Automatize parameter tuning in Ring-Learning-With-Errors-based leveled homomorphic cryptosystem implementations

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Abstract. Lattice-based cryptography offers quantum-resistant cryptosystems but there is not yet official recommendations to choose parameters with standard security levels. Some of these cryptosystems permit secure computations and aim at a wider audience than cryptographic community. We focus on one of them, a leveled homomorphic cryptosystem (LHE): Brakersi/Fan-Vercauteren's (BFV) one. The family of LHE cryptosystems needs to be well-instantiated not only to protect input and output ciphertexts and to perform efficiently computations, but also, for them, parametrization constrains the quantity of homomorphic computations that can be performed with guarantee of correctness. It demands to choose parameters accordingly. In addition, each implementation brings external constraints to optimize performance. All of this makes it tedious for the non-expert user to choose parameters. To solve this, we have developed CinguParam to help user to instantiate implementations of BFV in different libraries: Cingulata, FV-NFLlib and Microsoft SEAL (release 3.3). CinguParam permits to generate an up-to-date database of parameter sets in function of computation budget, security parameters and implementation choices. This tool includes a notion of budget to ensure correct homomorphic computations and the one of BKZ reduction cost model to grasp the gap from concrete security, nowadays. It makes use of the LWE-Estimator to obtain up-to-date security estimations. CinguParam permits to select automatically a suitable parameter set with Cingulata and it can be used to generate code snippets to set parameters with FV-NFLlib and Microsoft SEAL (release 3.3).

1 Introduction

Homomorphic encryption offers services to manipulate encrypted data without having access to their associated clear form. As data are encrypted from the beginning to the end of the process, it permits for instance, to share private data with powerful servers in order to delegate costly computations and to get valuable results. But choosing cryptographic parameters is a challenging task in homomorphic cryptography [1] based on the LWE (Learning with Errors) problem [2]. If parameters are inapropriately chosen, not only the attackers could access the data or the key, but also the result of homomorphic computations could be wrong.

In this document, we focus on parameters for the BFV (Brakerski/Fan-Vercauteren) [3,4], Leveled Homomorphic Encryption (LHE) scheme, based on the Ring-LWE problem [5], which has multiple open-source implementations [6,7,8,9] in C++.

We have developed CinguParam [10], a tool to ease the choice of parameters in such implementations. We propose different parameter settings rather than fixing one parameter set, in order to propose to the user different choices so that he could get the best set according to his precise needs. It aims to generate correct and secure parameters sets to perform homomorphic computations. It contains a database of parameter sets (see 4.1) which can be easily extended, if needed. In terms of security, LWE attacks are currently the best ones [11, p.8] against Ring-LWE-based cryptosystem. CinguParam makes use of the LWE-Estimator [12] which estimates an up-to-date security of LWE instances. We point that today, nobody has definitive concrete security parameter for lattice-based cryptosystems. Our tool helps to follow research progress. Indeed, there is still some way to go for that:

- Security reduction between lattice problems and LWE problem is not tight for lattice with dimension n < 187149 (less than 18 bits) [13].
- The best practical attacks (primal and dual ones) against Ring-LWE make use of BKZ-type algorithms employing enumeration or sieving algorithms. Their cost is unknown, there exist many BKZ cost models [14].
- For most of the models, security estimation is a lower bound under some hypotheses [14]. These are pessimistic views from the defender side.
- There are several recent progresses to better understand and estimate security against LWE attacks, for instance [13,15,16].

^{**} This work was been done while the author was a member of CEA LIST, Paris-Saclay, France.

Our tool CinguParam permits to compare the impact of parameter choices for homomorphic computations and to measure the uncertainty on concrete security level, at this stage, in the cryptographic community.

In Section 2, we establish notation for BFV scheme. In Section 3, we provide algorithms used to generate suitable parameters for BFV, and introduce a vocabulary with the willingness to clarify process with a high-level perspective. CinguParam could be extended to determine parameters for other schemes implementations. In Section 4, we present and analyze our experimental results in two parts: we consider CinguParam in standalone mode first, then we combine it with existing tools: Cingulata, FV-NFLlib, Microsoft SEAL (release 3.3).

