

# Homomorphic Encryption Implementation for Small SWaP Platforms

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# PROJECT INTRODUCTION & BACKGROUND

The background of the slide features abstract, flowing blue shapes that create a sense of movement and depth. These shapes are layered, with some appearing more prominent than others, set against a light blue gradient background.

# What is homomorphic encryption?

- Homomorphic encryption:
  - An encryption scheme that allows the processing of data in its encrypted form without access to the secret key
  - Preserves the structure of the underlying data
  - Some developed using public asymmetric key systems, such as
    - RSA
    - ElGamal
    - Paillier
- Encryption schemes can be fully homomorphic or partially homomorphic, with respect to one type of data operation (i.e. multiplication, addition)

# Types of Homomorphic Encryption

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Year	Acronym	Name	Description
2011	BGV	Brakerski-Gentry-Vaikuntanathan	RLWE lattice-based FHE
2011	BFV	Brakerski-Fan-Vercauteren	RLWE lattice-based FHE
2016	CKKS	Cheon-Kim-Kim-Song	Supports approx. arithmetics over complex numbers. Exploits ring isomorphism.

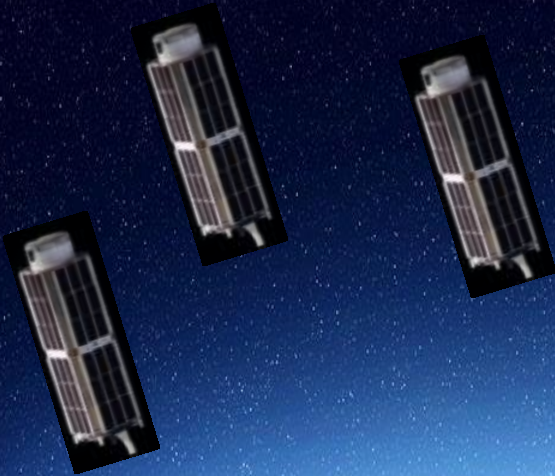
There are many homomorphic encryption schemes, including Enhanced Homomorphic Cryptosystem (EHC), Algebra Homomorphic Encryption (AHEE), and Non-interactive Exponential Homomorphic Encryption Scheme (NEHE); however, we focused most on BFV and CKKS, as they had the most materials available to researchers for free.



# Project Motivation: Space Applications, IoT Devices

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- Secure computation for space applications:
  - Cubesats for proprietary research missions with low-SWaP processors can transmit their encrypted data, and securely have data processed and transmitted back, using homomorphic encryption
  - Client-server model



# Project Objectives

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1. Analyze and implement a series of homomorphic encryption functions
  - Small SWaP Platform:
    - small SoCFPGA
    - FPGA-only
    - ARM processor only
2. Benchmark homomorphic encryption algorithms on cloud computing service
  - Amazon AWS EC2 cloud computing service
3. Implement client-server model
  - Modification of SEAL libraries to split computing requirements for implementation
    - Uses a combination of small SWaP device and a cloud computing service

# Homomorphic Encryption Library: Microsoft SEAL (MIT)

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- Simple Encrypted Arithmetic Library (SEAL):
  - Open-source software library developed by Microsoft
  - Implements various forms of homomorphic encryption
  - Standard C/C++ (no external dependencies)
  - Supports both asymmetric and symmetric encryption algorithms
  - Supports both BFV and CKKS encryption schemes



# Technical Approach

- Implement a low-SWaP client-server model for expanded use of homomorphic encryption schemes
  - Some existing experiments for IoT devices and high-performance computing platforms
    - Limited implementations achievable with Raspberry Pi in literature
  - No current literature on low-SWaP space applications
    - Desire for better encryption for cubesat and small sats, resistance to quantum computing for forced decryption
  - No library of functions for FPGA implementation commercially available
    - Some in-work at IBM, pay for access



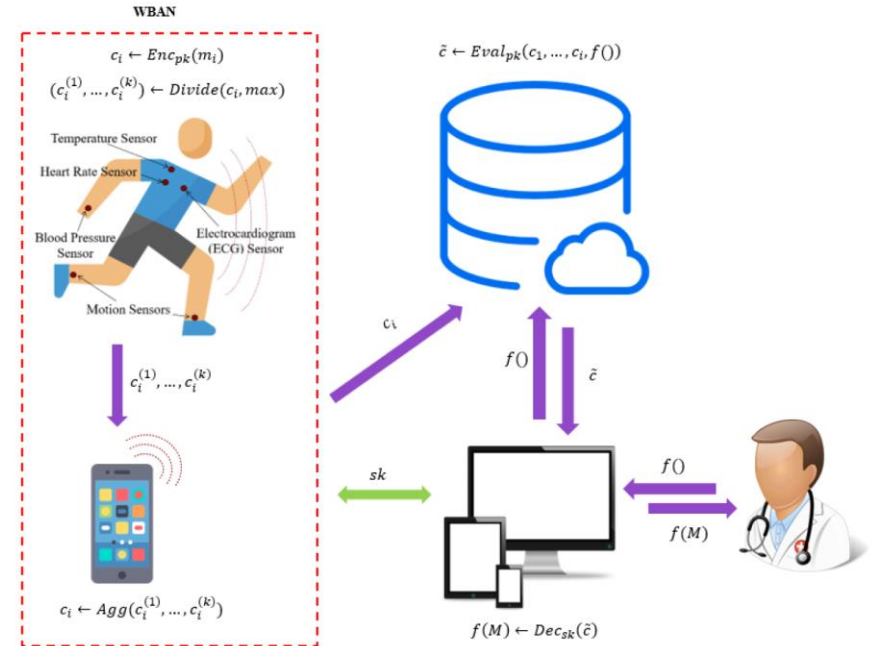
Two CubeSats, part of a constellation built and operated by Planet Labs Inc. to take images of Earth, were launched from the International Space Station on May 17, 2016.  
(Image: © NASA)

- Relevant current solicitation for research FY2020 – FY2023:
  - AFRL BAA *Capabilities for Cyber Resiliency*
  - AFRL BAA *Foundations of Trusted Computational Information Systems*
  - NAVAIR BAA *Cyber Warfare Detachment*
  - AFRL BAA *Next Generation Intelligence Collection and Analysis*
  - AFRL BAA *Measurement and Signatures Intelligence Exploitation*



# Related Work

- Simulated WBAN through OMNET++ and Castalia
  - 30s interval for reading and writing encrypted data
  - Simulation time 3600 s
  - Simulation with Helib and SEAL
    - Measured packet delay



**Table 5.** The running time (Milliseconds)

in healthcare systems.

