Client-Optimized Algorithms & Acceleration for Encrypted Compute Offloading

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<u>Client-aided</u> <u>H</u>E for <u>Opaque</u> <u>Compute</u> <u>Offloading</u>



- CHOCO enables privacy-preserving computation for resource-constrained devices
- CHOCO utilizes Homomorphic Encryption (HE) and Client-Aided Encrypted Computing
- CHOCO introduces client-optimized encrypted algorithms & hardware acceleration
- CHOCO makes client responsibility **competitive with local compute**
- CHOCO benefits generalize to diverse applications



Outline

- Introduction
- Resource-Constrained Devices & Privacy-Preserving Offload
- Encrypted Computing Background
- Fully Homomorphic Encryption & Client-Aided Encrypted Computing
- CHOCO System Implementation
- Algorithm Optimizations & Hardware Acceleration
- Results & Conclusions
- Active Client Computation, Communication & Applications



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• Introduction

• Resource-Constrained Devices & Privacy-Preserving Offload

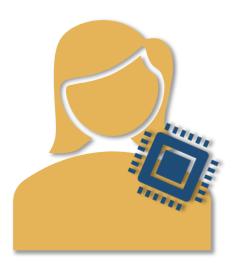
• Encrypted Computing Background

• CHOCO System Implementation

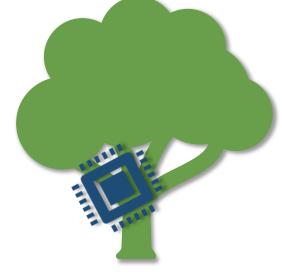
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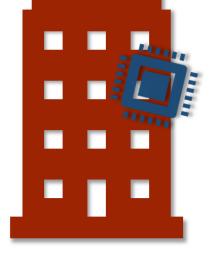


