CIPHERMATCH

Accelerating Homomorphic Encryption-Based String Matching via Memory-Efficient Data Packing and In-Flash Processing

Mayank Kabra

Rakesh Nadig, Harshita Gupta, Rahul Bera, Manos Frouzakis, Vamanan Arulchelvan, Yu Liang, Haiyu Mao, Mohammad Sadrosadati, and Onur Mutlu

SAFARI

ETH zürich

Executive Summary

<u>Problem</u>: Secure exact string matching using homomorphic encryption (HE) lacks scalability due to performance bottlenecks in **two key areas**:

- a) Use of complex homomorphic multiplication resulting in high computation cost
- b) Data movement bottleneck from large encrypted database stored in solid-state drive (SSD)

<u>Goal:</u> Develop an algorithm-hardware co-design to provide scalable, parallelizable and efficient HE-based secure *exact* string-matching

Key Idea: Use (a) only homomorphic addition and (b) perform in-flash processing by exploiting the operational principles of NAND-flash memory to accelerate secure exact string matching

<u>CIPHERMATCH</u>: A new algorithm-hardware co-design

that significantly improves the performance of HE-based secure exact string matching by

- a) using only homomorphic addition to reduce the high computation cost
- b) optimizing the data packing scheme to reduce memory footprint
- c) designing a new in-flash-processing (IFP) architecture to reduce data movement

Key Results:

- a) CIPHERMATCH algorithm: 42.9x speedup & 39.4x energy savings than best software
- b) CIPHERMATCH with IFP: **136.9x speedup** & **256.4x energy savings** over CM-SW

Talk Outline

Background, Problem & Goal

Key Idea

CIPHERMATCH System: Overview

CIPHERMATCH: Algorithm

CIPHERMATCH: Hardware

Evaluation Results



Talk Outline

Background, Problem & Goal

Key Idea

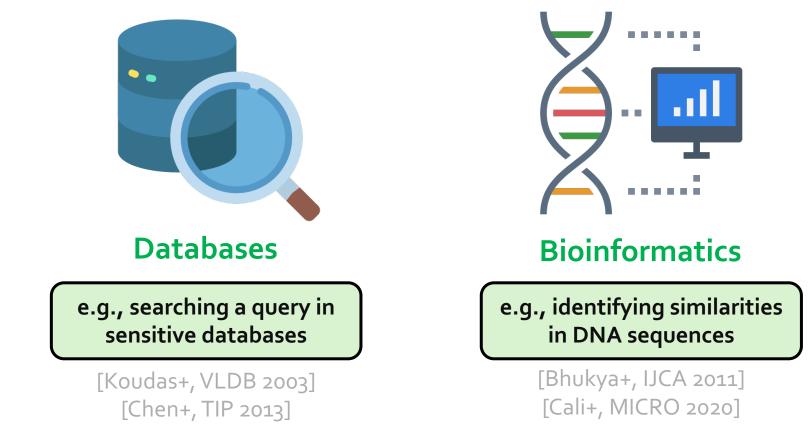
CIPHERMATCH System: Overview

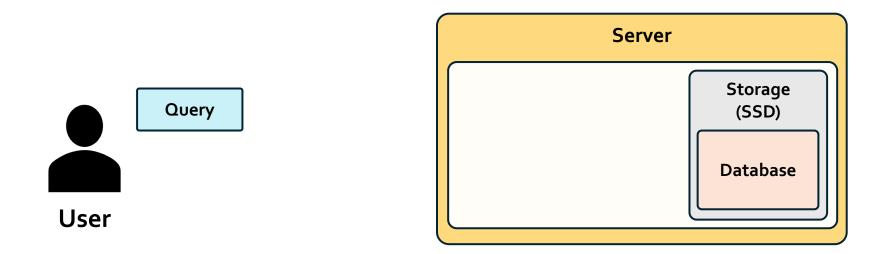
CIPHERMATCH: Algorithm

CIPHERMATCH: Hardware

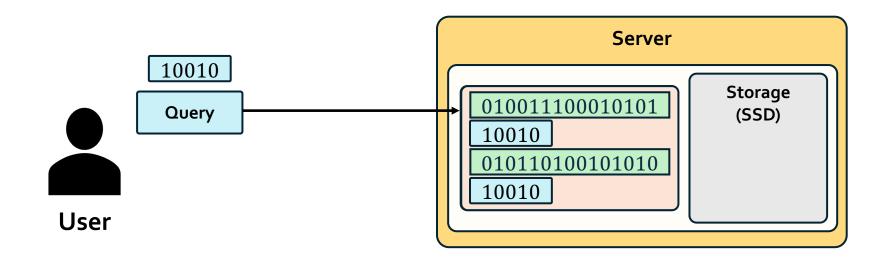
Evaluation Results

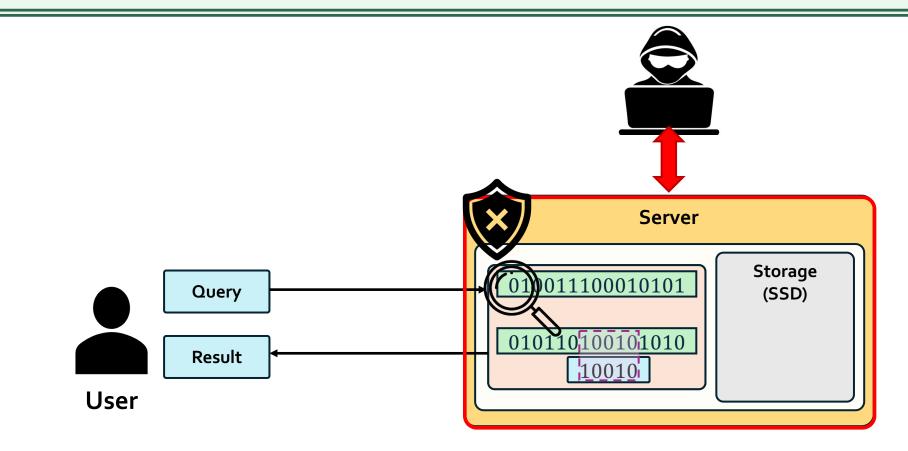
Exact string matching is used in many security critical applications, such as





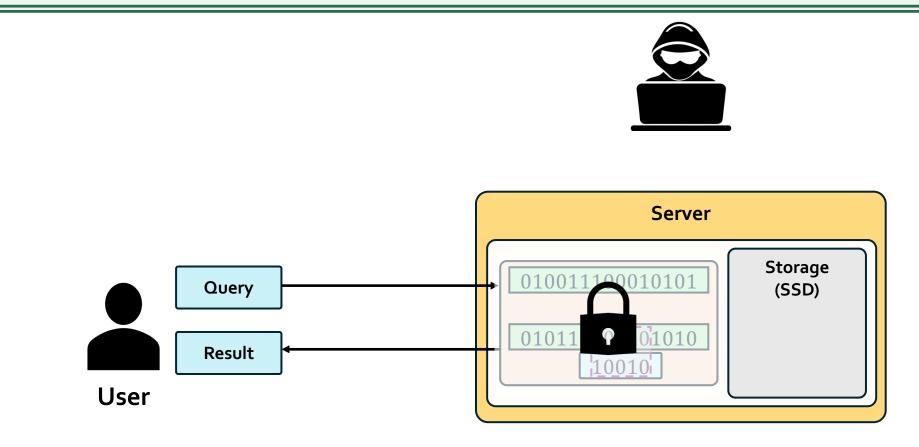






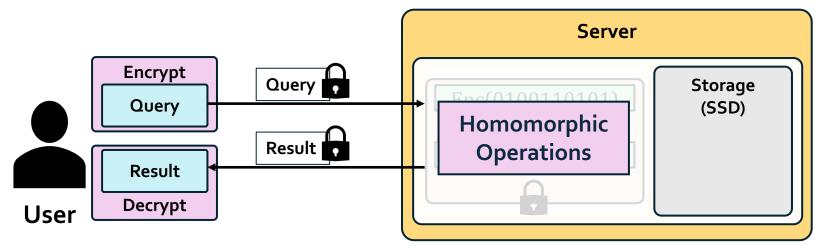
Performing computation on plaintext can lead to data leakage





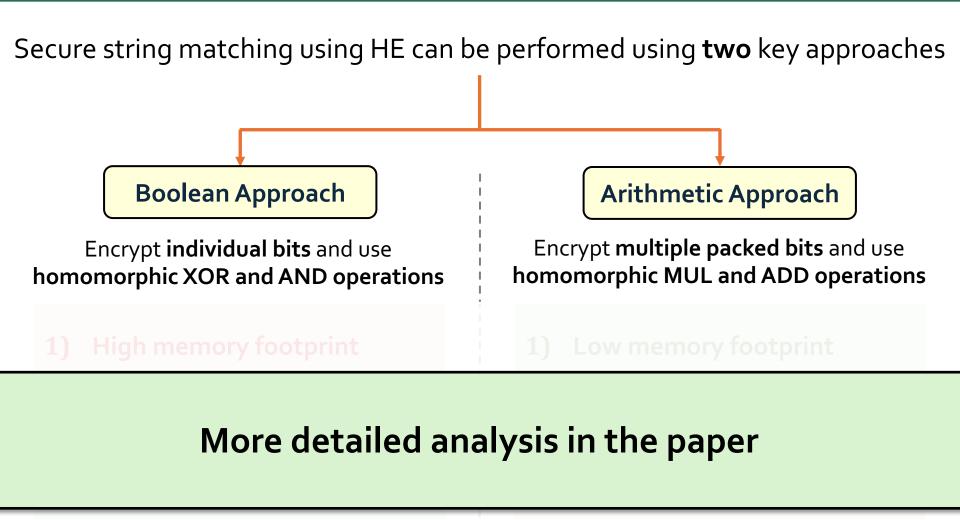
Homomorphic encryption (HE) can be leveraged to perform secure *exact* string matching

Secure Exact String Matching



Allows users to compute on encrypted data without decrypting it

Approaches to HE-based String Matching



3) Supports flexible query size

Supports limited query size



Prior Works on HE-based String Matching

Arithmetic Approach [Yasuda+, CCSW 2013; Kim+, TDSC 2017; Bonte+, CCS 2020]

Boolean Approach [Pradel+, TrustCom 2021 ; Aziz+, Information 2024]



Approaches to HE-based String Matching

Secure string matching using HE can be performed using **two** key approaches

Boolean Approach

Encrypt **individual bits** and use **homomorphic XOR and AND operations**

1) High memory footprint

2) High computation cost

3) Supports flexible query size

Arithmetic Approach

Encrypt multiple packed bits and use homomorphic MUL and ADD operations

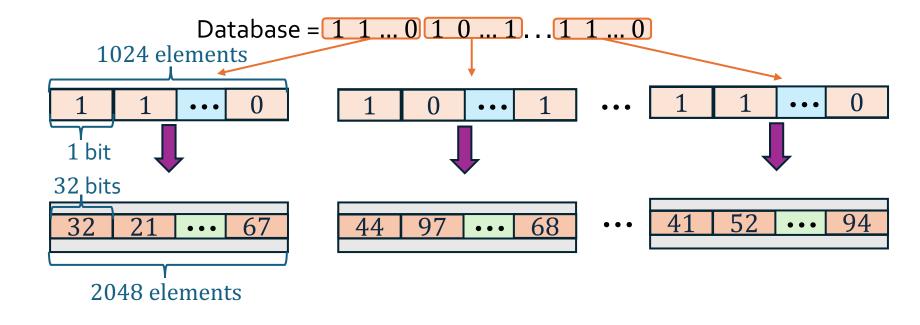
1) Low memory footprint

2) Low computation cost

3) Supports limited query size

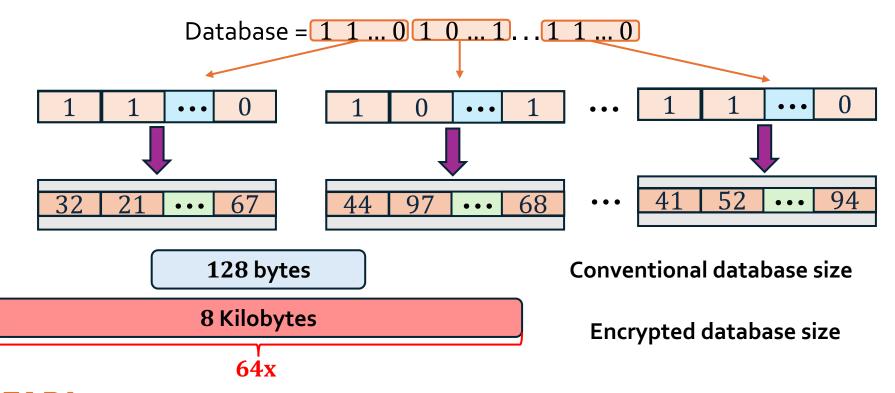
Arithmetic Approach

1. Encrypt multiple packed bits



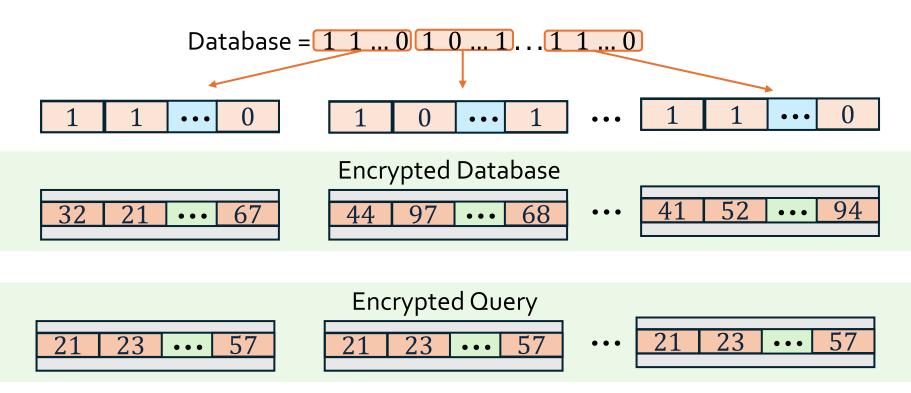
Arithmetic Approach

1. Encrypt multiple packed bits



Arithmetic Approach

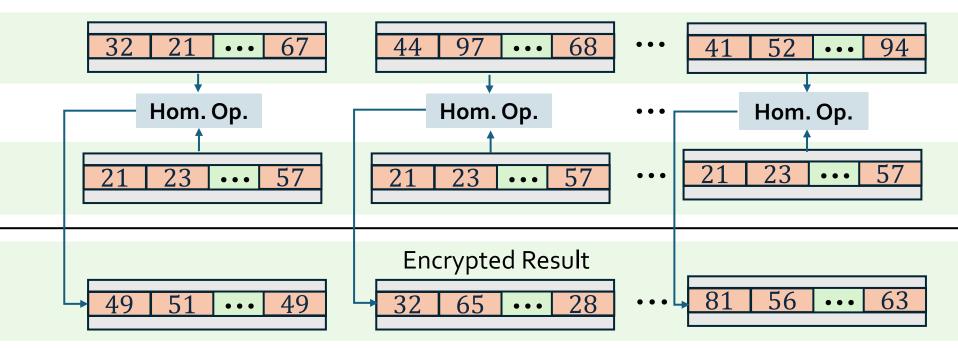
- 1. Encrypt multiple packed bits
- 2. Perform homomorphic MUL and ADD operations



SAFARI Yasuda et al. "Secure Pattern Matching Using Somewhat Homomorphic Encryption," in CCSW, 2013 16

Arithmetic Approach

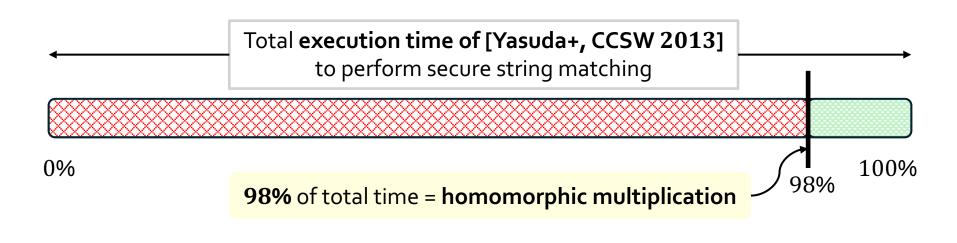
- 1. Encrypt multiple packed bits
- 2. Perform homomorphic MUL and ADD operations



