

**Camilla Hollanti**

Department of Mathematics  
and Systems Analysis  
Aalto University, Finland

# Private Information Retrieval:

Chasing Capacity  
with Algebraic Codes

Simons Institute  
March 8 2024



Institute for Advanced Study  
Technical University of Munich



**Aalto University**  
School of Science

Joint with/Thanks go to:

R.Freij-Hollanti, O.Gnilke, L.Holzbaur  
D.Karpuk, J.Li, R.Tajeddine, A.Wachter-Zeh,...

# Presentation outline

---

- Introduction to private information retrieval (PIR)
- PIR capacity: known results and our conjecture ( $\approx$  theorem)
- Star product PIR
- Full support-rank PIR and that theorem
- Current and future directions (beyond PIR)

# Collaborators

---



Ragnar  
Aalto



Dave  
WithSecure



Oliver  
Aalborg



Razane  
Helsinki



Lukas  
Infineon



Jie  
Huawei



Antonia  
TU Munich



# Private information retrieval (PIR)

---

- With PIR, a user is able to download one out of  $m$  files  $\{x^1, \dots, x^m\}$  from a database without revealing the identity  $i \in [m]$  of the file to the database holder.

## Theorem

*If the data is stored on only one server, perfect privacy cannot be achieved except by downloading the entire data.*

- Distributed storage system (DSS) with  $n$  servers...
- ...encoded with an  $[n, k]$  (MDS) code!

# Short history of PIR

---

- PIR from replicated databases was introduced by Chor in 1995 and was an active topic thereafter [Cho+95; Bei+02; Dvi+16].  
→ A sequence of papers reduced the communication cost to be sub-linear in  $m$ .

# Short history of PIR

---

- PIR from replicated databases was introduced by Chor in 1995 and was an active topic thereafter [Cho+95; Bei+02; Dvi+16].  
→ A sequence of papers reduced the communication cost to be sub-linear in  $m$ .
- More recently, a lot of renewed interest towards PIR from coded storage systems from various perspectives [Aug+14; Sha+14; Faz+15].

# Short history of PIR

---

- PIR from replicated databases was introduced by Chor in 1995 and was an active topic thereafter [Cho+95; Bei+02; Dvi+16].
  - A sequence of papers reduced the communication cost to be sub-linear in  $m$ .
- More recently, a lot of renewed interest towards PIR from coded storage systems from various perspectives [Aug+14; Sha+14; Faz+15].
- 100s of papers since 2015; hits on IEEE Xplore:
  - 1995–2004: 7
  - 2005–2014: 72
  - 2015–2024: 393

## A toy example with $[n, k]_q = [3, 2]_2$ code

---

$$\begin{array}{c} x_1^1 \\ x_1^2 \\ x_1^3 \end{array}$$

$$\begin{array}{c} x_2^1 \\ x_2^2 \\ x_2^3 \end{array}$$

$$\begin{array}{c} x_1^1 + x_2^1 \\ x_1^2 + x_2^2 \\ x_1^3 + x_2^3 \end{array}$$

- $m = 3$  files. Want  $x^1$ .  
Server  $j \in [n]$  stores  
 $y_j^i = (x^i G)_j, i \in [n]$ .





## A toy example with $[n, k]_q = [3, 2]_2$ code

---

$$\begin{array}{c} x_1^1 \\ x_1^2 \\ x_1^3 \end{array}$$

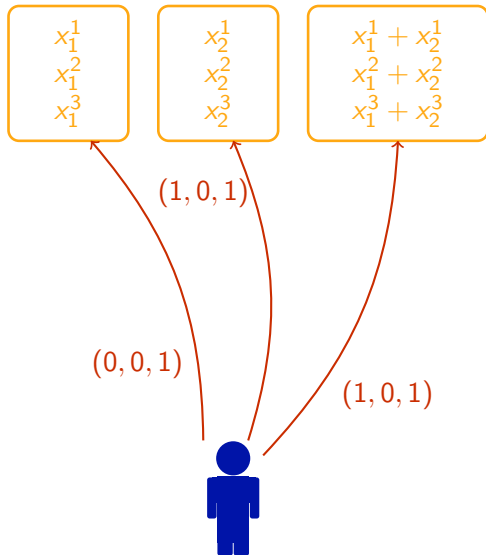
$$\begin{array}{c} x_2^1 \\ x_2^2 \\ x_2^3 \end{array}$$

$$\begin{array}{c} x_1^1 + x_2^1 \\ x_1^2 + x_2^2 \\ x_1^3 + x_2^3 \end{array}$$

- $m = 3$  files. Want  $x^1$ .  
Server  $j \in [n]$  stores  
 $y_j^i = (x^i G)_j, i \in [n]$ .
- 1st round: Choose  
**random**  $u = (1, 0, 1)$ .