		KeyGen	Encryption	Decryption	Addition	Multiplication	Bootstrapping
PC	HElib	0.760058	0.032094	0.013368	0.000094	0.066178	85.323830
	SEAL	1.930100	2.342723	0.177101	0.005329	2.187750	-
Raspberry Pi	HElib	79.933075	2.084733	1.258043	0.006370	4.707492	7,846.207000
	SEAL	181.319900	229.979548	46.673325	0.920642	480.622600	-

# OBJECTIVE 1

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# Objective 1:

## Small SWaP Platform Analysis

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Item No.	Manufacturer	Device Description	Platform Type
1	Intel	DE10 Nano SoC (Cyclone V)	SoC (ARM + FPGA)
2	Intel	DE0 Nano (Cyclone IV)	FPGA
3	Xilinx	Arty A7 (Zynq 7020)	SoC (ARM + FPGA)
4	Texas Instruments	MSP430 LaunchPad	Microcontroller
5	Raspberry Pi	Rpi 4	ARM Processor

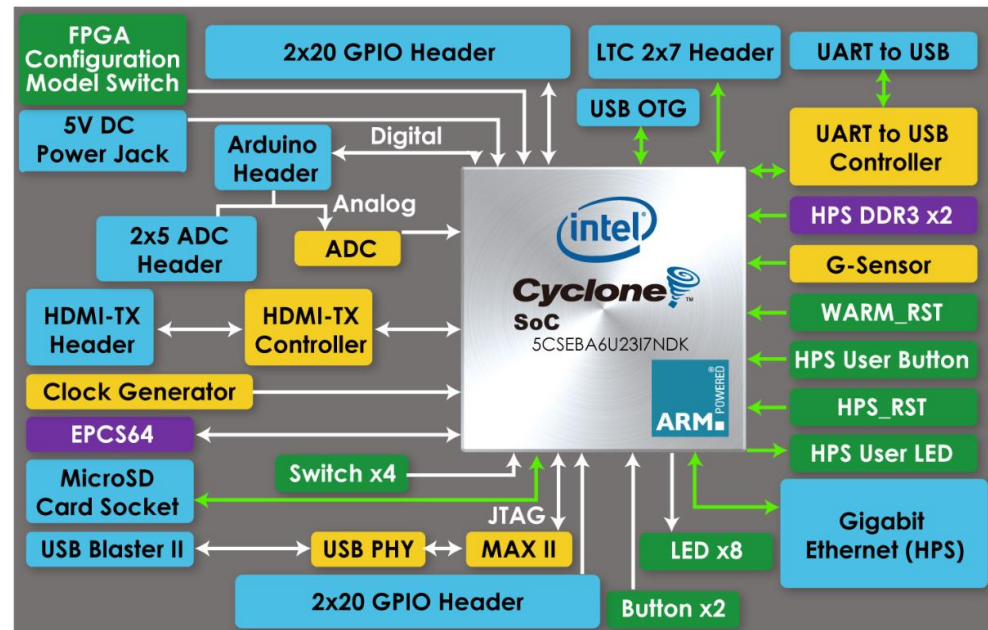
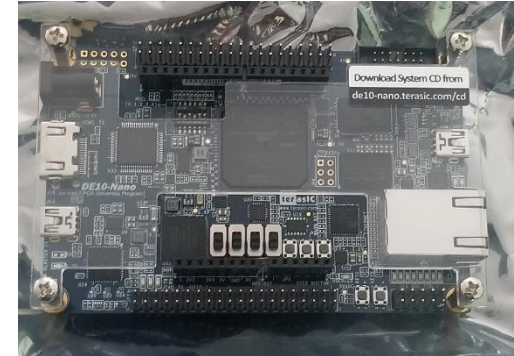
Criteria examined:

- Power consumption
- Processor type
- FPGA number of programmable logic elements
- Physical size
- Cost

# Objective 1: Small SWaP Platform Selection 1

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- Intel Cyclone V:  
DE10 Nano SoC
  - SoCFGPA
    - Dual-core Cortex A9  
32-bit ARM processor
    - 110k Programmable  
Logic Elements
- Linux OS:
  - Yocto Poky Distro
  - Boot from 32GB SD  
card



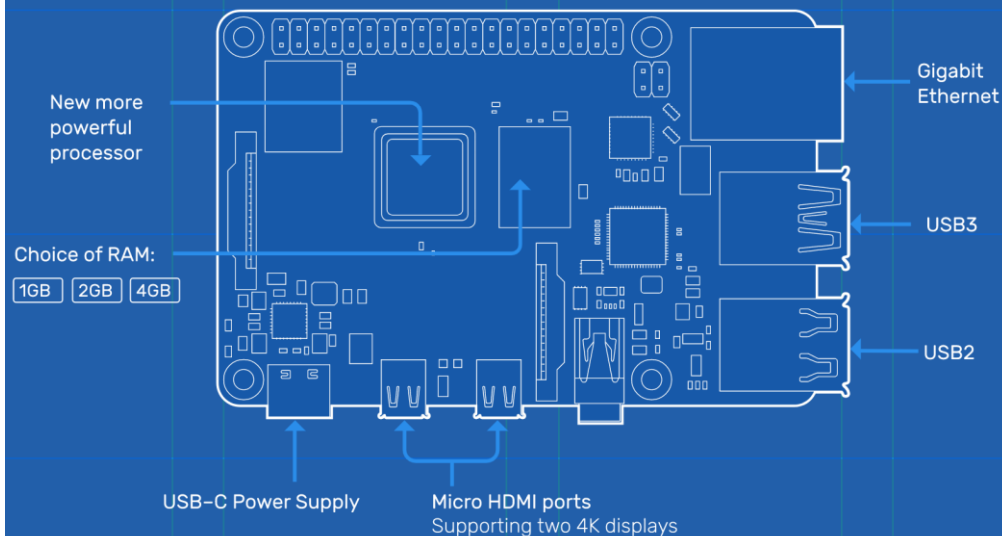
# Objective 1:

## Small SWaP Platform Selection 2

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- Raspberry Pi 4
  - Cortex A72 quad-core 64-bit ARM processor
  - 4GB DDR3 memory

### Raspberry Pi 4 Tech Specs





# Objective 1: Implementation

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- Using both DE10 Nano SoC and Raspberry Pi 4, we attempted to compile the SEAL library functions and run on the platform.
  - Using DE10 Nano SoC ARM processor
  - Porting C -> HDL on DE10 Nano SoC FPGA
  - Using Raspberry Pi 4 ARM processor

# Objective 1: Results

- 32-bit processor incompatible with SEAL and similar available homomorphic encryption libraries.
- Requires at least 64-bit processor.
- HDL  $\rightarrow$  C implementation too large for Cyclone V FPGA
- Success with Raspberry Pi 4 Cortex A78 ARM processor
  - All instances runnable for degrees  $< 32768$

# Objective 1: Results (2)

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#	Parameter	SEAL BFV Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.233	0.466	0.984	2.092	4.435
2	Average Unbatch	0.186	0.344	0.740	1.569	3.373
3	Average Encrypt	3.329	6.524	22.071	64.670	219.878
4	Average Decrypt	1.566	2.417	7.137	26.089	100.465
5	Average Add	0.025	0.036	0.116	0.498	2.047
6	Average Multiply	11.460	22.102	74.159	279.802	1,146.392
7	Average Multiply Plain	1.006	2.076	8.810	37.527	158.410
8	Average Square	8.252	16.237	55.181	209.815	867.853
9	Average Relinearize	N/A	N/A	16.476	72.648	396.322
10	Average Rotate 1 step	N/A	N/A	16.613	73.975	400.704
11	Average Rotate Random	N/A	N/A	53.823	341.561	1,765.455
12	Average Rotate Columns	N/A	N/A	16.624	73.992	400.718
13	CPU Usage [%]	5.0 %	10.6%	87.7%	99.7%	100%

# Objective 1: Results (3)

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#	Parameter	SEAL CKKS Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	1.801	2.969	7.514	20.257	61.416
2	Average Unbatch	2.594	4.250	14.735	57.825	257.827
3	Average Encrypt	3.016	4.455	21.417	66.505	232.468
4	Average Decrypt	0.148	0.215	0.842	3.325	13.085
5	Average Add	0.027	0.036	0.114	0.446	1.998
6	Average Multiply	0.427	0.729	2.951	12.627	48.889
7	Average Multiply Plain	0.186	0.316	1.243	4.921	19.506
8	Average Square	0.302	0.517	2.137	9.173	36.203
9	Average Relinearize	N/A	N/A	16.323	71.417	391.350
	Average Rescale	N/A	N/A	6.235	30.999	140.706
10	Average Rotate 1 step	N/A	N/A	16.734	73.160	399.765
11	Average Rotate Random	N/A	N/A	57.636	279.600	1,790.253
12	Average Complex Conjugate	N/A	N/A	16.684	73.036	398.881
13	CPU [%]	7.0%	11.6%	81.1%	98.3%	100%

## OBJECTIVE 2

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# Objective 2: AWS EC2 Selection

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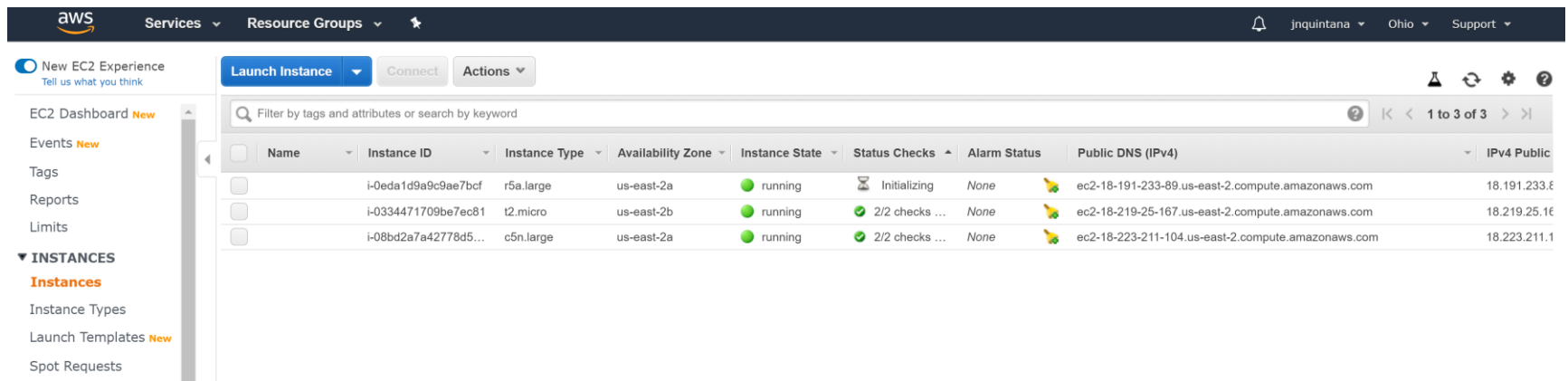
Item No.	Instance Name	Instance Type	CPU Cores	Memory
1	t1.micro	General Purpose	1	0.5 GB
2	c5n.large	Computation Optimized	2	4 GB
3	t2.micro	General Purpose	1	1 GB
4	r5a.large	Memory Optimized	2	16 GB

- Selected different instances based on processing properties:
  - Large, high-capability instances
  - Small, resource-limited instances

# Objective 2: Implementation (1)

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- Instantiated AWS EC2:
- Ubuntu 18.04 OS used for instantiations
- SSH to EC2 to run performance metrics for different functional implementations of SEAL benchmarking



The screenshot displays the AWS Management Console interface. The top navigation bar includes the AWS logo, 'Services', 'Resource Groups', and user information 'jnquintana' in 'Ohio'. The left sidebar shows navigation options like 'New EC2 Experience', 'Events', 'Tags', 'Reports', 'Limits', and 'INSTANCES'. The main content area features a 'Launch Instance' button and a table of existing EC2 instances. The table has columns for Name, Instance ID, Instance Type, Availability Zone, Instance State, Status Checks, Alarm Status, Public DNS (IPv4), and IPv4 Public. Three instances are listed, all in a 'running' state.