SAFARI Yasuda et al. "Secure Pattern Matching Using Somewhat Homomorphic Encryption," in CCSW, 2013 17

Execution Time of Arithmetic Approach

Arithmetic Approach [Yasuda+, CCSW 2013; Kim+, TDSC 2017; Bonte+, CCS 2020]



Homomorphic multiplication is **100x slower** than homomorphic addition on a CPU-system

Key Problem (I): Homomorphic multiplication

Arithmetic Approach [Yasuda+, CCSW 2013 ; Kim+, TDSC 2017 ; Bonte+, CCS 2020]

Homomorphic multiplication limits scalability of HE-based string matching algorithm

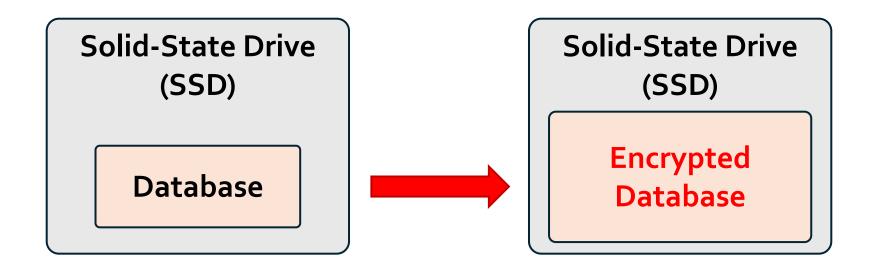
Homomorphic multiplication is 100x slower than homomorphic addition



Databases are Stored in Storage (SSD)

Databases are large and stored in SSDs

Homomorphic encryption **further increases** the database size





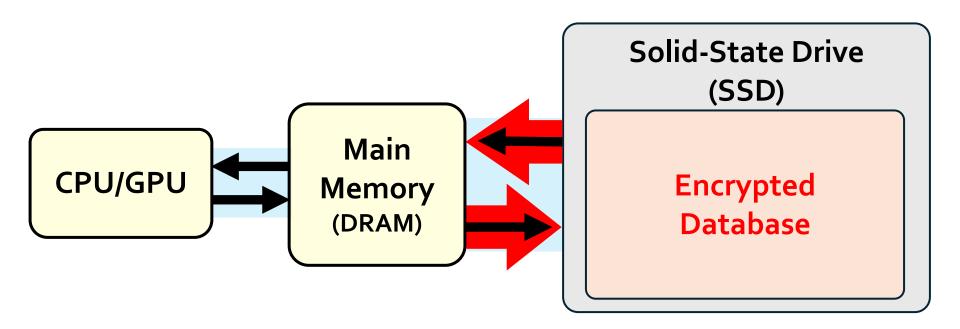
Key Problem (II): Data Movement Bottleneck

Databases are large and stored in SSDs

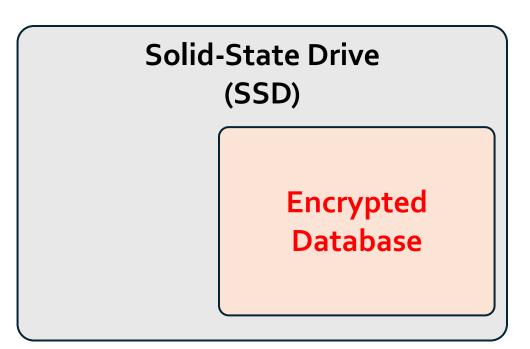
External I/O bandwidth of SSD

is the *main bottleneck* for reading large encrypted database

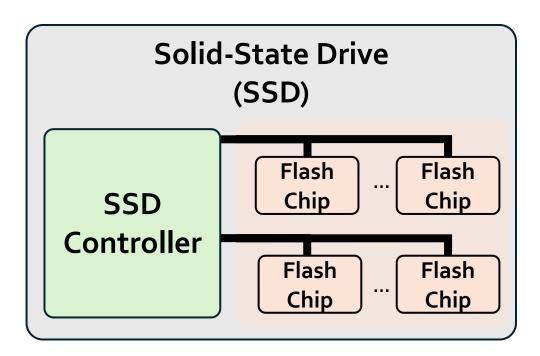
further increases the database size



Prior Works on Reducing Data Movement

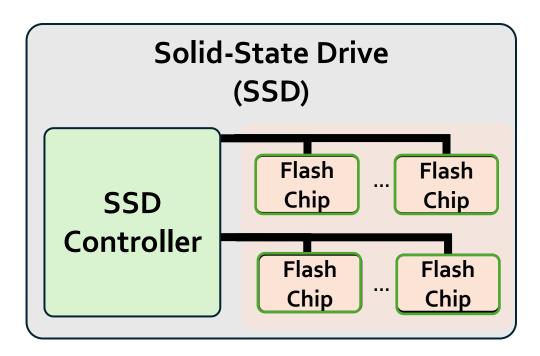


Prior Works on Reducing Data Movement



Prior Works on Reducing Data Movement

In-Flash Processing (IFP) [Park+, MICRO 2022; Gao+, MICRO 2021] enables computation inside SSD by exploiting the operational principles of NAND-flash memory



Develop an IFP-based algorithm-hardware co-designed system that can perform scalable, parallelizable and efficient secure *exact* string matching



Talk Outline

Background, Problem & Goal

Key Idea

CIPHERMATCH: System Overview

CIPHERMATCH: Algorithm

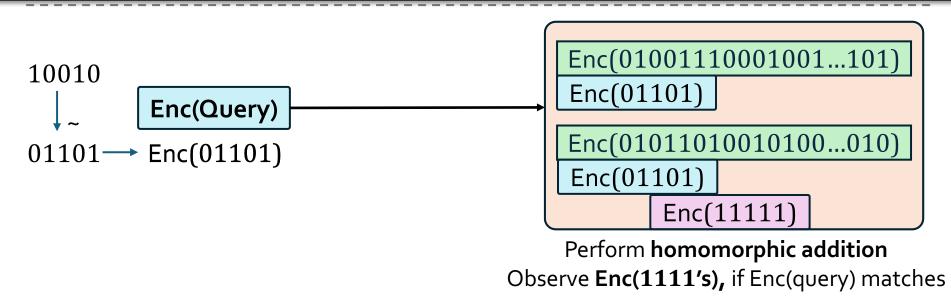
CIPHERMATCH: Hardware

Evaluation Results

Key Observation

In a conventional database, we perform *only* addition to get a string match

This observation can be extended to perform secure *exact* string matching using *only* homomorphic addition



Use only homomorphic addition to perform secure exact string matching



Homomorphic addition is highly parallelizable

Exploit inherent parallelism of NAND-flash memory

- Improves the **performance** of secure string matching
 - Reduces data movement



Use only homomorphic addition to perform secure exact string matching

Use in-flash processing (IFP) to reduce data movement and accelerate secure *exact* string matching



CIPHERMATCH

An algorithm-hardware co-design

Improves the performance of HE-based **secure** *exact* **string matching**

Reduces memory footprint

- by optimizing the data packing scheme used before encryption

Eliminates costly homomorphic multiplication

- by designing secure string-matching algorithm using only homomorphic addition

Reduces data movement

and leverages massive bit and array-level parallelism

- by designing an in-flash processing architecture

Talk Outline

Background, Problem & Goal

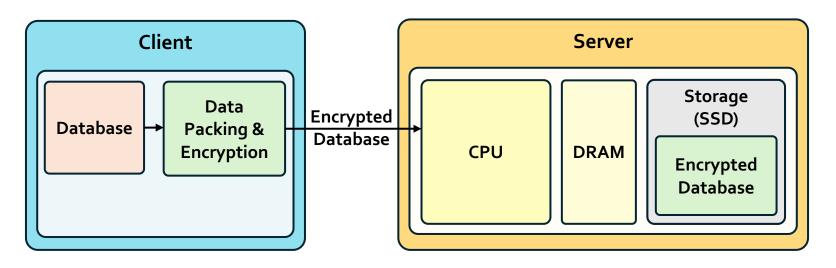
Key Idea

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CIPHERMATCH: Algorithm

CIPHERMATCH: Hardware

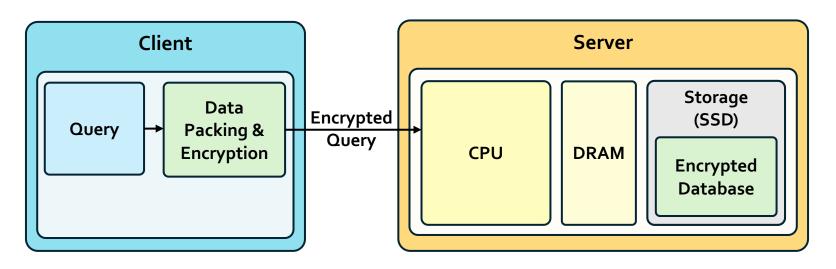
Evaluation Results



Efficiently pack the database

to reduce the memory footprint after encryption





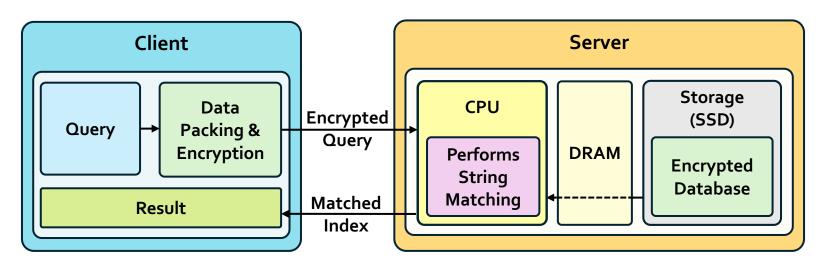
Efficiently pack the database

to reduce the memory footprint after encryption

Efficiently pack the query

to perform parallel secure string matching on encrypted database





Efficiently pack the database

to reduce the memory footprint after encryption

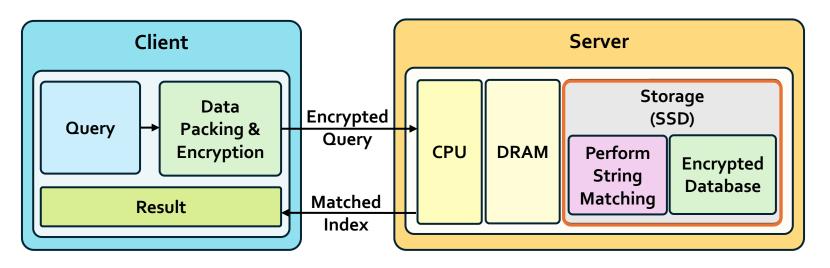
Efficiently pack the query

to perform parallel secure string matching on encrypted database

Perform secure exact string matching

using only homomorphic addition





Efficiently pack the database

to reduce the memory footprint after encryption

Efficiently pack the query

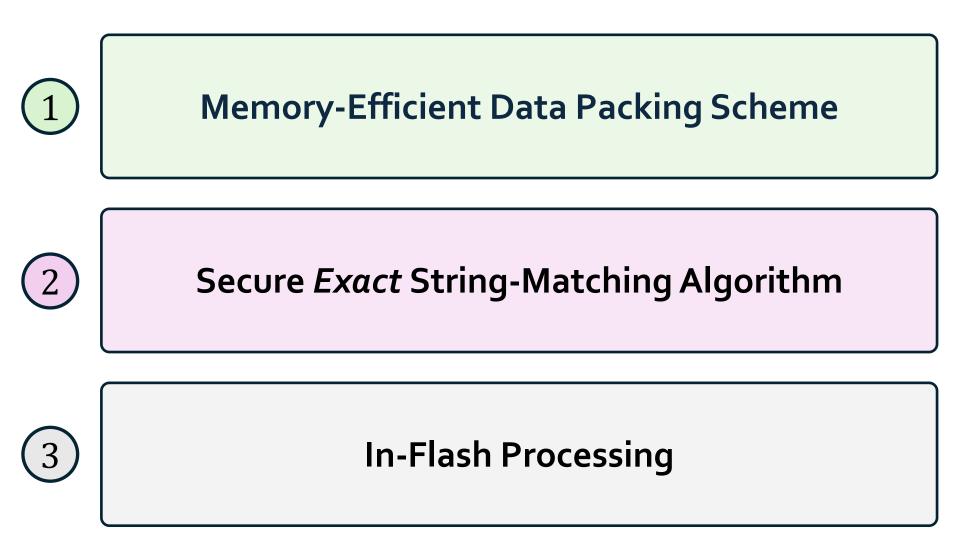
to perform parallel secure string matching on encrypted database

Perform secure exact string matching

using only homomorphic addition

Accelerate secure exact string matching by performing computations inside SSD by exploiting operational principles of NAND-flash memory