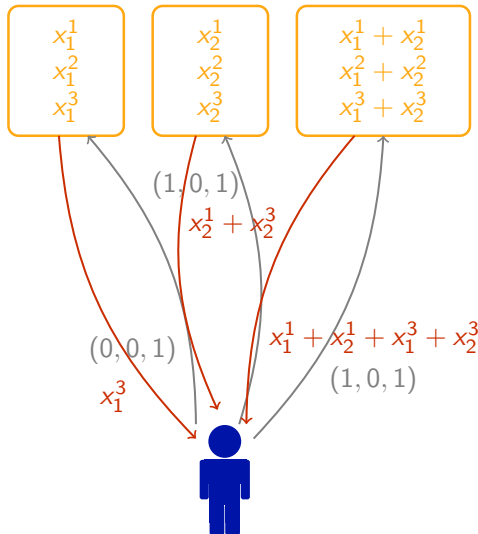


## A toy example with $[n, k]_q = [3, 2]_2$ code



- $m = 3$  files. Want  $x^1$ . Server  $j \in [n]$  stores  $y_j^i = (x^i G)_j, i \in [n]$ .
- 1st round: Choose **random**  $u = (1, 0, 1)$ .
- Send a query  $q_1 = u + e_1 = (0, 0, 1)$  to the **1st** server, and  $q_2 = q_3 = u = (1, 0, 1)$  to the other servers.

## A toy example with $[n, k]_q = [3, 2]_2$ code



- $m = 3$  files. Want  $x^1$ . Server  $j \in [n]$  stores  $y_j^i = (x^i G)_j, i \in [n]$ .
- 1st round: Choose **random**  $u = (1, 0, 1)$ .
- Send a query  $q_1 = u + e_1 = (0, 0, 1)$  to the **1st** server, and  $q_2 = q_3 = u = (1, 0, 1)$  to the other servers.
- Servers respond  $r_j = \langle y_j, q_j \rangle$ .
- Decode  $x_1^1 = \sum_{j=1}^n r_j$ .

## A toy example with $[n, k]_q = [3, 2]_2$ code

---

$$\begin{array}{c} x_1^1 \\ x_1^2 \\ x_1^3 \end{array}$$

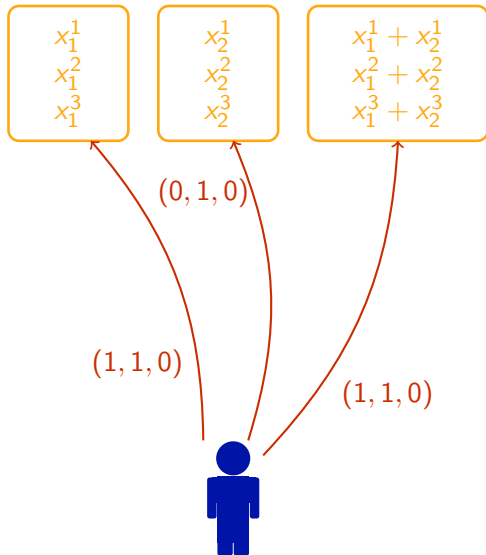
$$\begin{array}{c} x_2^1 \\ x_2^2 \\ x_2^3 \end{array}$$

$$\begin{array}{c} x_1^1 + x_2^1 \\ x_1^2 + x_2^2 \\ x_1^3 + x_2^3 \end{array}$$

- $m = 3$  files. Want  $x^1$ .  
Server  $j \in [n]$  stores  $y_j^i = (x^i G)_j$ ,  $i \in [n]$ .
- 2nd round: Choose **random**  $u = (1, 1, 0)$ .

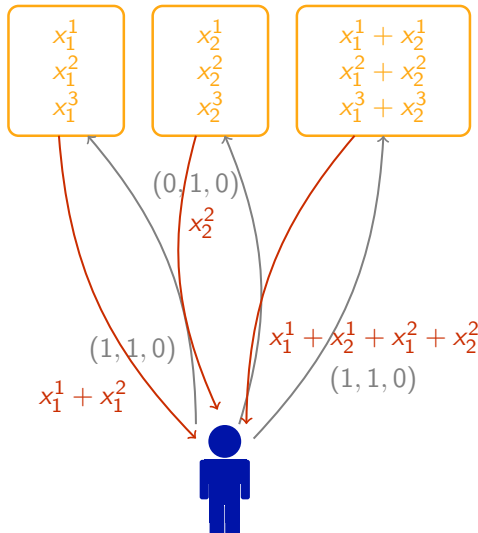


## A toy example with $[n, k]_q = [3, 2]_2$ code



- $m = 3$  files. Want  $x^1$ . Server  $j \in [n]$  stores  $y_j^i = (x^i G)_j, i \in [n]$ .
- 2nd round: Choose **random**  $u = (1, 1, 0)$ .
- $q_2 = u + e_1 = (0, 1, 0)$  to the **2nd** server,  $q_1 = q_3 = u$ .