2 Scheme parameters

Let us begin to categorize the different parameters of BFV scheme. We employ them in algorithms to generate secure parameters (see Section 3) that can be employed to process correctly a predetermined quantity of homomorphic computations.

BFV is a scheme based on a variant of LWE where data (plaintexts and ciphertexts) are in a polynomial ring. Let the defining polynomial f be a 2-power cyclotomic polynomial.

- ciphertext coefficient count n
- ciphertext modulus q
- Gaussian (noise/error) width σ
- private key distribution χ over $\mathbb{Z}[X]/(f)$

We retain two approaches to determine Gaussian noise width σ . Note that the first one depends on n and is hardness-reduction compliant [17,18]:

$$\sigma = \begin{cases} 2\sqrt{n} \text{ asymptotic approach} \\ \frac{8}{\sqrt{2\pi}} \text{ performance approach } [19,20] \end{cases}$$

Default choice of σ depends on cryptosystem implementation choices, it is indicated in Table 1.

2.1 Correctness parameters

Informally, each circuit¹ to be evaluated by the homomorphic encryption scheme is associated to a circuit cost depending on the cost and on the order of evaluated operations. Circuit budget defines the family of circuits that can be operated over fresh ciphertexts (the ones which are obtained directly from encryption, *i.e.*, without any previous homomorphic operations), the ones whose cost is less than the budget. There exist several approaches to define and estimate operation costs (see Section 3). They depend on:

- the homomorphic scheme
- the circuit budget
- the multiplicative depth L
- the plaintext modulus t

Relinearization Method 1

- integer base ω for gadget decomposition [21], also called decomposition bit count

Relinearization Method 2

- relinearization integer k
- $-B_k$ bound on relinearization error distribution
- ciphertext relinearization modulus p

¹ Cingulata take into account Boolean circuits whereas FV-NFLlib and Microsoft SEAL (release 3.3) perform computations associated with arithmetic circuits.

2.2 Security parameters

There is not yet satisfying solution to choose tightly parameters with a given security level for LWE-based homomorphic cryptosystems. This drives us to decline several notions around security levels, in order to highlight current situation.

- desired security level λ_0
- BKZ lattice reduction cost model
- estimated security level $\lambda(cost_model)$
- approximated security level $\lambda_1(cost_model)$ (see Table 3
- tolerance between desired and estimated security levels. Indeed, there can be important gaps between them. Implementations constraints restrain flexibility on parameter choices (e.g., n is a power of two). Other constraints are indicated in Table 1.
- compatibility with Regev quantum security reduction proof [17] which impacts on Gaussian noise width σ
- number of LWE samples $(n,q,relin_param)$
- success probability ϵ in distinguishing attack on LWE decision problem [22]

3 How to generate up-to-date, correct and secure parameters?

To derive parameters setting for LHE cryptosystems [23] is not a trivial question:

- Security based on variants of LWE is a burning issue. It asks more than ever to keep informed about cryptanalyses results.
- Correct decryption is an additional constraint with LHE cryptosystems compared to traditional cryptosystems. The quantity and the nature of operations that will be operated over the fresh ciphertexts must be known before parametrization. This is a strong constraint.
- Performance is an open problem, as we want to ensure concrete secure computations (*e.g.*, homomorphic computations) in everyday life (see Table 2).

CinguParam makes use of the LWE-Estimator [12] to estimate security against LWE attacks. The LWE-Estimator is also used to generate (n, q) parameters in a document provided by the consortium HomomorphicEncryption.org [19]. This document offer recommendations which consist in tight LWE attack cost estimations based on BKZ Sieve and Q-Core Sieve cost models (Table 4). These recommendations are independent with the scheme used and with implementation choices (Table 2). In particular, it does not take into account noise growth which depends on the chosen scheme and thus do not consider correctness as required for LHE cryptosystems. This explains why for same n, their parameters offer better security levels with smaller q, what may seem counterintuitive [19, pp. 26-29].

