

Comparing the Jacobi Method and LLL lattice reduction algorithms for cryptographic applications

IN Bachelor Semester Project

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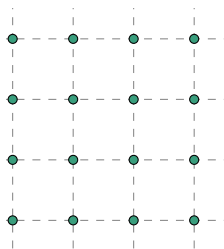
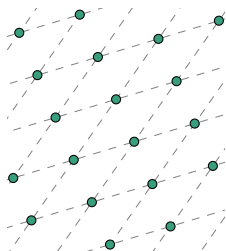
ÉCOLE POLYTECHNIQUE
FÉDÉRALE DE LAUSANNE

Fall 2014

Overview

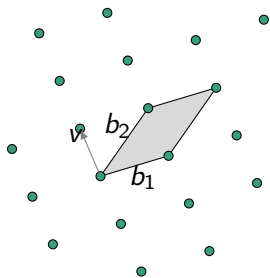
- 1 Reminders about lattices
- 2 Jacobi Method for lattice reduction
- 3 Experimental results
- 4 Acknowledgements and Bibliography

Lattice



- Discrete, additive subgroup of \mathbb{R}^m
- Intersecting points of an infinite regular n -dimensional grid in \mathbb{R}^m

Lattice



- Set $B = \{\mathbf{b}_1, \dots, \mathbf{b}_n\} \subset \mathbb{R}^m$, \mathbf{b}_i are linearly independent
- Full-rank lattices: $n = m$

Set of **integer** linear combinations

$$\text{Lattice } \mathcal{L} = \sum_i \mathbb{Z} \cdot \mathbf{b}_i$$

- B is called a basis of \mathcal{L} , it is not unique
- the volume of a full-rank lattice is given by $\text{vol}(\mathcal{L}) = |\det(B)|$

Random Lattice

We say that a lattice is a random lattice L of prime volume P if under HNF form its basis matrix B has the following properties:

- the diagonal has 1 for all it's entries except one position that is set to a prime number P . Hence, the $\det(B)$ is prime.
- All row entries of the matrix right to the position that is set to P are smaller than P in absolute value.

Without loss of generality, we hence restrict tests to random lattices of volume P whose basis in HNF form is as follows:

$$\begin{array}{ccccccc} P & \mathbf{a}_2 & \dots & \mathbf{a}_m \\ & 1 & & \\ & & \ddots & \\ & & & 1 \end{array}$$

where $a_i \in \mathbb{Z}/P\mathbb{Z}$.

Almost Orthogonal Lattice Bases

We define an *almost orthogonal lattice basis* M of dimension n and of bit length k as an $n \times n$ square matrix whose entries are k -bit integers picked at random.

Gram Schmidt orthogonalisation - GSO

- Basis $B = (\mathbf{b}_1, \dots, \mathbf{b}_n)$

- Compute GSO of B :

$$\mathbf{b}_1^* = \mathbf{b}_1$$

$$\mathbf{b}_2^* = \mathbf{b}_2 - \frac{\langle \mathbf{b}_2, \mathbf{b}_1^* \rangle}{\|\mathbf{b}_1^*\|^2} \mathbf{b}_1$$

$$\mathbf{b}_3^* = \mathbf{b}_3 - \frac{\langle \mathbf{b}_3, \mathbf{b}_1^* \rangle}{\|\mathbf{b}_1^*\|^2} \mathbf{b}_1^* - \frac{\langle \mathbf{b}_3, \mathbf{b}_2^* \rangle}{\|\mathbf{b}_2^*\|^2} \mathbf{b}_2^*$$

...