## A toy example with $[n, k]_q = [3, 2]_2$ code



- $m = 3$  files. Want  $x^1$ . Server  $j \in [n]$  stores  $y_j^i = (x^i G)_j$ ,  $i \in [n]$ .
- 2nd round: Choose **random**  $u = (1, 1, 0)$ .
- $q_2 = u + e_1 = (0, 1, 0)$  to the **2nd** server,  $q_1 = q_3 = u$ .
- $r_j = \langle y_j, q_j \rangle$ .
- $x_2^1 = \sum_{j=1}^n r_j$ .
- Privacy holds if no *collusion*.

## Collusion and $t$ -PIR

---

Servers in a *colluding set* may exchange their obtained queries in order to reveal the identity of the desired file.

### Definition

A PIR scheme *protects against* the colluding set  $J \subseteq [n]$ , if the projection of the overall query  $Q^i$  to  $J$  does not depend on the desired file  $i \in [m]$ .

## Collusion and $t$ -PIR

---

Servers in a *colluding set* may exchange their obtained queries in order to reveal the identity of the desired file.

### Definition

A PIR scheme *protects against* the colluding set  $J \subseteq [n]$ , if the projection of the overall query  $Q^i$  to  $J$  does not depend on the desired file  $i \in [m]$ .

### Definition

A  $t$ -PIR scheme protects against any colluding set of size  $\leq t$ .



# Capacity of PIR

---

- The rate  $\mathcal{R}$  of a PIR scheme is defined as

$$\mathcal{R} = \frac{\text{file size}}{\text{download size}}$$

# Capacity of PIR

---

- The rate  $\mathcal{R}$  of a PIR scheme is defined as

$$\mathcal{R} = \frac{\text{file size}}{\text{download size}}$$

- The capacity  $\mathcal{C}$  of PIR is the maximum possible rate for a given model.

# Capacity of PIR

---

- The rate  $\mathcal{R}$  of a PIR scheme is defined as

$$\mathcal{R} = \frac{\text{file size}}{\text{download size}}$$

- The capacity  $\mathcal{C}$  of PIR is the maximum possible rate for a given model.
- We call a scheme asymptotically capacity achieving if

$$\mathcal{R} = \lim_{m \rightarrow \infty} \mathcal{C}.$$

# PIR capacity and constructions

---

- Several capacity results and scheme constructions have been reported (non-exhaustive list!):
  - replication [Sun+17; Tia+19]
  - MDS-coded storage [Ban+18; Zhu+19]
  - colluding servers [Sun+18b]
  - **MDS and colluding** [Fre+17; Taj+18; D'O+18]

# PIR capacity and constructions

---

- Several capacity results and scheme constructions have been reported (non-exhaustive list!):
  - replication [Sun+17; Tia+19]
  - MDS-coded storage [Ban+18; Zhu+19]
  - colluding servers [Sun+18b]
  - **MDS and colluding** [Fre+17; Taj+18; D'O+18]
  - symmetric PIR (SPIR) [Wan+17b; Wan+17a; Wan+17c]
  - single-server PIR with side information [Hei+18]
  - non-MDS storage [Fre+19; Kum+19]
  - ...
  - graph-based PIR [Sad+23]

# Capacity of PIR

- A storage system with  $m$  files has the following (**conjectured**, **mostly proven**) capacities:

	replication	$[n, k]$ -coded
no collusion	$\dagger \frac{1-1/n}{1-(1/n)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{1}{n}$	$\P \frac{1-k/n}{1-(k/n)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k}{n}$
$t$ -collusion	$\ddagger \frac{1-t/n}{1-(t/n)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{t}{n}$	$\S \frac{1-\frac{t+k-1}{n}}{1-(\frac{t+k-1}{n})^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{t+k-1}{n}$

<sup>†</sup> Sun–Jafar [Sun+17].

<sup>¶</sup> Banawan–Ulukus [Ban+18].

<sup>‡</sup> Sun–Jafar [Sun+18b].

<sup>§</sup> Freij–Hollanti *et al.* [Fre+17]; Sun–Jafar [Sun+18a]; Holzbaur *et al.* [Hol+22].

# Capacity of PIR

- A storage system with  $m$  files has the following (**conjectured**, **mostly proven**) capacities:

	replication	$[n, k]$ -coded
no collusion	$\dagger \frac{1-1/n}{1-(1/n)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{1}{n}$	$\P \frac{1-k/n}{1-(k/n)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k}{n}$
$t$ -collusion	$\ddagger \frac{1-t/n}{1-(t/n)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{t}{n}$	$\S \frac{1-\frac{t+k-1}{n}}{1-(\frac{t+k-1}{n})^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{t+k-1}{n}$

<sup>†</sup> Sun–Jafar [Sun+17].