Name	Instance ID	Instance Type	Availability Zone	Instance State	Status Checks	Alarm Status	Public DNS (IPv4)	IPv4 Public
	i-0eda1d9a9c9ae7bcf	r5a.large	us-east-2a	running	Initializing	None	ec2-18-191-233-89.us-east-2.compute.amazonaws.com	18.191.233.89
	i-0334471709be7ec81	t2.micro	us-east-2b	running	2/2 checks ...	None	ec2-18-219-25-167.us-east-2.compute.amazonaws.com	18.219.25.167
	i-08bd2a7a42778d5...	c5n.large	us-east-2a	running	2/2 checks ...	None	ec2-18-223-211-104.us-east-2.compute.amazonaws.com	18.223.211.104

# Objective 2:

## Results for t2\_micro instance - BFV

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#	Parameter	SEAL BFV Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.024	0.050	0.129	0.273	0.539
2	Average Unbatch	0.026	0.048	0.105	0.214	0.453
3	Average Encrypt	0.537	1.034	3.034	8.262	26.371
4	Average Decrypt	0.118	0.234	0.764	2.671	11.488
5	Average Add	0.005	0.008	0.034	0.112	0.430
6	Average Multiply	1.103	2.180	7.349	27.757	116.834
7	Average Multiply Plain	0.113	0.234	0.931	4.047	19.624
8	Average Square	0.754	1.509	5.038	20.058	82.345
9	Average Relinearize	N/A	N/A	1.496	7.331	42.251
10	Average Rotate 1 step	N/A	N/A	1.549	7.354	43.365
11	Average Rotate Random	N/A	N/A	6.213	31.027	205.262
12	Average Rotate Columns	N/A	N/A	1.493	7.360	42.124
13	CPU Usage [%]	1%	2.3%	16.6%	51.8%	89.7%

# Objective 2:

## Results for t2\_micro instance - CKKS

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#	Parameter	SEAL CKKS Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.496	1.055	2.467	6.197	17.045
2	Average Unbatch	0.558	1.182	3.273	11.141	46.447
3	Average Encrypt	0.508	0.950	3.376	9.462	31.022
4	Average Decrypt	0.015	0.026	0.112	0.405	1.705
5	Average Add	0.006	0.009	0.030	0.105	0.418
6	Average Multiply	0.053	0.088	0.345	1.364	5.432
7	Average Multiply Plain	0.014	0.023	0.091	0.375	1.453
8	Average Square	0.035	0.064	0.251	1.015	4.191
9	Average Relinearize	N/A	N/A	1.530	7.811	43.702
	Average Rescale	N/A	N/A	0.733	3.516	15.965
10	Average Rotate 1 step	N/A	N/A	1.750	8.344	47.054
11	Average Rotate Random	N/A	N/A	7.205	31.284	227.630
12	Average Complex Conjugate	N/A	N/A	1.757	8.244	46.406
13	CPU [%]	1.0%	1.7%	15.0%	46.2%	100%

# Objective 2:

## Results for c5n\_large instance - BFV

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#	Parameter	SEAL BFV Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.025	0.054	0.105	0.213	0.425
2	Average Unbatch	0.023	0.039	0.083	0.168	0.352
3	Average Encrypt	0.421	0.786	2.274	6.227	19.964
4	Average Decrypt	0.082	0.165	0.541	1.920	7.527
5	Average Add	0.004	0.006	0.018	0.073	0.309
6	Average Multiply	0.747	1.571	5.255	20.001	81.804
7	Average Multiply Plain	0.081	0.169	0.712	2.999	12.707
8	Average Square	0.522	1.071	3.612	13.791	57.450
9	Average Relinearize	N/A	N/A	1.078	5.273	31.282
10	Average Rotate 1 step	N/A	N/A	1.083	5.310	31.056
11	Average Rotate Random	N/A	N/A	4.467	20.178	136.762
12	Average Rotate Columns	N/A	N/A	1.082	5.291	30.989
13	CPU Usage [%]	1.0%	1.7%	12.0%	36.7%	100%



# Objective 2:

## Results for c5n\_large instance - CKKS

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#	Parameter	SEAL CKKS Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.364	0.788	1.802	4.487	12.292
2	Average Unbatch	0.390	0.831	2.325	8.039	33.520
3	Average Encrypt	0.422	0.792	2.643	7.334	23.595
4	Average Decrypt	0.009	0.016	0.063	0.245	1.041
5	Average Add	0.004	0.006	0.017	0.068	0.284
6	Average Multiply	0.024	0.048	0.179	0.841	3.900
7	Average Multiply Plain	0.009	0.016	0.060	0.239	0.950
8	Average Square	0.016	0.032	0.120	0.552	2.763
9	Average Relinearize	N/A	N/A	1.097	5.357	31.338
	Average Rescale	N/A	N/A	0.519	2.574	11.756
10	Average Rotate 1 step	N/A	N/A	1.280	6.155	34.147
11	Average Rotate Random	N/A	N/A	4.458	28.515	145.624
12	Average Complex Conjugate	N/A	N/A	1.275	6.045	33.749
13	CPU [%]	0.7%	1.3%	11.0%	41.3%	100%

# Objective 2:

## Results for r5a\_large instance - BFV

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#	Parameter	SEAL BFV Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.056	0.056	0.124	0.273	0.513
2	Average Unbatch	0.022	0.046	0.091	0.196	0.451
3	Average Encrypt	0.482	0.902	2.604	7.491	24.381
4	Average Decrypt	0.102	0.186	0.632	2.374	9.425
5	Average Add	0.005	0.007	0.020	0.083	0.327
6	Average Multiply	1.002	1.986	6.581	25.472	103.115
7	Average Multiply Plain	0.102	0.206	0.930	3.827	16.475
8	Average Square	0.676	1.390	4.609	18.096	74.667
9	Average Relinearize	N/A	N/A	1.270	6.631	39.109
10	Average Rotate 1 step	N/A	N/A	1.322	6.585	39.271
11	Average Rotate Random	N/A	N/A	4.299	24.564	162.075
12	Average Rotate Columns	N/A	N/A	1.287	6.708	39.196
13	CPU Usage [%]	1.7%	2.3%	14.0%	63.1%	98%