CIPHERMATCH: Key Steps



Talk Outline

Background, Problem & Goal

Key Idea

CIPHERMATCH: System Overview

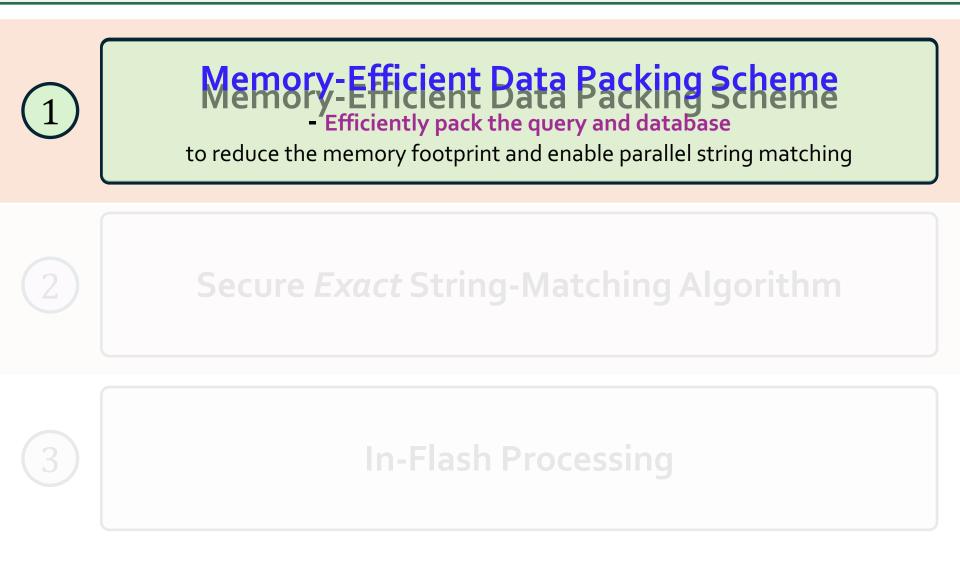
CIPHERMATCH: Algorithm

CIPHERMATCH: Hardware

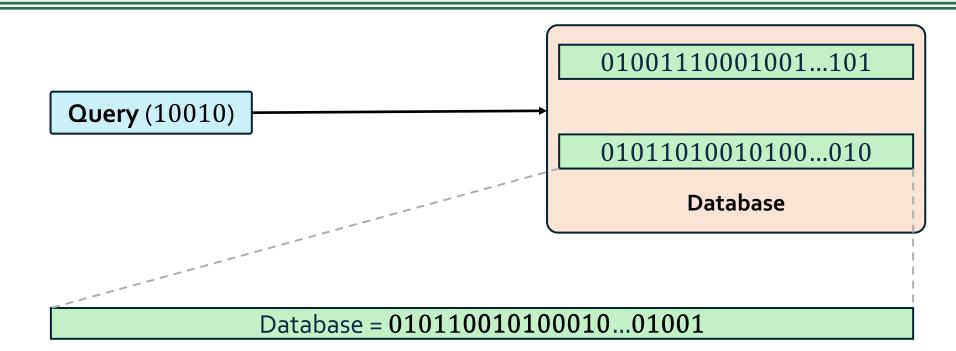
Evaluation Results



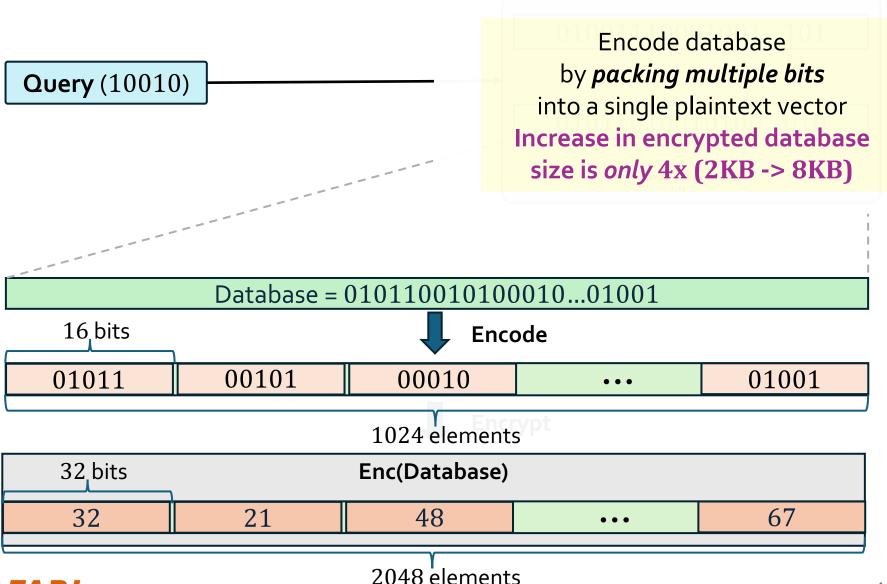
CIPHERMATCH: Key Steps

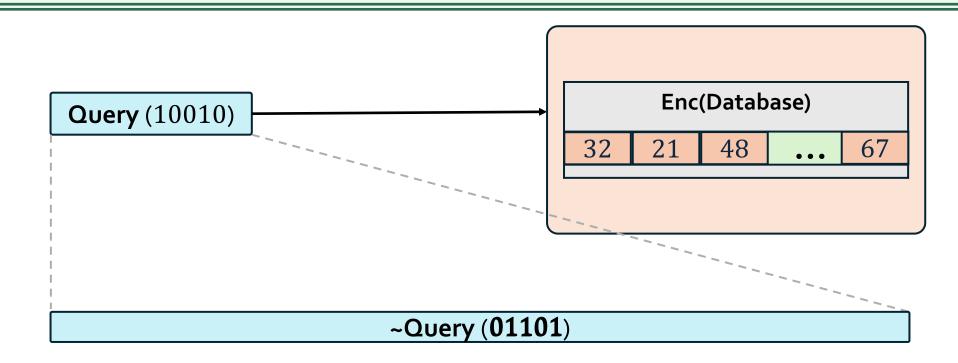






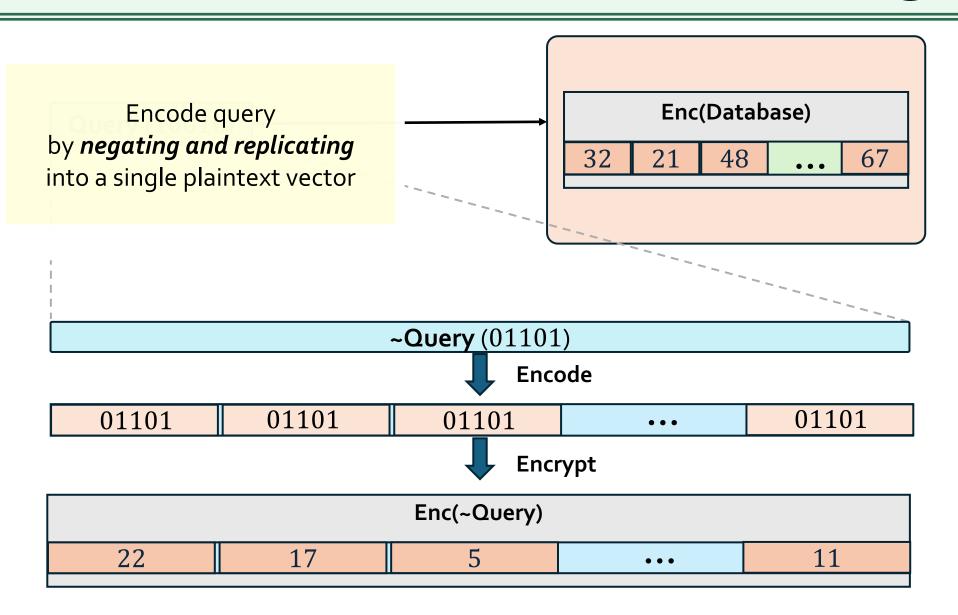






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1



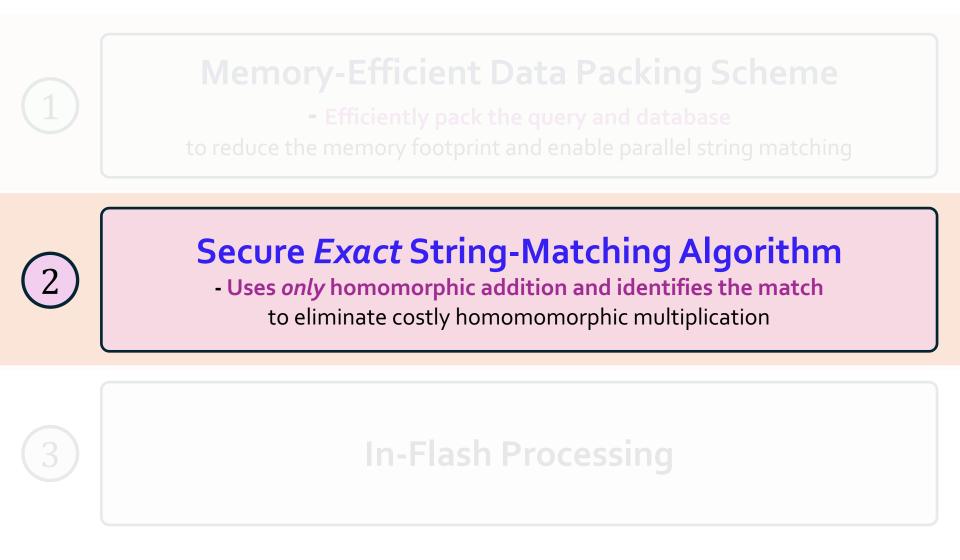
Enc(Database)				
32	21	48	•••	67

Enc(~Query)				
22	17	5	•••	11

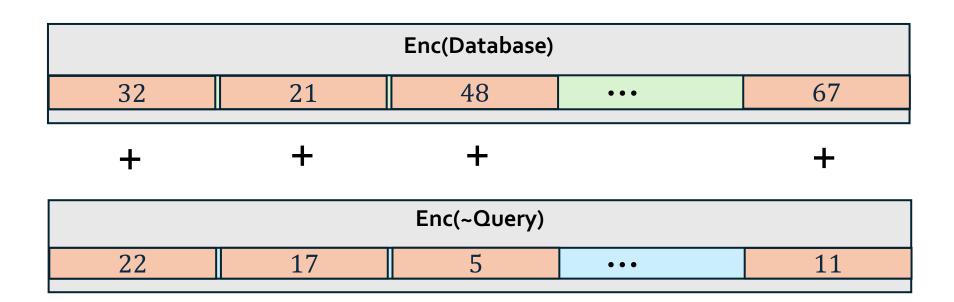


(1)

CIPHERMATCH: Key Steps



Secure Exact String-Matching Algorithm



Homomorphic addition is inherently element-wise addition



2

Secure Exact String-Matching Algorithm

Enc(Database)					
21	48	•••	67		
+	+		+		

Enc(~Query)				
22	17	5	•••	11

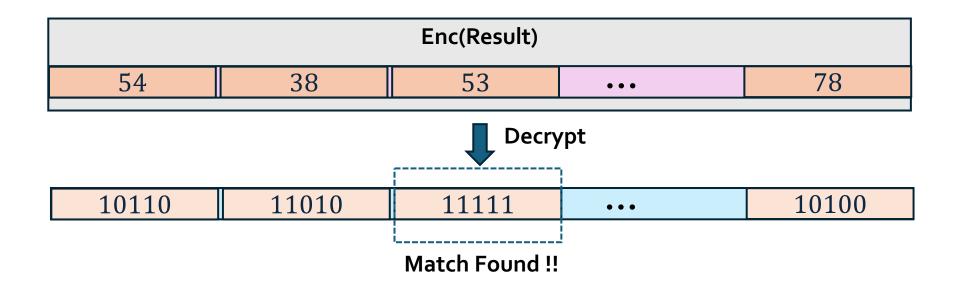
Enc(Result)				
54	38	53	•••	78

32

+

(2)

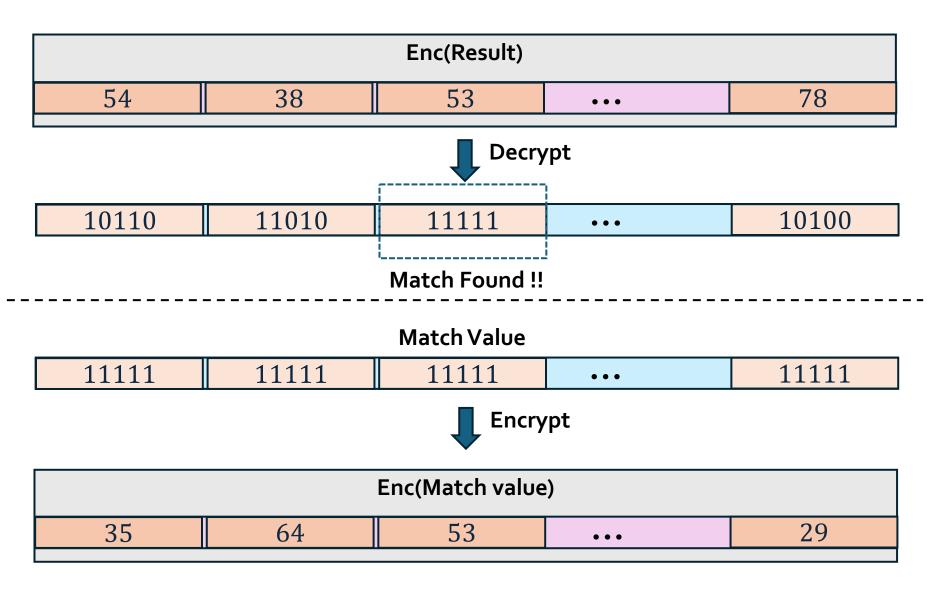
Secure Exact String-Matching Algorithm

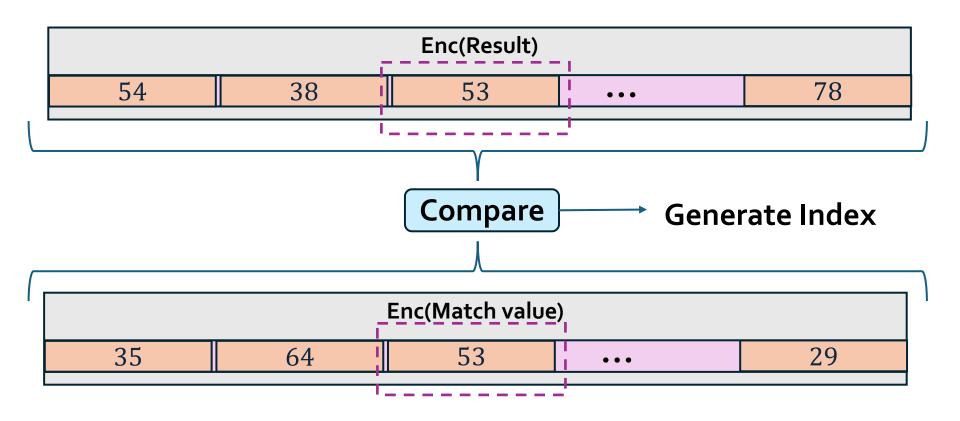


However, we want to find the match using Enc(Result) on the server



2

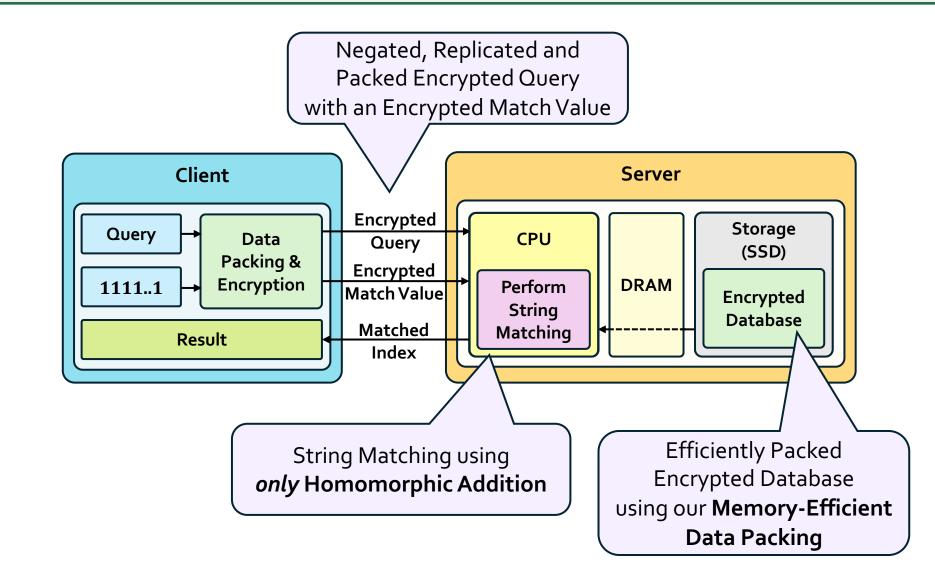




Compare and send the final index back to client



CIPHERMATCH: Algorithm (Summary)



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Key Idea

CIPHERMATCH: System Overview

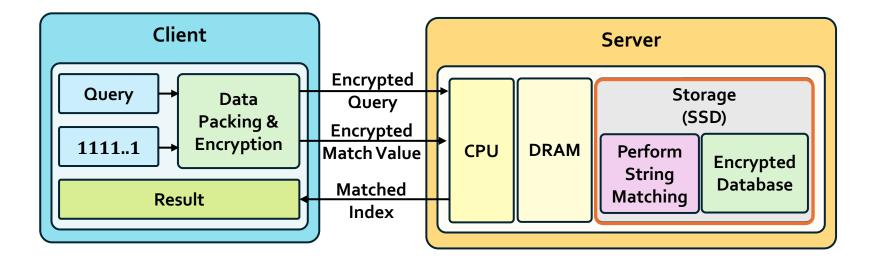
CIPHERMATCH: Algorithm

CIPHERMATCH: Hardware

Evaluation Results



CIPHERMATCH: Hardware Overview



Perform secure *exact* string matching inside SSD using in-flash processing (IFP)



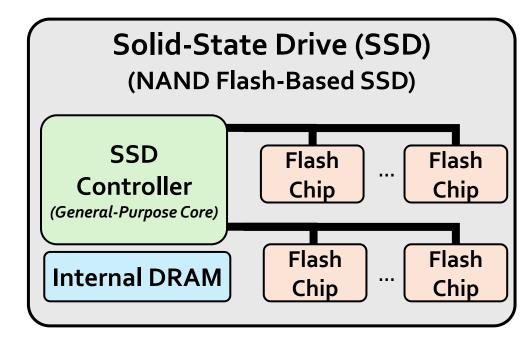
Overview of a Modern Solid State Drive (SSD)

Solid-State Drive (SSD)