Correct decryption depends on (noise/errors) added deliberately during encryption and amplified during each homomorphic operations to the ciphertexts in order to preserve confidentiality. Too much noise leads to decryption failures. Noise is a central notion in LWE-based cryptosystems. Ciphertext noise can be defined pessimistically (from the defender point of view) as the infinite norm of a polynomial which takes part in decryption step. It has different definitions[24, p.11]. Its computation requires the secret key [25]. For practical usage, with a private circuit, we use conservative upper bounds on the circuit noise. The choice of noise definition does not tighten significantly the bound to determine the quantity of secure computations we can operate correctly. In the current version of CinguParam, we employ the *inherent noise* as in the original paper [4].

We define the following terminology:

- max_encryption_noise for an upper bound on the noise of a fresh ciphertext,
- max_decryption_noise for an upper bound on the noise allowing a correct decryption,
- max_circuit_noise for an an upper bound on the noise after evaluating a homomorphic circuit.
- max_relin_noise for an an upper bound on the noise after relinearizing a ciphertext.

The ratio

 $max_decryption_noise(scheme)$

$max_encryption_noise(scheme)$

indicates how much noise, used to secure communications, can increase during homomorphic computations without degrading decryption correctness. It depends on the used scheme.

Algorithm 1: GenerateParam(*politic*, *L*, *cost_model*, λ_0 , *t*, *gen_method*)

 $first_pass = True$ $(n,q) = (n_0,q_0)$ $\sigma, \chi, \epsilon, relin_method \leftarrow politic$ $\mathbf{if} \ gen_method = MinCorrectModulus \ \mathbf{then}$ while first_pass or estimated security is inadequate do $first_pass = False$ $q = MinCorrectModulus(n, \sigma, \chi, \epsilon, L, t, relin_param)$ $noise_rate = \frac{\sigma}{q}$ $\lambda = \text{LWE-Estimator}(n, q, \chi, noise_rate, nr_samples,$ $cost_model$) $n = n \times 2$ n = n/2**else if** gen_method = MinSecureDegree **then** while first_pass or max_circuit_noise $\geq max_decryption_noise$ do $first_pass = False$ $(n, \frac{\sigma}{q}, \lambda) = MinSecureDegree(q, \lambda, \chi, cost_model, relin_param,$ $security_reduction)$ max_encryption_noise(scheme) max_circuit_noise(scheme, circuit) max_decryption_noise(scheme) $q = q \times scale$ q = q/scale

Algorithm 2: MinSecureDegree

$$\begin{split} n &= n_0 \\ first_pass = True \\ \textbf{while} \ first_pass \ or \ estimated \ security \ is \ inadequate \ \textbf{do} \\ & \left[\begin{array}{c} first_pass = False \\ noise_rate = \frac{\sigma}{q} \\ \lambda = \text{LWE-Estimator}(n,q,\chi,noise_rate,nr_samples, \\ cost_model) \\ n = n \times 2 \\ n = n/2 \\ \textbf{return} \ (n, \frac{\sigma}{q}, \lambda) \end{split} \end{split} \end{split}$$

Algorithm 3: MinCorrectModulus

 $\begin{array}{l} q = q_{0} \\ first_pass = True \\ \textbf{while } first_pass or max_circuit_noise \geq max_decryption_noise \ \textbf{do} \\ \\ first_pass = False \\ max_encryption_noise(scheme) \\ max_decryption_noise(scheme) \\ q = q \times scale \\ \\ \\ // \ scale \ \textbf{depends on implementation constraints and modify generation time, database size} \\ q = q/scale \\ \textbf{return } q \end{array}$

