

# FIT5124 Advanced Topics in Security

## Lecture 2: Lattice-Based Crypto. II

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# Plan for this lecture (and next)

- **How secure is lattice-based cryptography?**
  - Known cryptanalysis algorithms to break  $\gamma$ -SVP / SIS problem: LLL algorithm and variants.
  - Average-case hardness for SIS based on worst-case hardness of  $\gamma$ -SVP (only mention).
  - How to choose parameters for Ajtai's hash function for a given security level?
- **How to construct lattice-based encryption schemes? (start this week if sufficient time)**
  - Learning with Errors (LWE) Problem and Bounded Distance Decoding (BDD) problem
  - Symmetric-key encryption from LWE
  - Public-key encryption from LWE: Regev's cryptosystem (2005).

# Security of Lattice-Based Cryptography

- **Q1:** How should we choose the parameters  $q, m, n, d$  of Ajtai's hash function?
- **Q2:** How hard (secure) is SIS Problem?

We attempt to answer two subquestions for Q2 and return to Q1:

- **Q2a:** How hard is it to solve  $\gamma$ -SVP problem for an arbitrary lattice?
- **Q2b:** How hard is it to solve  $\gamma$ -SVP for random  $q$ -ary lattices  $L_q^\perp(A)$ , i.e. how do we know that SIS Problem is hard on 'average'? Is there a (non-negligible) subset of 'weak' matrices  $A$  for which problem is much easier than solving  $\gamma$ -SVP for arbitrary lattices?

# Security of Lattice-Based Cryptography

**Q2a:** How hard is it to solve  $\gamma$ -SVP problem for an arbitrary lattice?

**Ans:** Need to understand complexity of state of the art algorithms for these problems.

A difficult, not fully understood topic!

We briefly overview of two classical algorithms (foundation for current state of the art  $\gamma$ -SVP algorithm known as BKZ):

- **LLL** lattice reduction algorithm ( $\gamma = 2^{O(n)}$ , time =  $n^{O(1)}$ ).
- **Enumeration** algorithms, aka Fincke-Pohst enumeration ( $\gamma = 1$ , time =  $2^{O(n \log n)}$ ) – only mention.

Optimized  $\gamma$  vs. time tradeoff combination of those used in **BKZ** (aka Schnorr's block reduction) algorithm (state of the art alg. for  $\gamma = n^c$ , time  $\approx 2^{O(n/c)}$ ) – only mention.

## Algorithms for $\gamma$ -SVP: LLL

Recall: a given lattice  $L$  has an infinite number of bases  $B$ , but all have the **same** FP volume  $\det L$ :

- Most bases  $B$  are 'bad': **long** lattice vectors, far from orthogonal, FP of  $B$  is very 'skewed'
- Some bases  $B$  are 'good': **short** lattice vectors, close to orthogonal, FP of  $B$  is  $\approx$  an  $n$ -dim. cube of side length  $\approx \det L^{1/n}$ .

How to transform a 'bad' basis to a better one?

Use a **lattice basis reduction** algorithm: Given a basis  $B$  of lattice  $L$ , outputs a 'better' basis  $B'$  for  $L$

- Algorithm performs a sequence of unimodular operations on  $B$ 
  - Add integer multiple of one column to another column
  - Swap columns
- Each op. preserves basis property, 'improves' basis slightly

**First efficient (poly-time) reduction algorithm:** LLL (Lenstra Lenstra Lovasz, 1982)

## Algorithms for $\gamma$ -SVP: LLL

**Idea of LLL:** Make basis vectors 'approximately' orthogonal.

- GSO of a basis  $B$  tells us how 'orthogonal' the basis vectors are to each other:

$$\vec{b}_i^* = \vec{b}_i - \sum_{j=1}^{i-1} \mu_{i,j} \cdot \vec{b}_j^*, \text{ where } \mu_{i,j} = \frac{\langle \vec{b}_i, \vec{b}_j^* \rangle}{\langle \vec{b}_j^*, \vec{b}_j^* \rangle}.$$

What does 'approximately' orthogonal  $\vec{b}_i$ 's mean?

