

EUCLIDEAN LATTICES: An algorithm for Low- Density Sum Subset Algorithm

The subset sum problem is NP-Complete in the worst case [4], where a proof for NP-Completeness is shown. Therefore in this lecture we will be discussing an algorithm which in polynomial time solve almost all subset sum problems of sufficiently low density. Before, we need to define various terms briefly.

1 Definitions

Definition 1. Euclidean Lattices: Let $(\vec{b}_i)_{i \leq n}$ be linearly independent vectors in \mathbb{R}^m . The lattice L of rank n spanned by (\vec{b}_i) is

$$\begin{aligned} L((\vec{b}_i))_{i \leq n} &= \sum_{n=1}^{\infty} \mathbb{Z} \cdot \vec{b}_i \\ &= \left\{ \sum x_i \cdot \vec{b}_i, x_i \in \mathbb{Z} \right\} \end{aligned} \tag{1}$$

The \vec{b}_i are called a basis of L . Denote the lattice L by $L(B)$, where

$$\begin{bmatrix} | & . & . & . & . & | \\ \vec{b}_1 & \vec{b}_2 & .. & .. & .. & \vec{b}_n \\ | & . & . & . & . & | \end{bmatrix} \in \mathbb{R}^{m \times n}$$

(In this lecture, we consider basis-vector into a matrix column-wise (i.e, L is $Im_{\mathbb{Z}}(B)$.)

Bases of L are related by a unimodular transformation, ($B = B' \cdot U, U \in GL_n(\mathbb{Z})$).

Main lattice invariants:

- The determinant of L : $\det(L) = \sqrt{\det(B^T) \cdot B}$
If B is square i.e, $B \in \mathbb{R}^{n \times n}$, $\det(L) = |\det(B)|$.
Geometric interpretation : determinant is the inverse of the density of the lattice L points in \mathbb{R}^m .
Algebraic interpretation: $|\mathbb{R}^m / L(B)| = \det(L)$.
- The first minimum of a lattice $L(B)$:

$$\lambda_1(L) = \min \left\{ r : \exists \vec{b} \in L \setminus \{\vec{0}\} : \|\vec{b}\| \leq r \right\}, \text{ where } \|\cdot\| \text{ is an Euclidean norm.}$$

In other words, the length of shortest vector in $L(B)$.

Fact 1. (Minkowski's 1st Theorem)

Let $m = n$. for any L of rank n ,

$$\lambda_1 \leq \sqrt{n} \cdot (\det(L))^{\frac{1}{n}}$$

The bound is tight up to a constant.

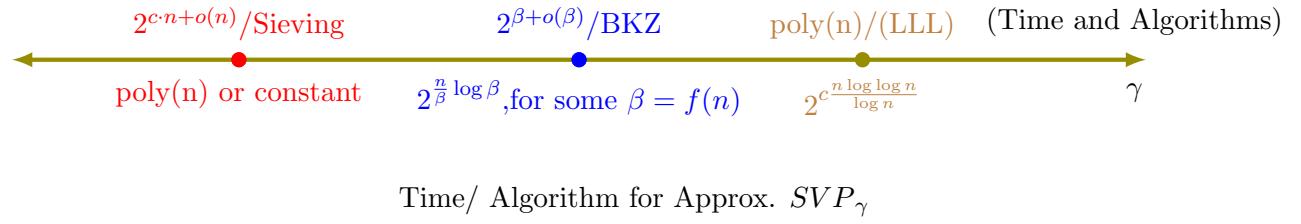
Definition 2. The Shortest Vector Problem (SVP) on L : It asks to find a non-zero vector $\vec{v} \in L$ such that $\|\vec{v}\| = \lambda_1(L)$.

- The solution to SVP is not unique.

Definition 3. The Approximation SVP with approximation factor $\gamma \geq 1$ (Approx SVP $_{\gamma}$) asks to find $\vec{v} \in L$ such that,

$$\|\vec{v}\| \leq \gamma \cdot \lambda_1(L)$$

In particular, the hardness of Approx SVP $_{\gamma}$ depends on γ .



At some point if $\gamma = 2^{\log^{1-\epsilon}}$, $\epsilon > 0$, SVP is NP-Hard.

Here,

- LLL - An algorithm due to Lenstra, Lenstra, Lovasz [5].
- BKZ - block Korkine- Zolotarev [6].