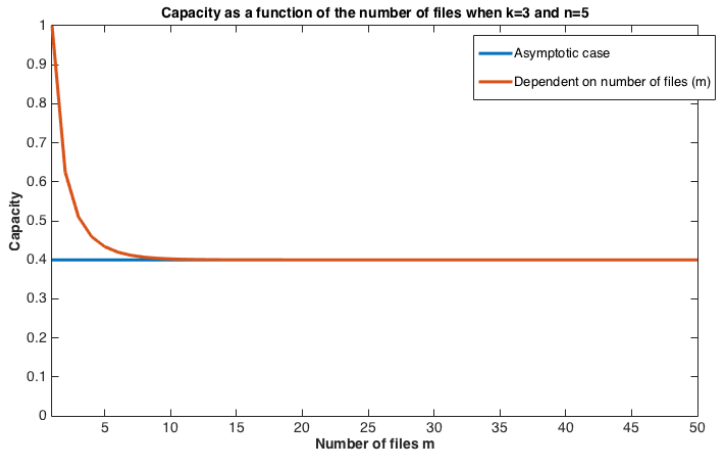
<sup>¶</sup> Banawan–Ulukus [Ban+18].

<sup>‡</sup> Sun–Jafar [Sun+18b].

<sup>§</sup> Freij–Hollanti *et al.* [Fre+17]; Sun–Jafar [Sun+18a]; Holzbaur *et al.* [Hol+22].

# Fast convergence: coded case ( $k > 1, t = 1$ )

---





# Capacity of uncoded PIR

## Theorem (Sun–Jafar 2016)

The capacity of replicated  $n$  server PIR for a storage system containing  $m$  files is given by

$$\left(1 + \frac{1}{n} + \cdots + \frac{1}{n^{m-1}}\right)^{-1} = \frac{1 - \frac{1}{n}}{1 - \frac{1}{n^m}} \xrightarrow{m \rightarrow \infty} 1 - \frac{1}{n}$$

Proof (sketch):

- Information theoretic argument to provide an upper bound.
- Scheme that achieves this bound.

## Sun–Jafar construction for replicated data

---

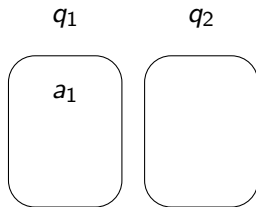
- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

## Sun-Jafar construction for replicated data

---

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

→ Download  $a_1$  from the 1st server.

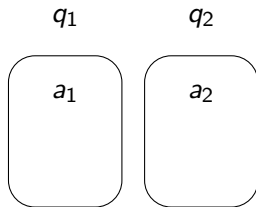


## Sun-Jafar construction for replicated data

---

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

- Download  $a_1$  from the 1st server.
- Symmetrize over both servers

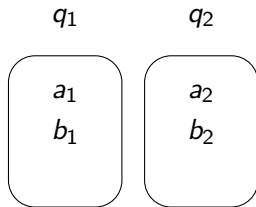


## Sun-Jafar construction for replicated data

---

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

- Download  $a_1$  from the 1st server.
- Symmetrize over both servers
- Symmetrize over files.

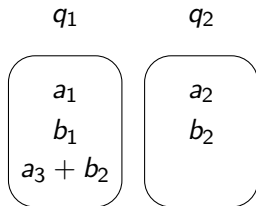


## Sun-Jafar construction for replicated data

---

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

- Download  $a_1$  from the 1st server.
- Symmetrize over both servers
- Symmetrize over files.
- Use  $b_i$  as interference.

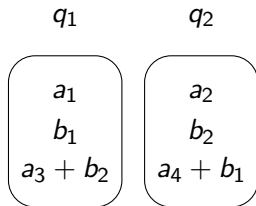


## Sun-Jafar construction for replicated data

---

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

- Download  $a_1$  from the 1st server.
- Symmetrize over both servers
- Symmetrize over files.
- Use  $b_i$  as interference.
- Symmetrize again.

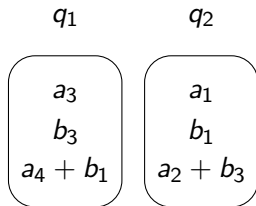


## Sun-Jafar construction for replicated data

---

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

- Download  $a_1$  from the 1st server.
- Symmetrize over both servers
- Symmetrize over files.
- Use  $b_i$  as interference.
- Symmetrize again.
- Random permutation provides privacy.

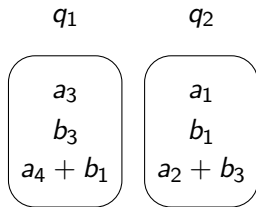




## Sun-Jafar construction for replicated data

- $m = 2$  files  $a$  and  $b$ , stored on  $n = 2$  servers via replication.
- “Sub-packetize” files into  $n^m = 4$  (!!!) symbols  $a_1, a_2, a_3, a_4$  and  $b_1, b_2, b_3, b_4$ . Assume a user wants  $a$ .
- A general retrieval strategy is as follows:

- Download  $a_1$  from the 1st server.
- Symmetrize over both servers
- Symmetrize over files.
- Use  $b_i$  as interference.
- Symmetrize again.
- Random permutation provides privacy.



- Rate =  $\frac{4}{6} = \frac{2}{3}$ .

