

A Full RNS Variant of Approximate Homomorphic Encryption

Jung Hee Cheon, Kyoohyung Han, Andrey Kim (Seoul National University)

Miran Kim (UTHealth), **Yongsoo Song** (UC San Diego)

SAC 2018

 Residue Number System (a.k.a. CRT)

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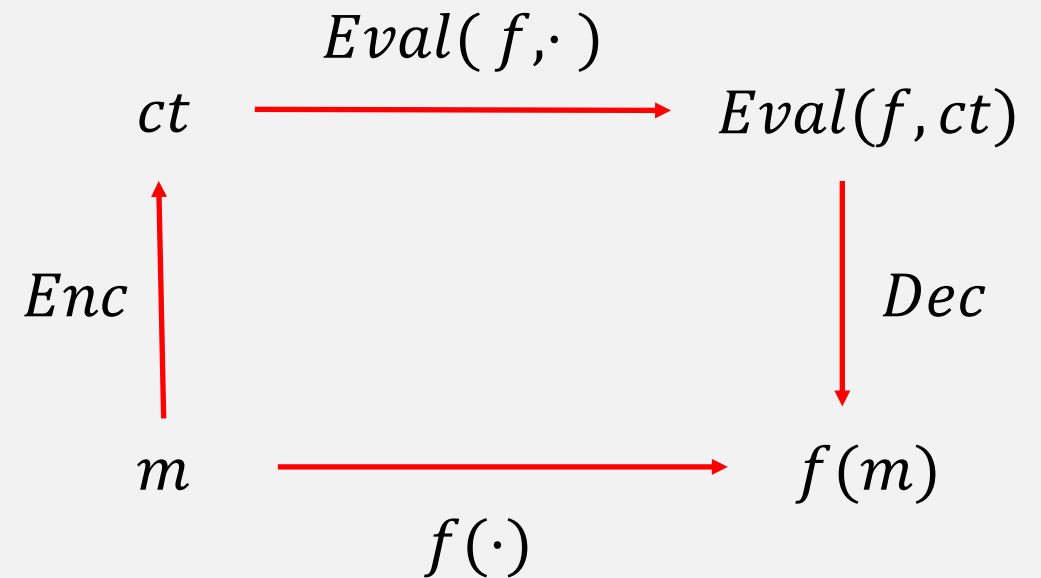
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Background

Secure Computation

- ❑ Differential Privacy
- ❑ (Secure) Multi-Party Computation
- ❑ (Fully) **Homomorphic Encryption**
 - Semantic security.
 - Non-interactive.
 - Reusable.
 - Long-term storage, Unlimited sources.



Landscape of HE Schemes

Scheme	Word Encryption	Bit Encryption	Approximate Encryption
Scheme (Library)			
Plaintext Space			
Operation			

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Plaintext Space	Finite field + Packing	Single Bit	Real / Complex + Packing
Operation	Addition, Multiplication	Binary Gate + Bootstrapping	Addition, Multiplication, Rounding

Approximate HE (HEAAN, 慧眼)

□ Design

- Homomorphic Encryption for Arithmetic of Approximate Numbers [CKKS (AC'17)]
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[KSW+ (JMI'18), KSK+ (iDASH'17, BMC'18), CKKS (IEEE Access'18)]

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[KSW+ (JMI'18), KSK+ (iDASH'17, BMC'18), CKKS (IEEE Access'18)]
- Matrix Computation & Evaluation of Neural Networks [JKLS (CCS'18)]

Approximate Computation

□ Numerical Representation

- $1.234 = 1234 \cdot 10^{-3}$.
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- Division by scaling factor p (a.k.a. Rounding operation).

(Leveled) Approximate HE

- Approximate Encoding / Encryption
 - (Ring) LWE-based.

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- $x \mapsto m = \lfloor p \cdot x \rfloor$. p : scaling factor. m : significant digits of x .

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- *Mult*: $(Enc(m_1), Enc(m_2)) \mapsto Enc(m \approx m_1 m_2 \approx p^2 \cdot x_1 x_2)$.

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- **Leveled** Structure : $(q_L = p^L) > (q_{L-1} = p^{L-1}) > \dots > (q_1 = p)$.