The BFV case

Maximal noise in a fresh ciphertext and maximal noise of a (homomorphic) ciphertext have been computed in [26] for some ring-LWE-based LHE schemes. In particular for the BFV scheme [3,4], noise growth has been well described in [27]. We only recall its bound (Equation 1) to compute maximal circuit noise after evaluating a circuit of given multiplicative depth L. In practice, two difficulties with BFV scheme occur:

- Multiplicative depth does not take into account noise growth caused by additions. How to take it into account to avoid incorrect decryption? This could be done by evaluating the variance of the noise at the end of the circuit rather than the infinite norm. This would be more optimistic, from the defender point of view.
- In most of the cases, multiplicative depth is sufficient to ensure correct decryption. But still, how to determine multiplicative depth? This question is solved with Cingulata compiler toolchain and runtime environmement [8]. It is employed with CinguParam, see Section 4.2 and serves to automatize parameter choice. This implementation focuses on binary ciphertexts manipulated with Boolean circuits and permits to interact easily with CinguParam. It is not done in implementations using arithmetic circuits, in our knowledge.

In the BFV scheme, the simplest way to determine the noise budget is derived from the multiplicative depth L. Let us recall, L is the maximal number of multiplications between any input and any output of the circuit we operate to evaluate secure computations. In the case of homomorphic encryption, inputs and outputs are ciphertexts.

In a more general setting, noise growth does not necessarily depend on the multiplicative depth. For instance, with the TFHE scheme [28], the focus is directed on the number of gates.

$$max_encryption_noise(BFV) = B_{error}(\epsilon, \sigma)(1 + 2\delta B_key)$$
(1)

where δ is the polynomial multiplication expansion factor, *i.e.*, $\max\{\frac{\|a \times b\|_{\infty}}{\|a\|_{\infty}\|b\|_{\infty}} : a, b \in \mathbb{Z}[X]/(X^n + 1)\}, B_{error}(\epsilon, \sigma) = \sqrt{2} \operatorname{erfc}^{-1}(\epsilon)\sigma$ and B_{key} are respectively an upper bound on the error distribution (resp. the private key distribution). The function erfc is the complementary error function and the quantity ϵ denotes the distinguishing attack advantage on LWE-decision problem, *i.e.*, the statictical distance within noise distribution and Gaussian distribution. For short, noise distribution is bounded by $B_{error}(\epsilon, \sigma)$ with probability $1 - \epsilon$.

 $max_circuit_noise(\text{BFV},L) = X^L \times max_encryption_noise + L \times X^{L-1}(Y + max_relin_noise)$

with

$$X = \delta t (4 + \delta B_{key}),$$
$$Y = \delta^2 B_{key} (B_{key} + t^2),$$

and

$$max_relin_noise(BFV) = \begin{cases} \delta \times \left[\frac{\log(q)}{\log(\omega)}\right] \times \omega \times B_{error}(\epsilon, \sigma) \text{ if } relin_param = 1\\ \frac{q \times B_k \times \delta}{p} + \frac{\delta \times B_{key} + 1}{2} & \text{ if } relin_param = 2 \end{cases}$$

A ciphertext is well decrypted if the circuit noise is smaller than the decryption noise. Circuit budget is used to generate secure and correct parameters with CinguParam. To ensure correct decryption, it has to be greater than decryption cost.

We can extend the vocabulary, similarly to [25], to express intuitively noise constraints:

- *min_decryption_cost*, the minimal cost to ensure correct decryption,

- *min_circuit_budget*, the minimal cost of the circuit evaluation,

$$\begin{split} & max_circuit_noise(\text{BFV},L) = w \\ & min_circuit_budget(\text{BFV},L) = -\log_2 2w \\ & max_decryption_noise(\text{BFV}) = \frac{\left\lfloor \frac{q}{t} \right\rfloor - (q \mod t)}{2} \\ & min_decryption_cost(\text{BFV}) = -\log_2(\left\lfloor \frac{q}{t} \right\rfloor - (q \mod t)) \end{split}$$

Parameters must fulfill these equivalent conditions to ensure a correct result after decryption:

$max_circuit_noise \le max_decryption_noise$ (2)

