PRIVATE INFORMATION RETRIEVAL

AND LATTICES:

ADVANCEMENTS AND FUTURE

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A Private Information Retrieval (PIR) scheme provides the ability for clients to **retrieve items from an online public (*) database of** *m elements*, without **revealing anything about their queries** to the untrusted host server(s)

- Parties:
 - a. Client(s)
 - b. Server (one or multiple)
- Steps:
 - Query
 - Response
 - Parse





Two types (sort of):

- 1. **Information-theoretic PIR:** client interacting with multiple non-colluding servers
- 2. **Computational-theoretic PIR**: client interacting with a single server, provides computational security based on cryptographic assumptions:
 - a. Stateless PIR:
 - The client does not store any (pre)information in order to launch queries
 - The schemes (a bunch!) perform worse than downloading the whole DB or they require computational overheads

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 - b. **Stateful PIR:** provides a "state" (or hint/digest) used as a "preprocessing" step amortised over *n* client queries

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 - b. Stateful PIR

Idea: encrypt the query instead of secret-sharing it

Limitations in **Computational-theoretic PIR**:

- Expensive pre-processing in terms of computation or communication
- High online-phase bandwidth consumption
- Lack of practical security parameters
- Lack of simple, open-source, available, verified implementations

Current look

□ Very active research area

□ Promising efficiency (computational/communicational/financial)

□ Variety of applications

Which applications?

Some deployments / related technologies exist:

- □ Brave (compromised credential-checking, TBD)
- Blyss (<u>https://github.com/blyssprivacy/sdk</u>)
- Google (<u>Device Enrollment</u>)
- Microsoft (<u>Password Monitor</u>)

More complex use-cases (not deployed):

- Approximate nearest-neighbor: <u>Brave News</u>
- Private search: <u>TipToe</u>
- □ Oblivious document ranking: <u>Coeus</u>

FRODOPIR

(but also *SimplePIR*) <u>https://eprint.iacr.org/2022/981</u> <u>https://eprint.iacr.org/2022/949</u>

Announcing FrodoPIR!



Just as the state of Sauron (its ring) moved to Frodo, we can move the mu and A to the client. The client then can then perform hidden queries to the server, just as Frodo remains hidden

Core ideas

- Built directly upon the learning with errors (LWE) problem *only* (similar to FrodoKEM)
 - Security relies on decisional LWE
 - Security is conservative (128 bits for 2^52 client queries): some parameters can be modified in order to make the scheme more efficient
- Highly configurable
 - Differences with SimplePIR: different pre-processing encoding, and the addition of a query pre-processing stage
- Tailored for efficiency and real-world applications







Notation

- DB is an array of *m* elements, each made up of *w* bits.
- Each entry is associated with the index *i* that corresponds to its position in the array.
- There are *C* clients that will each launch a maximum of *c* queries against DB.
- LWE:
 - a. *n* and *q* are the LWE dimension and modulus, respectively
 - b. ρ is the number of bits packed into each entry of the DB matrix (0 < ρ < q)
 - c. χ is the uniform ternary distribution over {-1, 0, 1}
 - d. λ is the concrete security parameter.
- $PRG(\mu, n, m, q)$ denotes a pseudorandom generator that expands a seed

 $\mu \in \{0,1\}^{\lambda} \in \mathbb{Z}_q^{x \times y}$

FrodoPIR (offline: server)

- Server setup: The server constructs their database containing *m* elements, and samples a seed $\mu \in \{0,1\}^{\lambda}$
- Server pre-processing: The server:
 - Derives a matrix $A \leftarrow PRG(\mu, n, m, q)$
 - Runs $\mathbf{D} \leftarrow \mathbf{parse}(DB, \rho)$
 - Runs $\ M \leftarrow A \cdot D$
 - Publishes the pair $(\mu, \mathbf{M}) \in \{0, 1\}^{\lambda} \times \mathbb{Z}_q^{n \times \omega}$

The "hint" is $\,M \leftarrow A \cdot D\,$

 $\mathbf{A} \in \mathbb{Z}_q^{n \times m}$ $\mathbf{D} \in \mathbb{Z}_q^{n \times \omega}$ $\omega = w/\log(\rho)$

FrodoPIR (offline: client)

Pre-processing. Each client:

