

# ON THE CONCRETE HARDNESS OF LEARNING WITH ERRORS

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Learning With Errors

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## Lattice

A lattice is a discrete additive subgroup of  $\mathbb{R}^m$ .

## Basis

Let  $\mathbf{B} = \{\mathbf{b}_1, \dots, \mathbf{b}_m\}$  be a set of  $m$  linearly independent vectors in  $\mathbb{R}^m$ . Then  $L(\mathbf{B}) = \{\sum_{i=1}^m x_i \mathbf{b}_i \mid x_i \in \mathbb{Z}\}$  is the lattice generated by this basis.

## Dual Lattice

Given a lattice  $L(\mathbf{B}) \subset \mathbb{R}^m$ , define its dual as  $\{\mathbf{x} \in \mathbb{R}^m \mid \mathbf{x}\mathbf{B} \in \mathbb{Z}^m\}$ .

We'll only use scaled-by- $q$  dual lattices, i.e.  $\{\mathbf{x} \in \mathbb{Z}_q^m \mid \mathbf{x}\mathbf{B} \equiv \mathbf{0}\}$

The Learning with Errors (LWE) problem was defined by Oded Regev [Reg05].

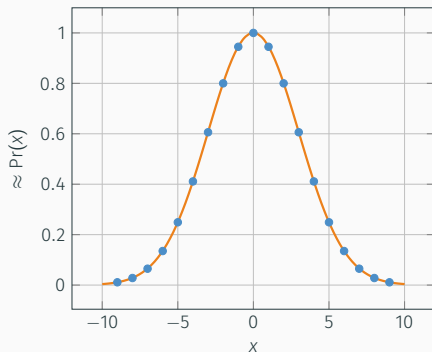
- Suppose a public matrix  $\mathbf{A}$  and a secret vector  $\mathbf{s}$ .
- If we were also given  $\mathbf{b} = \mathbf{As}$  we could compute  $\mathbf{s}$  by linear algebra.
- Now imagine this is noisy:  $\mathbf{c} = \mathbf{As} + \mathbf{e}$  with  $\mathbf{e}$  small.
- From  $\mathbf{A}$  and  $\mathbf{c}$  can we find  $\mathbf{s}$ ? Was  $\mathbf{c}$  even computed this way?

Given  $(\mathbf{A}, \mathbf{c})$  with  $\mathbf{c} \in \mathbb{Z}_q^m$ ,  $\mathbf{A} \in \mathbb{Z}_q^{m \times n}$ ,  $\mathbf{s} \in \mathbb{Z}_q^n$  and  $\mathbf{e} \in \mathbb{Z}_q^{m \times \ell}$  do we have

$$\begin{pmatrix} \mathbf{c} \end{pmatrix} = \begin{pmatrix} \leftarrow n \rightarrow \\ \mathbf{A} \end{pmatrix} \times \begin{pmatrix} \mathbf{s} \end{pmatrix} + \begin{pmatrix} \mathbf{e} \end{pmatrix}$$

or  $\mathbf{c} \leftarrow_{\$} \mathcal{U}(\mathbb{Z}_q^m)$ .

# PARAMETERS



- Parameters are:
  - dimension  $n$ ,
  - modulus  $q$  (e.g.  $q \approx n^2$ ),
  - noise size  $\alpha$  (e.g.  $\alpha q \approx \sqrt{n}$ ),
  - number of samples  $m$ .
- Elements of  $\mathbf{A}, \mathbf{s}, \mathbf{e}, \mathbf{c}$  are in  $\mathbb{Z}_q$ .
- $\mathbf{e}$  is sampled from a discrete Gaussian with width

$$\sigma = \frac{\alpha q}{\sqrt{2\pi}}.$$

## Search LWE

From samples  $(A, c)$  recover  $s$ .

## Decision LWE

Determine if samples  $(A, c)$  are LWE or uniformly random.

These problems are polynomial-time equivalent.



Given samples

$$(\mathbf{a}, c) = (\mathbf{a}, \langle \mathbf{a}, \mathbf{s} \rangle + e) \in \mathbb{Z}_q^n \times \mathbb{Z}_q$$

with  $\mathbf{a} \leftarrow \mathcal{U}(\mathbb{Z}_q^n)$ ,  $e \leftarrow D_{\alpha q, 0}$  and  $\mathbf{s} \in \mathbb{Z}_q^n$ , we can construct samples

$$(\mathbf{a}, c) = (\mathbf{a}, \langle \mathbf{a}, \mathbf{e} \rangle + e) \in \mathbb{Z}_q^n \times \mathbb{Z}_q$$

with  $\mathbf{a} \leftarrow \mathcal{U}(\mathbb{Z}_q^n)$ ,  $e \leftarrow D_{\alpha q, 0}$  and  $\mathbf{e}$  such that all components  $e_i \leftarrow D_{\alpha q, 0}$  in polynomial time.

