

# Implementing the FV FHE Scheme

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- 1 Introduction
- 2 Theoretical Description
- 3 Implementation Description
- 4 Practical Demonstration
- 5 Conclusion

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- Implemented everything but Bootstrapping

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- Alternatively, variant of the LPR Cryptosystem that uses relinearization-based multiplication
  - I will explain things from this perspective.

# Ring LWE

- Throughout, fix  $R_q = \mathbb{Z}_q[x]/(2^d + 1)$  for  $d = 2^n$ , and  $q \in \mathbb{N}$ .

## Ring LWE Assumption

For random  $s \in R_q$ , and a distribution  $\chi$  supported on  $R_q$ , the RLWE assumption is that for  $a(x) \leftarrow R_q$ ,  $u(x) \leftarrow R_q$ ,  $e(x) \leftarrow \chi$ :

$$(a(x), a(x)s(x) + e(x)) \approx_c (a(x), u(x)).$$

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- $\chi$  a discrete Gaussian of parameter  $\sigma$ .

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- KeyGen:  $sk \leftarrow \chi$ , and  $pk = (a(x), a(x)s(x) + b(x))$  is an RLWE sample



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$$u(x) \begin{pmatrix} a(x)s(x) + e(x) \\ -a(x) \end{pmatrix} + \begin{pmatrix} e_0(x) \\ e_1(x) \end{pmatrix} + \begin{pmatrix} \Delta m(x) \\ 0 \end{pmatrix}.$$

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- Easily supports additive homomorphism
- Decryption: For a ciphertext  $\begin{pmatrix} c_0(x) \\ c_1(x) \end{pmatrix}$ , compute  $c_0(x) + s(x)c_1(x)$ , and perform simple error-correction procedure.

# Relinearization

- Idea: view decryption  $\text{Dec}_s(c_0, c_1) = c_0 + c_1s = f_{c_0, c_1}(s)$  as a degree-1 polynomial in  $s$ , where

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- After scaling down by  $\Delta$ , one obtains an encryption of  $mm'$

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- 2 Relinearization is a technique to address both of these issues
  - I will focus on discussing the first issue

## Reducing Degrees

- The goal is to **linearize** the degree 2 polynomial, i.e. find  $c'_0, c'_1$  such that

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    - Issue:  $c_2e$  may be **large**



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- Each  $c_{2,i}$  **small** now
  - Downside: Relinearization key needs encryptions of  $T^i s^2$  for each  $i$

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  - Space complexity that scales poorly with  $\sigma$ .
- Relinearization via Digit Decomposition

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  - Relinearization works

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  - Faster polynomial arithmetic
  - Better Gaussian sampling
- Fun Implementation
  - The easy parts were hard, and the hard parts were easy