# Objective 2:

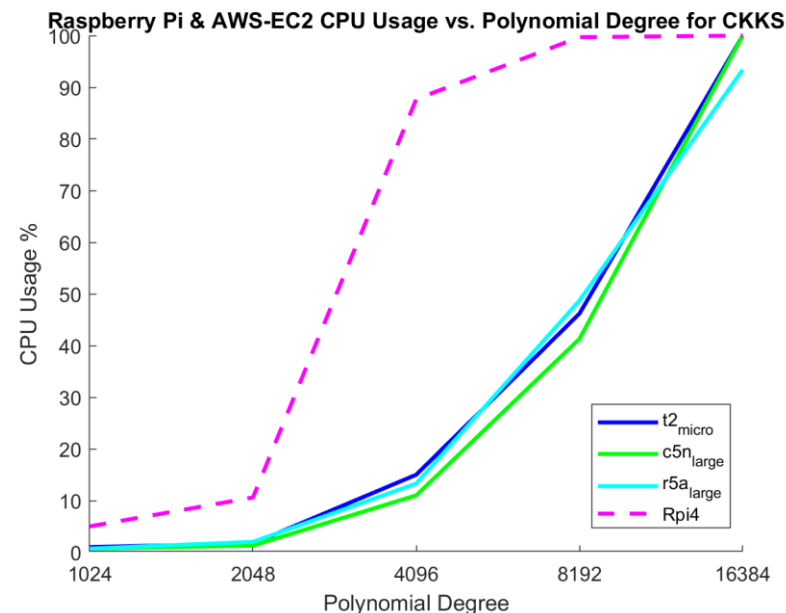
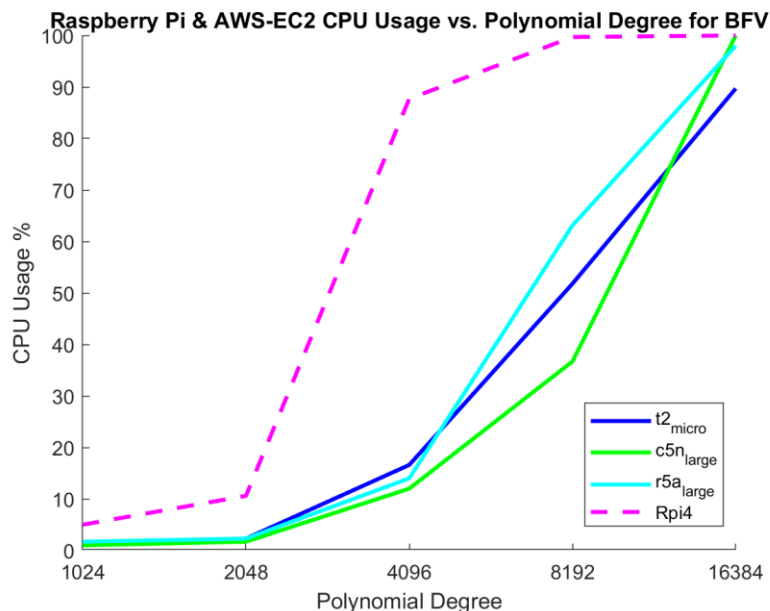
## Results for r5a\_large instance - CKKS

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#	Parameter	SEAL CKKS Degree Computation Time [ms]				
		1024	2048	4096	8192	16384
1	Average Batch	0.479	1.080	2.227	5.365	14.436
2	Average Unbatch	0.536	1.102	3250	10.403	44.064
3	Average Encrypt	0.481	0.855	3.145	8.735	28.172
4	Average Decrypt	0.012	0.022	0.091	0.302	1.229
5	Average Add	0.004	0.007	0.022	0.075	0.311
6	Average Multiply	0.033	0.066	0.268	0.984	4.578
7	Average Multiply Plain	0.013	0.024	0.100	0.351	1.500
8	Average Square	0.022	0.045	0.179	0.677	3.103
9	Average Relinearize	N/A	N/A	1.449	6.599	38.316
	Average Rescale	N/A	N/A	0.667	3.164	14.578
10	Average Rotate 1 step	N/A	N/A	1.623	7.275	41.515
11	Average Rotate Random	N/A	N/A	5.934	28.260	181.675
12	Average Complex Conjugate	N/A	N/A	1.579	7.215	41.180
13	CPU [%]	0.7%	2%	13.3%	48.7%	93.4%

# Objective 2 Results: CPU Usage

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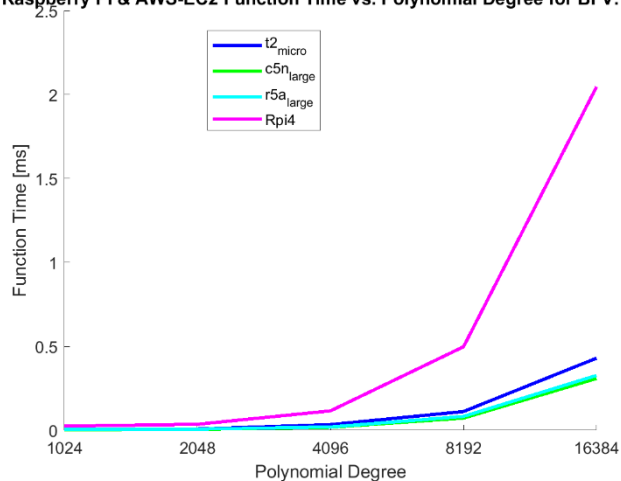
Raspberry Pi cannot complete 32k polynomial degree. AWS EC2 instances can complete without crashing but with long delay times (10s +) for each run.

CPU utilization approaches 100% for all current platforms, including computationally optimized and memory optimized AWS EC2 instances.