Let  $(\mathbf{a}, c) = (\mathbf{a}, \langle \mathbf{a}, \mathbf{s} \rangle + e) \in \mathbb{Z}_q^n \times \mathbb{Z}_q$  be an LWE sample and

$$p \approx \sqrt{\frac{2\pi n}{12}} \cdot \frac{\sigma_s}{\alpha},$$

where  $\sigma_s$  is the standard deviation of components of the secret  $\mathbf{s}$ . If  $p < q$  then

$$\left( \left\lfloor \frac{p}{q} \cdot \mathbf{a} \right\rfloor, \left\lfloor \frac{p}{q} \cdot c \right\rfloor \right) \text{ in } \mathbb{Z}_p^n \times \mathbb{Z}_p$$

follows a distribution close to an LWE distribution with parameters  $n, \sqrt{2}\alpha, p$ .

## Learning With Errors

- is assumed to be a hard problem like discrete logarithms, factoring, etc.
- reduces to hard problems on lattices, such as GapSVP.
- is assumed to have resistance against quantum computers, unlike discrete logarithms and factoring.
- is remarkably versatile for constructing cryptographic schemes.

## Identity-based encryption [GPV08]

Ciphertexts are of the form

$$(\mathbf{p}, c) = (\mathbf{A}\mathbf{s} + \mathbf{e}, \mathbf{u} \cdot \mathbf{s} + e + b \cdot \lfloor q/2 \rfloor)$$

where  $H(id) = \mathbf{u} = \mathbf{x}^T \mathbf{A}$  is the public key for the private key  $\mathbf{x}$ .

Decryption is done by

$$c - \langle \mathbf{x}, \mathbf{p} \rangle = -\langle \mathbf{x}, \mathbf{e} \rangle + e + b \cdot \lfloor q/2 \rfloor.$$

## Fully homomorphic encryption [BV11, AAFP11]

Think of LWE encryptions

$$(\mathbf{a}_i, c_i) = (\mathbf{a}_i, \langle \mathbf{a}_i, \mathbf{s} \rangle + e_i + b_i \cdot \lfloor q/2 \rfloor)$$

as noisy linear polynomials

$$-c_i + \sum a_{ij}x_j.$$

Add, multiply and decrypt as usual.

# HOW HARD IS LWE?

Given  $n$  (and  $\alpha, q$ ) how many operations does it take to solve?

- **Problem 1.** Algorithms/attacks are not well understood in terms of concrete running times.
  - Runtimes are given asymptotically.
  - Algorithms are better in practice than the theoretical bounds.
  - Many heuristic assumptions.
- **Problem 2.** Many variables
  - dimension, modulus, secret size
  - distribution of the secret
  - number of samples available to an attacker
  - variants of the problem (e.g. small secrets, BinaryError-LWE)

# WHAT DO PEOPLE DO CURRENTLY?

Often, in the literature, the following assumptions were made when estimating concrete security of an LWE-based scheme:

- the best attack is a lattice-based **distinguishing attack**;
- **BKZ** runs in roughly the time given in [LP11];
- the use of a **small secret** makes no difference for attacks.

All three of these assumptions turn out not to be correct.

## So, WHAT DID WE DO?

- Overview the **strategies** for attacking LWE.
- Analyse and present **running times**.
- Produce **concrete estimates** for attack timings for parameters sets.

The estimation code is available as a Sage module.