Resource-Constrained Devices are Everywhere



Health Monitoring



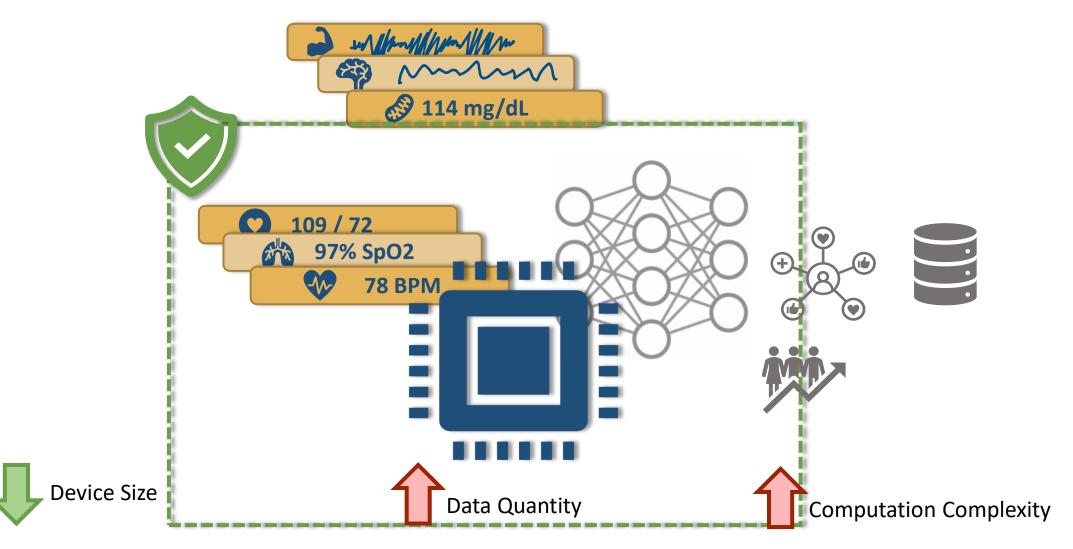


Wildlife Monitoring

Infrastructure Monitoring

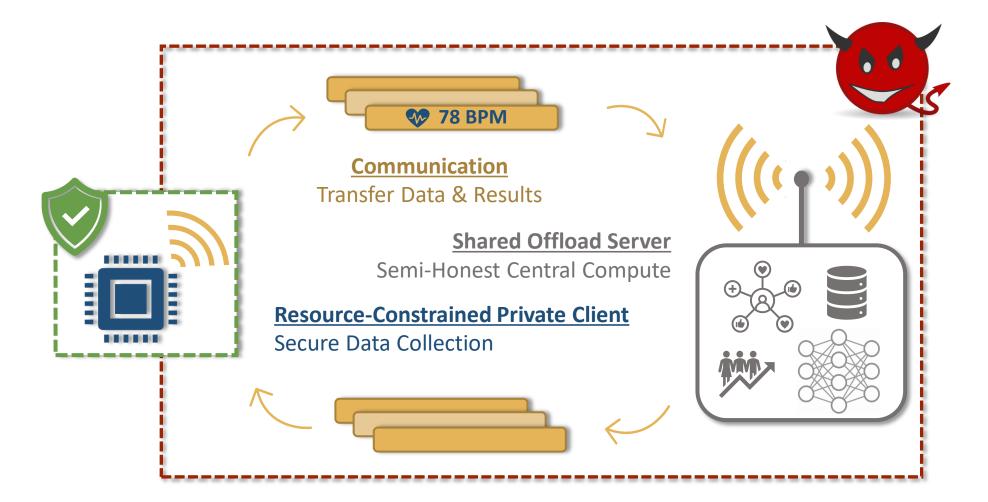


Computational Demands Exhaust Sensor Resources



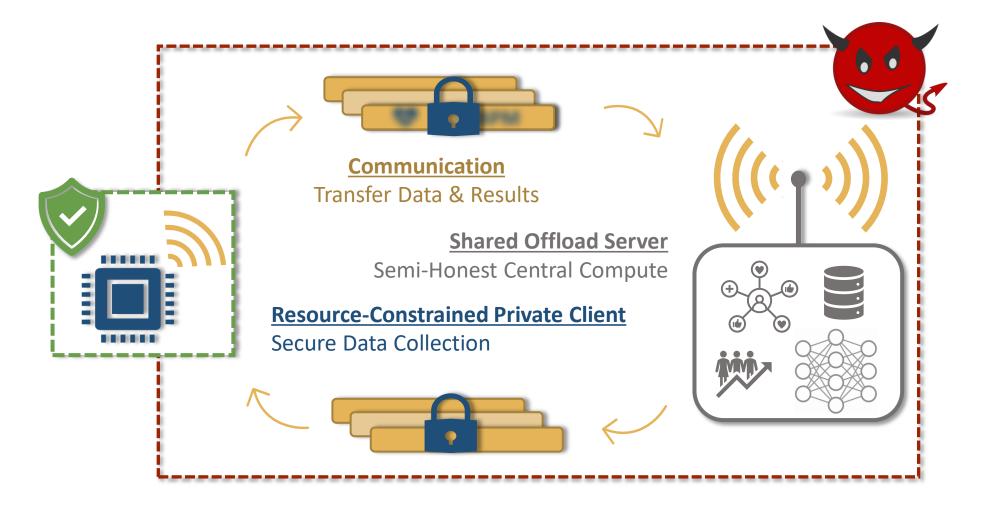


Privacy-Preserving Computation Offload





Privacy-Preserving Computation Offload



FHE: [Fully] Homomorphic Encryption



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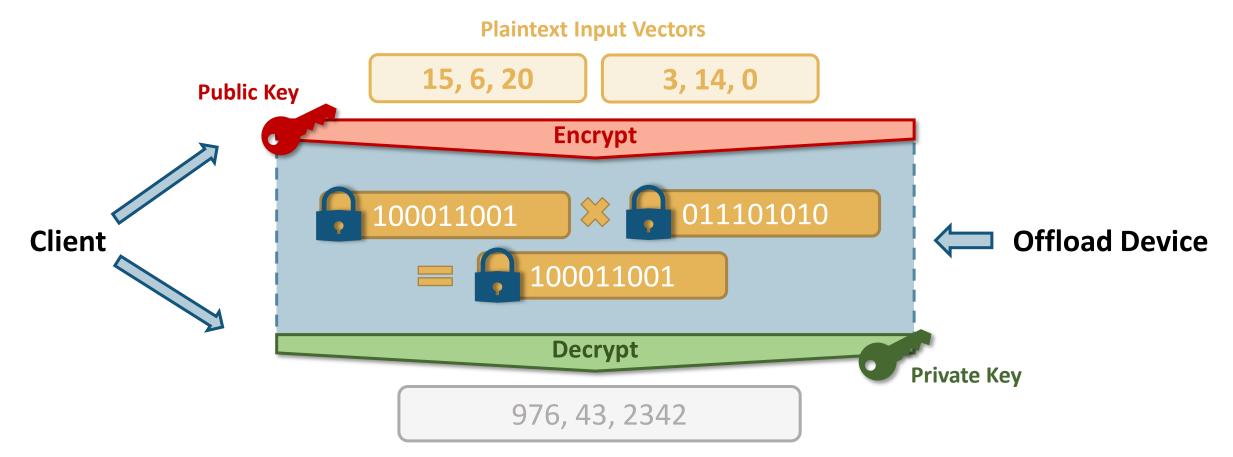
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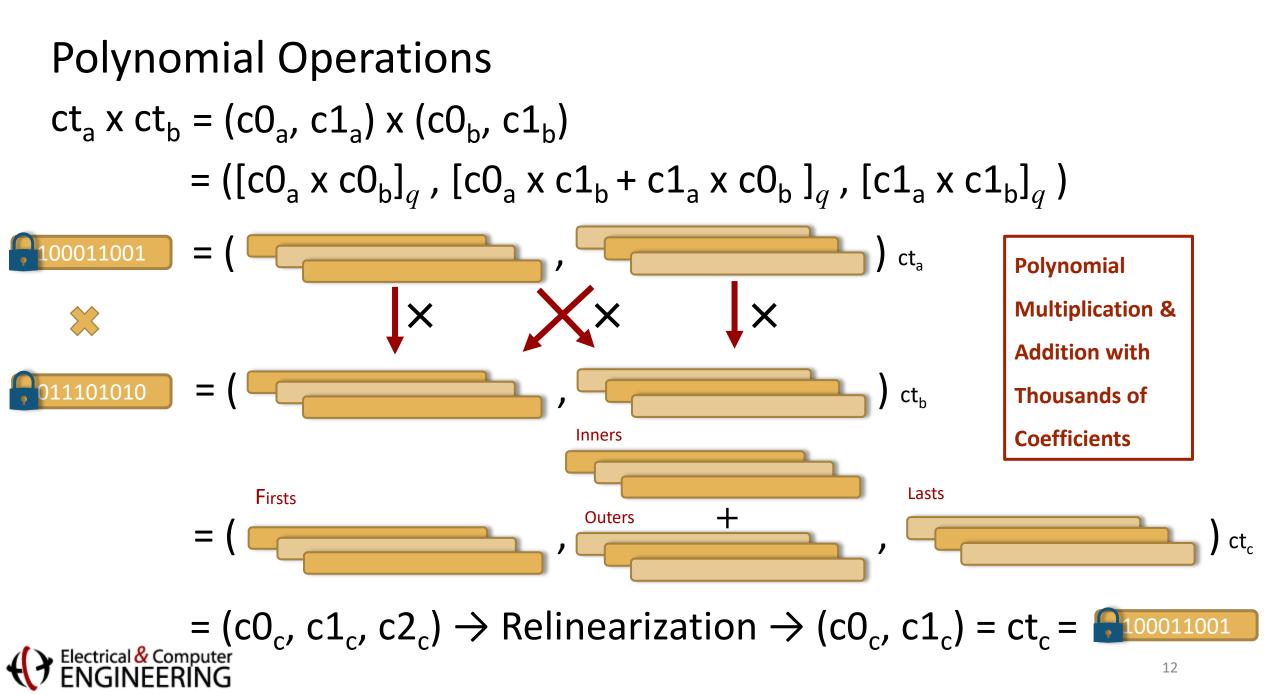


Homomorphic Encryption (HE)



Incorrect output due to noise growth





HE Challenges & Limitations





High Computation Costs

Linear Operations

Noise Growth & Arithmetic Depth

Parameter Selection

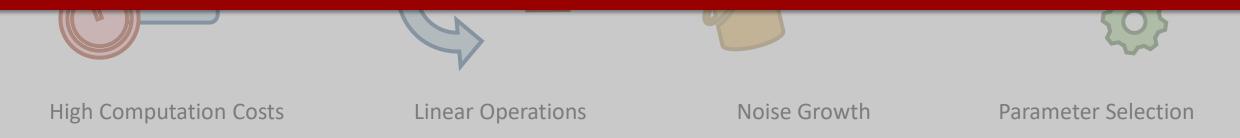




HE Challenges & Limitations



Offloading Entire Encrypted Applications is Still Infeasible in Many Scenarios

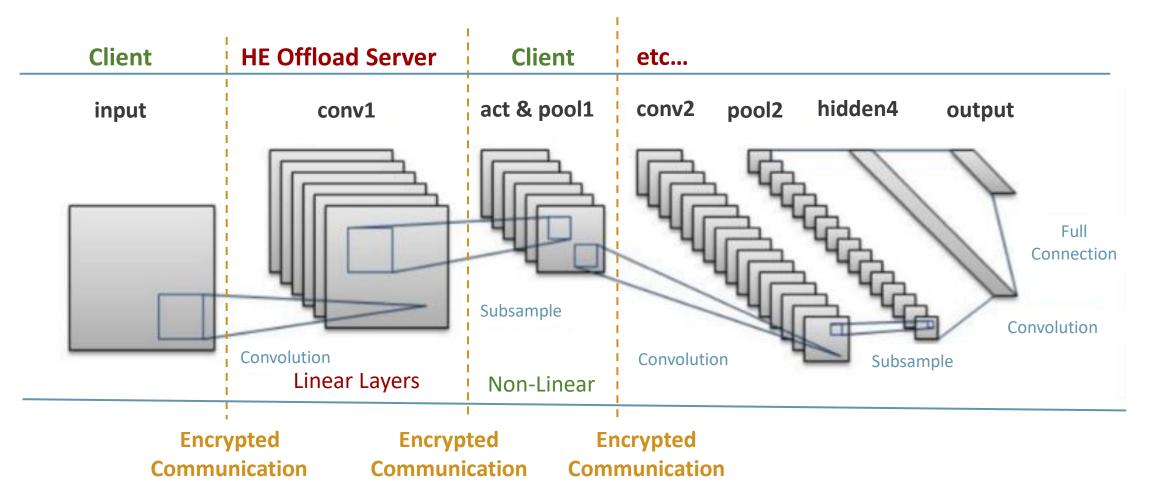




Client-Aided Encrypted Computing



Client-Aided Encrypted Inference

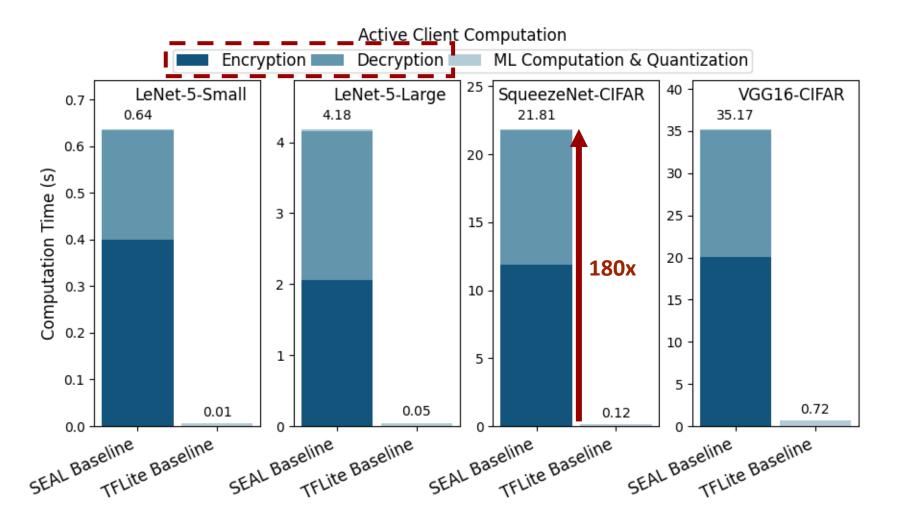


• Systematically limits arithmetic depth & regularly refreshes noise



Quantifying Client Responsibility

- ARM Cortex-A7 CPU Client
- Up to 180x overhead to offload compute
- Dominated by Homomorphic Encryption (HE) operations

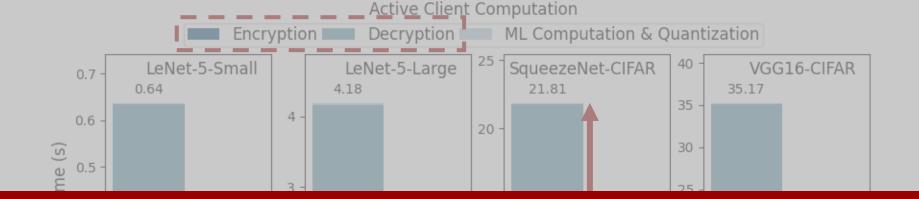




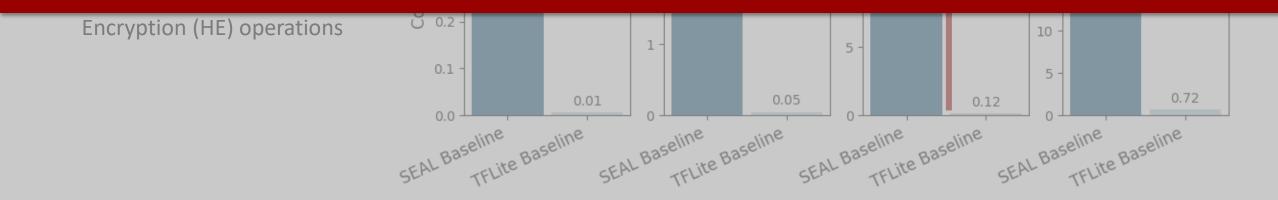
Microsoft Research. 2019. Simple Encrypted Arithmetic Library (release 3.4), <u>https://github.com/Microsoft/SEAL</u> M. Abadi et al. 2015. TensorFlow: Large-Scale Machine Learning on Heterogeneous Systems (release 2.2). <u>https://www.tensorflow.org/</u>.