- In general

$$\mathbf{b}_i^* = \mathbf{b}_i - \sum_{j < i} \mu_{ij} \mathbf{b}_j^* \text{ where } \mu_{ij} := \frac{\langle \mathbf{b}_i, \mathbf{b}_j^* \rangle}{\|\mathbf{b}_j^*\|^2}$$

The LLL Algorithm

- First polynomial-time reduction algorithm to be introduced outputting a nearly orthogonal basis
- LLL and BKZ 2.0 are the two reduction algorithms that are used in practice for applications in cryptology and digital signal processing (MIMO)

δ -LLL Reduced

δ -LLL Reduced

Ordered basis $b_1, \dots, b_n \in \mathbb{R}^m$ of \mathcal{L} , parameter $\delta \in (1/4, 1]$, s.t.
 $\forall i, j :$

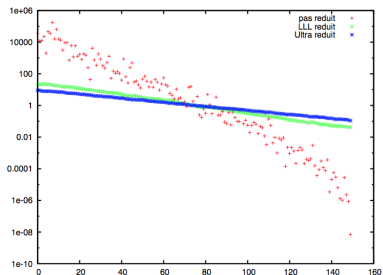
- $|\mu_{i,j}| \leq \frac{1}{2}$ for $1 \leq j < i \leq n$

δ -LLL Reduced

δ -LLL Reduced

Ordered basis $b_1, \dots, b_n \in \mathbb{R}^m$ of \mathcal{L} , parameter $\delta \in (1/4, 1]$, s.t. $\forall i, j$:

- $|\mu_{i,j}| \leq \frac{1}{2}$ for $1 \leq j < i \leq n$
- $\forall (b_i, b_{i+1})$, we have $(\delta - \mu_{i+1,i}^2) \|b_i^*\|^2 \leq \|b_{i+1}^*\|^2$



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Jacobi method for lattice reduction

- May 2012: Sanzheng Qiao publishes generic Jacobi paper[San12]
- June 2012: Complexity analysis [TQ12]
- July 2013: An Enhanced Jacobi Method for Lattice-Reduction-Aided MIMO Detection[TQ13]
- January 2014: A Hybrid Method for Lattice Basis Reduction[TQ14]
- Summer 2014: A Fast Jacobi-Type Method for Lattice Basis Reduction[Tia14]

Euclid's centered algorithm

Algorithm 1 Euclid's centered algorithm

Require: $(n, m) \in \mathbb{Z}^2$

Ensure: $\gcd(n, m)$

- 1: **if** $|n| < |m|$ **then**
 - 2: swap n and m
 - 3: **end if**
 - 4: **while** $m \neq 0$ **do**
 - 5: $r \leftarrow n - qm$ where $q = \lfloor \frac{n}{m} \rfloor$
 - 6: $n \leftarrow m$
 - 7: $m \leftarrow r$
 - 8: **end while**
 - 9: Output n
-

Lagrange algorithm

Algorithm 2 Lagrange algorithm

Require: Two basis $(\mathbf{b}_1, \mathbf{b}_2)$ vectors.

Ensure: a Lagrange reduced reduced basis $(\mathbf{b}_1, \mathbf{b}_2)$

1: **if** $\|\mathbf{b}_1\| < \|\mathbf{b}_2\|$ **then**

2: swap \mathbf{b}_1 and \mathbf{b}_2

3: **end if**

4: **repeat**

5: $q = \lfloor \frac{\langle \mathbf{b}_1, \mathbf{b}_2 \rangle}{\|\mathbf{b}_2\|^2} \rfloor$

$\mathbf{r} \leftarrow \mathbf{b}_1 - q\mathbf{b}_2$

$\mathbf{b}_1 \leftarrow \mathbf{b}_2$

$\mathbf{b}_2 \leftarrow \mathbf{r}$

6: **until** $\|\mathbf{b}_1\| \leq \|\mathbf{b}_2\|$

The generic Jacobi Method

Algorithm 3 Generic Jacobi Method

Require: a basis matrix $(\mathbf{b}_1, \dots, \mathbf{b}_n)$

Ensure: a generic-Jacobi reduced basis $(\mathbf{b}_1, \dots, \mathbf{b}_n)$

while not all pairs $(\mathbf{b}_i, \mathbf{b}_j)$ satisfy both generic-Jacobi reduction conditions **do**

for $i = 1$ **to** $n - 1$ **do**

for $j = i + 1$ **to** n **do**

$[\mathbf{b}_i, \mathbf{b}_j] = \text{Lagrange}(\mathbf{b}_i, \mathbf{b}_j)$

end for

end for

end while

ω -Lagrange reduced

There are two conditions for a basis to be ω -Lagrange-reduced.

$$\begin{cases} \lfloor \|\mathbf{a}_I^T \mathbf{a}_s / \|\mathbf{a}_s\| \rfloor \leq 1, \\ \omega \|\mathbf{a}_I\| \leq \|\mathbf{a}_I - \zeta \mathbf{a}_s\| \end{cases}$$

where $1/\sqrt{3} \leq \omega < 1$.