## Proof of converse (sketch)

---

### Theorem (Sun-Jafar, 2016)

*The rate of any replicated PIR scheme on  $n$  servers with  $m$  files satisfies*

$$\mathcal{R} = \frac{\text{file size}}{\text{download size}} \leq \frac{1 - \frac{1}{n}}{1 - \frac{1}{n^m}}.$$

## Proof of converse (sketch)

---

### Theorem (Sun-Jafar, 2016)

*The rate of any replicated PIR scheme on  $n$  servers with  $m$  files satisfies*

$$\mathcal{R} = \frac{\text{file size}}{\text{download size}} \leq \frac{1 - \frac{1}{n}}{1 - \frac{1}{n^m}}.$$

In other words,

$$\frac{\text{\#symbols downloaded}}{\text{\#desired symbols retrieved}} \geq 1 + \frac{1}{n} + \frac{1}{n^2} + \cdots + \frac{1}{n^{m-1}}$$

## Proof of converse (sketch)

---

### Theorem (Sun-Jafar, 2016)

*The rate of any replicated PIR scheme on  $n$  servers with  $m$  files satisfies*

$$\mathcal{R} = \frac{\text{file size}}{\text{download size}} \leq \frac{1 - \frac{1}{n}}{1 - \frac{1}{n^m}}.$$

In other words,

$$\frac{\text{\#symbols downloaded}}{\text{\#desired symbols retrieved}} \geq 1 + \frac{1}{n} + \frac{1}{n^2} + \cdots + \frac{1}{n^{m-1}}$$

File size  $:= L$ . *Claim:*

$$\text{\#symbols downloaded} \geq L + \frac{L}{n} + \frac{L}{n^2} + \cdots + \frac{L}{n^{m-1}}.$$

## Proof of converse (sketch)

---

Proof idea (induction over  $m$ ):

- Need to download  $\geq L$  symbols of the file we want.  
Clear if  $m = 1$ .

## Proof of converse (sketch)

---

Proof idea (induction over  $m$ ):

- Need to download  $\geq L$  symbols of the file we want.  
Clear if  $m = 1$ .
- Induction assumption ( $m$  files): We must download in total

$$D_m := L + \frac{L}{n} + \frac{L}{n^2} + \cdots + \frac{L}{n^{m-1}}$$

symbols from the files  $1, \dots, m$ .

## Proof of converse (sketch)

---

Proof idea (induction over  $m$ ):

- Need to download  $\geq L$  symbols of the file we want.  
Clear if  $m = 1$ .
- Induction assumption ( $m$  files): We must download in total

$$D_m := L + \frac{L}{n} + \frac{L}{n^2} + \cdots + \frac{L}{n^{m-1}}$$

symbols from the files  $1, \dots, m$ .

- From some server, need

$$\geq \frac{D_m}{n} = \frac{1}{n} \left( L + \frac{L}{n} + \frac{L}{n^2} + \cdots + \frac{L}{n^{m-1}} \right)$$

symbols from the files  $1, \dots, m$ .

## Proof of converse (sketch)

---

- Induction step: Assume we want file  $m + 1$ .
- Then by induction assumption, from some server need  $\frac{D_m}{n}$  symbols from the files  $1, \dots, m$ .



## Proof of converse (sketch)

---

- Induction step: Assume we want file  $m + 1$ .
- Then by induction assumption, from some server need  $\frac{D_m}{n}$  symbols from the files  $1, \dots, m$ .
- In addition, need to download  $L$  symbols from file  $m + 1$ .
- Total download:

$$L + \frac{D_m}{n} = L + \left( \frac{L}{n} + \frac{L}{n^2} + \frac{L}{n^3} + \dots + \frac{L}{n^m} \right) = D_{m+1}$$



The proofs for the coded storage case and for the colluding case are similar. Combine to get **coded AND colluding** case? Hard!

## Conjecture for PIR capacity with $t > 1, k > 1$

---

**Conjecture** [Fre+17, Conj. 1] Let  $C$  be a linear  $[n, k, d]$  code. Consider  $m$  files and let  $1 \leq t \leq n - k$ . Any  $t$ -PIR scheme has rate

$$\mathcal{R} \leq \frac{1 - \frac{k+t-1}{n}}{1 - \left(\frac{k+t-1}{n}\right)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k+t-1}{n}.$$

## Conjecture for PIR capacity with $t > 1, k > 1$

---

**Conjecture** [Fre+17, Conj. 1] Let  $C$  be a linear  $[n, k, d]$  code. Consider  $m$  files and let  $1 \leq t \leq n - k$ . Any  $t$ -PIR scheme has rate

$$\mathcal{R} \leq \frac{1 - \frac{k+t-1}{n}}{1 - (\frac{k+t-1}{n})^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k+t-1}{n}.$$

- Disproved in [Sun+18a] for  $m = 2, k = t = 2, n = 4$ .  
The PIR rate is  $3/5$ , while the conjecture states  $4/7$ .