Learning With Errors

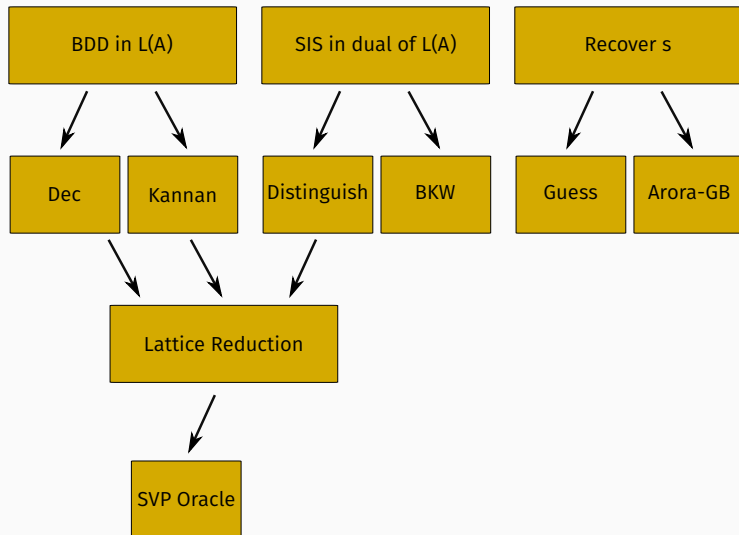
Strategies and Algorithms

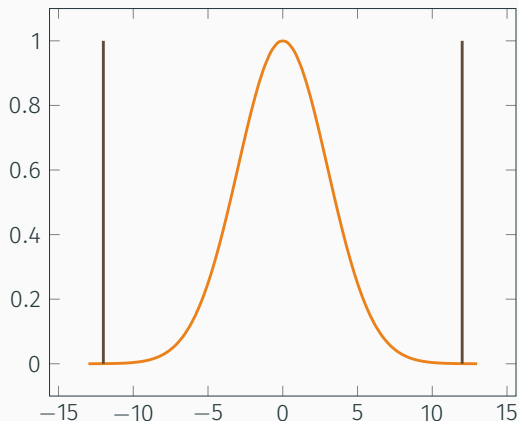
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# STRATEGIES FOR SOLVING LWE





- The error is from a small subset of  $\mathbb{Z}_q$ , say,  $(-\tau \cdot \sigma, \dots, \tau \cdot \sigma)$
- Each candidate gives rise to one linear equation.
- Construct equations of degree  $2\tau \cdot \sigma + 1$  encoding that one of these linear equations must hold.
- Solve the system using Gröbner bases.

Arora-Ge (Linearisation) with  $\sigma = \sqrt{n}$

$$\mathcal{O} \left( 2^{\omega n \log(8n \log n) - \omega n \log n} \right)$$

Gröbner Bases with  $\sigma = \sqrt{n}$

$$\mathcal{O} \left( 2^{2.82 \omega n} \right)$$

under some regularity assumption.



M.A., Carlos Cid, Jean-Charles Faugère, and Ludovic Perret.

**Algebraic algorithms for LWE.**

Cryptology ePrint Archive, Report 2014/1018, 2014.

<http://eprint.iacr.org/2014/1018>.

## Short Integer Solutions (SIS)

Given  $q \in \mathbb{Z}$ , a matrix  $\mathbf{B}$ , and  $t < q$ ; find  $\mathbf{y}$  with  $0 < \|\mathbf{y}\| \leq t$  and

$$\mathbf{y}\mathbf{B} \equiv \mathbf{0} \pmod{q}.$$

- Recall the dual lattice:  $L^* = \{\mathbf{x} \in \mathbb{R}^m \mid \mathbf{x}\mathbf{B} \in \mathbb{Z}^m\}$ .
- Then the scaled dual lattice,  $qL^*$  has the property that  $\mathbf{x}\mathbf{B} \equiv \mathbf{0} \pmod{q}$  for all  $\mathbf{x} \in qL^*$ .
- Therefore, a short vector of  $qL^*$  is equivalent to solving SIS on  $\mathbf{B}$ .

- Find a short  $\mathbf{y}$  solving SIS on  $\mathbf{A}$ .
- Given LWE samples  $\mathbf{A}, \mathbf{c}$  where either  $\mathbf{c} = \mathbf{A}\mathbf{s} + \mathbf{e}$  or  $\mathbf{c}$  uniformly random.
- Compute  $\langle \mathbf{y}, \mathbf{c} \rangle$ .
  - If  $\mathbf{c} = \mathbf{A}\mathbf{s} + \mathbf{e}$ , then  $\langle \mathbf{y}, \mathbf{c} \rangle = \langle \mathbf{y}\mathbf{A}, \mathbf{s} \rangle + \langle \mathbf{y}, \mathbf{e} \rangle \equiv \langle \mathbf{y}, \mathbf{e} \rangle \pmod{q}$ .
  - If  $\mathbf{c}$  is uniformly random, so is  $\langle \mathbf{y}, \mathbf{c} \rangle$ .
- If  $\mathbf{y}$  is sufficiently short, since  $\mathbf{e}$  is also small, then  $\langle \mathbf{y}, \mathbf{e} \rangle$  will also be short, and can be distinguished from uniform values.

# DISTINGUISH (LATTICE REDUCTION)

A **reduced lattice** basis is made of short vectors, in particular the first vector.