- **Small projection component:** 'small' relative projection length  $\mu_{i,j}$  of  $\vec{b}_i$  along previous  $\vec{b}_j^*$ 's ( $j < i$ ):
  - **LLL property 1:**  $|\mu_{i,j}| \leq 1/2$  for all  $i = 1, \dots, n$  and  $j < i$ .
- **Large orthogonal component:** 'large' remaining component  $\|\vec{b}_{i+1}^*\|$  of  $\vec{b}_{i+1}$  after removing components along  $\vec{b}_j^*$ 's ( $j < i + 1$ ):
  - **LLL property 2:**  $\|\vec{b}_{i+1}^* + \mu_{i+1,i} \vec{b}_i^*\|^2 \geq \delta \cdot \|\vec{b}_i^*\|^2$  for all  $i = 1, \dots, n - 1$  for some constant  $\delta$  ( $1/4 \leq \delta < 1$ ).

**Goal of LLL:** Perform elementary unimodular operations until both properties are satisfied.

# Algorithms for $\gamma$ -SVP: LLL

## Definition

A basis  $B$  for lattice  $L$  is  $\delta$ -LLL reduced if both LLL properties 1 and 2 are satisfied.

**LLL Algorithm.** Given  $n$ -dim. input basis  $B$ , do:

- **Start Step:** Compute GSO  $B^*$  for  $B$ .
- **Length Reduction Step:** (comment: after this step, LLL property 1 will be satisfied)
  - for  $i = 2$  to  $n$  do
    - for  $j = i - 1$  to  $1$  do
      - Update  $\vec{b}_i \leftarrow \vec{b}_i - c_{i,j}\vec{b}_j$ , where  $c_{i,j} = \left\lfloor \frac{\langle \vec{b}_i, \vec{b}_j^* \rangle}{\langle \vec{b}_j^*, \vec{b}_j^* \rangle} \right\rfloor$ .
- **Swap Step:** (comment: after this step, LLL property 2 will be satisfied by  $\vec{b}_i, \vec{b}_{i+1}$ )
  - If there is an  $i$  such that LLL property 2 is not satisfied (i.e.  $\|\vec{b}_{i+1}^* + \mu_{i+1,i}\vec{b}_i^*\| < \delta \cdot \|\vec{b}_i^*\|$ ), then:
    - Swap  $\vec{b}_i$  and  $\vec{b}_{i+1}$
    - Go back to Start Step.
  - Else, Return  $\delta$ -LLL reduced basis  $B$ .

# Algorithms for $\gamma$ -SVP: LLL

## Why does LLL work - Property 1?

- After length red. Step, LLL property 1 ( $|\mu_{i,j}| \leq 1/2$ ) is satisfied:
  - Throughout length red., GSO vectors  $\vec{b}_i^*$ 's do not change!
    - Adding a multiple of  $\vec{b}_j$  for  $j < i$  to  $\vec{b}_i$  only changes the projection of  $\vec{b}_i$  along  $\vec{b}_j$ , not the orthogonal component  $\vec{b}_i^*$ .
  - Recall (first lecture):  $\vec{b}_i^*$ 's coordinate matrix along the rotated coordinate system of normalized GSO vectors  $\vec{b}_i^* / \|\vec{b}_i^*\|$ :

$$\begin{bmatrix} \|\vec{b}_1^*\| & \|\vec{b}_1^*\| \cdot \mu_{2,1} & \cdots & \|\vec{b}_1^*\| \cdot \mu_{n,1} \\ 0 & \|\vec{b}_2^*\| & \cdots & \|\vec{b}_2^*\| \cdot \mu_{n,2} \\ 0 & 0 & \cdots & \|\vec{b}_3^*\| \cdot \mu_{n,3} \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & \|\vec{b}_n^*\| \end{bmatrix}$$

- $j$ th iteration of inner for loop: subtract integer multiple  $c_{i,j} = \lceil \mu_{i,j} \rceil$  of  $j$ th column above ( $\vec{b}_j$ ) from  $i$ th column  $\vec{b}_i - j$ th entry of  $i$ th column changes from  $\|\vec{b}_i^*\| \cdot \mu_{i,j}$  to  $\|\vec{b}_i^*\| \cdot (\mu_{i,j} - \lceil \mu_{i,j} \rceil)$ . So:  $\mu_{i,j} \rightarrow \mu'_{i,j} = \mu_{i,j} - \lceil \mu_{i,j} \rceil$  so  $|\mu'_{i,j}| \leq 1/2$ .