## Conjecture for PIR capacity with $t > 1, k > 1$

---

**Conjecture** [Fre+17, Conj. 1] Let  $C$  be a linear  $[n, k, d]$  code. Consider  $m$  files and let  $1 \leq t \leq n - k$ . Any  $t$ -PIR scheme has rate

$$\mathcal{R} \leq \frac{1 - \frac{k+t-1}{n}}{1 - \left(\frac{k+t-1}{n}\right)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k+t-1}{n}.$$

- Disproved in [Sun+18a] for  $m = 2, k = t = 2, n = 4$ . The PIR rate is  $3/5$ , while the conjecture states  $4/7$ .
- The query scheme in the counter-example is not full support-rank!
- Proved for full support-rank schemes in [Hol+22].
- *How*, and what is “full support-rank”?

# Codes from star products

---

- For two vectors  $x, y \in \mathbb{F}_q^n$ , define the *star product*

$$x \star y := (x_1 y_1, \dots, x_n y_n).$$

- Let  $C$  and  $D$  be linear codes in  $\mathbb{F}_q^n$ . Define the *star product code* as the linear span

$$C \star D := \langle \{c \star d \mid c \in C, d \in D\} \rangle.$$

---

<sup>†</sup>Mirandola–Zémor [Mir+15]: Apart from pairs  $C, C^\perp$  and their products, the only pairs that get to this bound are *generalized Reed–Solomon (GRS) codes*.

# Codes from star products

---

- For two vectors  $x, y \in \mathbb{F}_q^n$ , define the *star product*

$$x \star y := (x_1 y_1, \dots, x_n y_n).$$

- Let  $C$  and  $D$  be linear codes in  $\mathbb{F}_q^n$ . Define the *star product code* as the linear span

$$C \star D := \langle \{c \star d \mid c \in C, d \in D\} \rangle.$$

- *Product Singleton Bound*<sup>†</sup>:

$$d_{C \star D} \leq n - \dim(C) - \dim(D) + 2$$

---

<sup>†</sup>Mirandola–Zémor [Mir+15]: Apart from pairs  $C, C^\perp$  and their products, the only pairs that get to this bound are *generalized Reed–Solomon (GRS) codes*.

# Star-product PIR scheme

---

- We proposed a fully general coded retrieval scheme protecting against  $t$ -collusion [Fre+17]<sup>†</sup>.
- Asymptotically capacity achieving at the known points ( $k = 1$ ,  $t = 1$ ), when employed with GRS codes.
- Also achieves the (asymptotic) capacity of the above conjecture.

---

<sup>†</sup>You can find a couple of extra slides for details after the thank-you slide.

# Star-product PIR scheme

---

- We proposed a fully general coded retrieval scheme protecting against  $t$ -collusion [Fre+17]<sup>†</sup>.
- Asymptotically capacity achieving at the known points ( $k = 1$ ,  $t = 1$ ), when employed with GRS codes.
- Also achieves the (asymptotic) capacity of the above conjecture.
- **Novelty:**
  - *Earlier:*  $n$  queries from the entire space  $\mathbb{F}_q^m$ .
  - *Star product scheme:*  $m$  queries from an  $[n, t]$  code  $D \subseteq \mathbb{F}_q^n$ .  
→ smart (star-product) interplay of the  $[n, k]$  storage code  $C$  and the query code  $D$ .

---

<sup>†</sup>You can find a couple of extra slides for details after the thank-you slide.



## Rate vs. capacity

---

- What the user receives is a codeword in  $C \star D$  with errors in known positions.
- These errors can be treated as erasures and we know that the code  $C \star D$  can correct up to  $d_{C \star D} - 1$  erasures.

## Rate vs. capacity

---

- What the user receives is a codeword in  $C \star D$  with errors in known positions.
- These errors can be treated as erasures and we know that the code  $C \star D$  can correct up to  $d_{C \star D} - 1$  erasures.
- From the *product singleton bound* we know that

$$d_{C \star D} \leq n - \dim(C) - \dim(D) + 2$$

## Rate vs. capacity

---

- What the user receives is a codeword in  $C \star D$  with errors in known positions.
- These errors can be treated as erasures and we know that the code  $C \star D$  can correct up to  $d_{C \star D} - 1$  erasures.
- From the *product singleton bound* we know that

$$d_{C \star D} \leq n - k - t + 2$$

## Rate vs. capacity

---

- What the user receives is a codeword in  $C \star D$  with errors in known positions.
- These errors can be treated as erasures and we know that the code  $C \star D$  can correct up to  $d_{C \star D} - 1$  erasures.
- From the *product singleton bound* we know that

$$d_{C \star D} \leq n - k - t + 2$$

- For GRS pairs  $C, D$  this gives us a rate of

$$\mathcal{R} = \frac{n - k - t + 1}{n} = 1 - \frac{k + t - 1}{n}$$

## Rate vs. capacity

---

- What the user receives is a codeword in  $C \star D$  with errors in known positions.
- These errors can be treated as erasures and we know that the code  $C \star D$  can correct up to  $d_{C \star D} - 1$  erasures.
- From the *product singleton bound* we know that

$$d_{C \star D} \leq n - k - t + 2$$

- For GRS pairs  $C, D$  this gives us a rate of

$$\mathcal{R} = \frac{n - k - t + 1}{n} = 1 - \frac{k + t - 1}{n}$$

- Asymptotically capacity (and conjecture) achieving:

## Capacity of $\star$ -product schemes

---

Any “strongly linear” scheme can be replaced by a star product scheme for the same privacy model, without losing in the PIR rate:

### Theorem ([Hol+22])

*Consider a strongly linear PIR scheme from a storage code  $C$  and a query scheme as above. Then the rate is bounded by*

$$\mathcal{R} \leq 1 - \frac{k + t - 1}{n}$$

*for any number of files  $m$ .*

This bound coincides with the asymptotic capacity conjecture, which is achieved for any number of files with star product PIR.