1. Construct a basis of the dual from the instance.
2. Feed to a lattice reduction algorithm to obtain short vectors  $\mathbf{v}_i$ .
3. Check if  $\mathbf{v}_i \mathbf{A}$  are small.



Daniele Micciancio and Oded Regev.

**Lattice-based cryptography.**

In Bernstein et al. [BBD09], pages 147–191.

We revisit Gaussian elimination:

$$\left( \begin{array}{c|c|c|c|c|c} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} & \cdots & \mathbf{a}_{1n} & C_1 \\ \mathbf{a}_{21} & \mathbf{a}_{22} & \mathbf{a}_{23} & \cdots & \mathbf{a}_{2n} & C_2 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \mathbf{a}_{m1} & \mathbf{a}_{m2} & \mathbf{a}_{m3} & \cdots & \mathbf{a}_{mn} & C_m \end{array} \right)$$

$$\stackrel{?}{=} \left( \begin{array}{c|c|c|c|c|c} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} & \cdots & \mathbf{a}_{1n} & \langle \mathbf{a}_1, \mathbf{s} \rangle + \mathbf{e}_1 \\ \mathbf{a}_{21} & \mathbf{a}_{22} & \mathbf{a}_{23} & \cdots & \mathbf{a}_{2n} & \langle \mathbf{a}_2, \mathbf{s} \rangle + \mathbf{e}_2 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \mathbf{a}_{m1} & \mathbf{a}_{m2} & \mathbf{a}_{m3} & \cdots & \mathbf{a}_{mn} & \langle \mathbf{a}_m, \mathbf{s} \rangle + \mathbf{e}_m \end{array} \right)$$



$$\Rightarrow \left( \begin{array}{c|c|c|c|c|c} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} & \cdots & \mathbf{a}_{1n} & \langle \mathbf{a}_1, \mathbf{s} \rangle + \mathbf{e}_1 \\ 0 & \tilde{\mathbf{a}}_{22} & \tilde{\mathbf{a}}_{23} & \cdots & \tilde{\mathbf{a}}_{2n} & \langle \tilde{\mathbf{a}}_2, \mathbf{s} \rangle + \mathbf{e}_2 - \frac{\mathbf{a}_{21}}{\mathbf{a}_{11}} \mathbf{e}_1 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \\ 0 & \tilde{\mathbf{a}}_{m2} & \tilde{\mathbf{a}}_{m3} & \cdots & \tilde{\mathbf{a}}_{mn} & \langle \tilde{\mathbf{a}}_m, \mathbf{s} \rangle + \mathbf{e}_m - \frac{\mathbf{a}_{m1}}{\mathbf{a}_{11}} \mathbf{e}_1 \end{array} \right)$$

- $\frac{\mathbf{a}_{j1}}{\mathbf{a}_{11}}$  is essentially random in  $\mathbb{Z}_q$  wiping all “smallness”.
- If  $\frac{\mathbf{a}_{j1}}{\mathbf{a}_{11}}$  is 1 noise-size doubles because of the addition.

We considering  $a \approx \log n$  'blocks' of  $b$  elements each.

$$\left( \begin{array}{cc|ccc|c} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} & \cdots & \mathbf{a}_{1n} & C_0 \\ \mathbf{a}_{21} & \mathbf{a}_{22} & \mathbf{a}_{23} & \cdots & \mathbf{a}_{2n} & C_1 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \mathbf{a}_{m1} & \mathbf{a}_{m2} & \mathbf{a}_{m3} & \cdots & \mathbf{a}_{mn} & C_m \end{array} \right)$$

For each block we build a table of all  $q^b$  possible values indexed by  $\mathbb{Z}_q^b$ .

$$T^0 = \left[ \begin{array}{cc|ccc|c} -\lfloor \frac{q}{2} \rfloor & -\lfloor \frac{q}{2} \rfloor & \mathbf{t}_{13} & \cdots & \mathbf{t}_{1n} & c_{t,0} \\ -\lfloor \frac{q}{2} \rfloor & -\lfloor \frac{q}{2} \rfloor & \mathbf{t}_{23} & \cdots & \mathbf{t}_{2n} & c_{t,1} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \lfloor \frac{q}{2} \rfloor & \lfloor \frac{q}{2} \rfloor & \mathbf{t}_{q^2 3} & \cdots & \mathbf{t}_{q^2 n} & c_{t,q^2} \end{array} \right] + 1$$

For each  $\mathbf{z} \in \mathbb{Z}_q^b$  find row in  $\mathbf{A}$  which contains  $\mathbf{z}$  as a subvector at the target indices.