## Algorithms for $\gamma$ -SVP: LLL

**Why does LLL work - Property 2?** After swap step, LLL property 2 ( $\|\vec{b}_{i+1}^* + \mu_{i+1,i}\vec{b}_i^*\| \geq \delta \cdot \|\vec{b}_i^*\|$ ) is satisfied:

- Overall effect of swap step, swapping  $\vec{b}_i$  and  $\vec{b}_{i+1}$ :
  - $\vec{b}_j^*$  (and  $\mu_{i,j}$ ) for  $j < i$  stay the same:  $\vec{b}_j^{*new} = \vec{b}_j^*$  for  $j < i$ .
  - $\vec{b}_i^*$  and  $\vec{b}_{i+1}^* + \mu_{i+1,i}\vec{b}_i^*$  swap so property 2 at  $i$  is satisfied after swap:
    - $\vec{b}_i^{*new} = \vec{b}_{i+1}^* + \mu_{i+1,i}\vec{b}_i^*$ .
    - $\vec{b}_{i+1}^{*new} + \mu_{i+1,i}^{new}\vec{b}_i^{*new} = \vec{b}_i^*$

## Algorithms for $\gamma$ -SVP: LLL

### Why does LLL work - Run time?

Swap step may invalidate property 1 while length reduction step may invalidate property 2, but...

It can be shown that this cannot continue for very long - eventually both properties 1 and 2 are satisfied and algorithm terminates!

### Theorem

*The number of (length reduce, swap) iterations of LLL on input basis  $B$  before termination is at most*

$$n^2 \cdot \log(\max_i \|\vec{b}_i\|) / \log(1/\sqrt{\delta}).$$

*For any constant  $1/4 < \delta < 1$ , this is polynomial in bit length of the algorithm input. Moreover, the run-time for each iteration is also polynomial in the input bit length. Overall, run-time is polynomial in input length — LLL is efficient!*

## Algorithms for $\gamma$ -SVP: LLL

### How can we use LLL to solve $\gamma$ -SVP?

Intuitively, since LLL outputs an ‘approximately orthogonal’ basis for  $L$ , the basis vectors should be relatively short lattice vectors. A **practical** approach to solve  $\gamma$ -SVP for  $L(B)$ :

- Run LLL on  $B$  and get a  $\delta$ -LLL reduced LLL basis  $B'$  for  $L$ .
- Output the shortest vector among the  $n$  basis vectors in  $B'$ .

What approx. factor  $\gamma$  does this achieve? Not easy to predict theoretically!

But LLL properties of  $B'$  allow us to prove an upper bound on big  $\gamma$  can be.

# Algorithms for $\gamma$ -SVP: LLL

**Theoretical Upper bound on LLL Approx. factor  $\gamma$ .** If  $B$  is  $\delta$ -LLL reduced basis for  $L$ , LLL property 2 is:

$$\|\vec{b}_{i+1}^* + \mu_{i+1,i} \vec{b}_i^*\|^2 \geq \delta \cdot \|\vec{b}_i^*\|^2$$

By Pythagoras, LHS above is just  $\|\vec{b}_{i+1}^*\|^2 + \mu_{i+1,i}^2 \|\vec{b}_i^*\|^2$ , so we can rearrange to get:

$$\|\vec{b}_{i+1}^*\|^2 \geq (\delta - \mu_{i+1,i}^2) \cdot \|\vec{b}_i^*\|^2.$$

By LLL property 1,  $\mu_{i+1,i}^2 \leq 1/4$  so we get the successive ratio bound:

$$\frac{\|\vec{b}_{i+1}^*\|^2}{\|\vec{b}_i^*\|^2} \geq (\delta - 1/4), \text{ for all } i \geq 2.$$

Since the ratio of norms of all pairs of successive  $\vec{b}_i^*$ 's is at least  $\delta - 1/4$ , it immediately implies

$$\frac{\|\vec{b}_n^*\|^2}{\|\vec{b}_1^*\|^2} \geq (\delta - 1/4)^{n-1},$$

or  $\|\vec{b}_1\| = \|\vec{b}_1^*\| \leq (1/(\delta - 1/4))^{(n-1)/2} \cdot \|\vec{b}_n^*\|$ . But it can be shown that  $\|\vec{b}_n^*\| \leq \lambda_1(L)$ , so

$\|\vec{b}_1\| \leq (1/(\delta - 1/4))^{(n-1)/2} \cdot \lambda_1(L)$ . Conclusion (take  $\delta = 3/4$ ):

## Theorem

The LLL algorithm solves (in polynomial time)  $\gamma$ -SVP for  $n$ -dim. lattices, with  $\gamma \leq 2^{(n-1)/2}$ . Can also be shown

that **Hermite Factor**  $\gamma_{HF} \stackrel{\text{def}}{=} \frac{\|\vec{b}_1\|}{\det(L)^{1/n}} \leq 2^{(n-1)/4}$ .