Quantifying Client Responsibility

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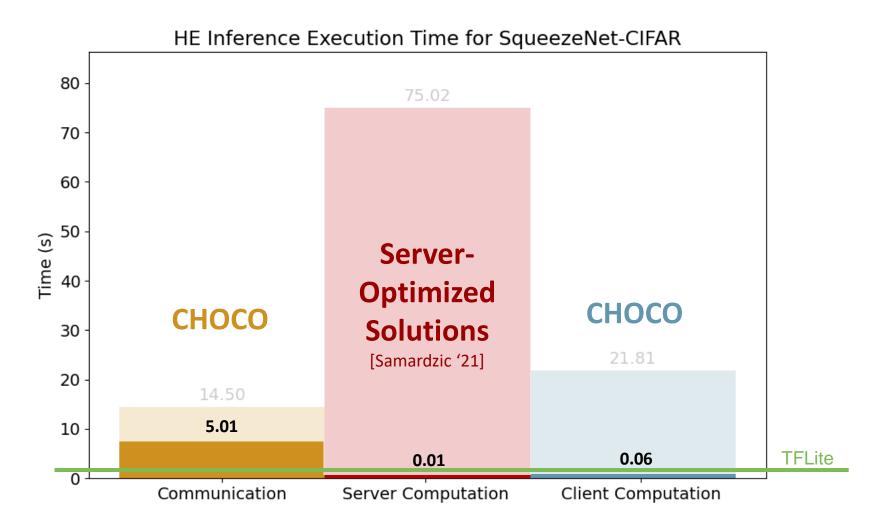
CHOCO Reduces Client-Side Computation by up to 341x through SW Algorithms & HW Acceleration





Microsoft Research. 2019. Simple Encrypted Arithmetic Library (release 3.4), <u>https://github.com/Microsoft/SEAL</u> M. Abadi et al. 2015. TensorFlow: Large-Scale Machine Learning on Heterogeneous Systems (release 2.2). <u>https://www.tensorflow.org/</u>.

Complete Client-Aided System Improvements





Nikola Samardzic, Axel Feldmann, Aleksandar Krastev, Srinivas Devadas, Ronald Dreslinski, Christopher Peikert, and Daniel Sanchez. 2021. *F1: A Fast and Programmable Accelerator for Fully Homomorphic Encryption*. Association for Computing Machinery, New York, NY, USA, 238–252.

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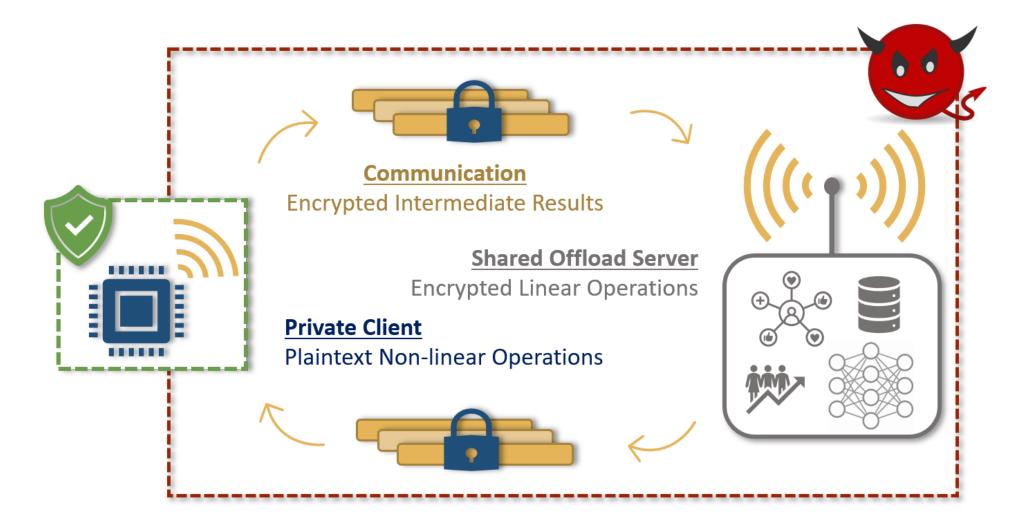
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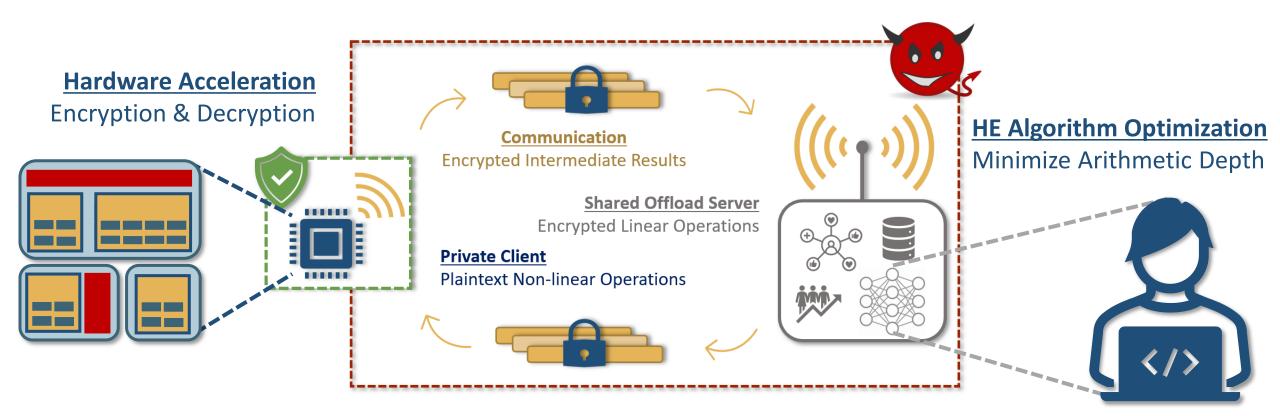