# Full support-rank PIR

- Clearly, it is suboptimal to send linearly dependent queries to servers.
- However, submatrices of the query matrix may be dependent [Sun+18a], *i.e.*, have supported columns that are linearly dependent.
- The technical assumption of full support-rank restricts all supported columns  $\mathcal{T}$ ,  $|\mathcal{T}| \leq t$ , to be independent:

## Definition

A linear PIR scheme is of *full support-rank* if for every query realization  $q \in \mathbb{F}^{\alpha m \times \beta n}$ , any subset  $\mathcal{T} \subseteq [n]$  of  $|\mathcal{T}| \leq t$  servers, and any file index  $i \in [m]$

$$\text{rank}(q[\psi_\alpha(i), \psi_\beta(\mathcal{T})]) = |\text{colsupp}(q[\psi_\alpha(i), \psi_\beta(\mathcal{T})])|.$$

# Capacity of full support-rank PIR

## Theorem ([Hol+22])

*The capacity of full support-rank linear PIR from  $[n, k]$ -MDS coded storage with  $t$  colluding servers is*

$$\mathcal{C} = \frac{1 - \frac{k+t-1}{n}}{1 - \left(\frac{k+t-1}{n}\right)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k+t-1}{n}.$$

- The converse follows the converse proof for the symmetric case [Wan+17a] with some additional lemmas.
- A capacity-achieving scheme can be constructed from [Fre+17; D'O+18].



# Capacity of full support-rank PIR

## Theorem ([Hol+22])

*The capacity of full support-rank linear PIR from  $[n, k]$ -MDS coded storage with  $t$  colluding servers is*

$$\mathcal{C} = \frac{1 - \frac{k+t-1}{n}}{1 - \left(\frac{k+t-1}{n}\right)^m} \xrightarrow{m \rightarrow \infty} 1 - \frac{k+t-1}{n}.$$

- The converse follows the converse proof for the symmetric case [Wan+17a] with some additional lemmas.
- A capacity-achieving scheme can be constructed from [Fre+17; D'O+18].
- The proof settles the earlier conjecture for linear PIR schemes for full support-rank schemes, which seems to cover *almost* everything.

## How to interpret the above result?

---

- Our definition of full support-rank PIR captures the linear independency of the queries that all general capacity-achieving schemes have in common.
- In order to exceed the conjectured capacity, it is *necessary* for some restrictions of the queries to subsets of  $t$  servers to be linearly dependent.

## How to interpret the above result?

---

- Our definition of full support-rank PIR captures the linear independency of the queries that all general capacity-achieving schemes have in common.
- In order to exceed the conjectured capacity, it is *necessary* for some restrictions of the queries to subsets of  $t$  servers to be linearly dependent.
- It is exactly this property that allows the scheme of [Sun+18a], which is *not* of full support-rank, to exceed the full support-rank capacity.
- It seems difficult to extend the counter-example for  $m > 2$  while maintaining a good rate.

# Conclusion

---

- Intro to star product PIR from coded storage.
- Overview of capacity results.
- Strongly linear and full support-rank PIR capacity (almost) proving earlier conjectures.
- Importance of star product schemes in terms of practical implementation: small field sizes and low sub-packetization.
- Highly generalizable
  - stragglers, adversaries, networks, streaming, distributed computation, interference alignment, quantum,...

# Beyond PIR: current and future directions

---



Okko



Elif



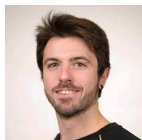
Syed



Masahito



Tefjol



Matteo

- Secure (analog) distributed matrix multiplication (SDMM):  
Okko Makkonen
- Cross-subspace alignment codes (CSA) for PIR and SDMM:  
cf. Jafar et al.
- Generalizations of the above using algebraic geometry codes:  
Okko, Dave, Elif Sacikara (+Gretchen)
- Quantum PIR: Matteo Allaix, Lukas, Tefjol Pllaha, Masahito Hayashi+group, Syed Jafar+group

# References I

---

- [Aug+14] D. Augot, F. Levy-Dit-Vehel, and A. Shikfa, “A storage-efficient and robust private information retrieval scheme allowing few servers”, in *Cryptology and Network Security*, Springer, 2014, pp. 222–239.
- [Ban+18] K. Banawan and S. Ulukus, “The capacity of private information retrieval from coded databases”, *IEEE Transactions on Information Theory*, vol. 64, no. 3, pp. 1945–1956, 2018.
- [Bei+02] A. Beimel, Y. Ishai, E. Kushilevitz, and J.-F. Raymond, “Breaking the  $O(n^{1/(2k-1)})$  barrier for information-theoretic private information retrieval”, in *The 43rd Annual IEEE Symposium on Foundations of Computer Science, 2002. Proceedings.*, IEEE, 2002, pp. 261–270.
- [Cho+95] B. Chor, O. Goldreich, E. Kushilevitz, and M. Sudan, “Private information retrieval”, in *Proceedings of IEEE 36th Annual Foundations of Computer Science*, IEEE, 1995, pp. 41–50.