$$\begin{aligned}
 & \left( \begin{array}{cc|ccc|c} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} & \cdots & \mathbf{a}_{1n} & C_0 \\ \mathbf{a}_{21} & \mathbf{a}_{22} & \mathbf{a}_{23} & \cdots & \mathbf{a}_{2n} & C_1 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \\ \mathbf{a}_{m1} & \mathbf{a}_{m2} & \mathbf{a}_{m3} & \cdots & \mathbf{a}_{mn} & C_m \end{array} \right) \\
 + & \left[ \begin{array}{cc|ccc|c} -\lfloor \frac{q}{2} \rfloor & -\lfloor \frac{q}{2} \rfloor & \mathbf{t}_{13} & \cdots & \mathbf{t}_{1n} & C_{t,0} \\ -\lfloor \frac{q}{2} \rfloor & -\lfloor \frac{q}{2} \rfloor + 1 & \mathbf{t}_{23} & \cdots & \mathbf{t}_{2n} & C_{t,1} \\ \vdots & \vdots & \ddots & \vdots & \vdots & \\ \lfloor \frac{q}{2} \rfloor & \lfloor \frac{q}{2} \rfloor & \mathbf{t}_{q^2 3} & \cdots & \mathbf{t}_{q^2 n} & C_{t,q^2} \end{array} \right] \\
 \Rightarrow & \left( \begin{array}{cc|ccc|c} \mathbf{a}_{11} & \mathbf{a}_{12} & \mathbf{a}_{13} & \cdots & \mathbf{a}_{1n} & C_0 \\ 0 & 0 & \tilde{\mathbf{a}}_{23} & \cdots & \tilde{\mathbf{a}}_{2n} & \tilde{C}_1 \\ \vdots & \vdots & \ddots & \vdots & \vdots & \\ \mathbf{a}_{m1} & \mathbf{a}_{m2} & \mathbf{a}_{m3} & \cdots & \mathbf{a}_{mn} & C_m \end{array} \right)
 \end{aligned}$$

Time and memory complexity of  $\mathcal{O}(2^{n/(2-1/c)})$  for  $q \approx n^c$ .



M.A., Carlos Cid, Jean-Charles Faugère, Robert Fitzpatrick, and Ludovic Perret.

**On the complexity of the BKW algorithm on LWE.**

*Designs, Codes and Cryptography*, 74:325–354, 2015.



Alexandre Duc, Florian Tramèr, and Serge Vaudenay.

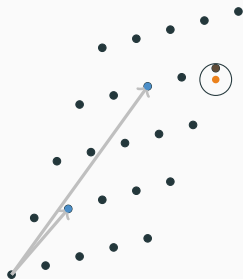
**Better Algorithms for LWE and LWR.**

In Elisabeth Oswald and Marc Fischlin, editors, *Advances in Cryptology – EUROCRYPT 2015*, volume 9056 of *LNCS*, pages 173–202. Springer, April 2015.

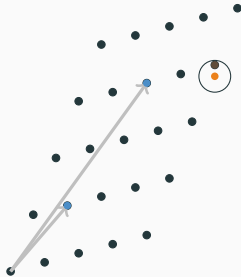
# WHAT IS BDD?

## Bounded Distance Decoding (BDD)

Given a basis of  $L$ , a target vector  $\mathbf{t}$ , and a distance parameter  $\beta > 0$  such that  $d(\mathbf{t}, L) < \beta \lambda_1(L)$ , find a  $\mathbf{y} \in L$  such that  $d(\mathbf{y}, \mathbf{t}) = d(L, \mathbf{t})$ .



here,  $\beta = 0.5$



- $\mathbf{b} = \mathbf{A}\mathbf{s}$  is a point in the lattice,
- $\mathbf{t} = \mathbf{A}\mathbf{s} + \mathbf{e}$  is a perturbed point.
- Solve the BDD instance to recover  $\mathbf{b}$ .
- Recover  $\mathbf{s}$  by linear algebra.

- Most basic is Babai's nearest planes.
- Lindner and Peikert: use multiple planes.
- Liu and Nguyen: use pruning strategy.
- No closed formula for runtime, can only be calculated numerically.



Richard Lindner and Chris Peikert.

**Better key sizes (and attacks) for LWE-based encryption.**

In Aggelos Kiayias, editor, *CT-RSA 2011*, volume 6558 of *LNCS*, pages 319–339. Springer, February 2011.



Mingjie Liu and Phong Q. Nguyen.

**Solving BDD by enumeration: An update.**

In Ed Dawson, editor, *CT-RSA 2013*, volume 7779 of *LNCS*, pages 293–309. Springer, February / March 2013.