<u>Client-aided</u> <u>H</u>E for <u>Opaque</u> <u>Compute</u> <u>Offloading</u>





<u>C</u>lient-aided <u>H</u>E for <u>Opaque</u> <u>Compute</u> <u>Offloading</u>

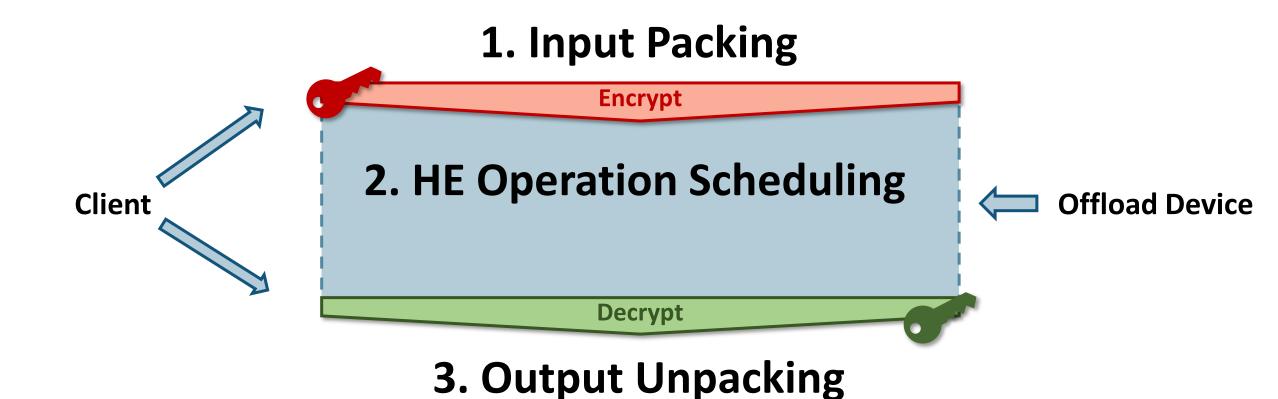




Encrypted Algorithm Optimization

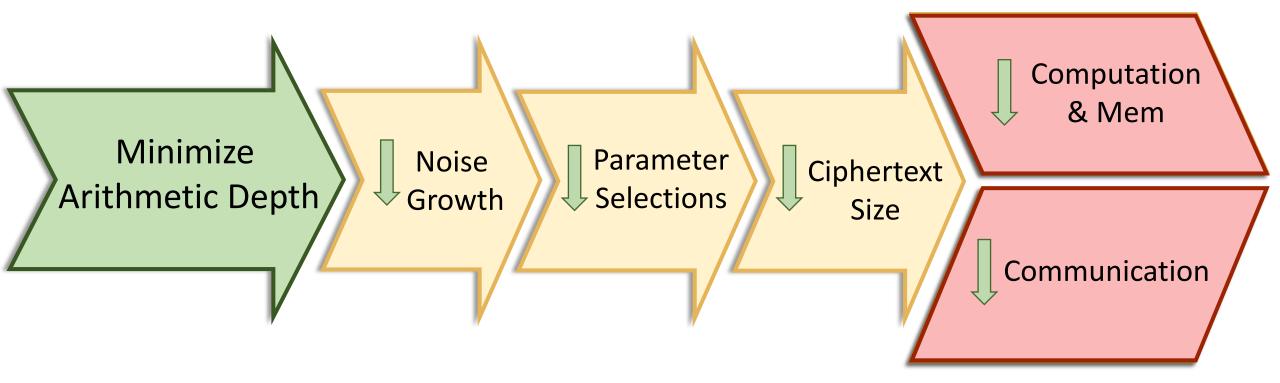


HE Algorithms



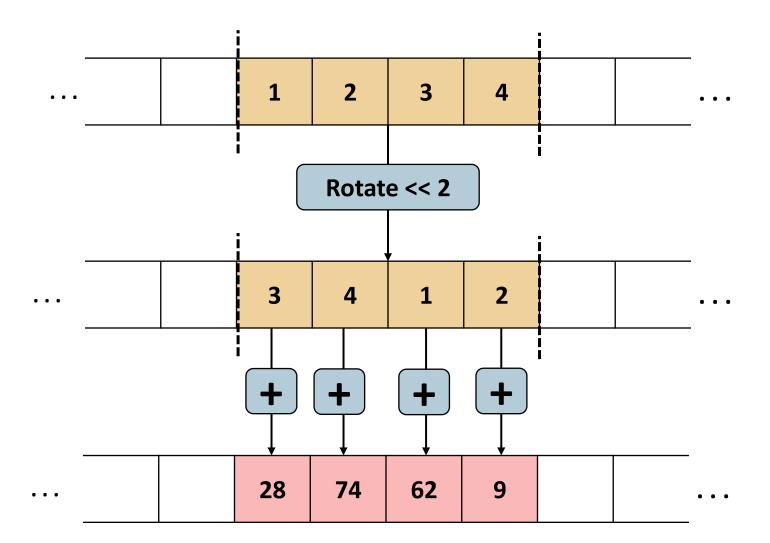


Algorithm Optimizations Impact Client Costs



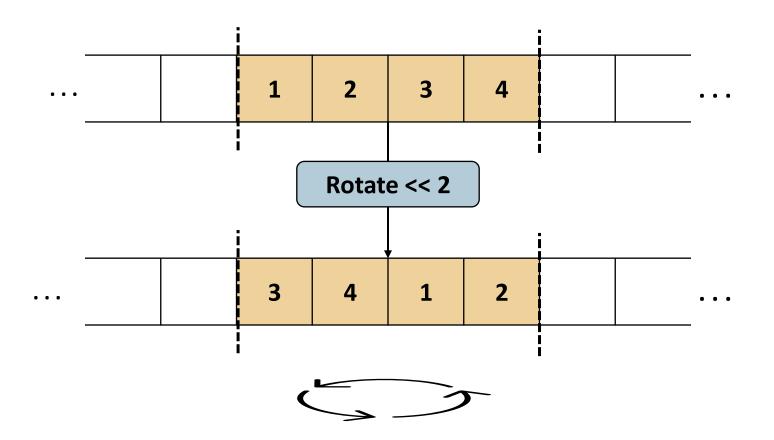


Windowed Rotations





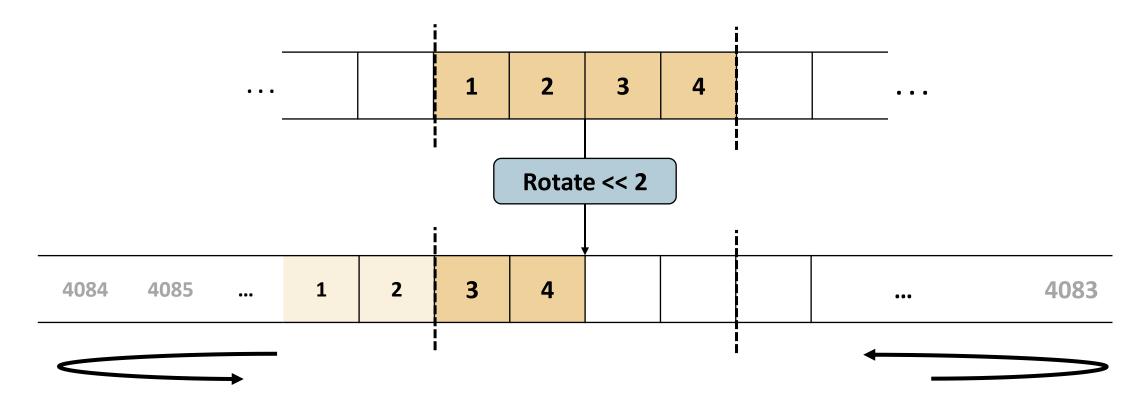
Windowed Rotations



Ideally: Values wrap around within a window of interest



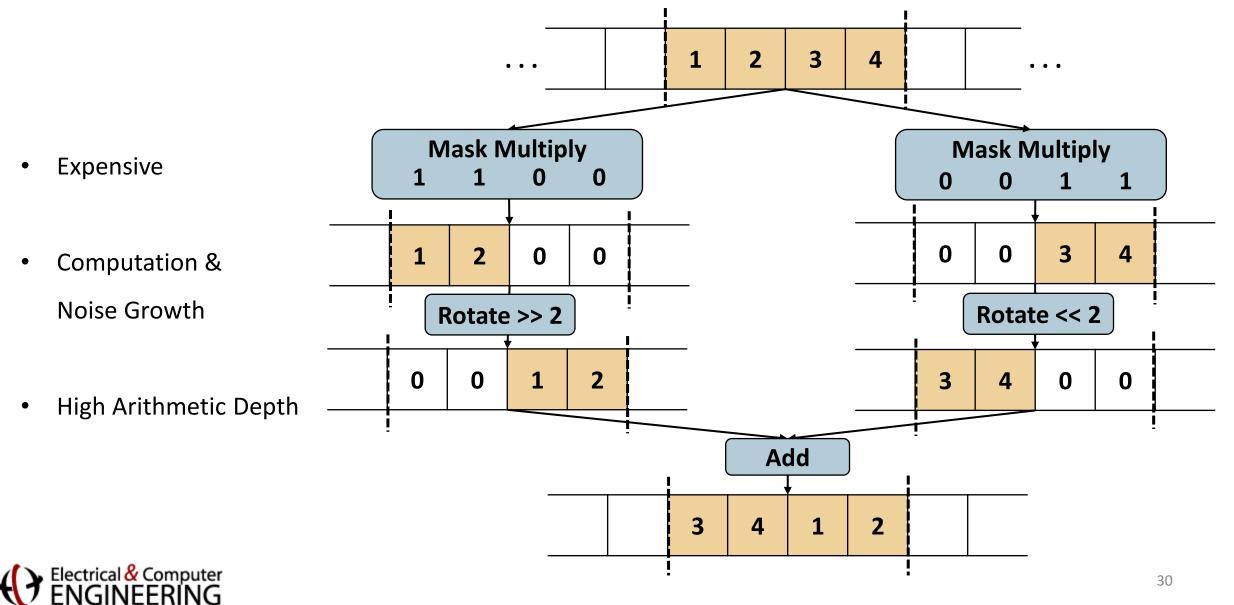
Windowed Rotations



Actually: Values wrap around the entire ciphertext vector

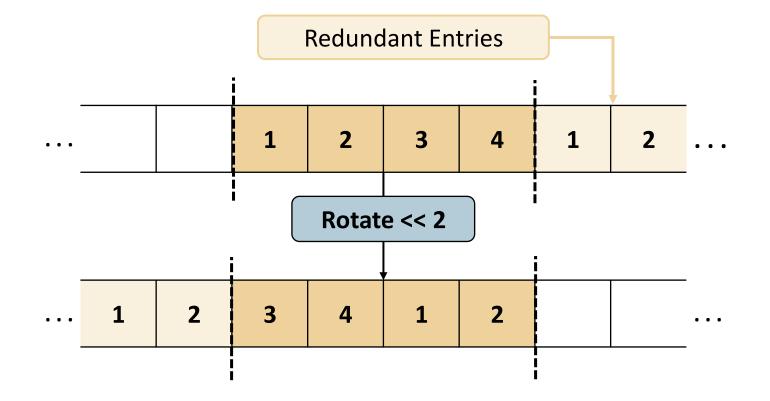


Standard Permutations



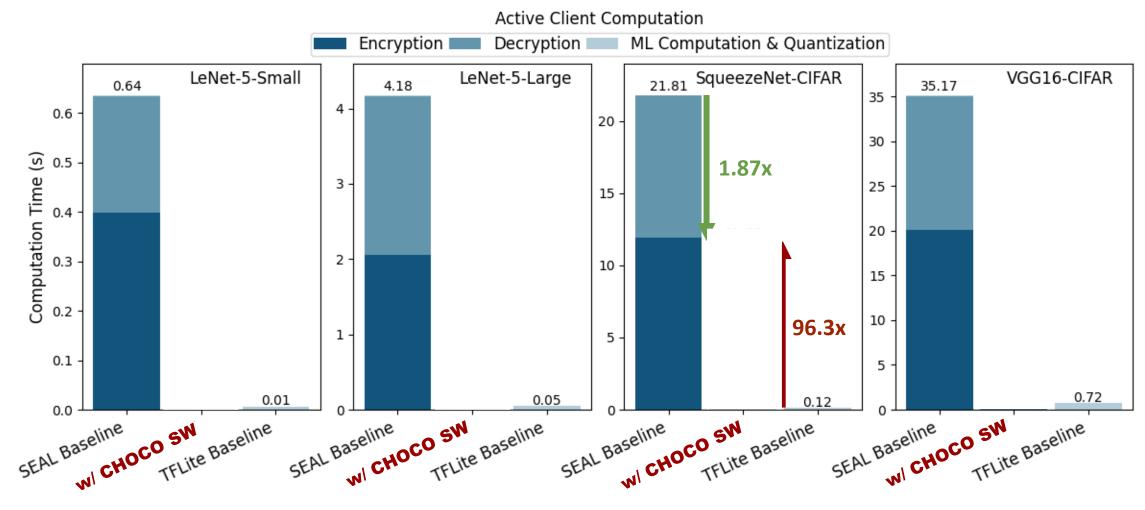
Rotational Redundancy

- Novel Input Packing
- Single HE Rotation
- Low arithmetic depth





CHOCO Algorithms Reduce Client Computation



•

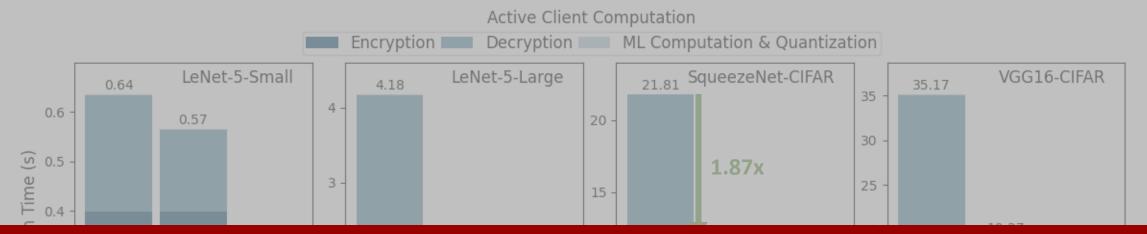
• **50%** Smaller Ciphertexts