## Algorithms for $\gamma$ -SVP: LLL

**LLL Approx. factor  $\gamma$  in practice.** For ‘random’ lattices, LLL experimentally performs **much better** than the theoretical upper bound  $\gamma \leq 2^{(n-1)/2}$ .

Experiments (see, e.g., [NS06]) show that for random lattices, LLL reduced bases tend to have, on average,

$$\frac{\|\vec{b}_{i+1}^*\|^2}{\|\vec{b}_i^*\|^2} \approx 1.04, \text{ for all } i \geq 2.$$

Consequently, for random lattices, LLL can (experimentally) solve  $\gamma$ -SVP for  $\gamma \approx 1.04^{n-1}$ . (and Hermite Factor  $\gamma_{HF} \approx 1.02^{n-1}$ ).

# Algorithms for $\gamma$ -SVP: Enumeration

**How to compute the shortest vector ( $\gamma = 1$ ).** If we really want the shortest vector, can always use a brute force search approach – **enumeration algorithms**.

- Enumerate all lattice vectors in a volume that is guaranteed to contain the shortest vector.
- Will not go into details here.

**Drawback:** enumeration run-time is (at least) exponential in dimension  $n$ !

Current state of the art enumeration algorithms (aka Fincke-Phost / Kannan) take time  $2^{O(n \log n)}$ .

**Remark:** Other algorithms (sieve algorithms – Ajtai et al 2001, Voronoi algorithms - Micciancio et al 2010, Gaussian sampling algorithms – Regev et al 2015) exist that trade off exponential memory  $2^{O(n)}$  for  $2^{O(n)}$  time.

- Large memory tends to make these algorithms less practical (but still being improved)...

## Algorithms for $\gamma$ -SVP: BKZ

**Trading off larger time for smaller  $\gamma$ .** In late 1980s, Schnorr introduced a hierarchy of generalizations of LLL, called **Block Korkhine Zolotarev (BKZ) algorithm**. Trades off larger run-time for smaller approx. factor  $\gamma$  – currently state of the art for attacking lattice-based crypto:

- Combines ideas of LLL and enumeration algorithms.
- **Idea:** introduce a 'block size' parameter  $k \in \{2, 3, \dots, n\}$  into LLL: generalize the  $2 \times 2$  GSO submatrix blocks in LLL property 2
- Gradual 'interpolation' between the extremes of LLL ( $k = 2$ ,  $\gamma = 2^{O(n)}$ ,  $T = n^c$ ) and enumeration ( $k = n$ ,  $\gamma = 1$ ,  $T = 2^{O(n \log n)}$ ).
- For general block size  $k$ , variants of BKZ [HPS'11] provably achieve  $\gamma(k) \leq k^{(n-1)/(k-1)}$  with run-time  $n^c \cdot k^{O(k)}$ .

# Complexity of $\gamma$ -SVP: Asymptotic Summary

**Summary: State of the art (BKZ) asymptotic  $\gamma$ -time tradeoff.** For future reference, we have the following (approximate) asymptotic relations:

- For security against attacks running time  $T = 2^\lambda$  – security parameter  $\lambda$ , need

$$2^\lambda = 2^{O(k \log k)}, \text{ so } \lambda = O(k \log k).$$

- At this run-time, achieve  $\gamma_{HF} = \delta(k)^n$  with  $\delta(k) = k^{1/(k-1)} \approx k^{1/k}$ , so

$$\log \gamma_{HF} = (n/k) \log k, \text{ so } \log \gamma_{HF} = \Omega\left(\frac{n \log^2 \lambda}{\lambda}\right).$$

Overall, get asymptotic **lattice 'rule of thumb'** for  $\gamma$ -SVP (using BKZ):

$$n = \Omega\left(\frac{\lambda}{\log^2 \lambda} \cdot \log \gamma_{HF}\right) \approx \lambda \cdot \log \gamma_{HF}.$$

**Remark:** Need lattice dim.  $n$  proportional to **product** of bit-security level  $\lambda$  and log. approx. factor.

- $\log \gamma_{HF}$  factor is a reason behind relatively long keys in lattice-based cryptosystems...