$\gamma$ -uSVP

Given a lattice  $L$  s.t.  $\lambda_2(L) > \gamma \lambda_1(L)$ , find a shortest nonzero vector in  $L$

1. Reduce BDD to uSVP via Kannan's embedding:

$$B = \begin{pmatrix} A^T & 0 \\ \mathbf{c} & t \end{pmatrix}$$

where in practice  $t = 1$ .

2. Use lattice reduction to solve uSVP instance.



M.A., Robert Fitzpatrick, and Florian Göpfert.

**On the efficacy of solving LWE by reduction to unique-SVP.**

In Hyang-Sook Lee and Dong-Guk Han, editors, *ICISC 13*, volume 8565 of *LNCS*, pages 293–310. Springer, November 2014.

- So far the secret vector was chosen as  $\mathbf{s}_{(i)} \leftarrow \mathbb{Z}_q$ .
- Some applications choose  $\mathbf{s}_{(i)} \leftarrow \{-1, 0, 1\}$  or  $\mathbf{s}_{(i)} \leftarrow \{0, 1\}$ .
- This is for efficiency or to make certain operations possible (FHE).
- We call such an LWE instance a **small secret** instance.

In most algorithms, a small secret makes the instance easier.

- **exhaustive search**: check  $2^n$  or  $3^n$  elements rather than  $(\alpha q)^n$ .
- **modulus switching**: we can improve many algorithms by switching to a smaller modulus.

Given LWE samples  $\mathbf{A}, \mathbf{c}$ :

- Recall that BKW finds short vectors  $\mathbf{y}$  such that  $\mathbf{yA} = 0$ .
- Instead, find short vectors  $\mathbf{y}$  such that  $\mathbf{yA} = \mathbf{w}$  is **small**.
- Then  $\langle \mathbf{y}, \mathbf{c} \rangle = \mathbf{y} \cdot \mathbf{A} \cdot \mathbf{s} + \langle \mathbf{y}, \mathbf{e} \rangle = \langle \mathbf{w}, \mathbf{s} \rangle + \langle \mathbf{y}, \mathbf{e} \rangle$ .
- If  $\mathbf{s}$  is small, so is  $\langle \mathbf{w}, \mathbf{s} \rangle$ .



M.A., J.-C. Faugère, R. Fitzpatrick, and L. Perret.

**Lazy modulus switching for the BKW algorithm on LWE.**

In Hugo Krawczyk, editor, *PKC 2014*, volume 8383 of *LNCS*, pages 429–445. Springer, March 2014.

- Embed LWE instance into different uSVP lattice.
- Exploits the difference between size of the secret and the size of the error by scaling.
- Has little effect for FHE case, because the noise is already very small, but dramatic effect for Regev's PKC parameters.



Shi Bai and Steven D. Galbraith.

**Lattice decoding attacks on binary LWE.**

In Willy Susilo and Yi Mu, editors, *ACISP 14*, volume 8544 of *LNCS*, pages 322–337. Springer, July 2014.

## Theory [BLP<sup>+</sup>13]

A small secret LWE instance as hard as standard LWE requires dimension  $n \log q = \mathcal{O}(n \log n)$ , for typical parameter choice  $q = n^c$  for some small  $c$ .

## Bai and Galbraith's Attack

Dimension  $n \log(\log n)$  may be sufficient.



Zvika Brakerski, Adeline Langlois, Chris Peikert, Oded Regev, and Damien Stehlé.

### **Classical hardness of learning with errors.**

In Dan Boneh, Tim Roughgarden, and Joan Feigenbaum, editors, *45th ACM STOC*, pages 575–584. ACM Press, June 2013.

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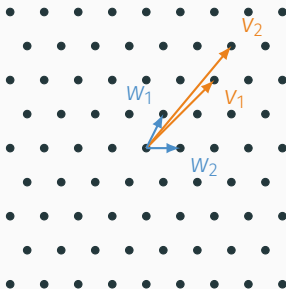
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# LATTICE REDUCTION

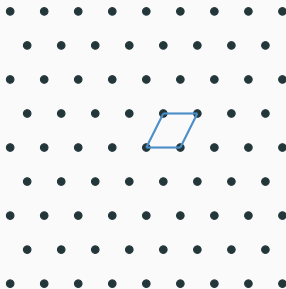
- Above we made reference to lattice basis reduction algorithms.
- Examples of lattice reduction algorithms: LLL, BKZ, BKZ 2.0.
- These take as input a lattice basis and outputs a reduced basis:





# LATTICE REDUCTION

- The success of a lattice reduction algorithm is characterised by the ‘root-Hermite factor’  $\delta_0$ .
- This is defined by  $\|\mathbf{b}_1\| = \delta_0^m \text{vol}(L)^{1/m}$ .