CHOCO SW = SEAL baseline + Rotational Redundancy

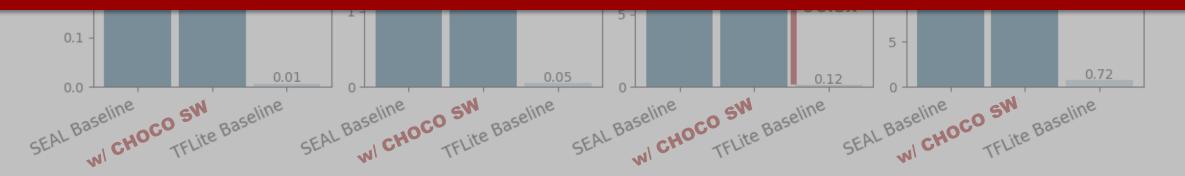
Electrical & Computer

- Average 1.7x improvement over SEAL
- Average 62.5x remaining overhead vs TFLite

CHOCO Algorithms Reduce Client Computation



CHOCO Algorithm Optimizations Provide a Critical but Insufficient Reduction in Client Computation



• **CHOCO SW** = SEAL baseline + Rotational Redundancy • **50%** Smaller Ciphertexts

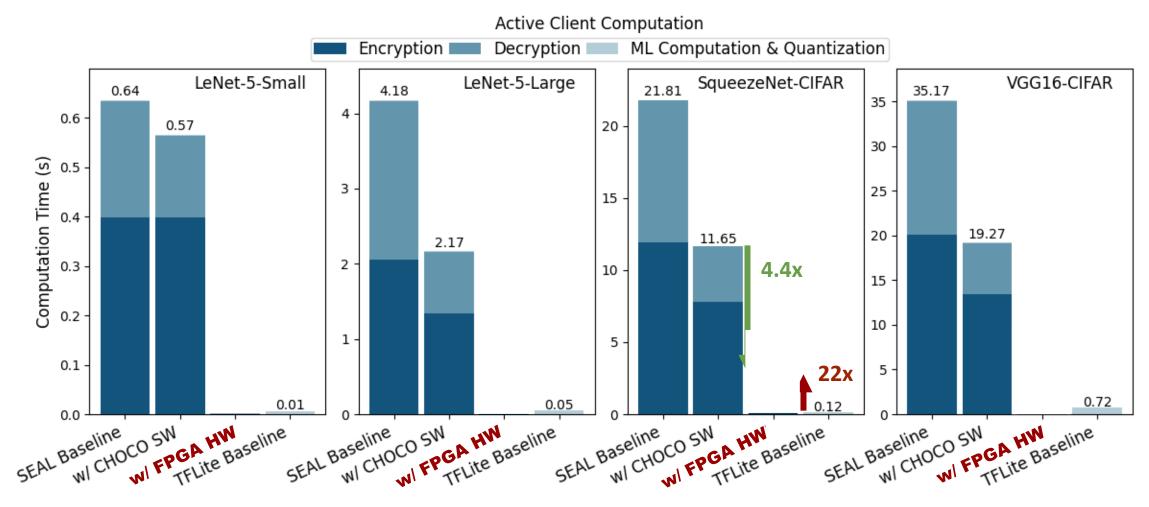


- Average 1.7x improvement over SEAL
- Average **62.5x** remaining overhead vs TFLite

Hardware Acceleration



Existing FPGA Acceleration is Incomplete

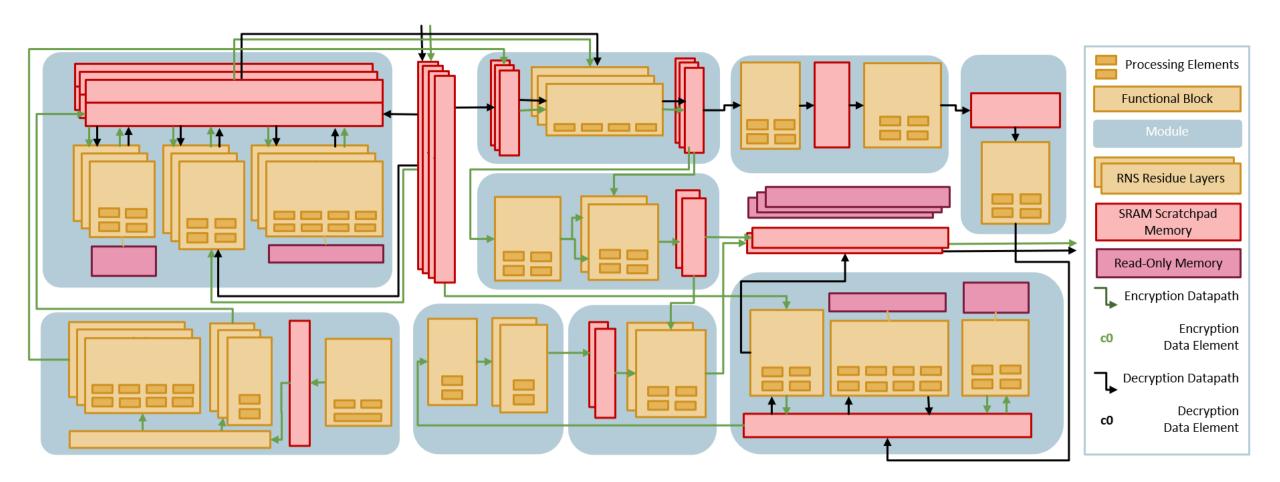


• **FPGA HW** = CHOCO SW + Encryption/Decryption FPGA • Average **14.5x** remaining overhead vs TFLite

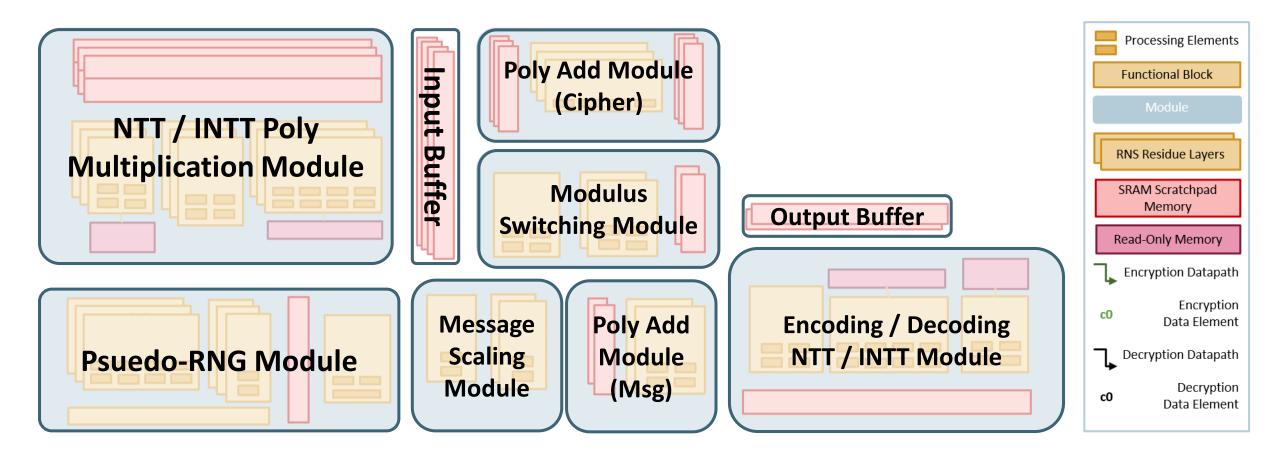


A. Mert, E. Ozturk, and E. Savas. 2020. Design and Implementation of Encryption/Decryption Architectures for BFV Homomorphic Encryption Scheme. IEEE Transactions on Very Large Scale Integration (VLSI) Systems 28, 02 (feb 2020), 353–362.