# Complexity of $\gamma$ -SVP: Numerical Summary

**Numerical estimates of optimized BKZ time versus  $\gamma$ .** Chen and Nguyen [CN11] gave numerical estimates for Hermite Factor and time for 'random' lattices versus block size for optimized (state of the art) BKZ variants:

**Table 2.** Approximate required blocksize for high-dimensional BKZ, as predicted by the simulation

Target Hermite Factor	$1.01^n$	$1.009^n$	$1.008^n$	$1.007^n$	$1.006^n$	$1.005^n$
Approximate Blocksize	85	106	133	168	216	286

**Table 3.** Upper bound on the cost of the enumeration subroutine, using extreme pruning with aborted-BKZ preprocessing. Cost is given as  $\log_2$ (number of nodes).

Blocksize	100	110	120	130	140	150	160	170	180	190	200	250
BKZ-75-20%	41.4	47.1	53.1	59.8	66.8	75.2	84.7	94.7	105.8	117.6	129.4	204.1
Simulation of BKZ-90/100/110/120	40.8	45.3	50.3	56.3	63.3	69.4	79.9	89.1	99.1	103.3	111.1	175.2

Can be used to estimate concrete numerical parameters for cryptosystems!

# Parameters for Ajtai's hash Function: Hardness of SIS

**Recall:** Ajtai's hash function collision-resistance security (provably) depends on hardness of SIS problem: finding vectors of length  $\leq \beta = 2d\sqrt{m}$  in SIS lattice  $L_q(A)$  (dimension  $m$ ,  $\det L_q(A) = q^n$  – see tute).

How to choose parameters  $q, n, m, d$  for given security parameter  $\lambda$  based on hardness of SIS?

To get security level  $\approx 2^\lambda$  (enum. cost) against BKZ attacks, possible approach (see [MR08] survey):

- Assume attacker runs BKZ with block length  $k$  such that enumeration cost is  $\approx 2^\lambda$  (e.g. use [CN11] tables).
- Find corresponding BKZ Hermite factor  $\gamma_{HF} = \delta^m$  (e.g. use [CN11] tables).
- Attacker can compute a non-zero vector  $\vec{v}$  in SIS lattice  $L_q(A)$  of norm  $\leq \ell = \min(q, \delta^m \cdot \det(L_q(A))^{1/m})$ . Breaks  $\text{SIS}_\beta$  if  $\min(q, \delta^m \cdot \det(L_q(A))^{1/m}) \leq \beta$ .

# Parameters for Ajtai's hash Function: Hardness of SIS

- Attack optimization ([MR08]): Attacker uses only a subset of  $m' \leq m$  of columns of  $A$ , where  $m'$  is chosen to an optimal value  $m^*$  minimizing  $\ell(m') = \min(q, \delta^{m'} \cdot \det(L_q(A))^{1/m'})$ . Turns out that  $m^* = \sqrt{\frac{n \log q}{\log \delta}}$  and  $\ell(m^*) = \min(q, 2^{2\sqrt{n \log q \log \delta}})$ .

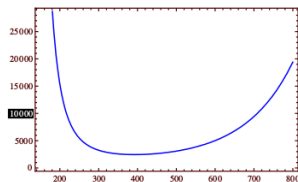


Figure 2: Estimated length of vector found with  $\delta = 1.01$ ,  $q = 4416857$ , and  $n = 100$  as a function of  $m$ .

- For  $\text{SIS}_\beta$  hardness, choose hash parameters such that  $\ell(m^*) > \beta^* = 2d\sqrt{m^*}$ , so:

$$q \geq \beta^* = 2d\sqrt{m^*} \text{ and } n \geq \frac{\log^2(\beta^*)}{4 \log q \log(\delta)}.$$

# Ajtai's hardness proof for SIS

Why do we think that SIS is 'hard on average' (no weak instances occur with non-negligible probability)?

Ajtai's average-case to worst-case connection Theorem (1996, improved by Gentry et al [GPV08]).

## Theorem

If there is an algorithm  $A$  that solves  $SIS_{q(n),m(n),\beta(n)}$  in poly-time, for some **non-negligible fraction** of input matrices  $G \in \mathbb{Z}_q^{mn \times n}$ ,

Then there is an algorithm  $B$  that solves  $\gamma(n)$ -SIVP in polynomial time for **all** input lattices  $L$  of dimension  $n$  with:

$$\gamma = O(\beta\sqrt{n}), q(n) = \omega(\gamma\sqrt{\log n}).$$

- $\gamma$ -SIVP is a variant of  $\gamma$ -SVP that asks for a  $\gamma$  approximation to the  $n$  linearly independent shortest lattice vectors.
- We won't study this proof, but it gives us a theoretical foundation for security of SIS.

## References referred to in the Slides

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