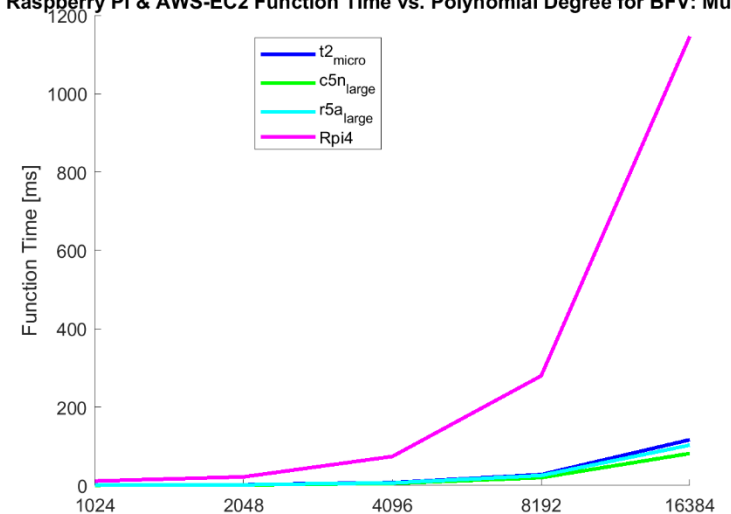
# Objective 2 Results: Function Computation Time

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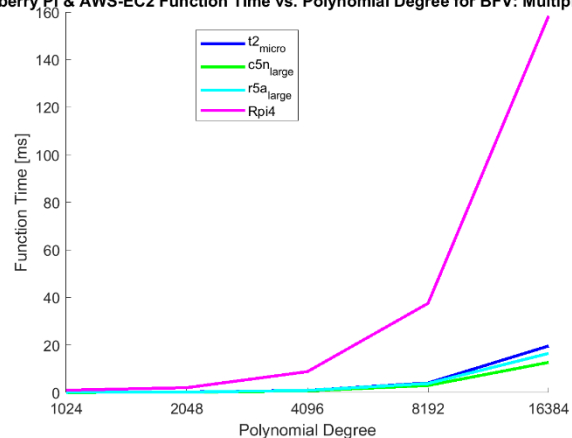
Raspberry Pi & AWS-EC2 Function Time vs. Polynomial Degree for BFV: Add



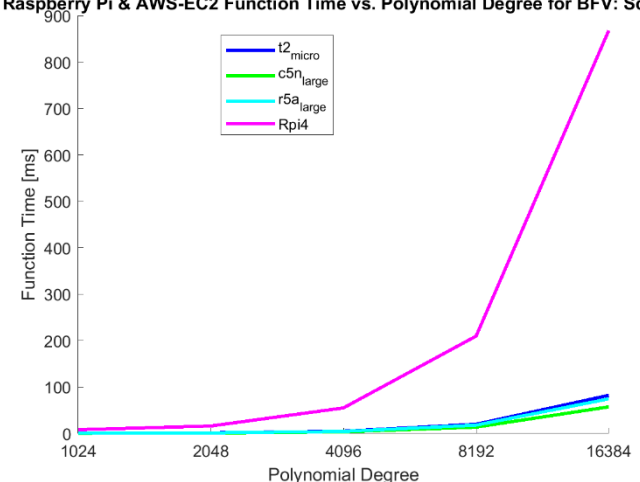
Raspberry Pi & AWS-EC2 Function Time vs. Polynomial Degree for BFV: Multip



Raspberry Pi & AWS-EC2 Function Time vs. Polynomial Degree for BFV: Multiply Pi



Raspberry Pi & AWS-EC2 Function Time vs. Polynomial Degree for BFV: Squar





## OBJECTIVE 3

The background of the slide features abstract, flowing blue shapes that create a sense of movement and depth. These shapes are layered, with some appearing more prominent than others, set against a light, off-white background.

# Objective 3: Client-Server Model

- Client-server model will move the bulk of the calculations from the resource-constrained Raspberry Pi to the AWS EC2 server and from micro AWS EC2 client and high performance AWS EC2 server.
  - Intended to lessen the computational cost on each device to allow a homomorphic solution for small SWaP devices

# Objective 3: Implementation

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- Client BFV with 4096 default degree test and server BFV with 4096 default degree test added to SEAL library functions

```
| The following examples should be executed while reading |
| comments in associated files in native/examples/.      |
+-----+-----+
| Examples          | Source Files          |
+-----+-----+
| 1. BFV Basics    | 1_bfv_basics.cpp      |
| 2. Encoders       | 2_encoders.cpp        |
| 3. Levels         | 3_levels.cpp          |
| 4. CKKS Basics   | 4_ckks_basics.cpp     |
| 5. Rotation       | 5_rotation.cpp        |
| 6. Performance Test | 6_performance.cpp     |
+-----+-----+
[ 785 MB] Total allocation from the memory pool

> Run example (1 ~ 6) or exit (0): 6

+-----+-----+
| Example: Performance Test |
+-----+-----+

Select a scheme (and optionally poly_modulus_degree):
 1. BFV with default degrees
 2. BFV with a custom degree
 3. CKKS with default degrees
 4. CKKS with a custom degree
 5. Client BFV with default degrees
 6. Server BFV with default degrees
 0. Back to main menu

> Run performance test (1 ~ 6) or go back (0):
```

# Objective 3: Client-Server Implementation

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```
+-----+
| Server BFV Performance Test with Degrees: 4096, 8192, 16384, and 32768 |
+-----+
/
| Encryption parameters :
|   scheme: BFV
|   poly_modulus_degree: 4096
|   coeff_modulus size: 109 (36 + 36 + 37) bits
|   plain_modulus: 786433
|
+-----+
Running tests ..... Done

Average add: 18157 microseconds
Average multiply: 61216 microseconds
Average multiply plain: 35296 microseconds
Average square: 42591 microseconds

/
| Encryption parameters :
|   scheme: BFV
|   poly_modulus_degree: 8192
|   coeff_modulus size: 218 (43 + 43 + 44 + 44 + 44) bits
|   plain_modulus: 786433
|
+-----+
Running tests ..... Done