<u>CHOCO</u> – <u>Through</u> <u>Accelerated</u> <u>Cryptographic</u> <u>Operations</u>



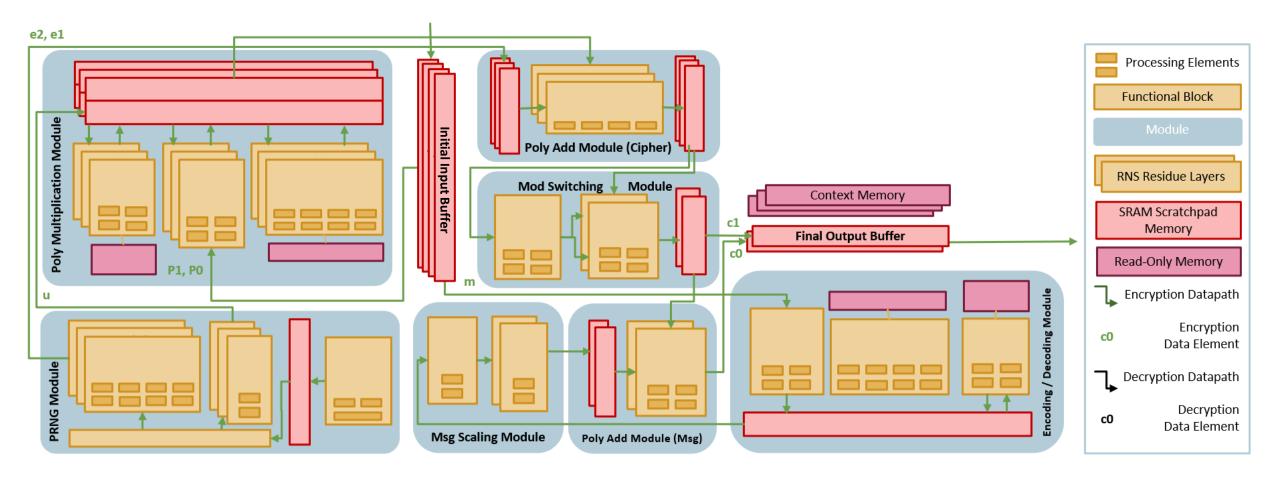




• Encrypt(pk, m): For $m \in R_t$, let $pk = (p_0, p_1)$. Sample $u \stackrel{\$}{\leftarrow} R_2$, and $e_1, e_2 \leftarrow \chi$. Compute

$$\mathtt{ct} = ([\Delta m + p_0 u + e_1]_q, [p_1 u + e_2]_q)$$

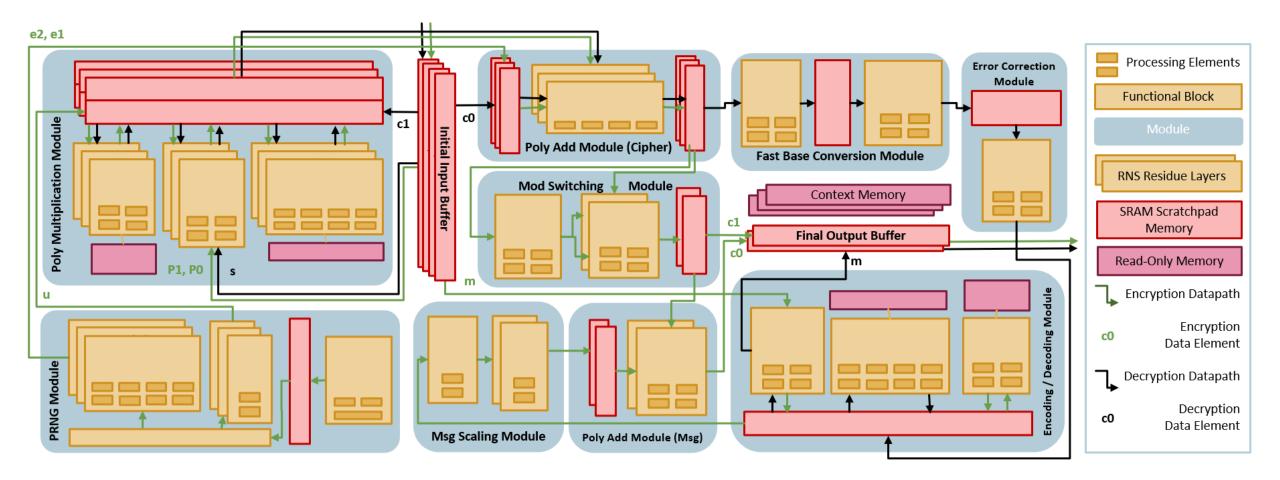




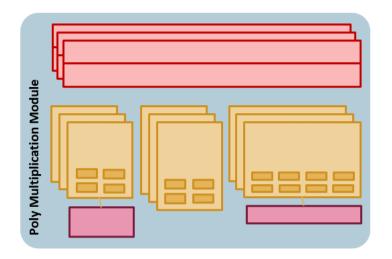
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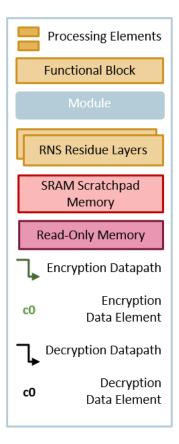
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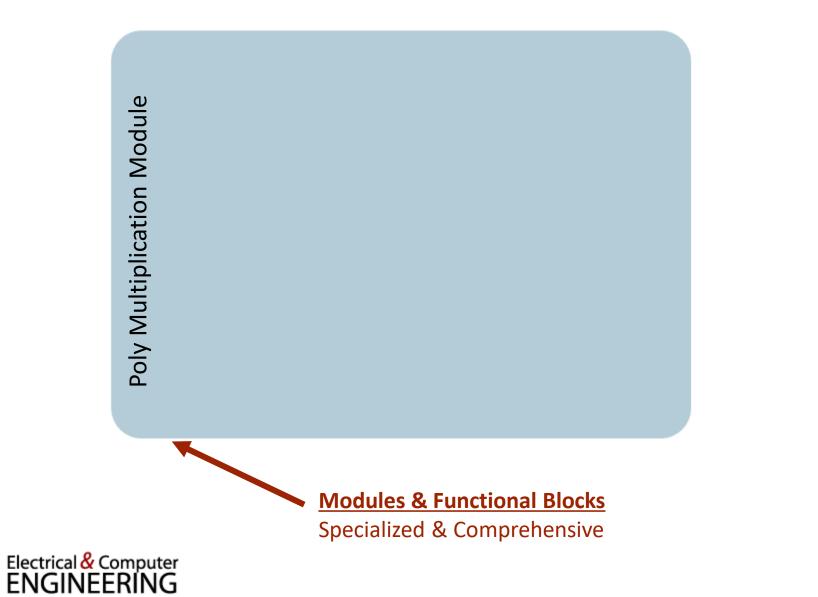