- Best known lattice reduction algorithm.
- BKZ is parametrised by blocksize  $k$ : bigger  $k$  mean better quality but more time.
- It can be seen as generalised LLL, which has  $k = 2$ .
- Literature disagrees on:
  - limiting value of  $\delta_0$  which BKZ can achieve (as a function of  $k$ );
  - runtime of BKZ (as a function of  $\delta_0$ ,  $k$ , or both).

We estimate BKZ as follows:

- Blocksize: Solve  $\delta_0 \approx \left( \frac{k}{2\pi e} (\pi k)^{\frac{1}{k}} \right)^{\frac{1}{2(k-1)}}$  for  $k$ .
- CPU cycles for one SVP call in dimension  $k$ :  $t_k = 2^{0.27k \log k - 1.02k + 16}$
- Required number of rounds:  $\rho \approx \frac{n^2}{k^2} \log n$ .
- Overall cost:  $\rho \cdot t_k$ .



Yuanmi Chen.

*Réduction de réseau et sécurité concrète du chiffrement complètement homomorphe.*

PhD thesis, Paris 7, 2013.



Guillaume Hanrot, Xavier Pujol, and Damien Stehlé.

**Analyzing blockwise lattice algorithms using dynamical systems.**

In Phillip Rogaway, editor, *CRYPTO 2011*, volume 6841 of *LNCS*, pages 447–464. Springer, August 2011.

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- What is better for SIS: BKW or lattice reduction?
- What is the best use for lattice reduction: Decoding, Kannan und solving SIS?
- Is there a best algorithm overall?
- What is the best small secret strategy?

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Short answer: it depends ....

- For most algorithms, there is no sufficiently precise closed formula for runtime.
- We provide a Sage module for estimating how long various algorithms take to run.
- It takes as input parameters  $n, \alpha, q$ .
- It outputs estimates of bit operations, memory requirements and number of calls to the LWE oracle.

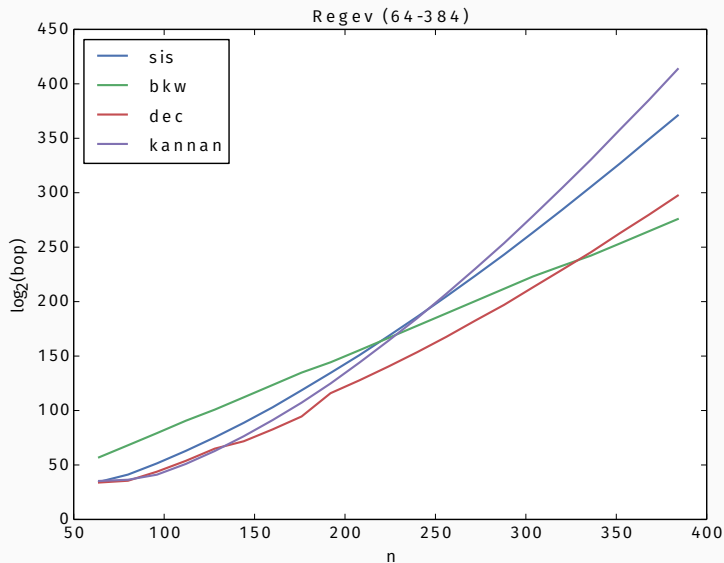
We consider some ‘typical’ parameter sets.

**Regev** These are Regev’s example choices for parameters from [Reg09]. We use [AFC<sup>+</sup>13] to pick  $q \approx n^2$  and  $\alpha = 1/(\sqrt{2\pi n} \log_2^2 n)$ .

