$\left\{ \begin{array}{l} \text{Gen} \rightarrow \text{pk}, \text{sk} \\ \text{Enc}_{pk}(m, r) \rightarrow c \\ \text{Dec}_{sk}(c) \rightarrow m \end{array} \right. \quad (\text{Still, } \{\text{pk}, \text{Enc}_{pk}(m_i)\} \approx_c \{\text{pk}, \text{Enc}_{pk}(m_i)\} \text{ for security.})$$

$$\text{Eval}_{pk}(f, c_1, \dots, c_k \in D_f) = C_{f, c_1, \dots, c_k} \quad \text{and} \quad \text{Dec}_{sk}(C_{f, c_1, \dots, c_k}) = f(m_1, \dots, m_k)$$

$\forall f \in \text{poly}$

Only show $f = \text{NAND}$.

[Gentry, Sahai, Waters, 13] Learning with Error \rightarrow leveled homomorphic encryption

modulus $q = 2^l$ and $\log n < l < n$

Gadget Matrix

$$g = [1 \ 2 \ 4 \ \dots \ 2^{l-1}]$$

$$G = \begin{pmatrix} g & & \\ & g & \\ & & \ddots \\ & & & g \end{pmatrix} \in \mathbb{Z}^{n \times nl}$$

Think about $Gx = t \pmod q$ as a SIS problem (is trivial)

Define $G^{-1}(t) = \begin{bmatrix} t_1 \\ t_2 \\ \vdots \\ t_n \end{bmatrix}^t \rightarrow$ the bit decomposition of $t_i \in \{0, \dots, 2^{l-1}\}$

$\Rightarrow G \cdot G^{-1}(t) = t \pmod q$

sk: $s \in \mathbb{Z}_q^n$ (LWE secret)

pk: $A \in \mathbb{Z}_q^{n \times m}, y = s^T A + e^T \pmod q$

where $e \leftarrow X_{\sigma}^m, q = 2^l, \sigma = \sqrt{n}, m = \Omega(n \log q)$

$m \in \{0, 1\}$

Enc_{pk}(μ):

1. $R \leftarrow \{0, 1\}^{m \times (n+1)l}$

2. $C = \begin{pmatrix} A \\ y \end{pmatrix} \cdot R + \mu G \pmod q$

Gadget $\in \mathbb{Z}^{(n+1) \times (n+1)l}$

Dec_{sk}(C):

$$(s^T, -1) \cdot C = (s^T, -1) \begin{pmatrix} A \\ y \end{pmatrix} \cdot R + \mu (s^T, -1) G$$

$$= (s^T A - y) R + \mu (s^T, -1) G$$

$$= \underbrace{-e^T R}_{\text{small}} + \mu (s^T, -1) G$$

if $\mu = 0$ 0.

$\mu = 1, (s, -1) G$ some entries $= \frac{q}{2}$

• $+ \text{ mod } q$:

$$C_1 + C_2 = C_+ = \underbrace{\begin{pmatrix} A \\ y \end{pmatrix}}_{\text{Small}} (R_1 + R_2) + (M_1 + M_2)G$$

• AND: C_1, C_2

$$\begin{aligned} \text{Eval: } C_1 \cdot G^{-1}(C_2) &\rightarrow \left[\begin{pmatrix} A \\ y \end{pmatrix} R_1 + M_1 G \right] G^{-1}(C_2) \\ &= \begin{pmatrix} A \\ y \end{pmatrix} \underbrace{R_1}_{\substack{\text{small} \\ q \gg m}} \underbrace{G^{-1}(C_2)}_{\substack{\text{small} \\ q \gg m}} + \underbrace{M_1 C_2}_{M_1 B R_2 + M_1 M_2 G} \\ &= B \tilde{R} + M_1 M_2 G. \\ &\quad (\text{where } \|\tilde{R}\|_\infty < m+1) \end{aligned}$$

Note that:

$$\begin{array}{ccccccc} \|\tilde{R}\|_\infty & \longrightarrow & \|\tilde{R}\|_\infty & \longrightarrow & \|\tilde{R}\|_\infty & \xrightarrow{\tilde{r}} & \\ 1 & & m+1 & & (m+1)^2 & & (m+1)^{\tilde{r}} \\ & & & & \text{Need } q > m^{\tilde{r}} & & \end{array}$$

level: (Reduce noise)

$$C, \text{ Enc}_{pk}(sk, r) = Csk$$

$$\text{Eval}(\text{Dec-and-ReEnc}, Csk, C_m) \rightarrow \text{Crefresh}$$