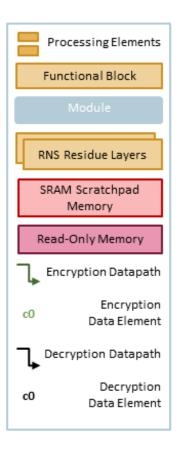


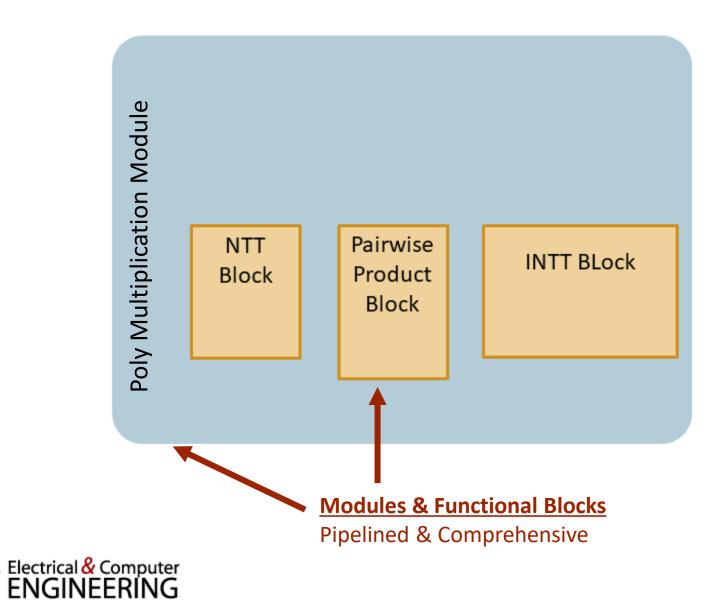


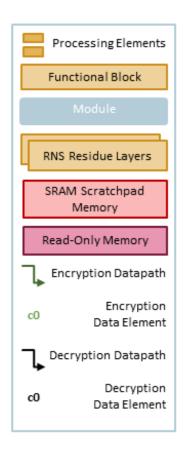


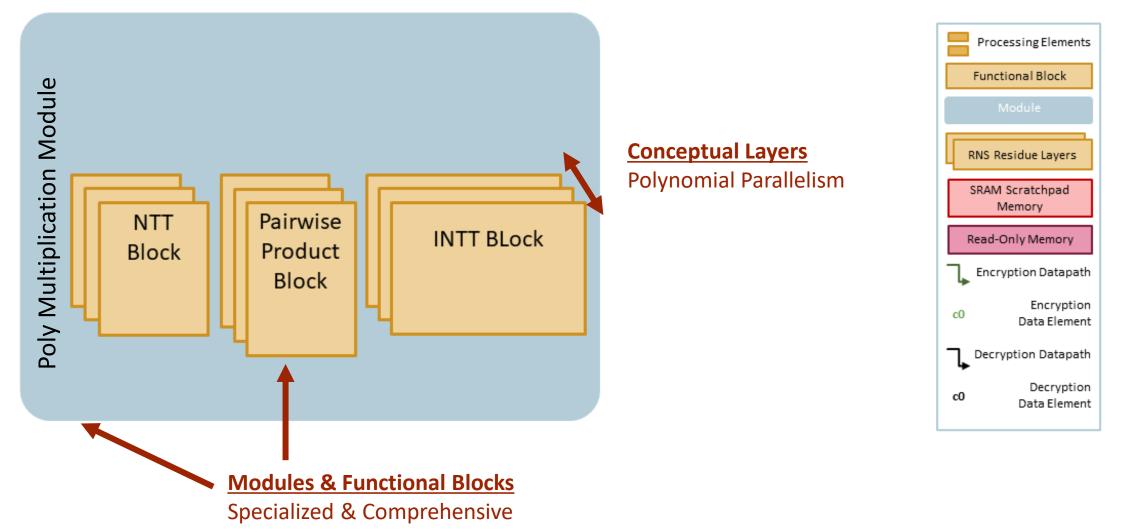




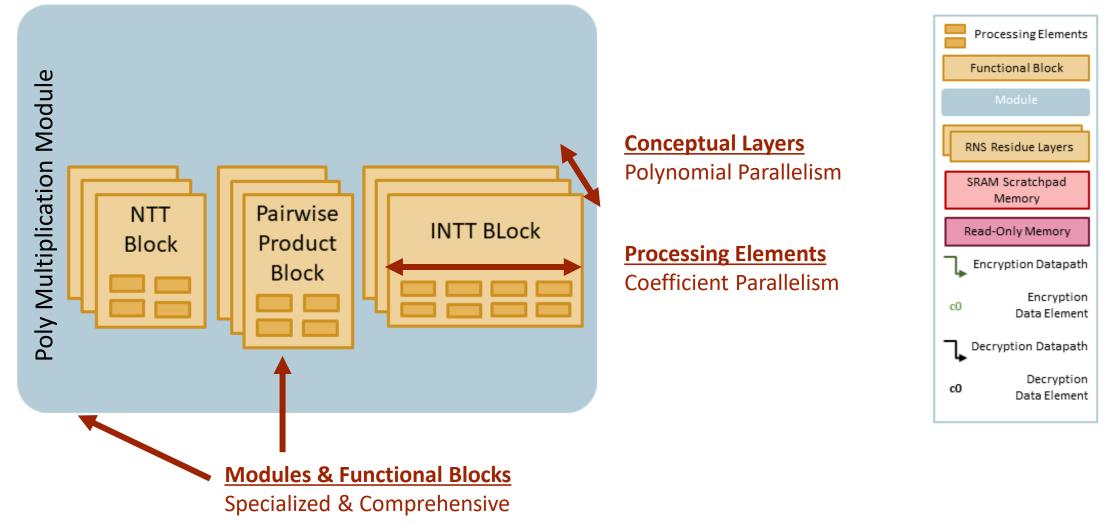




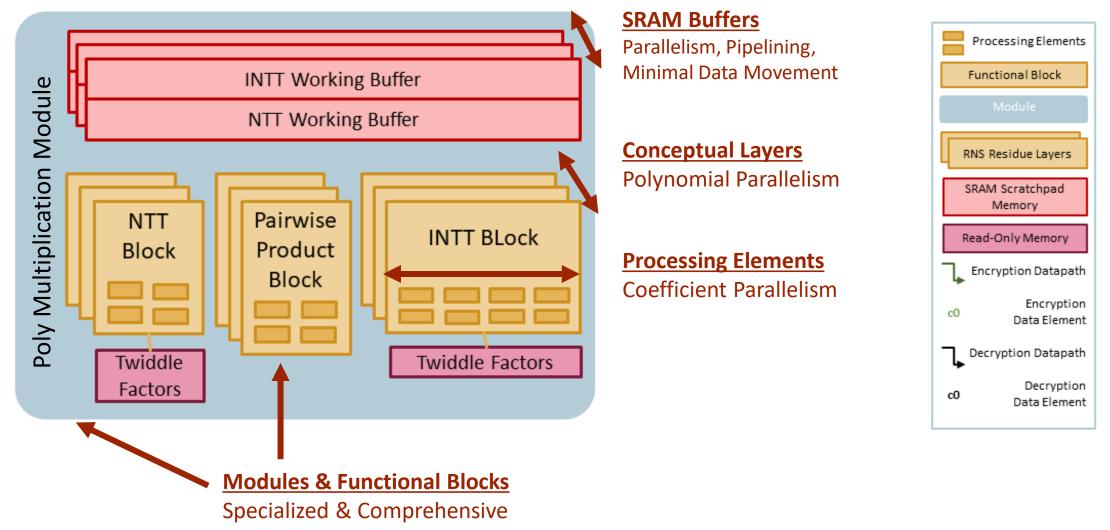






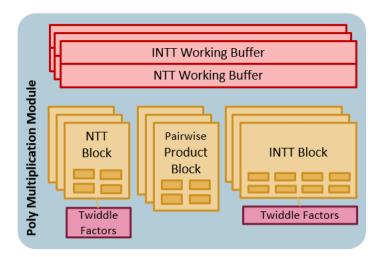


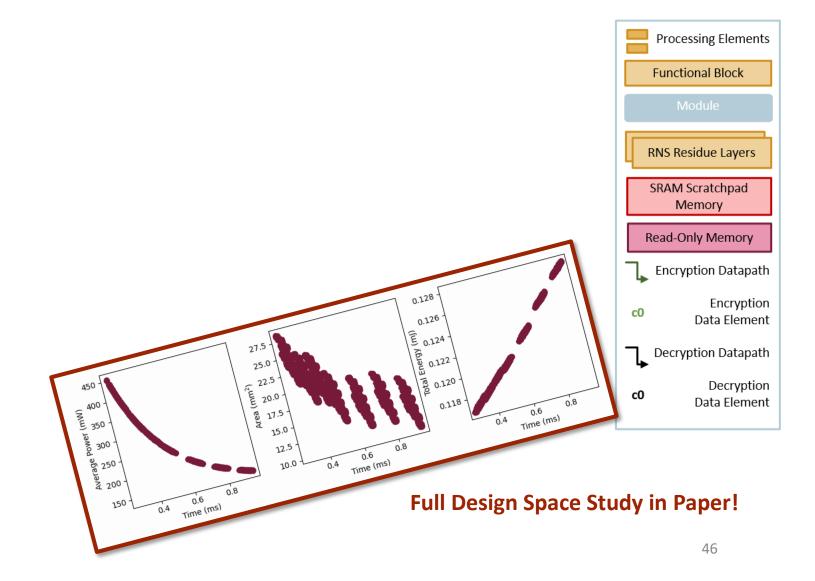


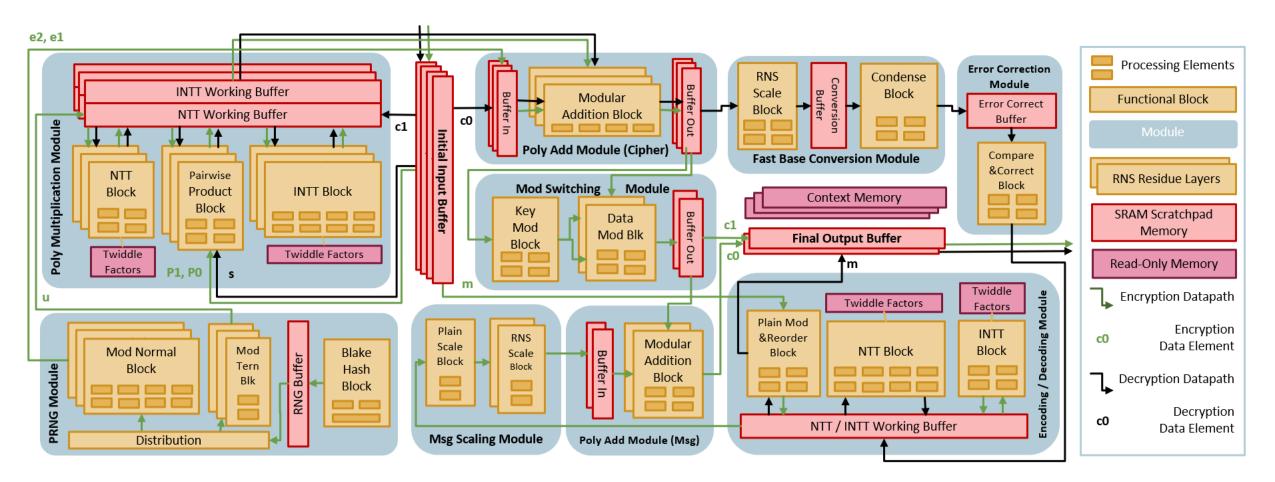




Electrical & Computer







19.3 mm² area. Consumes 200 mW power, .1228 mJ to perform a single encryption in .66 ms.



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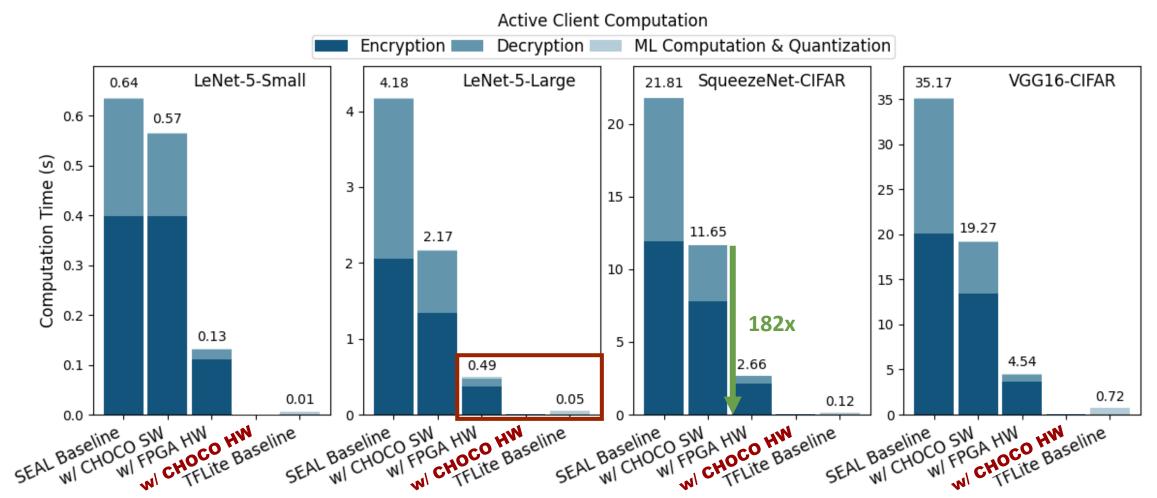
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CHOCO-TACO Accelerates Client Compute