2 Estimate, Upper bounds.

2.1 Number of integers vectors in a ball:

$$\text{Denote } S_n(R) = \left\{ \vec{X} \in \mathbb{Z}^n : \left\| \vec{X} \right\|^2 \leq R \right\}$$

Number of vectors inside or on the n -dimensional sphere of radius \sqrt{R} centered at \vec{O} .

Theorem 1. For all $n \geq 1$, $S_n(\frac{n}{2}) \leq 2^{C_o \cdot n}$, where $C_o = 1.54725$.

Proof. Consider $\theta(z) = 1 + 2 \sum_{n=1}^{\infty} z^{2^n} = 1 + 2z + 2z^4 + \dots$, ($\theta_3(z)$: Jacobi θ_3 function), $\theta(z)$ is a θ -series of 1-dim integer lattice \mathbb{Z} .

In general for a lattice Λ , θ is defined as:

$$\theta_n(z) = \sum_{X \in \Lambda} q^{\|X\|^2 \cdot z}, \text{ where, } q = e^{+\pi i} = \sum_{m=1}^{\infty} N_m q^m \quad (2)$$

where N_m counts the number of lattice points at distance m from the origin.

Let,

$$\gamma_n(k) = \left\{ X \in \mathbb{Z}^n : \|X\|^2 \leq k \right\}$$

then

$$(\theta(z))^n = \sum_{k=1}^{\infty} \gamma_n(k) z^k.$$

Now to relate $S_n(R)$ to $(\theta(z))^n$.

for $x \geq 0$, it holds $e^{ndx} \cdot e^{-kx} \geq 1$, $k \leq nd$ then,

$$S_n(\alpha_n) = \sum_{k \leq \alpha n} \gamma_n(k) \leq e^{ndx} \sum_{k=0}^{\infty} \gamma_n(k) \cdot e^{-kx} = e^{ndx} [\theta(z)]^n$$

Taking ln, obtain

$$\ln S_n(\alpha_n) \leq n \cdot dx + n \cdot \ln(\theta(z)) = n(\alpha x + \ln \theta(z)), \text{ Take, } \alpha x + \ln \theta(z) = \delta(x)$$

Minimizing δ for $\alpha = \frac{1}{2}$, gives

$$S_n(\alpha_n) \leq 2^{1.54725n}$$

□

3 Low- Density Subset Sum(S)

Goal: Reduce an instance of low-density Subset Sum to SVP.

Remainder: The Subset Sum Problem: Given $a_1, a_2, \dots, a_n \xleftarrow{\$} [1, A]$, and $s = \sum_{i=1}^n e_i a_i$ with $e_i \in \{1, 0\}$, find $\vec{e} = (e_1, e_2, \dots, e_n)$.

The density of Subset Sum is defined as:

$$d = \frac{n}{\log (\max\{a_i\})} = \frac{n}{\log A}$$

This problem is known to be NP-complete [3](in its feasibility recognition form), and so is thought to be very hard in general. However, there are two algorithms, One by Brickell [1] and the other

by Lagarias and Odlyzko [2], which in polynomial time solve almost all subset sum problems of sufficiently low density. Both methods rely on basis reduction algorithms to find short non-zero vectors in special lattices.

The interesting case is $d < 1$. For $d < 0.9408$, it can be reduced to SVP in $\text{poly}(n)$, for $1 < d < 1.003$, the problem is NP-hard.



Lemma 2. *The solution vector $\hat{e} = (0, e_1, e_2 \dots, e_n) \in \mathbb{Z}^{n+1}$ is in $L(B)$.*

Proof. $B \cdot (-1, e_1, e_2 \dots, e_n)^t = (0, e_1, e_2 \dots, e_n)$

Consider a lattice $L(B)$ spanned by the columns of $B \in \mathbb{Z}^{(n+1)(n+1)}$

$$\begin{bmatrix} Ns & Na_1, & \dots & \dots & \dots & Na_n \\ 0 & 1 & 0 & 0 & \dots & 0 \\ 0 & 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \dots & 1 \end{bmatrix} \in \mathbb{R}^{m \times n}$$

$L(B)$ is of rank $(n + 1)$.

It is not enough to show that $\hat{e} \in L(B)$.

We would like \hat{e} to be the shortest vector in L . In the other words, we are interested in vectors $\hat{X} \in L$:

$$\begin{cases} \|\hat{X}\| \leq \|\hat{e}\|, \\ \hat{X} \in L, \\ \hat{X} \neq \{\vec{0}, \pm \hat{e}\} \end{cases} \quad (3)$$

Lets fix some notation,

$$\sum_{i=1}^n e_i \leq \frac{n}{2}$$

(otherwise replace $s = \sum_{i=1}^n a_i - s$ and the first basis vector by $(N(\sum_{i=1}^n a_i - s), 0, 0, \dots, 0)^t$).