- Downloads (μ, \mathbf{M})
- Derives $A \leftarrow PRG(\mu, n, m, q)$
- Samples *c* vectors:
 - $s_j \leftarrow \chi^n \qquad e_j \leftarrow \chi^m$
- Computes:

$$b_{j} \leftarrow s_{j}^{T} \cdot \mathbf{A} + e_{j}^{T} \qquad \in \mathbb{Z}_{q}^{m}$$
$$c_{j} \leftarrow s_{j}^{T} \cdot \mathbf{M} \qquad \in \mathbb{Z}_{q}^{\omega}$$

- Stores the pairs as the set $X = (b_j, c_j)_{j \in [c]}$

Essentially, computes c sets of preprocessed query parameters (optional step).

 $j \in [c]$

FrodoPIR (online: client)

Query generation. For the index *i* that the client wishes to query, the client generates a vector (the all-zero vector except where $f_i[i] = q/\rho$):

$$f_i = (0, \cdots, 0, q/\rho, 0, \cdots, 0)$$

It then pops a pair (b, c) from internal state and computes:

$$b' = b + f_i$$

The client uses a single set of preprocessed query parameters to produce an "encrypted" query vector, which is sent to the server

FrodoPIR (online: server)

Response. The server receives *b*' from the client, and responds with:

 $c' \leftarrow b' \cdot \mathbf{D}$

Essentially, the server responds by multiplying the vector with their DB matrix

 $\in \mathbb{Z}_a^{\omega}$

Post-processing. The client receives *c*', and calculates:

 $v \leftarrow \lfloor (c' - c)_{\rho} \rceil$

Essentially, the client get the value by "decrypting" using their pre-processed query parameters)

Efficiency. PIR schemes require a communication overhead smaller than the solution of having clients download *the entire server database*. In the stateful PIR case, it should hold when amortizing costs over the number of client queries.

Definition 5. (Efficiency) For a single client launching c queries, a PIR scheme is efficient if the total client communication overhead is smaller than |DB|.

Therefore, for stateful schemes, the total client communication cost is calculated using the equation: $comms(offline) + c \cdot comms(online)$.

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	Offline	Online
Client upload		$m\log(q)$
Client download	$128 + n\omega \log(q)$	$\omega \log(q)$

Efficiency.

	Offline	Online	
Client upload Client download	$128 + n\omega \log(q)$	$m\log(q)\ \omega\log(q)$	

Holds for large *c*: c > 18000 for $m = 2^{16}$

	Number of DB items $(\log(m))$	16	17	18	19	20
	Client download (KB)	5682.47	5682.47	5682.47	6313.07	6313.07
Offline	Database preprocessing (s)	92.409	185.30	374.56	825.50	1679.8
	Client derive params (s)	0.5208	1.042	2.1	4.29	8.39
	Client query preprocessing (s)	0.134	0.265	0.532	1.058	2.111
	Client query (KB)	256	512	1024	2048	4096
	Server response (KB)	3.203	3.203	3.203	3.556	3.556
Online	Client query (ms)	0.0177	0.0454	0.0813	0.1565	0.3328
	Server response (ms)	45.74	89.57	179.3	397.06	779.75
	Client output (ms)	0.418	0.4182	0.416	0.4559	0.4627

https://github.com/brave-experiments/frodo-pir

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Security: Indistinguishability of client queries. It assumes a semi-honest server that follows the protocol correctly and attempts to learn more based on the client queries they receive:

Server view: b' is distributed uniformly in \mathbb{Z}_q^m

under the assumption that decional-LWE is difficult to solve

• Regev encryption remains secure even when the same matrix *A* is used to encrypt many messages, provided that each ciphertext uses an independent secret vector *s* and error vector *e*

Chris Peikert, Vinod Vaikuntanathan, and Brent Waters. A framework for efficient and composable oblivious transfer.

Security: Indistinguishability of client queries.

We use the *decisional Matrix LWE problem*: extended form of the problem in which the secrets and errors are also matrices to prove *l*-query *indistinguishability* (with $l = poly(\lambda)$)

 A standard hybrid argument shows that any adversary that can distinguish the two distributions with advantage ε can be used to construct an efficient adversary breaking the decision LWE problem with advantage at least ε/l

J. W. Bos, C. Costello, L. Ducas, I. Mironov, M. Naehrig, V. Nikolaenko, A. Raghunathan, and D. Stebila. *Frodo: Take the ring! Practical, quantum-secure key exchange from LWE*.