$min_circuit_budget \ge min_decryption_cost$ (3)

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Implementation politic	Cingulata _BFV	$SEAL_BFV^a$	FV-NFLlib		
ainhantant madulus a	power of 2	upper bounded ^b product of distinct primes	product of particular primes ^c of size of the type		
cipilertext modulus q		each prime congruent to 1 modulo $2n$ (NTT multiplications) ^d	used to store the polynomial (i.e., 16,32 or 64 bits) minus two		
plaintext modulus t	2	≥ 2 (integer encoder)	> 9		
		a prime congruent to 1 modulo $2n$ (batch encoder)	2		
default Gaussian noise width σ	$2\sqrt{n}$	3.2	3.2		
standard deviation $\sigma \sqrt{2\pi}$	$2\sqrt{2\pi n}$	8	8		
statistical distance ϵ within	2-64	0-28	2^{-128} (and [20])		
noise distribution and Gaussian distribution	2	2 2	2 (see [29])		
noise max width $B_{error}(\epsilon, \sigma)$	10σ (as recommended in [4])	6σ	14σ		
private key distribution χ	uniform distribution in $\{0, 1\}^n$	uniform distribution in $\begin{bmatrix} 1 & 0 & 1 \end{bmatrix}^n$	noise distribution		
	with Hamming weight 63	uniform distribution in {-1, 0, 1}	noise distribution		
compatible with Regev's security reduction	yes	no	no		
relinearization method [4]	2	1	1		

Table 1. Implementation politics are parameter constraints in different implementations of BFV scheme. The parameter n is the ciphertext polynomial degree, a power of two, for performance reason.

^b See the file native/src/seal/util/hestdparms.h in Microsoft SEAL (release 3.3) for upper bounds on size depending on modulus degree.

^c See the file include/nfl/params.hpp for exact form in FV-NFLlib library

^d The congruence is required for batching with CRT.

4 Analysis and presentation of experimental results

4.1 CinguParam in standalone mode

CinguParam permits to:

- generate and store a database of parameter sets for different
 - desired levels of security
 - multiplicative depths
 - implementation politic

- estimate with hindsight the security of parameter sets against LWE attacks

Input parameters Database contains files whose names indicate main input parameters in the following order:

- The circuit multiplicative depth, an integer $0 \le L \le 20$
- The BKZ lattice reduction cost model (see Table 4)
- The approximated security level λ_1 (80,128,192,256)
- The plaintext modulus t (see Table 1)

The database can be easily extended to customized parameters. It is important to note that there is no parameter set for all $(L, cost_model, \lambda_0, t)$. This is due to the gap between desired security level λ_0 and estimated security level λ which can be greater than 64 bits. For this reason, at the end of database generation, each file is renamed by approximating estimated security level λ (see 3 with standard security levels. We only consider parameter sets in the range [[80, 256]]. For this reason, the database can contain less parameter sets than expected and we employ two methods: Algorithm 2 and 3 to select parameters inside Algorithm 1. Our experimental results indicate Algorithm 3 generate more performant parameters in terms of ciphertext size, *i.e.*, $n \log_2(q)$. Note, it is not possible to employ Algorithm 3 for FV-NFLlib and old versions of SEAL (up to release 3.2) since we can not ensure output modulus would have a bitsize with particular properties, as imposed by these implementations (see Table 1 and 2). Parameter set contain a decomposition of ciphertext modulus bitsize suited for Microsoft SEAL (release 3.3) to ease its use in the library.

^a Up to release 3.2 of this library, ciphertext modulus bitsize is restricted to multiple of 10. This impeaches tight parametrization in a similar manner as FV-NFLlib, in Table 2.