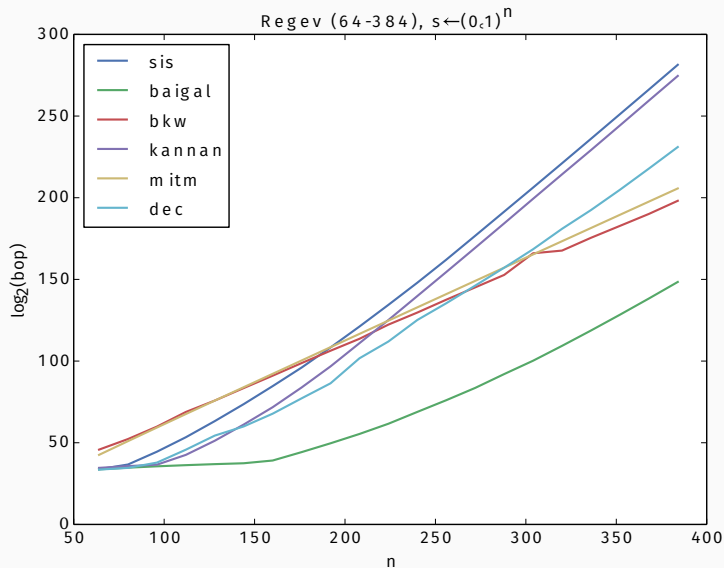
**FHE** Given  $n$  and the multiplicative depth  $L$  we set  $q = 2^{16.5 \cdot L + 5.4} \cdot 8^{2L-3} \cdot n^L$  and  $\alpha = \sqrt{2\pi} \cdot 3.2/q$  inspired by parameters suggested in [GHS12].

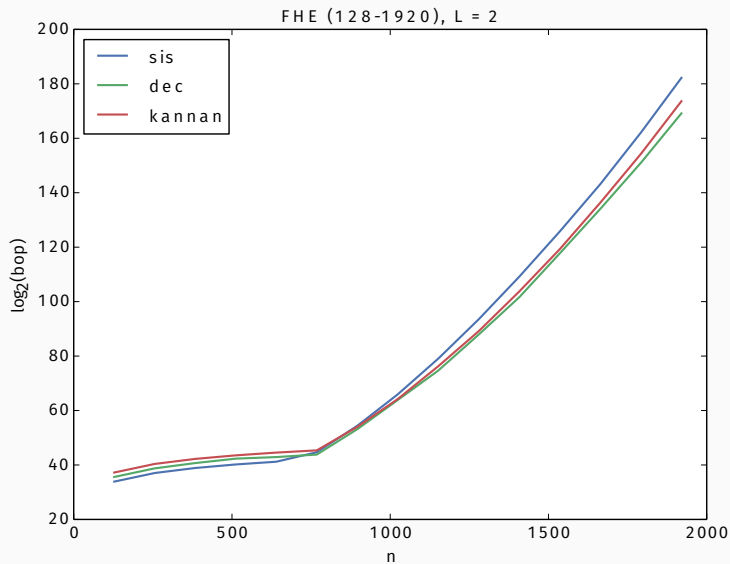
For small secrets we always assume  $\mathbf{s}_{(i)} \in \{0, 1\}$ .





# REGEV: SMALL SECRET





Learning With Errors

Strategies and Algorithms

Lattice Reduction

Estimator

Conclusion

## Results

- No one algorithm always outperforms all others.
- Parameters are paramount.
- Small secrets matter.

## Open Problems

- Is there an algorithm in  $2^{\mathcal{O}(n)}$  time but less than  $2^{\mathcal{O}(n)}$  memory?
- How long does lattice reduction actually take?
- Can we bridge the gap between theory and practice for small secrets?

## Questions?

survey <http://eprint.iacr.org/2015/046>  
estimator <https://bitbucket.org/malb/lwe-estimator>



M.A., Robert Fitzpatrick, Daniel Cabracas, Florian Göpfert, and Michael Schneider.

**A generator for LWE and Ring-LWE instances, 2013.**

available at <http://www.iacr.org/news/files/2013-04-29lwe-generator.pdf>.



M.A., Pooya Farshim, Jean-Charles Faugère, and Ludovic Perret.  
**Polly cracker, revisited.**

In Dong Hoon Lee and Xiaoyun Wang, editors, *ASIACRYPT 2011*, volume 7073 of *LNCS*, pages 179–196. Springer, December 2011.



Daniel J. Bernstein, Johannes Buchmann, and Erik Dahmen, editors.

***Post-Quantum Cryptography.***

Springer, 2009.



Zvika Brakerski, Adeline Langlois, Chris Peikert, Oded Regev, and Damien Stehlé.



Richard Lindner and Chris Peikert.

**Better key sizes (and attacks) for LWE-based encryption.**

In Aggelos Kiayias, editor, *CT-RSA 2011*, volume 6558 of *LNCS*, pages 319–339. Springer, February 2011.



Oded Regev.

**On lattices, learning with errors, random linear codes, and cryptography.**

In Harold N. Gabow and Ronald Fagin, editors, *37th ACM STOC*, pages 84–93. ACM Press, May 2005.



Oded Regev.

**On lattices, learning with errors, random linear codes, and cryptography.**

*J. ACM*, 56(6), 2009.