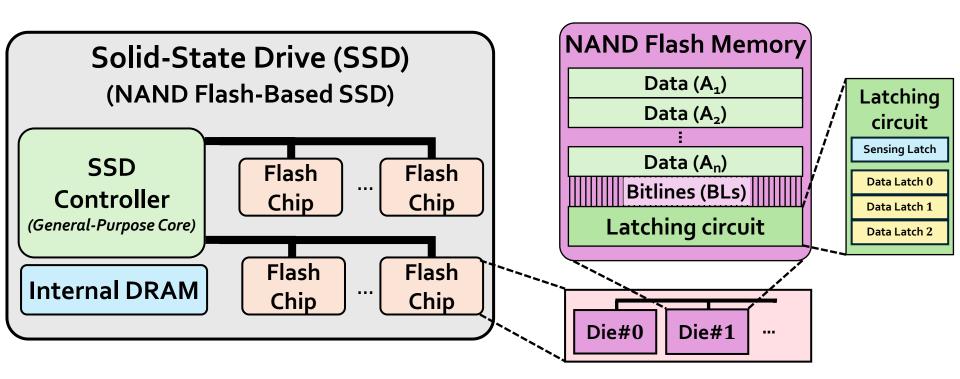
(NAND Flash-Based SSD)



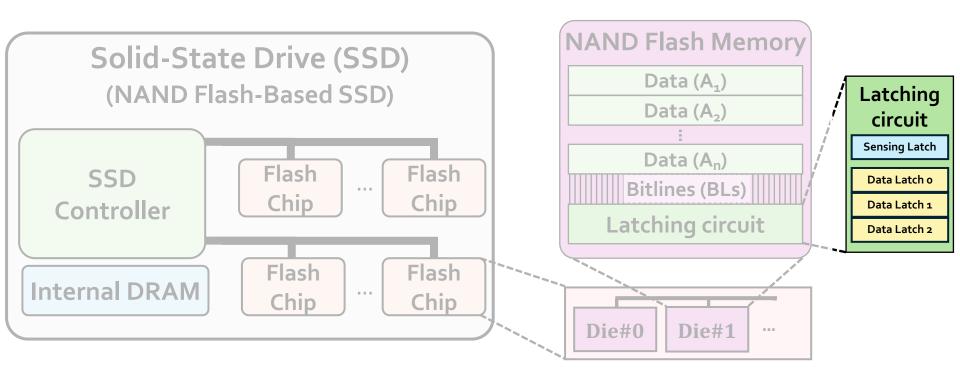
Overview of a Modern Solid State Drive (SSD)



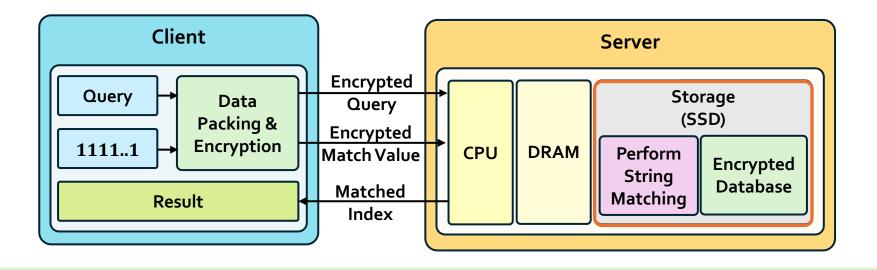
Overview of a Modern Solid State Drive (SSD)



Prior work [Gao+, MICRO 2021] Uses latching circuit to perform *only* bitwise operations



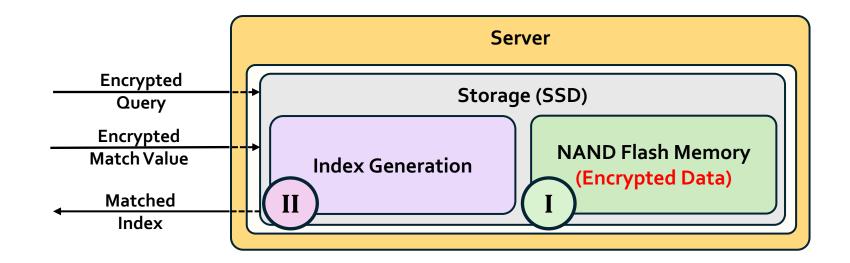
Advantages of Secure String Matching in SSD



Homomorphic addition can be parallelized

Exploit bit-level and array-level parallelism of NAND-flash memory

CIPHERMATCH: Hardware Overview

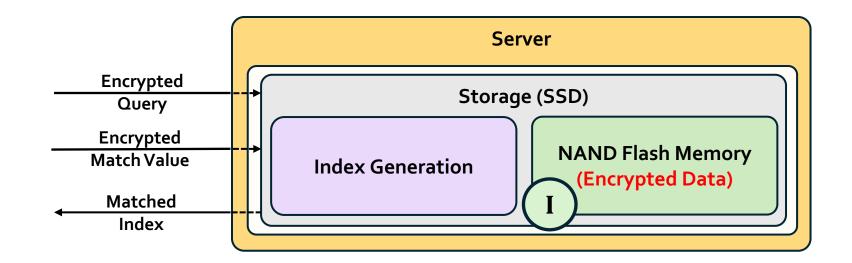


Perform homomorphic additions inside NAND-flash memory



Generate the final index by comparing it with match value

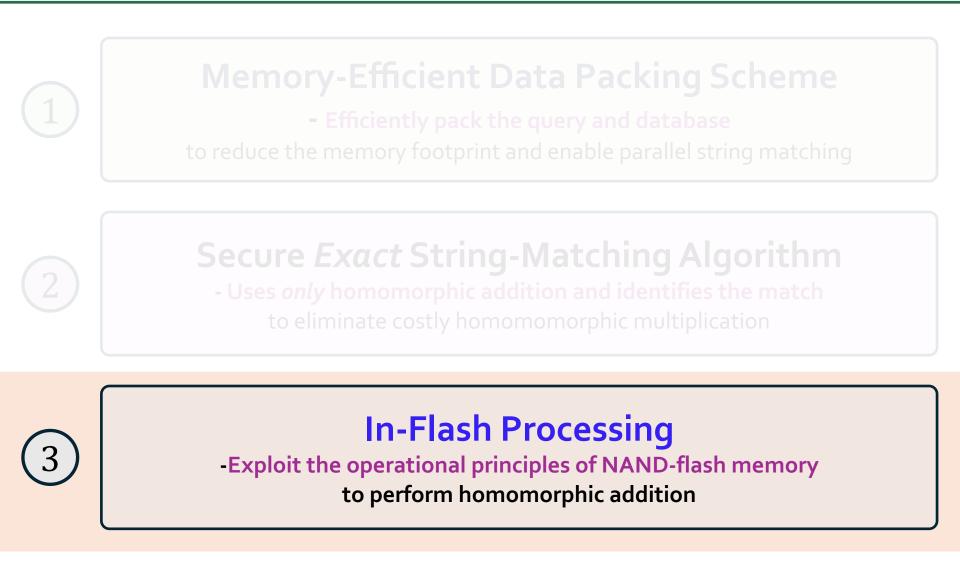
CIPHERMATCH: Hardware Overview



Perform homomorphic additions inside NAND-flash memory

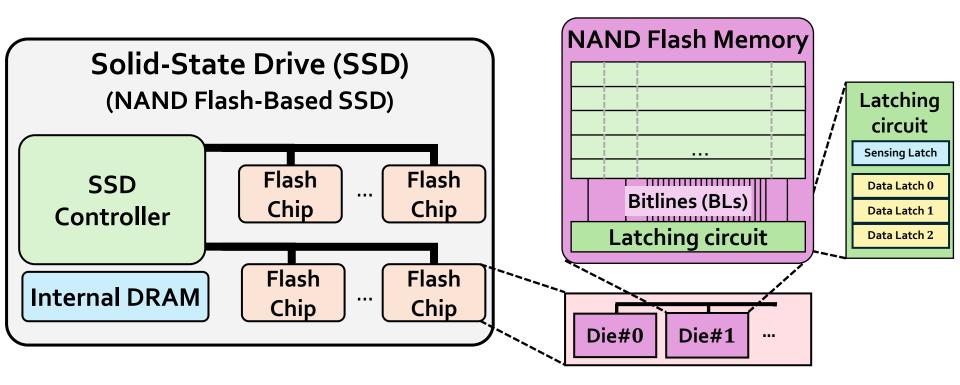
Perform element-wise addition inside NAND-flash memory

CIPHERMATCH: Key Steps

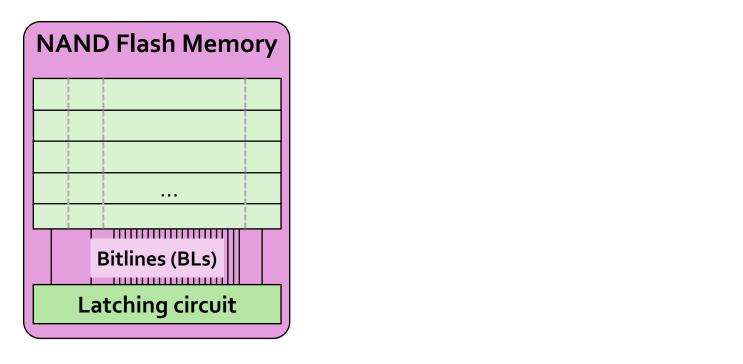




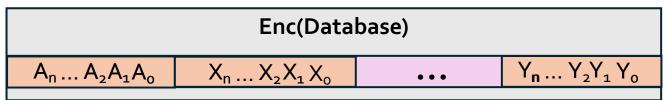
We use **bit-serial addition** to **avoid carry propagation** across different bitlines



CIPHERMATCH: Bit-Serial Addition

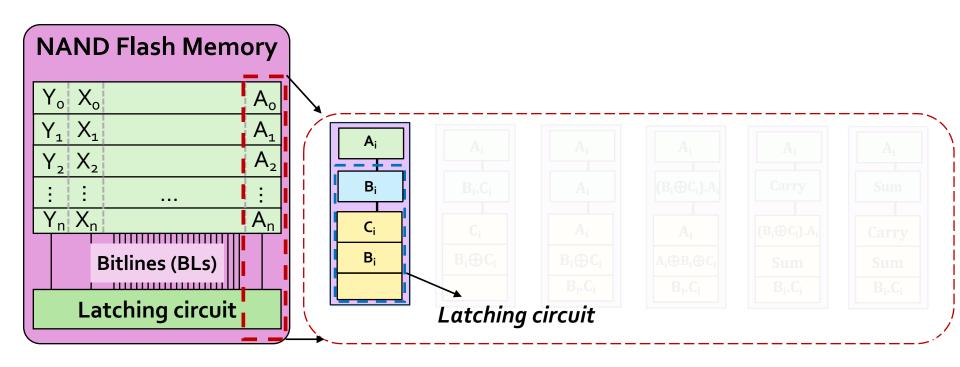


Lay out the data vertically in NAND-flash memory





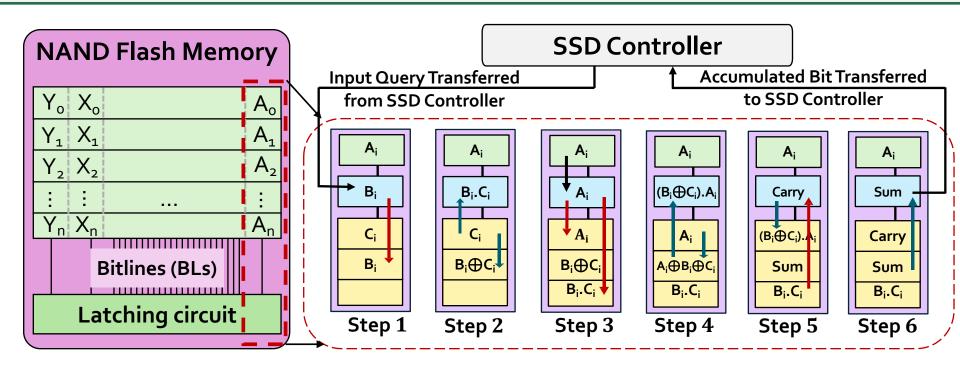
CIPHERMATCH: Bit-Serial Addition



Lay out the data vertically in NAND-flash memory



CIPHERMATCH: Bit-Serial Addition



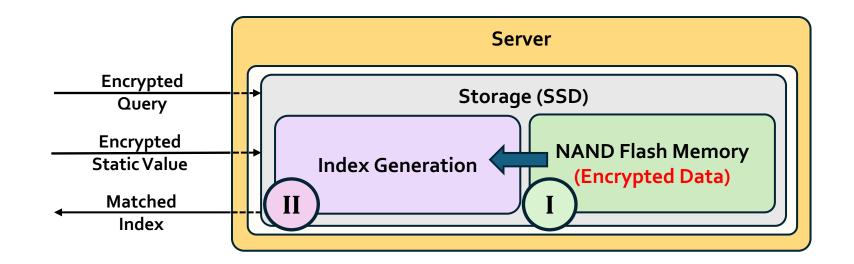
Lay out the data vertically in NAND-flash memory

Send the query from SSD controller to the latches

Perform Steps 1-6 to perform bit-serial addition

Accumulate the sum the account (b) to bat to SSD controller

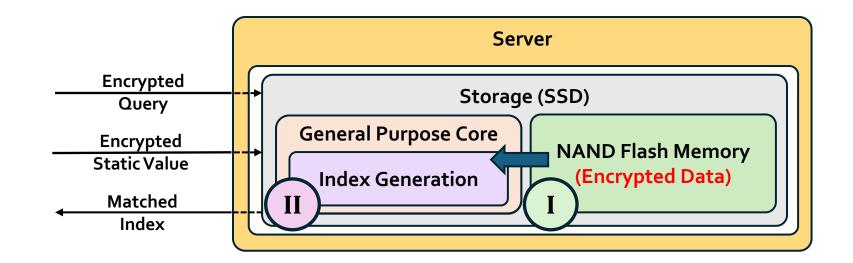
CIPHERMATCH: Hardware (Summary)



Perform homomorphic additions inside NAND-flash memory

Generate the **final index** by **comparing it with match value**

CIPHERMATCH: Hardware (Summary)



Generate the **final index** by **comparing it with match value**

Use general purpose cores to identify the final match



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Talk Outline

Background, Problem & Goal

Key Idea

CIPHERMATCH: System Overview

CIPHERMATCH: Algorithm

CIPHERMATCH: Hardware

Evaluation Results

Evaluation Methodology (1/2): Real System

Our Implementation

Intel Xeon, 6 cores, 3.2 GHz 32GB DDR4 DRAM 2TB PCIe 4.0 SSD

We evaluate software-based CIPHERMATCH implementation (CM-SW) by modifying the Microsoft SEAL library

Baselines

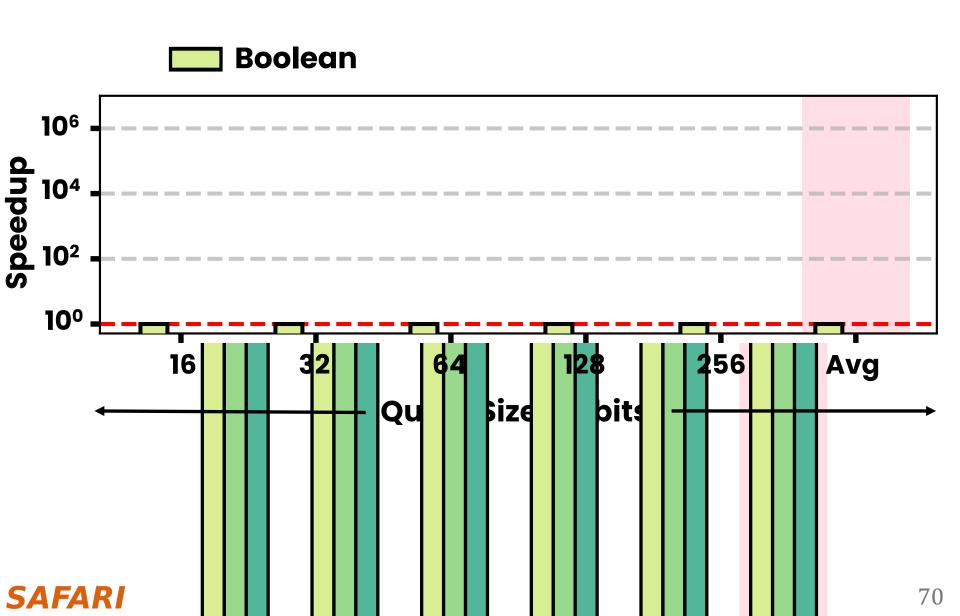
- Arithmetic (using SEAL): State-of-the-art arithmetic approach [Yasuda+, CCSW 2013]
- Boolean (using TFHE-rs): State-of-the-art Boolean approach [Aziz+, Information 2024]