Implementation politic	Cingulata_BFV		SEAL_BFV		FV-NFLlib	
Generation method	2 MinCorrectModulus		65537 MinCorrectModulus		65537 MinSecureDegree	
$L \cos t \mod \lambda_1$	$\log_2(a)$	n	$\log_2(a)$	n	$\log_2(a)$	n
0 a core sieve 80	82(1)				82(4)	
0_core_sieve_80		20.10	- 1	20.10		
0_bkz_sieve_128	54	2048	54	2048		
0_bkz_enum_192						
0_bkz_sieve_80	'				62	2048
0_bkz_enum_128						2048
0_paranoid_sieve_128	54	4096	54	4006	62	
0_q_core_sieve_192			- 04	4090	02	4096
0 bkz sieve 192						
5_bkz_sieve_80	150	1000	254	8192	010	0100
5_bkz_enum_80	158	4096			310	8192
5_bkz_enum_128			254	8192		
5_paranoid_sieve_80					310	16384
5_core_sieve_128	166	8192			310	10304
5_bkz_sieve_128	100	0101		-		
5_q_core_sieve_128					310	16384
5 paranoid siova 128			262	16384		
5 core sieve 192						
5_paranoid_sieve_192	174	16384				
10_bkz_sieve_80	000	0100	470	16904		10004
10_bkz_enum_128	302	8192	478	16384	558	16384
10_paranoid_sieve_80						
10_bkz_sieve_128	326	16384				
10_core_sieve_128	020	10001				
10_q_core_sieve_128					558	32768
10 bkz sieve 192						
10_core_sieve_192			502	32768		
10_q_core_sieve_192						
10_paranoid_sieve_192	342	32768				
15_core_sieve_80	470	16384	478	16384	868	32768
15_bkz_enum_80			702	10001		
15_bkz_enum_192	470	16384			868	32768
15_bkz_sieve_80						
15_DKZ_SIEVE_128					868	32768
15 paranoid sieve 80			734	32768	I	
15_core_sieve_128						
15_paranoid_sieve_128					000	GEE2G
15_q_core_sieve_192	502	32768			000	00000
15_bkz_sieve_192	002	52100				
15_core_sieve_192			700	05500		
15_paranoid_sieve_192			766	65536		
20_Dkz_enum_80	614	16384				
20_bkz_enum 128			966	32768	1116	32768
20_core_sieve_80			000	02100	I	
20_paranoid_sieve_80						
20_bkz_sieve_128	654	32768				
20_core_sieve_128	001	02100				
20_q_core_sieve_128					1116	65536
20_paranoid_sieve_128						
20_DKZ_SIeve_192			1006	65536		
20_core sieve 192						
20_paranoid_sieve_192	694	65536				

Table 2. Some parameters from CinguParam database for different politics and plaintext moduli. Let us recall, ciphertexts are represented by polynomials of degree n with coefficients of bitsize $\log_2(q)$, in BFV cryptosystem. Computation budget (multiplicative depth, in this case) has an important impact on memory/time cost. Different files have same parameters, the approximated security level λ_1 reflects the degree of pessimism/optimism of each BKZ cost model and the difficulty to determine security level compared to traditional cryptosystems.

Security and correctness In general, security strength is defined in terms of bit of security (*e.g.*, 80 for legacy standard protection level, 128 for security at least ten-years, 192 for long-term protection). For post-quantum systems, 5 different categories [30, pp.15-19] have been defined by NIST because of uncertainties on security estimations. The first approach is known from a wider audience. For this reason, we make use of it, in CinguParam.

In LWE-based cryptography, one important difficulty is to estimate the attack cost. Lattice reduction algorithm are employed in LWE attacks, it makes use of an SVP oracle (two families: enumeration and sieving one). There

exist many models in the literature². Each model correspond to a bound on the cost of BKZ-family lattice reduction algorithm. There exist important gaps between estimations as shown in [14]. In other words, it is not possible to obtain desired security level with certitude, we only have estimations according to a BKZ cost model.

The database in CinguParam contains parameter sets considering this selection of BKZ lattice reduction cost models and can be extended for any model of your choice. It is stored in xml files. We select five cost models which are commonly used in the literature. They are given in Table 4 from the most optimistic to the most pessimitic, from the defender point of view.