Further, denote $T = \sum_{i=1}^n a_i$. Then wlog, $s \geq \frac{T}{n}$ (otherwise if $s < \frac{T}{n}$, any $a_i > \frac{T}{n}$ can not be in the subset), $s \leq (1 - \frac{1}{n}) \cdot T$ (otherwise, any $a_i \geq \frac{T}{n}$ must be in the subset). That is, $\frac{T}{n} \leq s \leq \frac{n-1}{n}T$

□

Lemma 3. For any $a_i \xleftarrow{\$} [1, A]$ and $N > \sqrt{n}$ and C_o as in Theorem 1, it holds

$$P := \Pr_{\substack{a_i \xleftarrow{\$} A}} [\exists \hat{X} \text{ that satisfies(3)}] \leq n(2n\sqrt{\frac{n}{2}} + 1) \frac{2^{C_o n}}{A}$$

In particular, if $A = 2^{Cn}$ with $C \geq C_o$, $\lim_{n \rightarrow \infty} P = 0$

It means that for

$$d \leq \frac{n}{C_o \cdot n} = 0.6463$$

the Subset Sum Problem can be reduced to SVP.

Proof. Since $N > \sqrt{n}$, for \hat{X} to satisfy (3), it must hold that $\hat{X}_1 = 0$, (otherwise, $\|\hat{X}\| \geq |\hat{X}_1| \geq N > \sqrt{n} > \|\hat{e}\|$, ($\|\hat{e}\| \leq \sqrt{\frac{n}{2}}$)).

Denote, $\vec{X} = X_1, X_1 \dots X_n, \vec{e} = e_1, e_1 \dots e_n$,

$$\begin{aligned} P &\leq \Pr_{a_i} (\exists (\vec{X}, y) \in \mathbb{Z}^n \times \mathbb{Z} : \|\vec{X}\| \leq \|\vec{e}\|, |y| < n \cdot \sqrt{\frac{n}{2}}, \vec{X} \notin \{0, \vec{e}, -\vec{e}\}, \sum X_i a_i = sy) \\ &\leq \Pr [\sum X_i a_i = sy, \text{for fixed } (\exists (\vec{X}, y), \|\vec{X}\| \leq \|\vec{e}\|, |y| < n \cdot \sqrt{\frac{n}{2}}, \vec{X} \notin \{0, \vec{e}, -\vec{e}\}) \cdot K \cdot K'] \end{aligned} \quad (4)$$

where,

$$K = |\{\vec{X} : \|\vec{X}\| \leq \|\vec{e}\|\}| \quad (5)$$

and

$$K' = |\{y : \|y\| < n \cdot \sqrt{\frac{n}{2}}\}| \quad (6)$$

In equation(4)

$$\sum_{i=1}^n a_i \cdot x_i = y \sum_{i=1}^n e_i \cdot x_i \iff \sum_{i=1}^n a_i \cdot z_i = 0, \text{ for, } z_i = x_i - y e_i$$

Since $\vec{X} \neq \vec{0}$, wlog, we assume $z_i = 0$,

$$\implies \Pr [\sum_{i=1}^n a_i \cdot z_i = 0] = \Pr [a_i = \sum_{i=1}^n \frac{a_i}{z_1}]$$

Take, $z' = \sum_{i=1}^n \frac{a_i}{z_1}$, we get,

$$\sum_{j=1}^A \Pr [a_i = j \mid z' = j] \cdot \Pr [z' = j]$$

Since a_i 's are choosen uniformly at random thus,

$$\sum_{j=1}^A \Pr [a_i = j] \cdot \Pr [z' = j] = \sum_{j=1}^A \frac{1}{A} \Pr [z' = j]$$

As $\Pr[z' = j] < 1$,
This implies

$$\sum_{j=1}^A \Pr[a_i = j] \cdot \Pr[z' = j] \leq \frac{1}{A}$$

Now consider (5) i.e. $K = |\{\vec{X} : \|\vec{X}\| \leq \|\vec{e}\|\}| \leq |\{\vec{X} : \|\vec{X}\| \leq \|\vec{e}\|\}| \leq s \cdot (\frac{n}{2}) \leq 2^{C_{\text{on}}}$

Consider (5) i.e $K = |\{\vec{X} : \|\vec{X}\| \leq \|\vec{e}\|\}| \leq 1 + 2 \cdot n \sqrt{\frac{n}{2}}$