Workloads

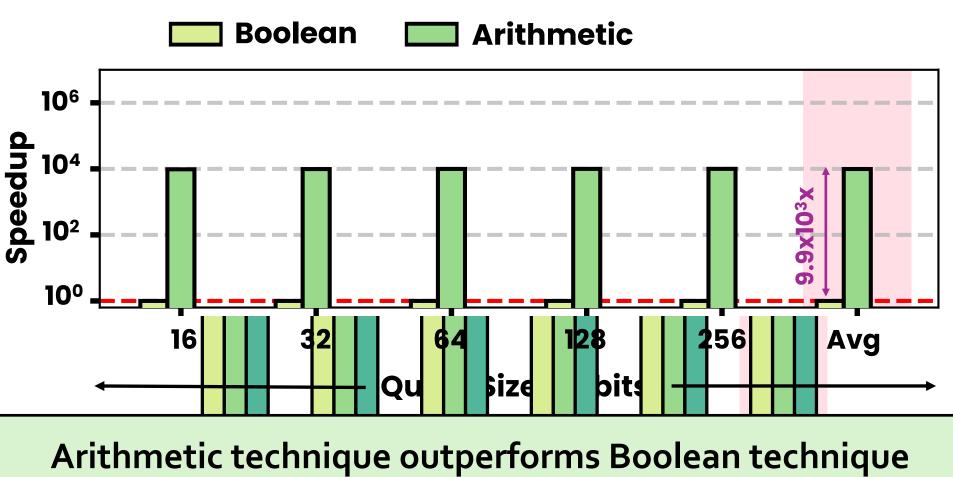
- Varying query size (16-256 bits)* for encrypted database size of 128 GB
- Varying encrypted database size (8-128 GB)* for 16-bit query and 1000 queries

* including all circular shifted queries

Speedup for Different Query Sizes

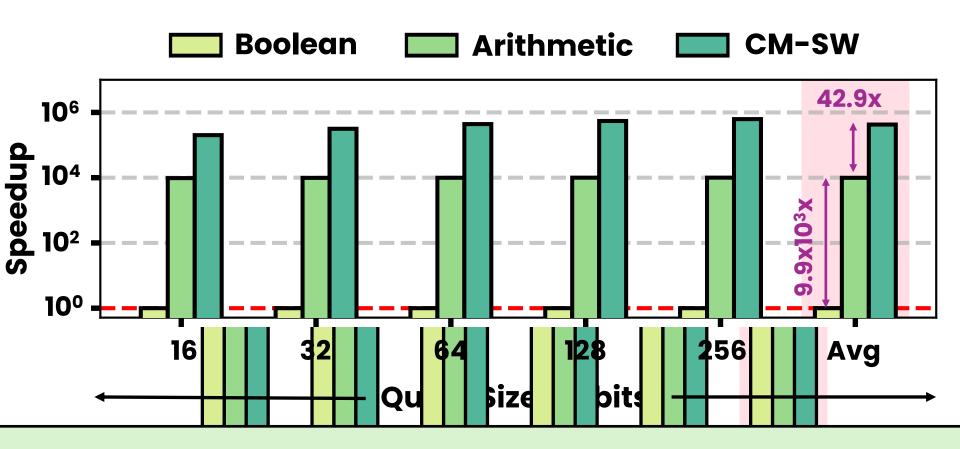


Speedup for Different Query Sizes (1/3)



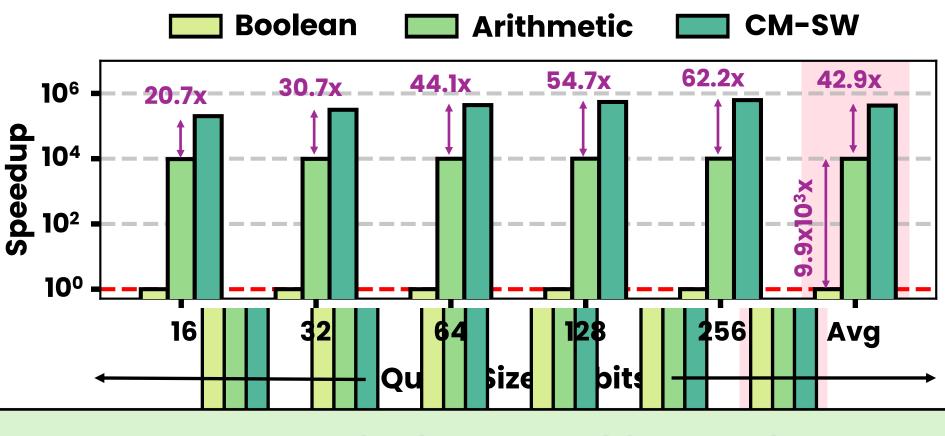
by orders of magnitude

Speedup for Different Query Sizes (2/3)



CM-SW outperforms the best prior arithmetic technique by 42.9x

Speedup for Different Query Sizes (3/3)



CM-SW speedup increases with query size (due to the elimination of homomorphic multiplication)

Evaluation Methodology (2/2): Simulation

Our Implementation

We evaluate IFP-based CIPHERMATCH implementation (CM-IFP) by modeling the characteristics of the NAND-flash memory

Baselines

- CM-SW: CIPHERMATCH on compute-centric system [same as real system]
- CM-PuM: CIPHERMATCH on memory-centric system [*, 32GB DDR4-2400]
- **CM-PuM-SSD:** CIPHERMATCH on storage-centric system [*, SSD DRAM 2GB LPDDR4-1866]

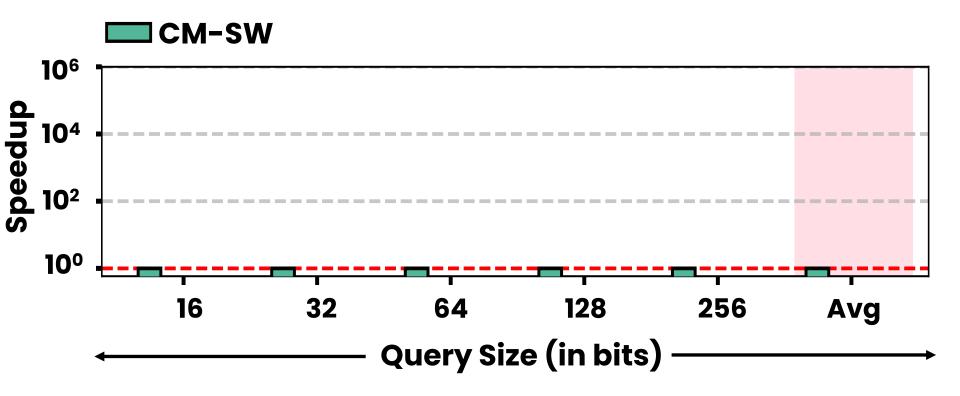
[*] - SIMDRAM framework [Hajinazar+, ASPLOS 2021]

Workloads

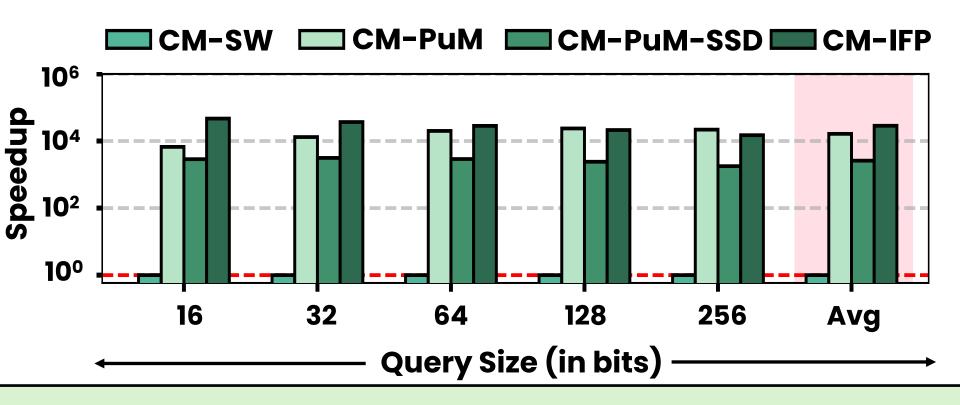
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Speedup for Different Query Sizes

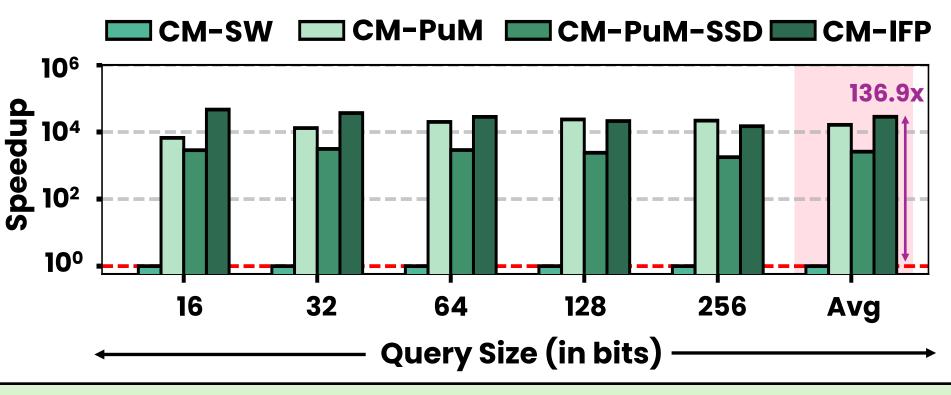


Speedup for Different Query Sizes (1/3)



All three near-data processing systems improve performance by reducing data movement

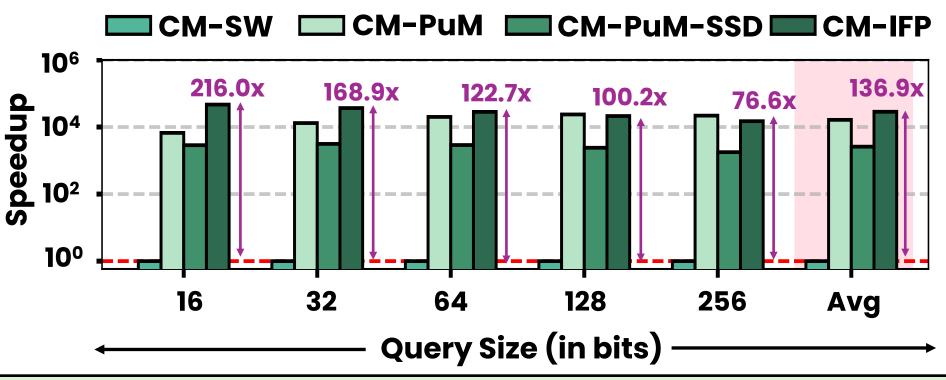
Speedup for Different Query Sizes (2/3)



CM-IFP outperforms CM-SW by 136.9x

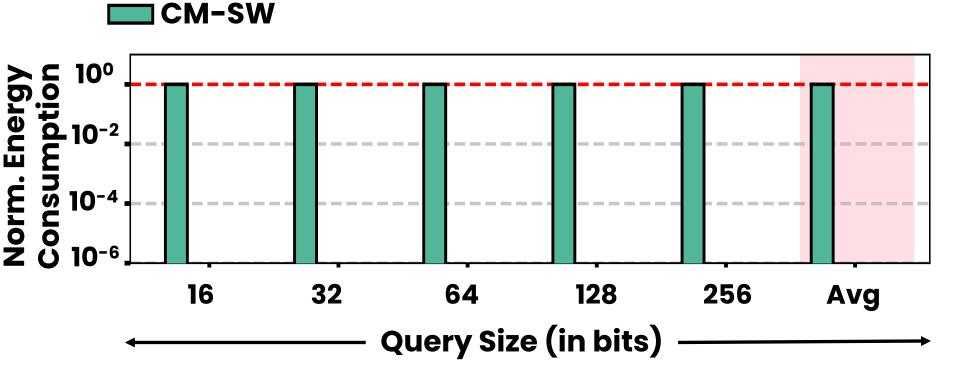
CM-IFP outperforms other near-data processing systems

Speedup for Different Query Sizes (3/3)

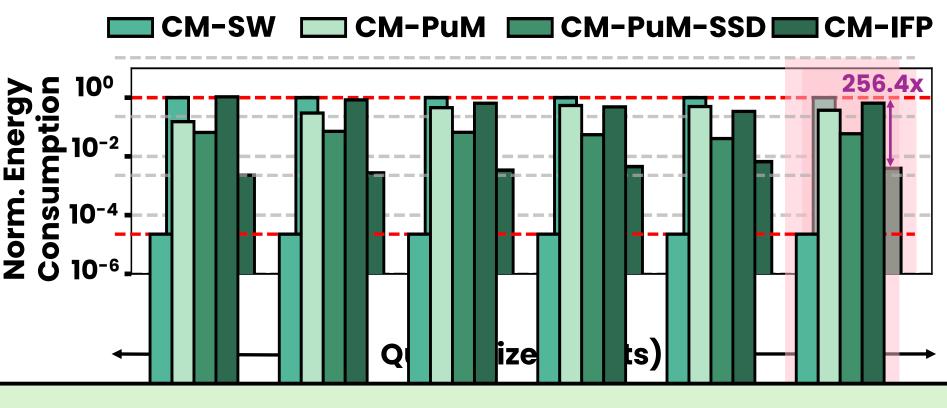


CM-IFP speedup decreases with query sizes due to repeated flash read operations on same data for circularly shifted queries

Energy Consumption for Different Query Sizes



Energy Consumption for Different Query Sizes



All three near-data processing systems provide large energy savings over CM-SW

CIPHERMATCH: Accelerating Homomorphic Encryption-Based String Matching via Memory-Efficient Data Packing and In-Flash Processing

Mayank Kabra† Rakesh Nadig† Harshita Gupta† Rahul Bera† Manos Frouzakis† Vamanan Arulchelvan† Yu Liang† Haiyu Mao‡ Mohammad Sadrosadati† Onur Mutlu† ETH Zurich† King's College London‡

Homomorphic encryption (HE) allows secure computation on encrypted data without revealing the original data, providing significant benefits for privacy-sensitive applications. Many cloud computing applications (e.g., DNA read mapping, biometric matching, web search) use exact string matching as a key operation. However, prior string matching algorithms that use homomorphic encryption are limited by high computational latency caused by the use of complex operations and data movement bottlenecks due to the large encrypted data size. In this work, we provide an efficient algorithm-hardware codesign to accelerate HE-based secure exact string matching. We propose CIPHERMATCH, which (i) reduces the increase in memory footprint after encryption using an optimized software-based data packing scheme, (ii) eliminates the use of costly homomorphic operations (e.g., multiplication and rotation), and (iii) reduces data movement by designing a new in-flash processing (IFP) architecture.

format). Since cloud servers can be shared among multiple users, sensitive user data can become vulnerable to security threats and leaks [24-26]. HE can significantly benefit privacy-sensitive applications [27-31] that require exact string matching [13-17, 19, 21-23] as the fundamental operation by directly operating on encrypted data without requiring decryption.

Unfortunately, homomorphic operations are typically $10^4 \times$ to $10^5 \times$ slower than their traditional unencrypted counterparts in existing systems [32]. Prior works propose two main approaches to perform secure string matching: (1) the Boolean approach (e.g., [17, 33]), and (2) the arithmetic approach (e.g., [27,29,34]). The Boolean approach [17,33] packs individual bits into a polynomial, encrypts it, and uses homomorphic XNOR and AND operations to perform secure string matching on a search pattern of any size. In contrast, the arithmetic approach [27,29,34] packs multiple bits into a polynomial, encrypts it, and employs homomorphic multiplication and addition

https://arxiv.org/pdf/2503.08968

To Summarize ...