Iterative Lagrange

Algorithm 4 LagrangeIT

Require: The matrices G, Z , a pair of indices $(i, j) : i < j$ and a parameter ω

Ensure: Updated G, Z where one Lagrange iteration was performed on the i th and j th basis vectors.

$s \leftarrow i$

$l \leftarrow j$

if $g_{ii} > g_{jj}$ **then**

$s \leftarrow j; l \leftarrow i$

end if

$q \leftarrow \lfloor \frac{g_{ij}}{g_{ss}} \rfloor$

if Verify both ω -Lagrange-reduced conditions **then**

$z_{l-} = q * z_s$

$g_{l-} = q * g_s$

 Updating entries of the Gram matrix

end if

The Fast Jacobi method

Algorithm 5 Fast-Jacobi Reduction

Require: a basis matrix ($\mathbf{B} = \mathbf{b}_1, \dots, \mathbf{b}_n$) and ω

Ensure: a reduced basis ($\mathbf{b}_1, \dots, \mathbf{b}_n$) where each pair of vectors is ω -Lagrange reduced

$$G = B^T B, Z = I_n$$

while LagrangeT method reduced the basis vectors **do**

for $i = 1$ **to** $n - 1$ **do**

for $j = i + 1$ **to** n **do**

$$[G, Z] = \text{LagrangeT}(G, Z, i, j, \omega)$$

end for

end for

end while

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Our Implementation

- Generic and Fast-Jacobi implemented
- Written in C++ with newNTL
- ZZ and double implementations
- Benchmarked against FPLLL ($\delta = 0.99$)

Reduction quality indicators

Orthogonality Defect

The *orthogonality defect* of a basis $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n$ of a lattice L is defined by:

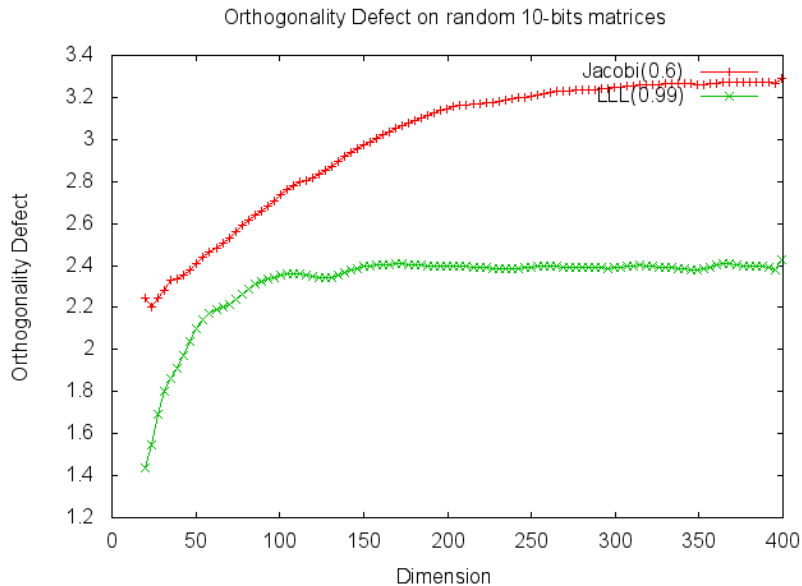
$$\text{OrthDefect}(L) := \sqrt[n]{\frac{\prod_{i=1}^n \|\mathbf{b}_i\|}{\det(L)}}$$