Average add: 65073 microseconds
Average multiply: 228081 microseconds
Average multiply plain: 127401 microseconds
Average square: 162559 microseconds

/
| Encryption parameters :
|   scheme: BFV
|   poly_modulus_degree: 16384
|   coeff_modulus size: 438 (48 + 48 + 48 + 49 + 49 + 49 + 49 + 49 + 49) bits
|   plain_modulus: 786433
|
+-----+
Running tests ..... Done

Average add: 270092 microseconds
Average multiply: 961099 microseconds
Average multiply plain: 529304 microseconds
Average square: 683535 microseconds
```

```
+-----+
| Client BFV Performance Test with Degrees: 4096, 8192, 16384, and 32768 |
+-----+
/
| Encryption parameters :
|   scheme: BFV
|   poly_modulus_degree: 4096
|   coeff_modulus size: 109 (36 + 36 + 37) bits
|   plain_modulus: 786433
|
+-----+
Generating secret/public keys: Done
Running tests + Plaintext polynomial: 2
. + Plaintext polynomial: 2x^1 + 2
. + Plaintext polynomial: 4x^1 + 2
. + Plaintext polynomial: 2x^2 + 2x^1 + 2
. + Plaintext polynomial: 4x^2 + 2
. + Plaintext polynomial: 4x^2 + 2x^1 + 2
. + Plaintext polynomial: 4x^2 + 4x^1 + 2
. + Plaintext polynomial: 2x^3 + 2x^2 + 2x^1 + 2
. + Plaintext polynomial: 4x^3 + 2
. + Plaintext polynomial: 4x^3 + 2x^1 + 2
. Done

Average add: 15929 microseconds
Average multiply: 61121 microseconds
Average multiply plain: 34737 microseconds
Average square: 43107 microseconds

/
| Encryption parameters :
|   scheme: BFV
|   poly_modulus_degree: 8192
|   coeff_modulus size: 218 (43 + 43 + 44 + 44 + 44) bits
|   plain_modulus: 786433
|
+-----+
Generating secret/public keys: Done
Running tests + Plaintext polynomial: 2
. + Plaintext polynomial: 2x^1 + 2
. + Plaintext polynomial: 4x^1 + 2
. + Plaintext polynomial: 2x^2 + 2x^1 + 2
. + Plaintext polynomial: 4x^2 + 2
. + Plaintext polynomial: 4x^2 + 2x^1 + 2
. + Plaintext polynomial: 4x^2 + 4x^1 + 2
. + Plaintext polynomial: 2x^3 + 2x^2 + 2x^1 + 2
. + Plaintext polynomial: 4x^3 + 2
. + Plaintext polynomial: 4x^3 + 2x^1 + 2
. Done