Conclusion

CIPHERMATCH

A new algorithm-hardware codesign that significantly improves the performance of secure *exact* string matching algorithm



Pack multiple bits of database and thus eliminate the use of homomorphic multiplication

Use in-flash processing (IFP) to accelerate secure exact string-matching

- + Reduces memory footprint
- + Provides scalable
 - secure exact string-matching
- + Reduces data movement
- + Leverages bit-level and array-level parallelism

Key Results

- CIPHERMATCH-SW: 42.9x speedup & 39.4x lower energy than best software
- CIPHERMATCH-IFP: 136.9x speedup & 256.4x lower energy than CM-SW

CIPHERMATCH

Accelerating Homomorphic Encryption-Based String Matching via Memory-Efficient Data Packing and In-Flash Processing



ETH zürich

Mayank Kabra

Rakesh Nadig, Harshita Gupta, Rahul Bera, Manos Frouzakis, Vamanan Arulchelvan, Yu Liang, Haiyu Mao, Mohammad Sadrosadati, and Onur Mutlu

Backup slides

Summary

CM-SW provides 42.9x speedup over the state-of-the-art approach in real systems

Due to our new memory-efficient data packing scheme and use of *only* homomorphic additions

CM-IFP provides 136.9x speedup over CM-SW and outperforms three near-data processing systems

Due to our new IFP design to perform in-flash operations and exploiting large-scale bit-level parallelism



Executive Summary

Problem: Secure exact string matching using homomorphic encryption (HE) operations face performance bottlenecks in two key areas:

- (a) High computation cost due to use of complex homomorphic operations (e.g., multiplication)
- (b) data movement bottleneck due to large homomorphically encrypted data

Motivation: Reducing memory expansion from HE and performing computation where the database resides can improve the performance of secure exact string matching algorithm

Opportunity: (a) Perform memory-efficient packing of the database to reduce the increase in memory footprint after encryption and (b) perform simple computations (e.g., HE addition) inside solid state drives (SSDs – i.e., where the database is stored) to reduce data movement

<u>CIPHERMATCH</u>: A novel algorithm-hardware co-design that significantly improves the performance of HE-based secure string matching by using *only homomorphic addition* and leveraging the operational principles of NAND-flash memory.

Key Idea:

- 1) To pack multiple bits of data in the each coefficient of ciphertext
- 2) Use in-flash processing (IFP) to perform string matching inside NAND-flash memory

Key Benefits:

- + Reduce memory expansion after encryption
- + Eliminates the use of complex HE operations
- + Reduces data movement bottleneck

Key Results: (i) Software-based CIPHERMATCH implementation (CM-SW) achieves 42.9x speedup over state-of-the-art software approaches

(ii) CIPHERMATCH IFP implementation futher improves upon CM-SW, achieving 136.9x better performance and 256.4x lower energy consumption

Executive Summary

<u>Problem</u>: Secure exact string matching using homomorphic encryption (HE) operations face performance bottlenecks in two key areas:

- (a) High computation cost due to use of complex homomorphic operations (e.g., multiplication)
- (b) Data movement bottleneck due to large homomorphically encrypted data size

Motivation: Reducing memory expansion from HE and performing computation where the database resides can improve the performance of secure exact string matching algorithm

Opportunity: (a) Optimize memory usage by **packing encrypted data efficiently** and (b) perform secure string matching using **simple HE operations (addition) inside SSDs**, reducing data movement.

<u>CIPHERMATCH: A novel algorithm-hardware co-design</u> that significantly improves the performance of HE-based secure exact string matching (a) by using *only homomorphic addition* and (b) leveraging the *operational principles of NAND-flash memory* to perform secure exact string matching

Key Idea:

- 1) Pack multiple bits of data in the each coefficient of ciphertext
- 2) Use in-flash processing (IFP) to perform string matching inside NAND-flash memory

Key Benefits:

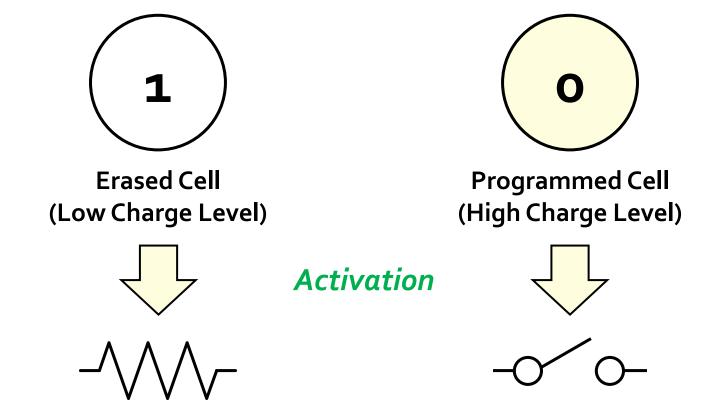
- + Reduce memory expansion after encryption
- + Eliminates the use of complex HE operations
- + Reduces data movement bottleneck

Key Results:

(i) Software-based CIPHERMATCH (CM-SW): 42.9× speedup over existing state-of-the-art approaches (ii) CIPHERMATCH with IFP: 136.9× faster and 256.4× lower energy consumption than CM-SW

NAND Flash Basics: A Flash Cell

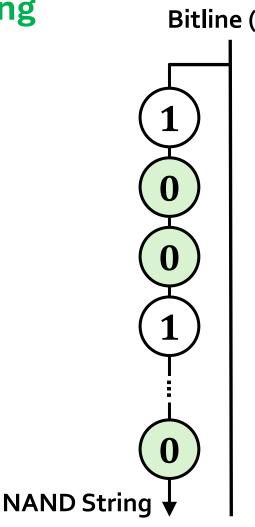
 A flash cell stores data by adjusting the amount of charge in the cell



Operates as a resistor

NAND Flash Basics: A NAND String

• A set of flash cells are serially connected to form a NAND String Bitline (BL)



NAND Flash Basics: A NAND Block

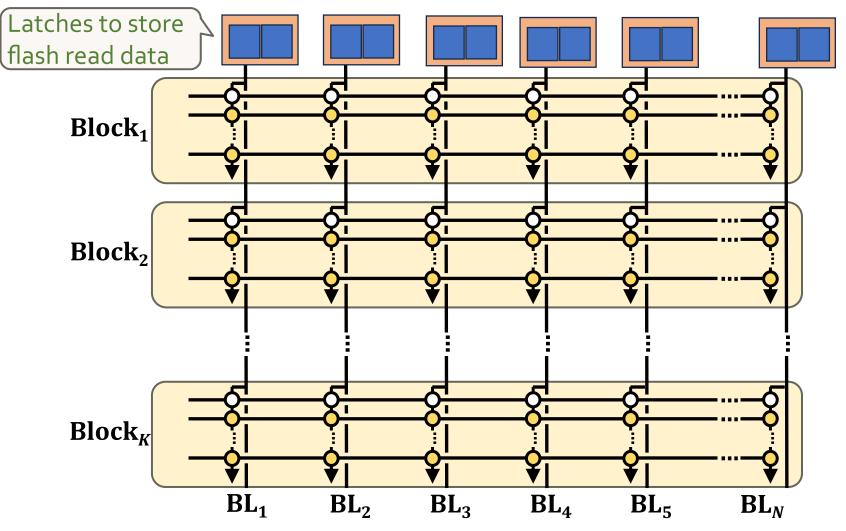
 NAND strings connected to different bitlines comprise a NAND block BL₁ **BL**_N BL₄ BL, BL₃ BLς **Block** WL₁ 1 A single wordline (WL) controls a large number of WL, O flash cells: High bit-level parallelism WL₃ 0 Ο O WL_4 WL_M 0 0 0 0 0 0

SAFAR

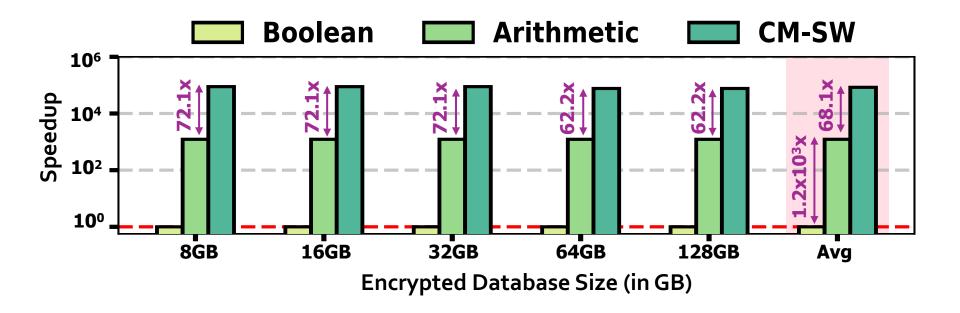
91

NAND Flash Basics: A NAND Plane





Speedup for Different Database Size



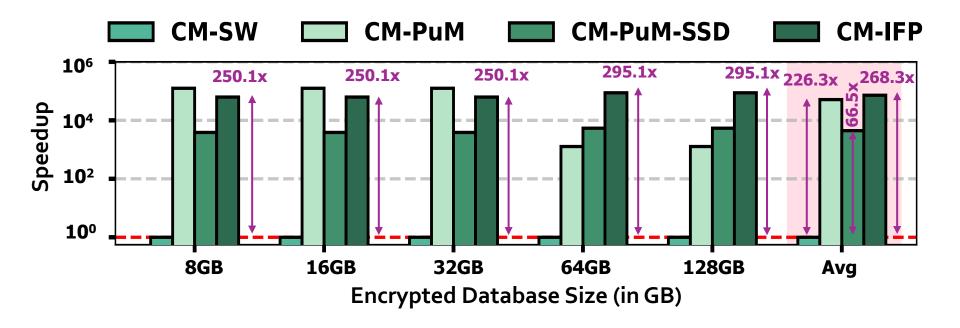
CM-SW shows average speedup of 68.1x over prior arithmetic approach

CM-SW speedup decreases as data size exceeds DRAM capacity, primarily due to increased data movement between storage and DRAM.



Query Size of 16 bits

Speedup for Different Database Size



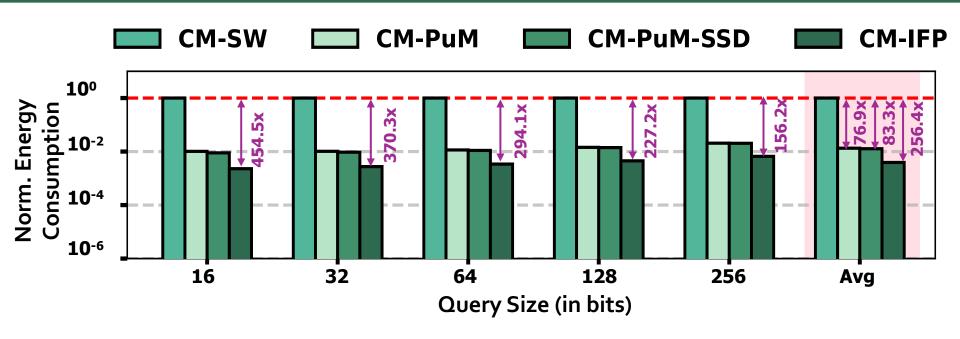
CM-IFP shows highest average speedup of 268.3x over CM-SW

CM-SW speedup decreases when data size goes beyond DRAM size due to frequent data movement between storage and DRAM



Query Size of 16 bits

Energy Consumption for Different Query Size

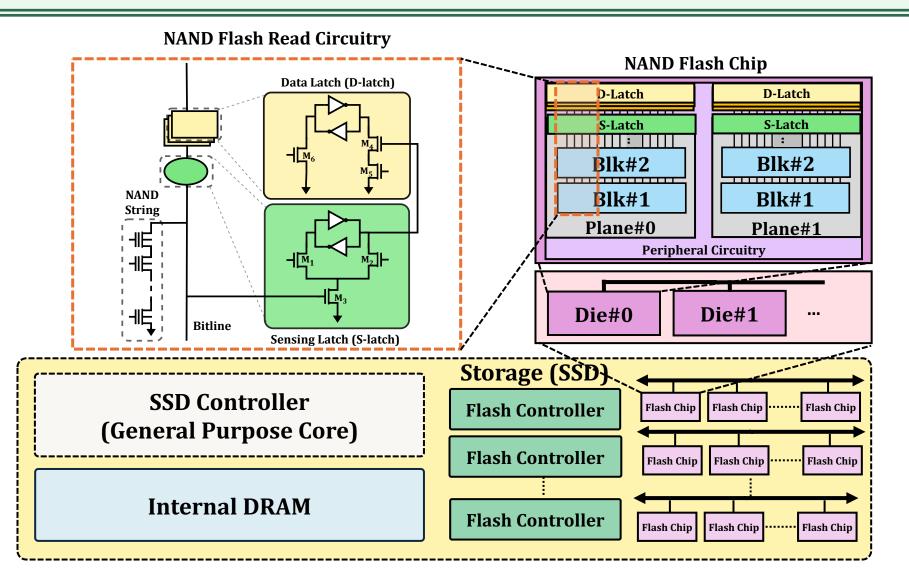


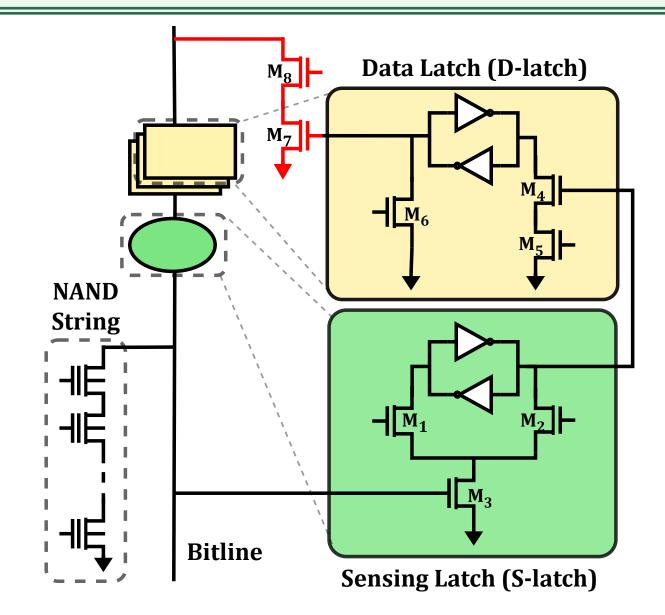
CM-IFP shows highest average energy savings of 256.4x over CM-SW

CM-IFP energy efficiency decreases with increasing query sizes due to expensive flash reads