 $\kappa = (log(\rho) * m) / (n * log(q))$ denote the improvement factor relative to the offline client download, compared to the DB size.

128 bits of security for 2^{52} queries: security can increase by increasing the dimension but this will impact κ

q	2^{32}	2^{32}	2^{32}	2^{32}	2^{32}
n	1774	1774	1774	1774	1774
m	2^{16}	2^{17}	2^{18}	2^{19}	2^{20}
ρ	2^{10}	2^{10}	2^{10}	2^{9}	2^9
ĸ	13.028	26.056	52.112	93.802	187.603
λ	128	128	128	128	128

How we come up with it?

- Based on Regev's encryption where the expansion factor is roughly $F = n \approx$ 1024, where *n* is the lattice security parameter
 - Do a lot of the work in advance and re-use
 - Additively homomorphic

$$(a,c) = (a, a^T s + e + \lfloor q/\rho \rfloor \cdot \mu) \in \mathbb{Z}_q \times \mathbb{Z}_q$$

• We use a ternary uniform distribution (bounded by $4\sqrt{m}$, with *m* being the number of samples taken from the distribution)

Z. Brakerski, A. Langlois, C. Peikert, O. Regev, and D. Stehle. *Classical hardness of learning with errors*.

What are the advantages?

1. It is simple: easy to explain, easy to push to production

2. LWE-based PIR schemes are simpler to implement: they require no polynomial arithmetic or fast Fourier transforms

3. LWE-based PIR schemes do not require the server to store any extra per-client state. In contrast, many schemes based on Ring LWE rely on optimizations that require the server to store one "key-switching hint" for each client

4. LWE-based PIR schemes are faster and cheaper: the encryption scheme needs to be linearly (not fully) homomorphic, so we can use smaller and more efficient lattice parameters

What is still limiting?

- The pre-processing stage is still dependent on the database size
 - \circ Can be reduce, but it is still dependent from the size of element in the database
- Many steps are dependent on A
 - We make it independent for client with $\mu \in \{0,1\}^{\lambda}$
 - But it is still costly during client parsing

But, is this enough?

- Databases are not structured in this simple way
 - They are indexed by keywords
 - They are structured as JSON, Graphs, Excel spreadsheets
- The queries we are interested in are not simple:
 - Complex queries with AND/OR statements
 - Combination of database systems
 - Approximate nearest neighbor (ANN) elements
- Databases are constantly updated
- Is the security we assume enough?
 - What about malicious security?
 - What about private databases?

Not all systems are created equally

VERIPIR

Leo de Castro, Keewoo Lee https://eprint.iacr.org/2024/341

Important security properties

Authenticated/verifiable/malicious PIR:

• The DB "hint" (a commitment) is accompanied by a non-interactive proof-of-knowledge of the DB, and then every answer from the server is verified against this proof to ensure that it is consistent with the commitment

Important security properties

Authenticated/verifiable/malicious PIR:

- The DB "hint" (a commitment) is accompanied by a non-interactive proof-of-knowledge of the DB, and then every answer from the server is verified against this proof to ensure that it is consistent with the commitment
 - Use extractable SIS-based commitments

Carsten Baum, Jonathan Bootle, Andrea Cerulli, Rafael del Pino, Jens Groth, and Vadim Lyubashevsky. *Sub-linear lattice-based zeroknowledge arguments for arithmetic circuits*.

SIS-based commitments

- Extractable SIS-based commitments:
 - Without the "zero-knowledge":



Carsten Baum, Jonathan Bootle, Andrea Cerulli, Rafael del Pino, Jens Groth, and Vadim Lyubashevsky. *Sub-linear lattice-based zeroknowledge arguments for arithmetic circuits*.

SIS-based commitments

• Extractable SIS-based commitments:



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Limitations and future

- Gives a solid base but...
- Assumes a public, honest digest
 - Can we expand it to symmetric PIR?
 - Future work! (all with lattices and, hence, post-quantum)

WHAT ELSE?

So, you want to research on this?

- Expand the security model:
 - How does leakage impact it?
 - Is it attackable?
- Introduce 'updatable' techniques
- Look at other applications of DB:
 - Do we fulfil them?
- How do we deal with variable-length elements?
 - Is padding enough?
- Can we make it simple with the ring?
- Can we look at state-of-the-art data structures/graphs/matrix theory?