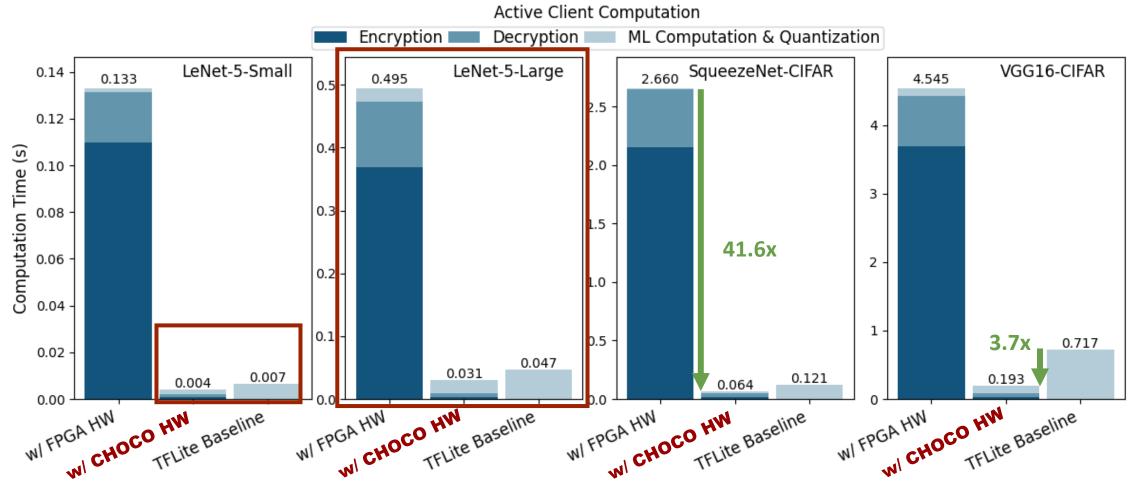


• **CHOCO HW** = CHOCO SW + CHOCO-TACO Encryption/Decryption Simulated ASIC



• Average 123.3x Improvement over CHOCO software alone

CHOCO-TACO Accelerates Client Compute



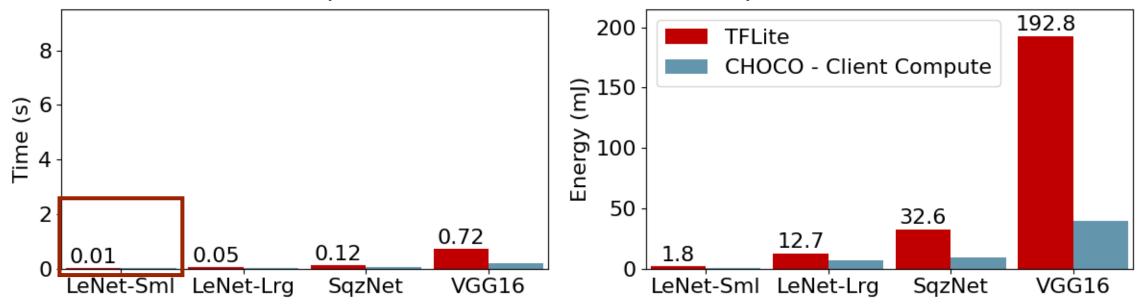
Average 29x better than FPGA accelerators



• Average 2.2x better than local compute via

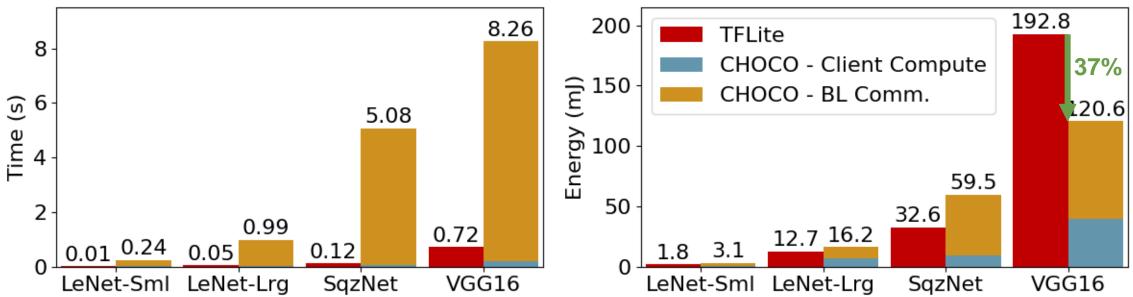
CHOCO Makes Client End-to-End Costs Feasible

Local Compute w/ TFLite vs Offloaded Compute w/ CHOCO





CHOCO Makes End-to-End Client Costs Feasible

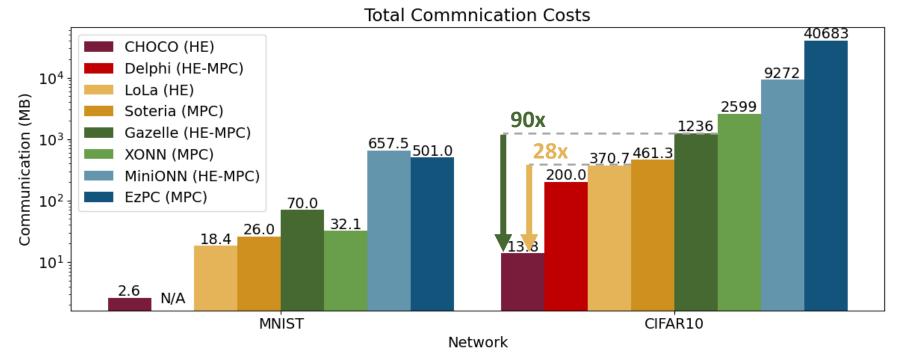


Local Compute w/ TFLite vs Offloaded Compute w/ CHOCO

- Privacy-Preserving Offload can be **Competitive with Local Compute**
- 37% decrease in energy consumption for VGG16
- Up to 66% communication reduction from SEAL baseline



CHOCO Algorithms Reduce Communication



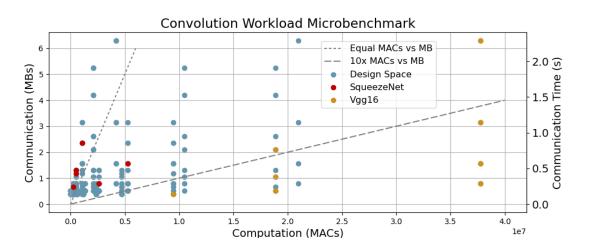
- Up to three orders of magnitude improvement in communication
- Nearly 90x improvement over Gazelle [Juvekar `18]
- 28x better than LoLa (not client-aided) [Brutzkus `19]

- C. Juvekar, V. Vaikuntanathan, and A. Chandrakasan. 2018. *GAZELLE: A Low Latency Framework for Secure Neural Network Inference*. In Proceedings of the 27th USENIX Conference on Security Symposium (Baltimore, MD, USA) (SEC'18). USENIX Association, USA, 1651–1668.

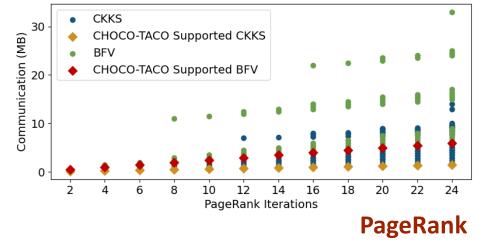
- Alon Brutzkus, Ran Gilad-Bachrach, and Oren Elisha. 2019. Low Latency Privacy Preserving Inference. In Proceedings of the 36th International Conference on Machine Learning (Proceedings of Machine Learning Research, Vol. 97), Kamalika Chaudhuri and Ruslan Salakhutdinov (Eds.). PMLR, Long Beach, California, USA, 812–821.



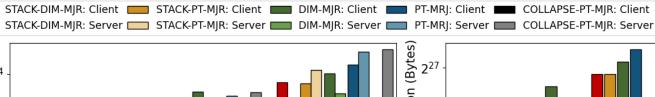
See Paper for More Applications & Results



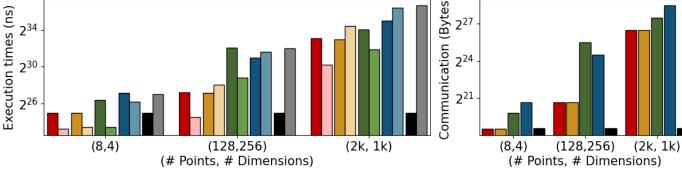
Communication Requirement for Varying PR Iterations



Unmodified DNN Networks



Performance of Encrypted Distance Calculation Variations





Encrypted Distance Calculations (K-Means & KNN)

Conclusions

- CHOCO motivates and prioritizes **client-aware optimizations**
- CHOCO alogorithm optimizations reduce communication by orders of magnitude over prior work
- CHOCO-TACO hardware comprehensively accelerates client-side cryptographic primitives
- CHOCO enables participation from resource-constrained devices in client-aided encrypted computation
- CHOCO makes client responsibility competitive with local compute
- CHOCO benefits generalize to diverse applications



Client-Optimized Algorithms & Acceleration for Encrypted Compute Offloading

Thank You! Questions?

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