Average add: 59338 microseconds
Average multiply: 228033 microseconds
Average multiply plain: 125155 microseconds
Average square: 164971 microseconds
```

# Objective 3: Client Server Results

## Table (c5n shown)

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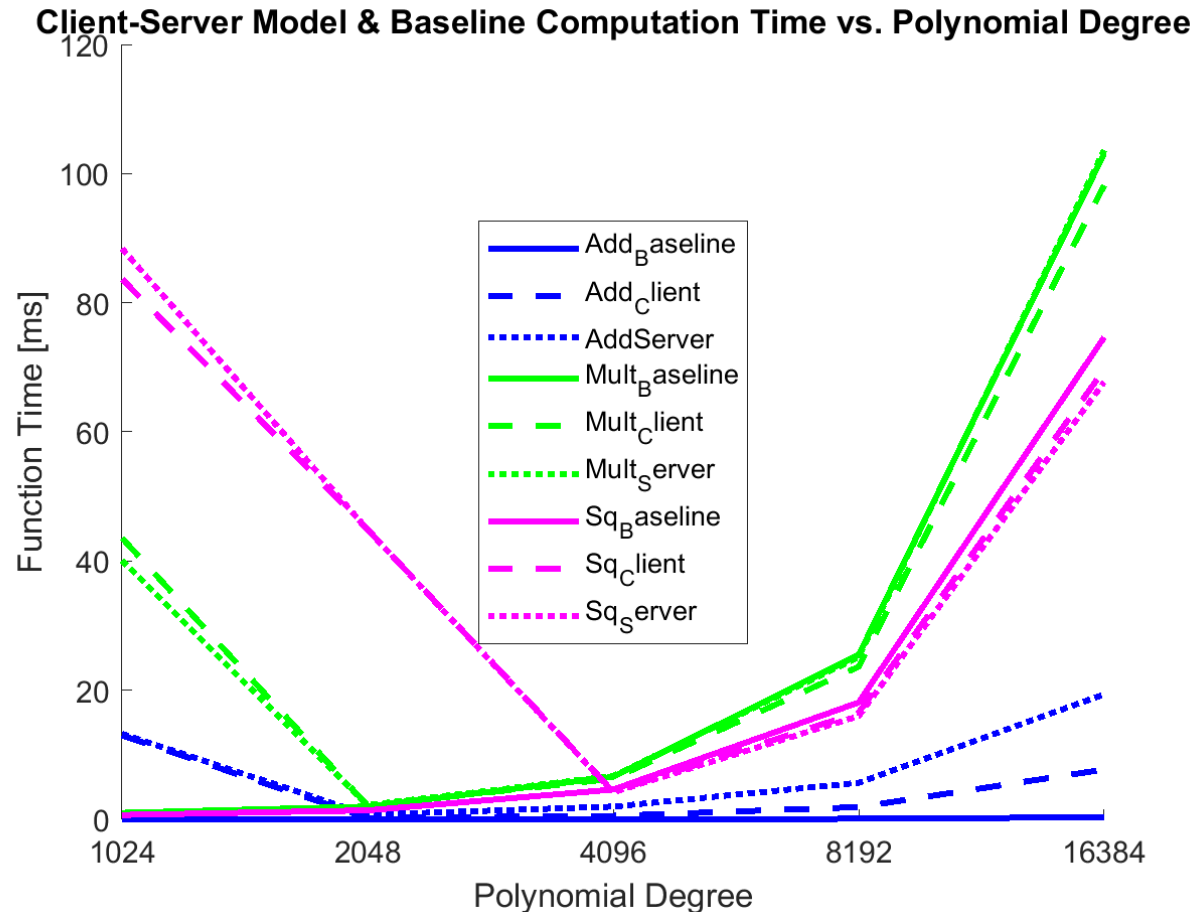
#	Parameter	SEAL BFV Degree (Client) Computation Time [ms]				
Degree		1024	2048	4096	8192	16384
5	Average Add	13.002	0.226	0.539	1.895	7.71
6	Average Multiply	43.559	2.106	6.294	23.639	98.195
7	Average Multiply Plain	87.966	42.817	18.324	5.619	23.462
8	Average Square	83.679	44.901	4.321	16.539	69.679
13	CPU Usage [%]	15.1%	20.4%	34.2%	38.7%	42.4%

#	Parameter	SEAL BFV Degree (Server) Computation Time [ms]				
Degree		1024	2048	4096	8192	16384
5	Average Add	13.245	0.712	1.971	5.667	19.364
6	Average Multiply	40.085	2.221	6.708	25.129	103.67
7	Average Multiply Plain	86.588	42.386	18.429	6.14	25.528
8	Average Square	88.393	44.865	4.19	15.996	67.72
13	CPU Usage [%]	27.3%	30.5%	45.7%	60.2%	72.4%

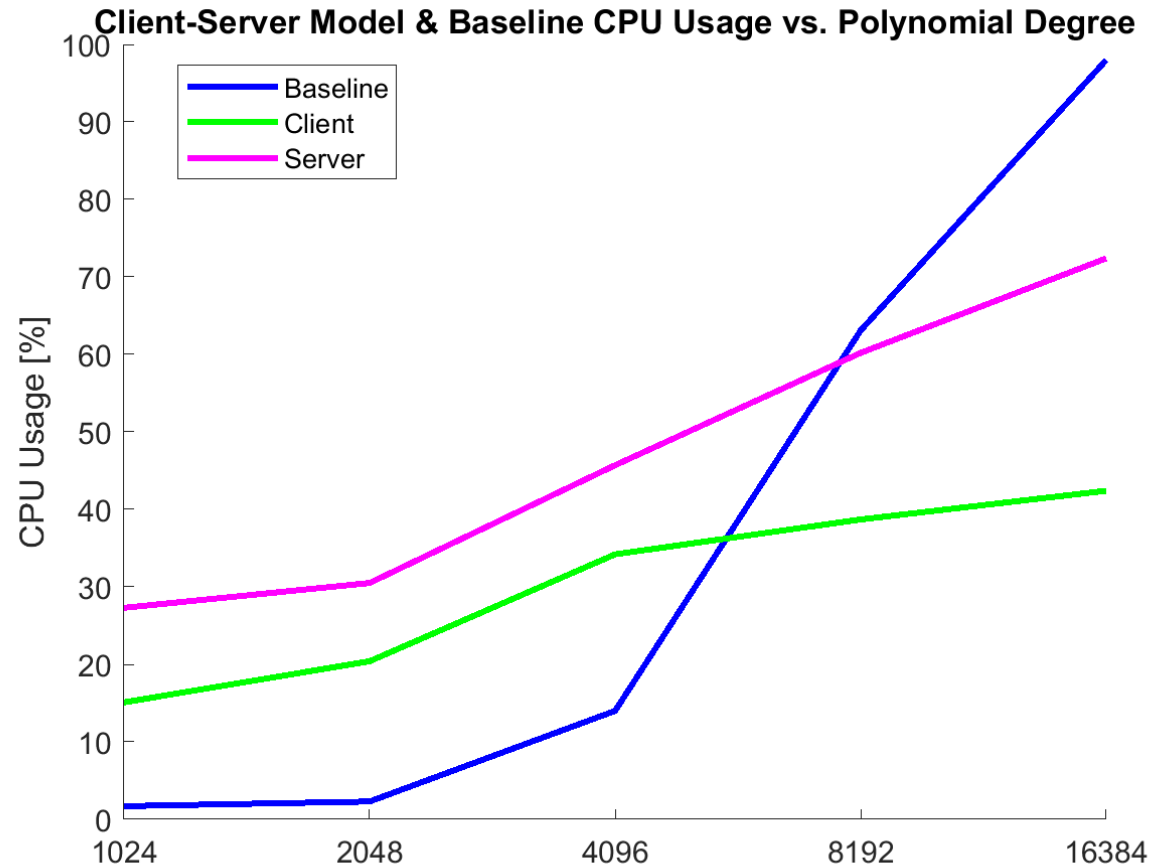
# Objective 3: Results – Function Time vs. Baseline

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# Objective 3: Results – CPU Usage

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# SUMMARY

The background of the slide features abstract, flowing blue shapes that create a sense of movement and depth. These shapes are layered, with some appearing more prominent than others, set against a light, off-white background.

# Future Work

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- Optimization of the SEAL library functions for to improve resource freeing and reallocation
- Improvement of Raspberry Pi platform resource allocation
- Implementation on IoT device app for in-situ encrypted data usage

# Summary

- Homomorphic encryption has made advancements in the past decade for implementation.
  - Understood from mathematical perspective
  - Can now optimize for resource-constrained environments
- Client-server model prototype can allow for future IoT device platforms to make use of homomorphic encryption and optimize for practical use.

# GitHub File Structure

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- README.txt
- 209AS\_FinalPresentation.pptx
- 209AS\_FinalReport.docx
- Client-Server Model <software folder>
- Console data logs <text file data folder>
- <https://github.com/daniel-achee/ece209-project.git>

# Division of Labor

- Presentation & Report:
  - Primary: J. Quintana
  - Supporting: D. Achee
- Objective 1:
  - Primary: J. Quintana
  - Supporting: D. Achee
- Objective 2:
  - Primary: D. Achee
  - Supporting: J. Quintana
  - Initial AWS-EC2 instance implementation done by D. Achee, data shown from J. Quintana AWS-EC2 instances.
- Objective 3:
  - Server side: D. Achee
  - Client side: J. Quintana
  - Client-server model done by D. Achee, Rpi platform side done by J. Quintana

# References

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