Thank you Henry Corrigan-Gibbs, Alex Davidson, Alexandra Henzinger, Stefano Tessaro, Eli Richarson, Daniel Escudero, Luiza Barros, Abdelraham Aly for input and discussing all of this!

Building steps

- Keyword-based PIR:
 - "Call Me By My Name: Simple, Practical Private Information Retrieval for Keyword Queries": <u>https://eprint.iacr.org/2024/092</u>
 - "Binary Fuse Filters: Fast and Smaller Than Xor Filters": <u>https://arxiv.org/abs/2201.01174</u>
- Security:
 - "Fully Malicious Authenticated PIR": <u>https://eprint.iacr.org/2023/1804</u>
 - "VeriSimplePIR: Verifiability in SimplePIR at No Online Cost for Honest Servers": <u>https://eprint.iacr.org/2024/341</u>
- Complex queries:
 - "Private Web Search with Tiptoe": <u>https://eprint.iacr.org/2023/1438</u>
 - "Coeus: A System for Oblivious Document Ranking and Retrieval": <u>https://eprint.iacr.org/2022/154</u>
- Updatability:
 - "Incremental Offline/Online PIR"

https://www.cis.upenn.edu/~sga001/papers/incpir-sec22.pdf

An announcement

PIR workshop at PETS: <u>https://github.com/private-retrieval/wip</u>





https://www.womenincryptography.com/

https://criptolatino.org/

THANK YOU!

@claucece <u>www.sofiaceli.com</u>

CHALAMET-PIR

(one solution) https://eprint.iacr.org/2024/092

Core ideas

• Very simple (®) idea



Core ideas

- 1. Have a DB structured as a Key-Value (KV) map (size *m*, where each element *v* is indexed by a key *k*)
- 2. Convert this map into a filter (*F*) structure (think on a Bloom Filter) with a set of *k* hash functions and some false positive probability
 - a. The filter has a function that allows to recover *v*: $fpt_{\epsilon}(v) \leftarrow F.check(k)$
 - b. The filter is broken into d columns: interpret it as a matrix with cm (*) rows
- 3. Query for an element with a long vector where there are 1s on $h_i(k)$

Basic construction

- Same ideas as previous in literature, but:
 - We leverage the usage of *Binary Fuse Filters*
 - Minimise the space and query overheads of key-value filters, while maintaining quick access times
 - Reconstruct using XOR
 - Divide the filter into many more segments
 - We can use *any* LWE-based PIR scheme

https://lib.rs/crates/haveibeenpwned

https://sts10.github.io/2023/01/11/playing-with-binary-fuse-filters.html

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	$2^{16} \times 1024$ B	0.010597	6.5508	0.22001
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	$2^{19} \times 1024$ B	0.14442	54.053	0.24204
	$2^{20} \times 1024 \mathrm{B}$	0.24049	116.89	0.24384
	$2^{16} \times 1024$ B	0.050048	37.830	0.47251
	$2^{17} \times 1024 \mathrm{B}$	0.1787	74.733	0.47046
EC2 "t2.t2xlarge"	$2^{18}\times1024\mathrm{B}$	0.19739	143.82	0.46782
	$2^{19} \times 1024 \mathrm{B}$	0.4219	319.82	0.50735
	$2^{20} \times 1024 \mathrm{B}$	0.8471	634.21	0.56381
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	$2^{14} \times 100 \mathrm{kB}$	0.0040465	760.64	35.485
Table 2: Online pe LWEPIR = FrodoP	$2^{14} \times 100 \text{ kB}$ erformance IR, $k = 3$). I	0.0040465 (millisecon Response is	760.64 ds) of Cha s a server	35.4 alametl operat

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Table 2: Online performance (milliseconds) of ChalametPIR (LWEPIR = FrodoPIR, k = 3). Response is a server operation, while Query and Parsing are run by the client.

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Properties

- Security: Same as FrodoPIR (LWE-based), but:
 - We allow for false-positives, as we assume a public database. What impact does this have?
 - We provide a random value in case of non-inclusion -> leakage impact
- Efficiency: Same as FrodoPIR (LWE-based), but:
 - Blow-up due to filter: ç

- Is it sufficient?
 - Assumes the same length of elements