Hermite Factor

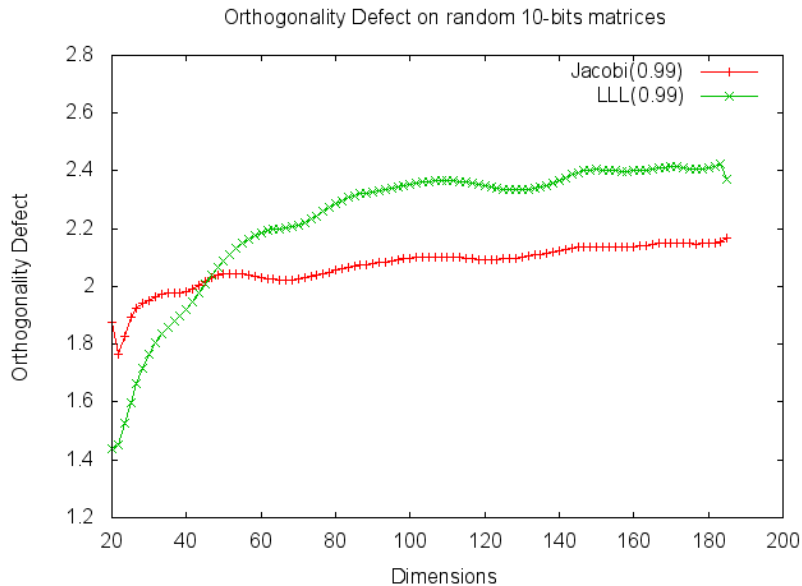
The *Hermite factor* of basis vectors $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n$ of a lattice L is defined by

$$\text{HF}(L) := \frac{\|\mathbf{b}_1\|}{\sqrt[n]{\det(L)}}$$

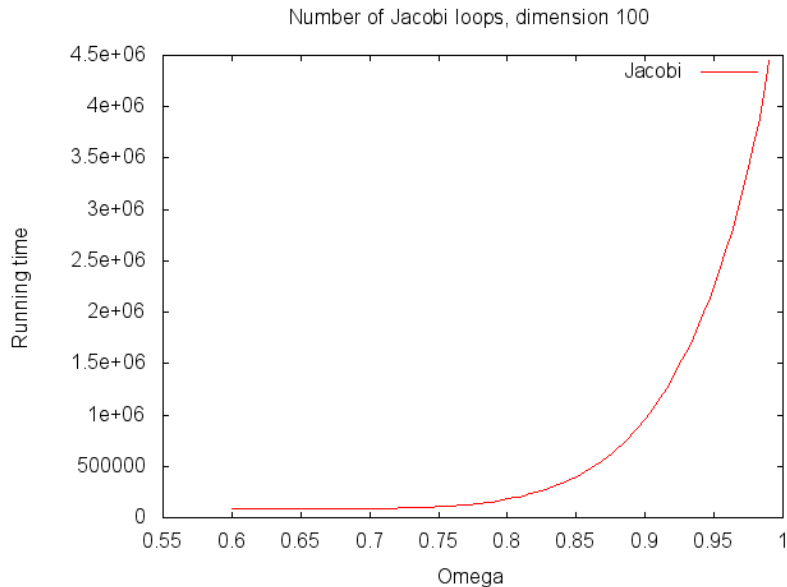
Almost orthogonal basis, $\omega = 0.6$



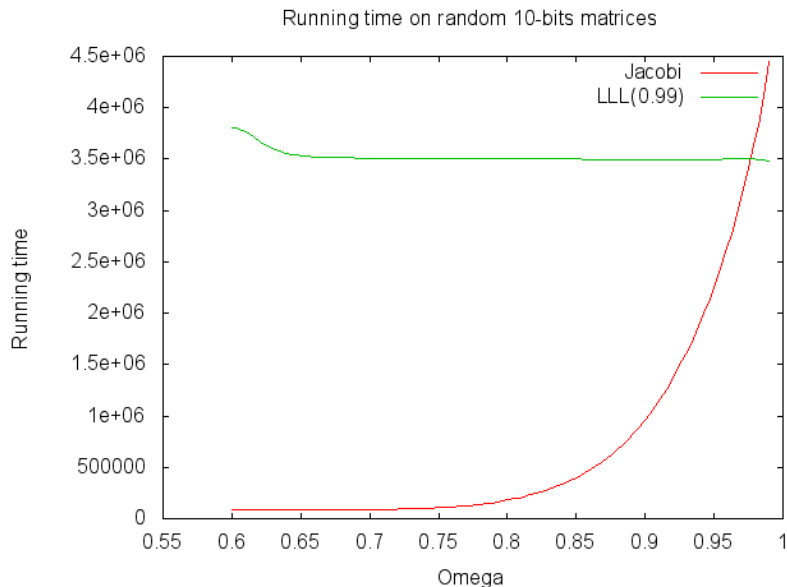
Almost orthogonal basis, $\omega = 0.99$