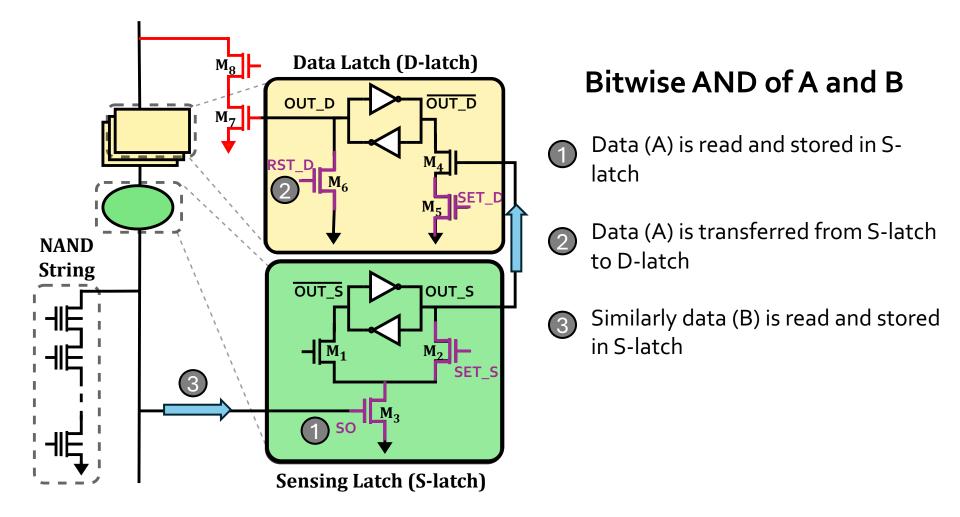
Database Size of 128 GB

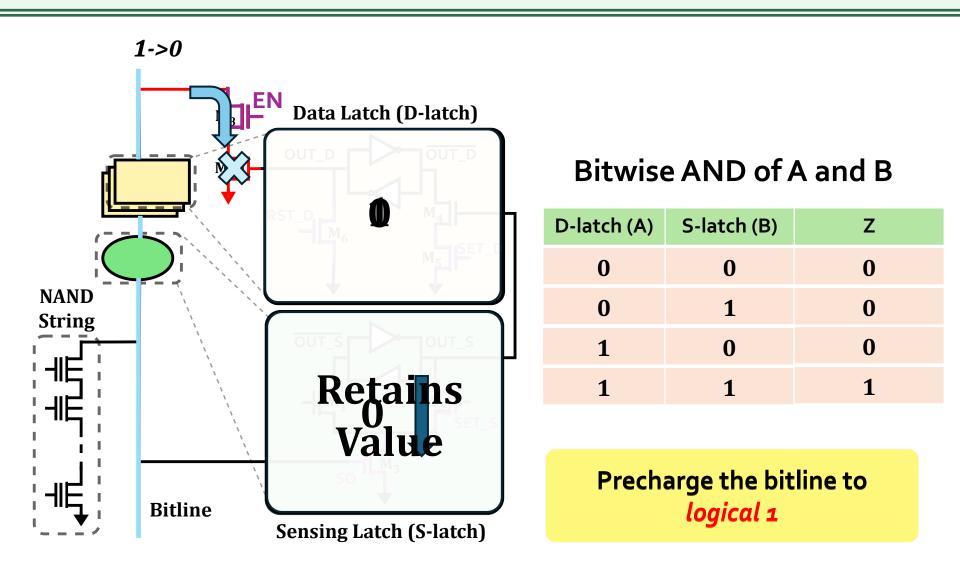




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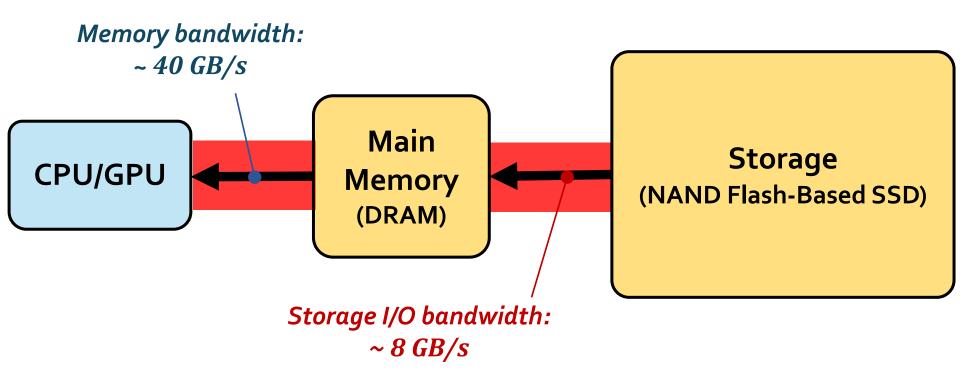
Design proposed by prior work [Cho+, Patent 2022]





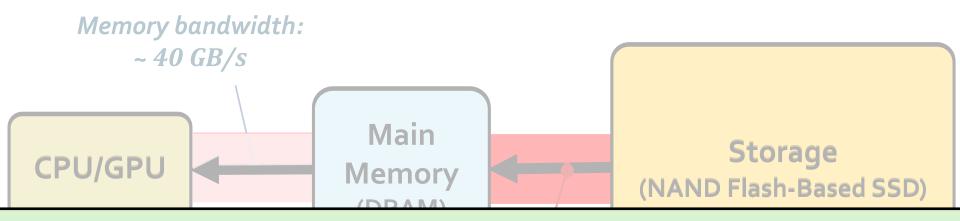
Data Movement Bottleneck

• Compute-centric systems: Move entire data from storage to CPU/GPU



Motivation (II) – Data Movement Bottleneck2

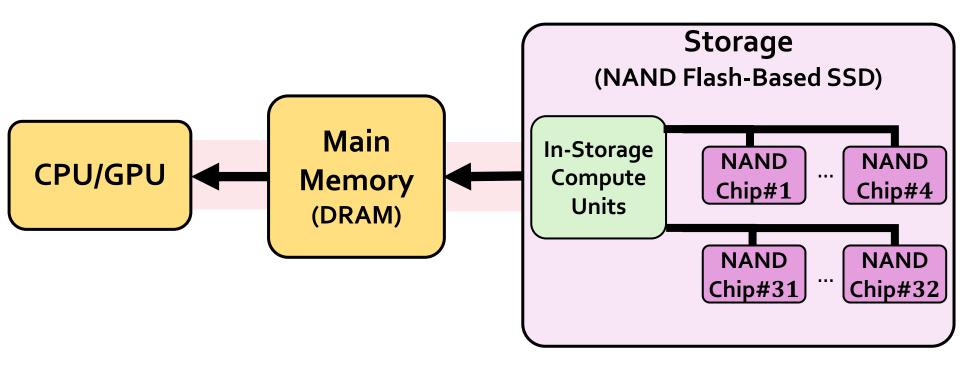
- Compute-centric systems: Move entire data from storage to CPU/GPU
- Memory-centric systems: Perform computations in main memory



External I/O bandwidth of storage systems is the *main bottleneck* for memory intensive application

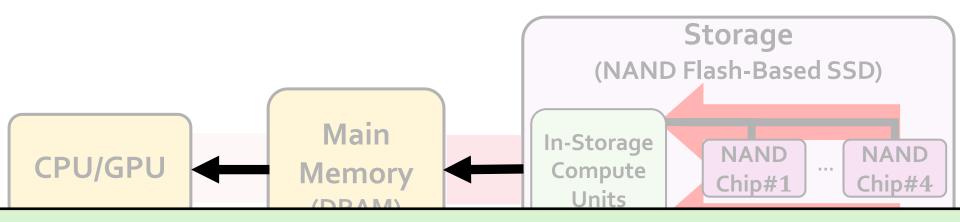
Motivation (II) – Data Movement Bottleneck2

- **Compute-centric systems:** Perform computations in CPU/GPU
- Memory-centric systems: Perform computations in main memory
- **Storage-centric systems:** Perform computations inside storage system



Motivation (II) – Data Movement Bottleneck2

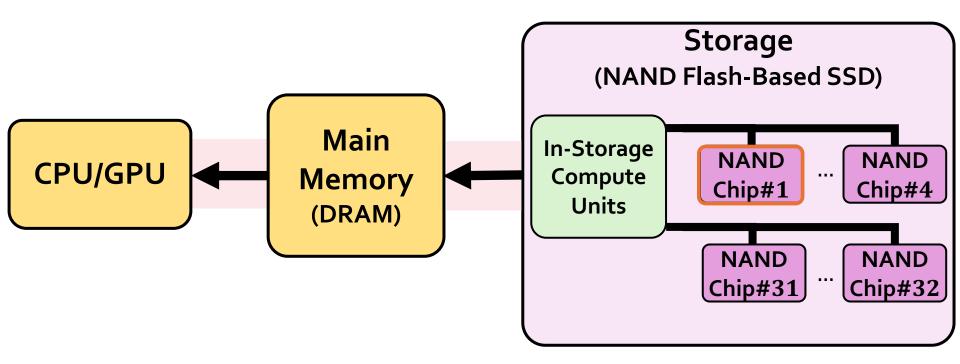
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- Memory-centric systems: Perform computations in main memory
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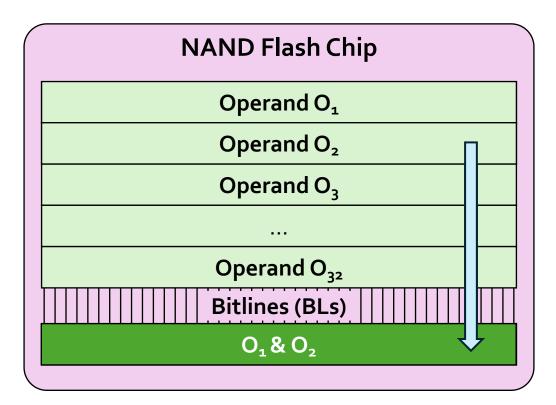
SSD-internal bandwidth becomes the new bottleneck for computations



Perform computations inside NAND-flash chips by using operational principles of NAND-flash memory

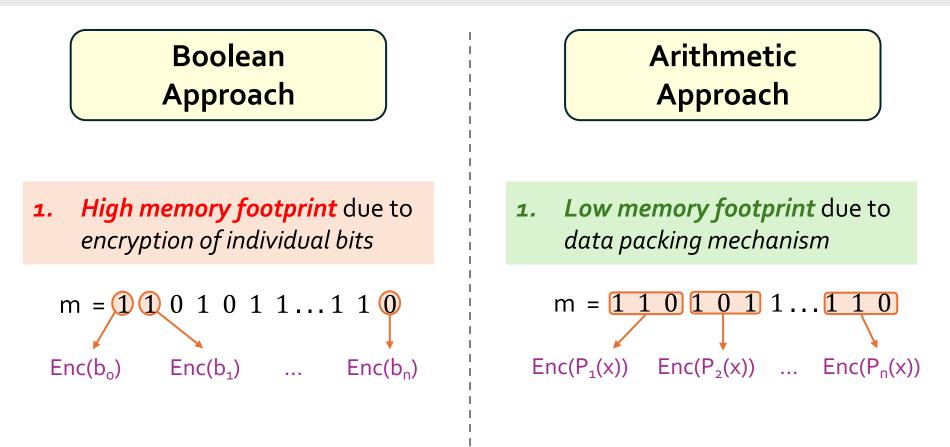


Prior Works ([Gao+, MICRO 2021] , [Park+, MICRO 2022]) perform *bitwise operations using the latching circuit*



Limitations of Prior Works

- Boolean Approach [Pradel+, TrustCom 2021; Aziz+, Information 2024]
- Arithmetic Approach [Yasuda+, CCSW 2013 ; Kim+, TDSC 2017 ; Bonte+, CCS 2020]



Limitations of Prior Works

- Boolean Approach [Pradel+, TrustCom 2021; Aziz+, Information 2024]
- Arithmetic Approach [Yasuda+, CCSW 2013; Kim+, TDSC 2017; Bonte+, CCS 2020]

Boolean Approach

- High memory footprint due to encryption of individual bits
- **2.** High computation cost due to large number of HE operations

 $Enc(b_0) \bigoplus Enc(b_1) \bigoplus Enc(b_2) \dots 1000x$

Arithmetic Approach

- Low memory footprint due to data packing mechanism
- 2. Low computation cost due to small number of HE operations

 $Enc(P_1(x)) \times Enc(P_2(x)) \times \dots 10x$

Limitations of Prior Works

- Boolean Approach [Pradel+, TrustCom 2021; Aziz+, Information 2024]
- Arithmetic Approach [Yasuda+, CCSW 2013 ; Kim+, TDSC 2017 ; Bonte+, CCS 2020]

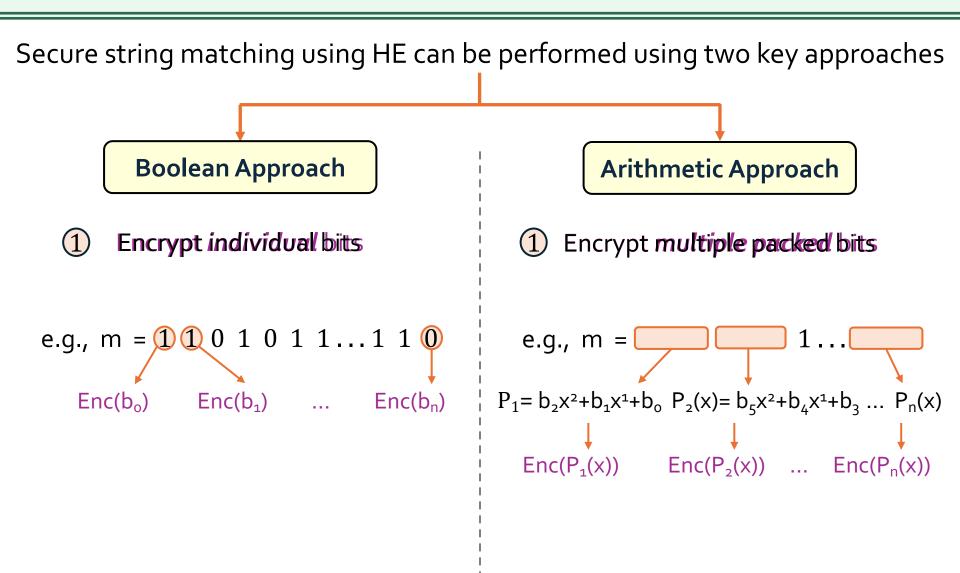
Boolean Approach

- High memory footprint due to encryption of individual bits
- **2.** High computation cost due to large number of HE operations
- 3. Support flexible query sizes due to unlimited computations

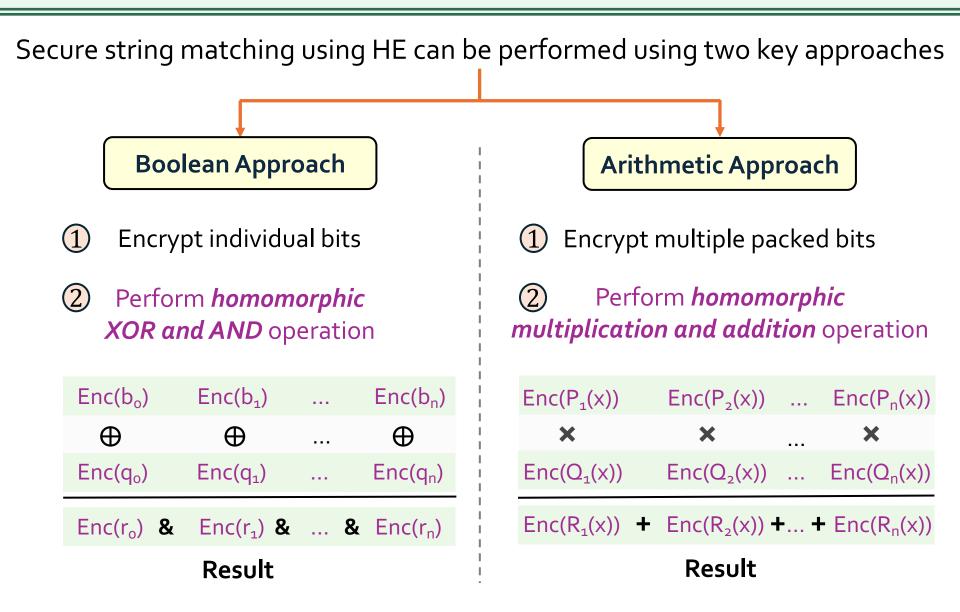
Arithmetic Approach

- Low memory footprint due to data packing mechanism
- 2. Low computation cost due to small number of HE operations
- **3.** Support limited query sizes due to limited computations