Hence,

$$\implies P \leq n(1 + 2n\sqrt{\frac{n}{2}}) \cdot \frac{2^{C_{\text{on}}}}{A}$$

Define $y \in \mathbb{Z}$ such that,

$$y \cdot s = \sum_{i=1}^n \hat{X}_{i+1} a_i, \text{i.e}$$

y is the coefficient of \hat{X} in its first basis vector b_1 . Since $s > 0$,

$$\begin{aligned} \|y\| \cdot s &= \left| \sum_{i=1}^n \hat{X}_{i+1} \right| \leq \|\hat{X}\| \cdot \left| \sum_{i=1}^n a_i \right| \\ &\leq T \cdot \sqrt{\frac{n}{2}} \\ &\leq s \cdot n \sqrt{\frac{n}{2}} (\|\hat{X}\| \leq \|\hat{e}\|, T \leq s \cdot n) \\ \iff |y| &\leq n \cdot \sqrt{\frac{n}{2}}. \end{aligned}$$

□

Remarks:

- Setting $b_i = (Ns, \frac{1}{2}, \dots, \frac{1}{2})^t$ improves d to $d < 0.9408$ (it allows to consider only $\vec{X}'s$) such that $\|\vec{X}\| < \frac{\sqrt{n}}{2}$, in particular the optimization in Theorem(1) when $d = \frac{1}{4}$, gives $C_o = 1.0628$; hence $d < \frac{1}{C_o} = 0 \cdot 9408$.
- Applying Minkowski's bound to $L(B)$, we expect the length of the shortest vector in $L(B)$ to be

$$\begin{aligned} \lambda_1(L(B)) &\leq \sqrt{n+1} \cdot \det(L(B))^{\frac{1}{n+1}} \\ &\approx \sqrt{n} \cdot (Ns)^{\frac{1}{n}} \approx \sqrt{n} \cdot (\sqrt{n}A)^{\frac{1}{n}}, (\text{since, } s \approx A, N \approx \sqrt{n}) \end{aligned}$$

In $\text{poly}(n)$, LLL applied to $L(B)$, returns a vector $\vec{e}' \in L(B)$ such that $\|\vec{e}'\| \leq 2^{\frac{n \lg \lg n}{\lg n}} \lambda_1(L(B))$. As soon as $\|\vec{e}\|$ for \hat{e} the Subset Sum solution, is by a factor of $2^{\frac{-n \lg \lg n}{\lg n}}$ shorter than $\lambda_1(L(B))$, LLL will return $\|\hat{e}\|$:

$$\begin{aligned} \|\hat{e}\| &\leq 2^{\frac{n \lg \lg n}{\lg n}} \cdot \lambda_1(L(B)) \\ \sqrt{\frac{n}{2}} &\leq 2^{\frac{n \lg \lg n}{\lg n}} \cdot \sqrt{n}(\sqrt{n}A)^{\frac{1}{n}} \end{aligned}$$

on solving for A, we get:

$$\begin{aligned} A &\geq O(2^{\frac{n^2 \lg \lg n}{\lg n}}) \\ \implies \alpha &= \frac{n}{\log A} < \frac{\lg n}{n \lg \lg n} \end{aligned}$$

One can extend these arguments to BKZ algorithm

- This algorithm can be extended to non-zero weights e_i . We only require $\|e\|$ to be short.

References

- [1] E. F. Brickell, *Solving low density knapsacks*, Advances in Cryptology, Proceedings of Crypto 83, Plenum Press, New York, 1984, 25-37.
- [2] A. M. Frieze, *On the Lagarias-Odlyzko algorithm for the subset sum problem*, SIAM J. Comput. 15(2) (May 1986), 536-539.
- [3] B. A. LaMacchia, *Basis Reduction Algorithms and Subset Sum Problems* SM Thesis, Dept. of Elect. Eng. and Comp. Sci., Massachusetts Institute of Technology, Cambridge, MA (1991).
- [4] M.R. Garey, David S.Johnson, *Computer and Tractability: A Guide to theory of NP-Completeness*, W.H.Freeman and Company (1979)
- [5] A. K. Lenstra, H. W. Lenstra, and L. Lovasz, *A hierarchy of polynomial time lattice basis reduction algorithms*, Theoretical Computer Science 53 (1987), 201-224.
- [6] C. P. Schnorr, *Factoring polynomials with rational coefficients*, Math. Ann. 261 (1982), 515-534.