Main Result

Motivation

Ring structure $R_q = \mathbb{Z}_q[x]/(x^n + 1)$.

Expensive operation & High-precision library ($\log q = 250 \sim 800$).

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Homo. Operations	Full RNS B/FV Variants [BEHZ17, HPS18]	
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Ideal : Approx RNS Basis

□ Rounding Operation

- $Enc(m) \pmod{q_l} \mapsto Enc(p_l^{-1} \cdot m) \pmod{q_{l-1}}$ for $p_l = q_l/q_{l-1}$.

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$$R_{q_L} \cong R_{p_1} \times R_{p_2} \times \dots \times R_{p_L} \text{ for } q_L = p_1 p_2 \dots p_L.$$

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□ Example ($p = 2^{55}$, $n = 2^{15}$)

$p_1 = 80000000080001, p_2 = 80000000130001, p_3 = 7FFFFFFFE90001, \dots$

$$R_{p_1} \times R_{p_2} \times \dots \times R_{p_L} \cong \mathbb{Z}_{p_1}^n \times \mathbb{Z}_{p_2}^n \times \dots \times \mathbb{Z}_{p_L}^n.$$

Idea2 : Approx Modulus Switching

□ Non-Polynomial Algorithms

- Key-switching process (e.g. Homomorphic multiplication)

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Alternative algorithms **without** RNS conversions?

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$$\text{RNS}_{p_i}^{-1}(a_i) \equiv \sum_i [a_i \cdot \hat{p}_i^{-1}]_{p_i} \cdot \hat{p}_i \pmod{q_l} \quad \text{for } \hat{p}_i = q_l/p_i.$$
$$\sum_i [a_i \cdot \hat{p}_i^{-1}]_{p_i} \cdot \hat{p}_i = q_l \cdot t + a \quad \text{for a small } t.$$

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□ Our Approx Mod Raising Algorithm (from q_l to $\Delta \cdot q_l$)

$$R_{p_1} \times \cdots \times R_{p_l} \rightarrow R_{p_1} \times \cdots \times R_{p_l} \times R_{\Delta_1} \times \cdots \times R_{\Delta_k},$$

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RNS Friendly Computation & **Correctness** of Homo Operations
(w/ additional noise)

Summary

□ Idea I: Approximate Basis

- $q_L = p_1 p_2 \dots p_L$ with $p_L \approx p$ for RNS decomposition.
- Approximate error ($p_l^{-1} m \approx p^{-1} m$) of the Rounding algorithm.

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vs **Precision loss** of computation

HEAAN vs RNS HEAAN

Variant	L	N	$\log q$	$\lceil \log Q_L \rceil$	Enc (ms)	Dec (ms)	Add (ms)	Cmult (ms)	Mult&RS (ms)
HEAAN	5	2^{15}	55	336	332	106	30	204	740
	10	2^{15}		611	530	135	32	281	1,355
	15	2^{16}		886	1,465	344	70	762	4,169
HEAAN-RNS	5	2^{15}	55	336	31	4.6	2.9	25	85
	10	2^{15}		611	58	7.8	4.3	44	164
	15	2^{16}		886	177	10.0	15.5	125	563

- 8x ~ 12x speed up

HEAAN vs RNS HEAAN

Function	N	$\log q$	$\log p$	Consumed levels	Input precision	Total time	Amortized time
x^{16}	2^{13}	155	30	4	14 bits	0.31s	0.07ms
x^{-1}						0.45s	0.11ms
$\exp(x)$						0.65s	0.16ms
x^{1024}	2^{15}	620	56	10	36 bits	7.46s	0.43ms

HEAAN
- 14 bits precision

Function	Degree	N	$\log q$	$\lceil \log Q_L \rceil$	Total time	Amortized time
x^{-1}	15	2^{14}	55	281	167ms	$21\mu s$
$\exp x$	7	2^{14}	55	281	164ms	$20\mu s$
Sigmoid	7	2^{14}	55	281	161ms	$19\mu s$

RNS HEAAN
- 32 bits precision

<https://github.com/HanKyoohyung/HEAAN-dev>

Questions?