Approaches for HE-based String Matching

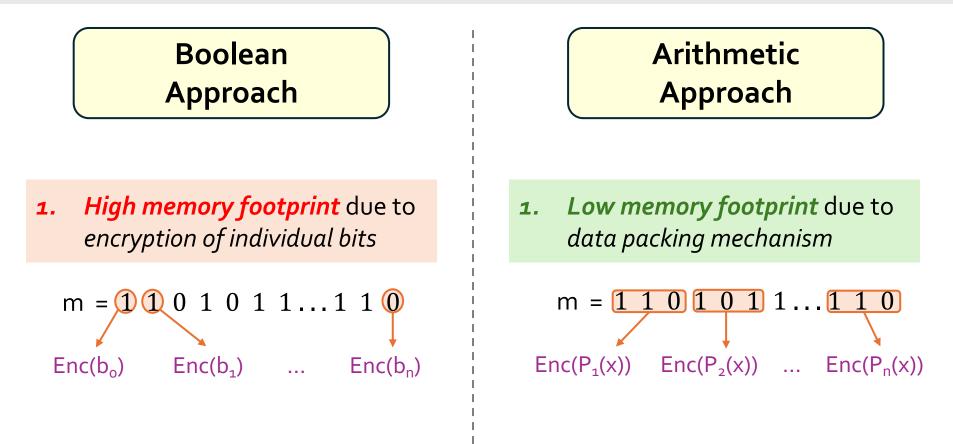


Approaches for HE-based String Matching



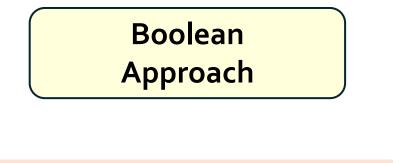
Prior Works on HE-based String Matching

- Boolean Approach [Pradel+, TrustCom 2021; Aziz+, Information 2024]
- Arithmetic Approach [Yasuda+, CCSW 2013 ; Kim+, TDSC 2017 ; Bonte+, CCS 2020]



Prior Works on HE-based String Matching

- Boolean Approach [Pradel+, TrustCom 2021; Aziz+, Information 2024]
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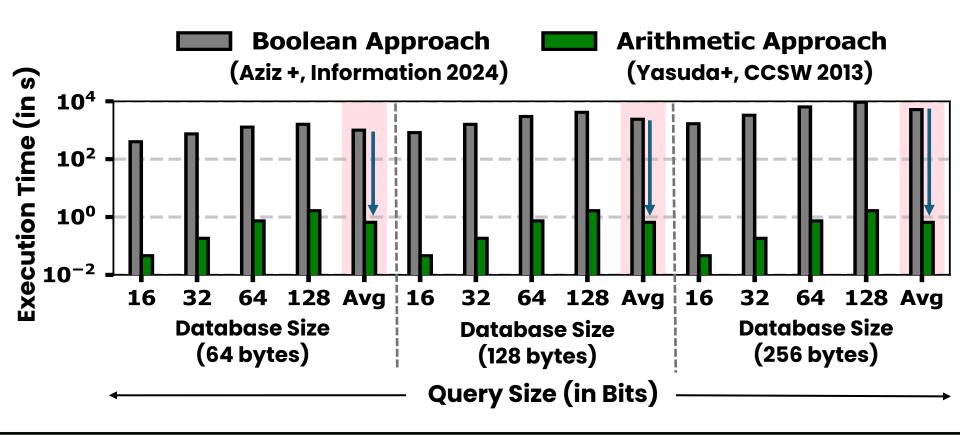


- High memory footprint due to encryption of individual bits
- 2. Support flexible query sizes due to unlimited computations



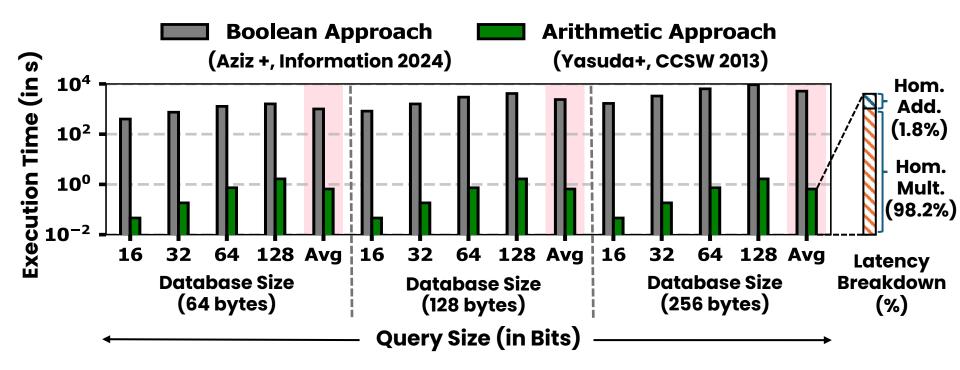
- Low memory footprint due to data packing mechanism
- 2. Support limited query sizes due to limited computations

Arithmetic Approach Performs Better

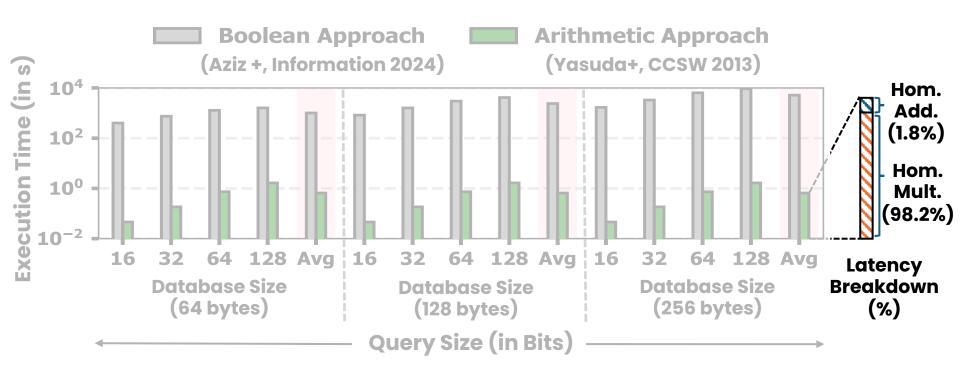


Arithmetic approach *performs better with larger database sizes* due to fewer HE operations

Latency Breakdown of Arithmetic Approach



Key Problem (I): Complex HE Operations



Prior arithmetic approaches

use costly homomorphic multiplication operations which limits the scalabilty of HE-based string matching

Key Observation

String matching can be performed using addition operation

If we negate the data, add it to the original data, we get a string of 1 1 1 1's

Key Observation

String matching can be performed using addition operation $m = 0 1 0 1 0 0 1 0 1 0 1 0 1 0 1 \\
+ \sim m = 10 10 1 1 0 1 0 1 1 0 1 0 \\
11 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 \\
Value which can be checked$

Secure string matching can be performed using HE addition operation

Enc(m) =
$$(5 x^{1024} + 10 x^{1023} + + 19, ...)$$

+ Enc(~m) = $(6 x^{1024} + 11 x^{1023} + + 3, ...)$
(11 $x^{1024} + 21 x^{1023} + + 22, ...)$
Decrypt
(1111...1 $x^{1024} + 1111...1 x^{1023} + + 1111...1$)

Key Observation

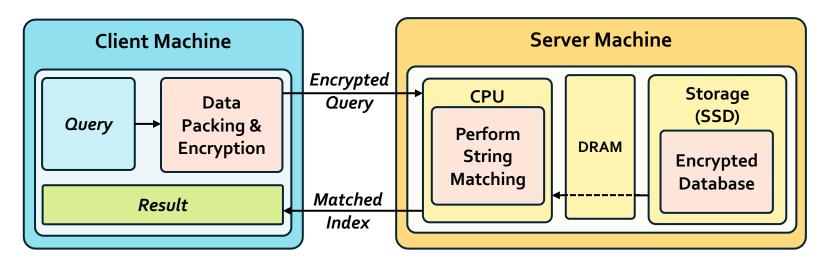
String matching can be performed using addition operation

This output after homomorphic addition is an encrypted value of 1111's *that can be used for matching*

+ Enc(\sim m) = (6 x¹⁰²⁴ + 11 x¹⁰²³ + ... + 3 , ...)

 $(11 x^{1024} + 21 x^{1023} + + 22, ...)$

Memory-Efficient Data Packing Scheme

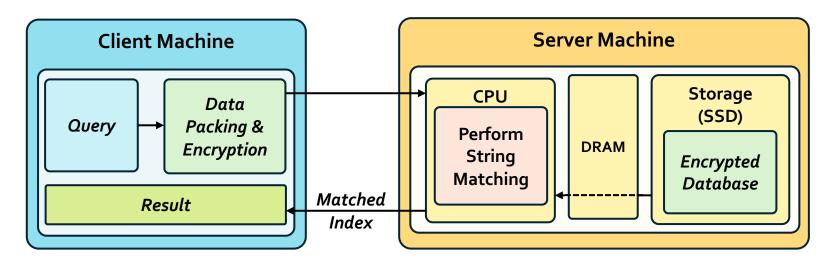


Assume, **Database (d)** = 10101111001011111 ... 10101

Encode database into multiple **plaintext polynomials (P(x))** by **packing multiple bits** into a single polynomial coefficient

e.g.,
$$P(x) = 10101...1x^{1024} + 10010...1x^{1023} + + 11...10101$$

CIPHERMATCH: Data Packing Scheme



Assume,

Query (q) = 1110011 > 0001100

The query (q) is *negated, replicated* and encoded into the **plaintext polynomials (Q(x))**

e.g.,
$$Q(x) = 0\ 0\ 0\ 1\ 1...0\ x^{1024} + 0\ 0\ 0\ 1\ 1...0\ x^{1023} + \ + \ 0\ 0\ 0\ 1\ 1...0$$



2

String matching can be performed using addition operation

Value which can be checked

This output after homomorphic addition is an encrypted value of 1111's *that can be used for matching*

+ $Enc(\sim m) = (6 x^{1024} + 11 x^{1023} + ... + 3, ...)$

 $(11 x^{1024} + 21 x^{1023} + + 22, ...)$

String matching can be performed using addition operation

Secure String matching can be performed using HE addition operation

Enc(m) =
$$(5 x^{1024} + 10 x^{1023} + + 19, ...)$$

Enc(~m) = $(6 x^{1024} + 11 x^{1023} + + 3, ...)$
(11 $x^{1024} + 21 x^{1023} + + 22, ...)$
Decrypt
(1010...1 $x^{1024} + 1111...1 x^{1023} + + 1001...0)$

SAFARI

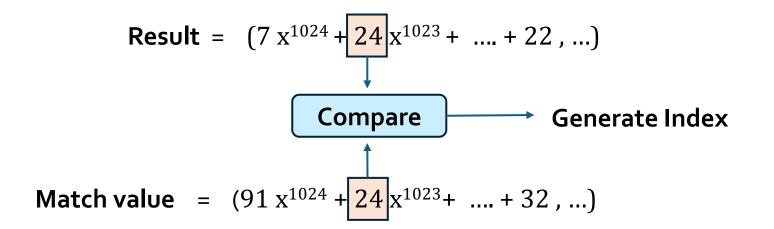
+

Encrypted Database Enc(m) =
$$(5 x^{1024} + 10 x^{1023} + + 19, ...)$$

Encrypted Query +Enc(~q) = $(2 x^{1024} + 14 x^{1023} + + 3, ...)$
Result = $(7 x^{1024} + 24 x^{1023} + + 22, ...)$
Decrypt
(1010...1x¹⁰²⁴ + 111...11 x¹⁰²³ + + 1001...0)
Match polynomial = 111...11 x¹⁰²⁴ + 111...11 x¹⁰²³ + + 111...11
Match value = $(91 x^{1024} + 24 x^{1023} + + 32, ...)$

Encrypted Database Enc(m) =
$$(5 x^{1024} + 10 x^{1023} + + 19, ...)$$

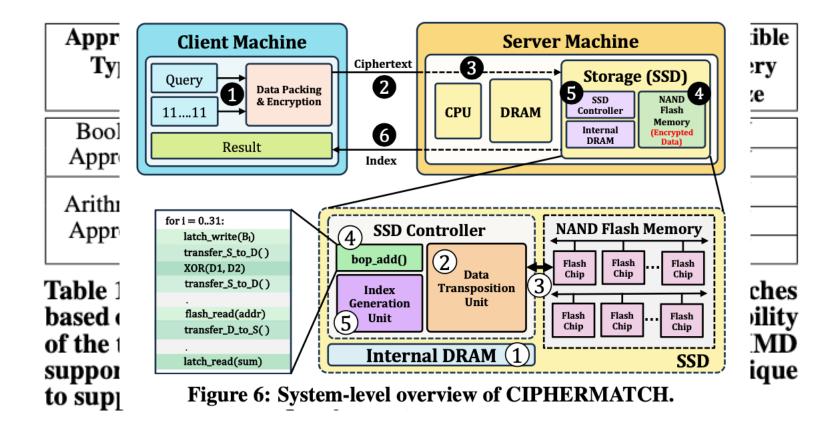
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Match polynomial = 111...11x¹⁰²⁴ + 111...11x¹⁰²³ + + 111...11
Encrypt
Match value = $(91 x^{1024} + 24 x^{1023} + + 32, ...)$



Compare the *match value*



Qualitative Analysis of Prior Work



System-Level Overview of CIPHERMATCH

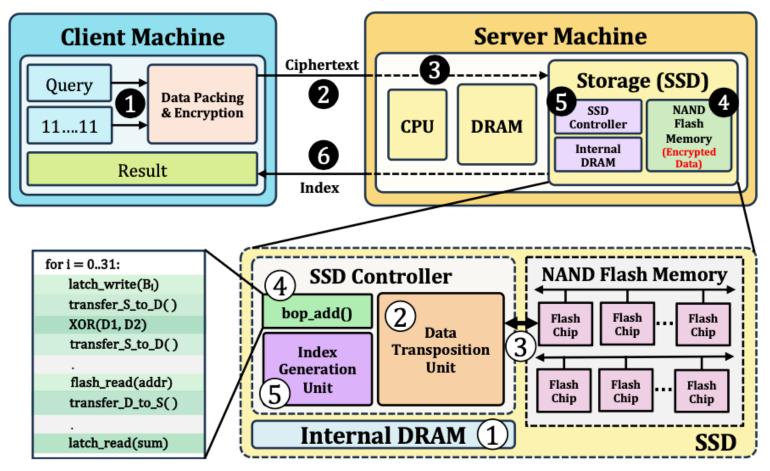


Figure 6: System-level overview of CIPHERMATCH.

Evaluation Configuration

CPU: Intel(R) Xeon(R) Gold 5118	Microarchitecture: Intel Skylake [149]x86-64 [150], 6 cores, out-of-order, 3.2 GHzL1 Data + Inst. Private Cache: 32kB, 8-way, 64B lineL2 Private Cache: 256kB, 4-way, 64B lineL3 Shared Cache: 8MB, 16-way, 64B line
Main Memory	32GB DDR4-2400, 4 channels
Storage (SSD)	Samsung 980 Pro PCIe 4.0 NVMe SSD 2 TB [102]
Operating System (OS)	Ubuntu 22.04.1 LTS

Table 2: Real CPU system configuration.

Evaluation Configuration

CM-PuM	32 GB DDR4-2400, 4 channel, 1 rank, 16 banks;
	Peak throughput: 19.2 GB/s
	Latency: T _{bbop} : 49 ns; Energy: E _{bbop} : 0.864 nJ;
	where <i>bbop</i> is bulk bitwise operation
	SSD External-Bandwidth: 7-GB/s external I/O bandwidth;
	(4-lane PCIe Gen4)
CM-IFP and CM-PuM-SSD	48-WL-layer 3D TLC NAND flash-based SSD; 2 TB
	SSD Internal DRAM: 2GB LPDDR4-1866 DRAM cache;
	1 channel, 1 rank, 8 banks
	NAND-Flash Channel Bandwidth: 1.2-GB/s Channel IO rate
	Controller Cores: ARM Cortex-R5 series @1.5GHz; 5 Cores [153]
	NAND Config: 8 channels; 8 dies/channel; 2 planes/die;
	2,048 blocks/plane; 196 (4×48) WLs/block; 4 KiB/page
	Latency: T _{read} (SLC mode): 22.5 µs [60]; T _{AND/OR} : 20 ns [62];
	$T_{latchtransfer}$: 20 ns [62]; T_{XOR} : 30 ns [60]; T_{DMA} : 3.3 μ s;
	T _{bit_add} (CM_IFP): 29.38µs
	Energy : E _{read} (SLC mode): 20.5µJ/channel [60];
	E _{AND/OR} : 10nJ/KB [62]; E _{latchtransfer} : 10nJ/KB [62];
	E _{XOR} : 20nJ/KB [60]; E _{DMA} : 7.656µJ/channel;
	E _{index_gen} (SSD controller): 0.18µJ/page size;
	E _{bit_add} (CM_IFP): 32.22µJ/channel

Table 3: Simulated system configurations.