Security is estimated a priori against primal attack [11] via uSVP. A posteriori, we estimate security against dual-lattice attack and small/sparse secret variant [31] as well as lattice decoding attacks [32]. All estimations are operated with LWE-Estimator [12].

In addition, homomorphic cryptosystem is costly in terms in performance. This forces to reduce the parameter range to lessen costs. For instance, the ciphertext polynomial degree is a power of two. This kind of restriction prevents to have a parameter set with a chosen security level $\lambda(cost_model)$.

To face the situation, we associate a desired security level with an interval of estimated security in Table 3.

Estimated security range	Approximated security level
$\llbracket 80, 128 - tolerance_gap \rrbracket$	80
$\llbracket 128 - tolerance_gap, 192 - tolerance_gap \rrbracket$	128
$\llbracket 192 - tolerance_gap, 256 - tolerance_gap \rrbracket$	192
$\llbracket 256 - tolerance_gap, 312 \rrbracket$	256

Table 3. There is a gap between concrete security and desired security in lattice based cryptography. For the time being, we only associate an approximated security level with an estimated security range. Estimations are bounds on attack cost given a BKZ lattice reduction cost model. Arbitrarily, we choose $tolerance_gap = 8$ during database generation. This can be easily modified if desired.

BKZ cost models	Short description	
BKZ Enum	Enumeration algorithm prevail for dimension < 70 , sieving one for greater dimensions.	[33, Table 4]
BKZ Sieve	The number of calls to an SVP oracle is polynomial and estimated heuristically.	[34, Figure 3]
Core Sieve (mode classical)	Core models evaluate one call to an SVP oracle	
Q-Core Sieve (mode quantum)	The attacker benefit from Grover's search algorithm	[11]
Paranoid Sieve (mode paranoid)	Best plausible attack cost	

Table 4. Default BKZ lattice reduction cost models in CinguParam database. We can easily extend it for better models to come. The costs are exponential in the blocksize β for each model, they are defined in https://bitbucket.org/malb/lwe-estimator/raw/HEAD/estimator.py.

4.2 CinguParam combined with cryptosystems implementations

We focus here on implementations with recent activities. Let us present them briefly. On one hand, Microsoft SEAL is a library permitting to use two homomorphic encryption schemes: full-RNS BFV [35] and CKKS [36]. The release 3.3 (june 2019) increases flexibility on parameter choices by removing implementation constraints on bitsize of ciphertext moduli q. On the other hand, Cingulata is a compiler toolchain and runtime environment for running programs over encrypted data with two homomorphic encryption schemes since june 2019: BFV and TFHE [28]. With Microsoft SEAL (release 3.3) and potentially any implementation of BFV, we can generate automatically snippets to initialize cryptographic parameters by selecting any parameter set in CinguParam. We demonstrate with Cingulata, a more practical usage of CinguParam. Parameters are selected automatically using CinguParam as a submodule for any program written for BFV implementation in Cingulata. During compilation of the program describing the homomorphic computations, CinguParam makes use of it to select automatically parameter set in

² https://estimate-all-the-lwe-ntru-schemes.github.io/docs/

the database. Default security parameters are $\lambda_1 = 128$, cost_mode=BKZ Sieve. Plaintext modulus is t = 2 in Cingulata. If no parameter set is found given input parameters $(L, cost_model, \lambda_1, t)$, a static choice is made by default to find some approaching parameter set. For instance, a parameter set with multiplicative depth L + 1 or L + 2 is chosen if no parameter set is found for the quadruplet of input parameters. An interactive mode is also proposed to select manually a parameter set in a filtered database if nothing corresponds to input parameters.

4.3 Future works

There exist several trails to extend our work :

- Tighten noise estimation to improve performance (see Table 2) using noise variance, in the average case (see Section 3). In counterpart, this approach requires additional hypotheses.
- Take into account hybrid relinearization technique introduced in [37].
- Extend CinguParam with other LWE-based homomorphic cryptosystems.

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