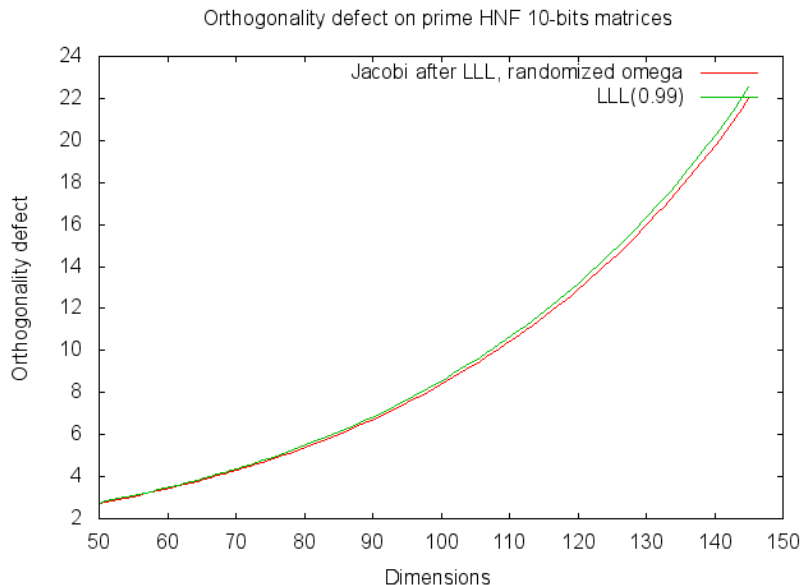
Average number of inner loops by ω



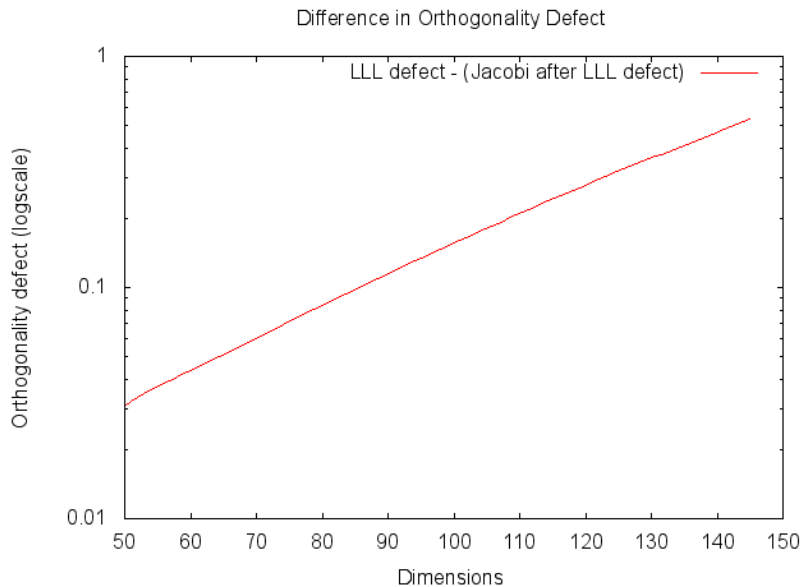
Note on running time depending on Ω



Jacobi after LLL



Jacobi after LLL



Jacobi after LLL

Example of LLL-reduced basis but not Jacobi-reduced

$$B = \begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \\ \mathbf{b}_3 \end{bmatrix} = \begin{bmatrix} 0 & 2 & 0 \\ 0 & 1 & 2 \\ 2 & 0 & 0 \end{bmatrix}$$

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Acknowledgements

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Thank you