## References II

---

- [D'O+19] R. G. D'Oliveira and S. El Rouayheb, "One-shot PIR: Refinement and lifting", *IEEE Transactions on Information Theory*, vol. 66, no. 4, pp. 2443–2455, 2019.
- [Dvi+16] Z. Dvir and S. Gopi, "2-server PIR with sub-polynomial communication", *Journal of the ACM (JACM)*, vol. 63, no. 4, 2016.
- [Faz+15] A. Fazeli, A. Vardy, and E. Yaakobi, "Codes for distributed PIR with low storage overhead", in *Information Theory (ISIT), 2015 IEEE International Symposium on*, 2015, pp. 2852–2856.
- [Hol+22] L. Holzbaur, R. Freij-Hollanti, J. Li, and C. Hollanti, "Toward the capacity of private information retrieval from coded and colluding servers", *IEEE Transactions on Information Theory*, 2022.
- [Mir+15] D. Mirandola and G. Zémor, "Critical pairs for the product Singleton bound", *IEEE Transactions on Information Theory*, vol. 61, no. 9, pp. 4928–4937, 2015.

## References III

---

- [Sad+23] B. Sadeh, Y. Gu, and I. Tamo, “Bounds on the capacity of private information retrieval over graphs”, *IEEE Transactions on Information Forensics and Security*, vol. 18, pp. 261–273, 2023.
- [Sha+14] N. B. Shah, K. V. Rashmi, and K. Ramchandran, “One extra bit of download ensures perfectly private information retrieval”, in *2014 IEEE International Symposium on Information Theory*, 2014.
- [Sun+17] H. Sun and S. A. Jafar, “The capacity of private information retrieval”, *IEEE Transactions on Information Theory*, vol. 63, no. 7, pp. 4075–4088, 2017.
- [Taj+18] R. Tajeddine, O. W. Gnille, and S. El Rouayheb, “Private information retrieval from MDS coded data in distributed storage systems”, *IEEE Transactions on Information Theory*, vol. 64, no. 11, pp. 7081–7093, 2018.



## References IV

---

- [Taj+19] R. Tajeddine, O. W. Gnilke, D. Karpuk, R. Freij-Hollanti, and C. Hollanti, "Private information retrieval from coded storage systems with colluding, Byzantine, and unresponsive servers", *IEEE Transactions on Information Theory*, vol. 65, no. 6, pp. 3898–3906, 2019.
- [Tia+19] C. Tian, H. Sun, and J. Chen, "Capacity-achieving private information retrieval codes with optimal message size and upload cost", *IEEE Transactions on Information Theory*, vol. 65, no. 11, pp. 7613–7627, 2019.
- [Zhu+19] J. Zhu, Q. Yan, C. Qi, and X. Tang, "A new capacity-achieving private information retrieval scheme with (almost) optimal file length for coded servers", *IEEE Transactions on Information Forensics and Security*, vol. 15, pp. 1248–1260, 2019.
- [D'O+18] R. G. L. D'Oliveira and S. E. Rouayheb, "Lifting private information retrieval from two to any number of messages", in *2018 IEEE International Symposium on Information Theory (ISIT)*, 2018, pp. 1744–1748.

## References V

---

- [Fre+17] R. Freij-Hollanti, O. W. Gnilke, C. Hollanti, and D. A. Karpuk, "Private information retrieval from coded databases with colluding servers", *SIAM Journal on Applied Algebra and Geometry*, vol. 1, no. 1, pp. 647–664, 2017.
- [Fre+19] R. Freij-Hollanti, O. W. Gnilke, C. Hollanti, A. Horlemann-Trautmann, D. Karpuk, and I. Kubjas, " $t$ -private information retrieval schemes using transitive codes", *IEEE Transactions on Information Theory*, vol. 65, no. 4, pp. 2107–2118, 2019.
- [Hei+18] A. Heidarzadeh, F. Kazemi, and A. Sprintson, "Capacity of single-server single-message private information retrieval with coded side information", in *2018 IEEE Information Theory Workshop (ITW)*, 2018, pp. 1–5.
- [Kum+19] S. Kumar, H. Lin, E. Rosnes, and A. G. i. Amat, "Achieving maximum distance separable private information retrieval capacity with linear codes", *IEEE Transactions on Information Theory*, pp. 1–1, 2019.

## References VI

---

- [Sun+18a] H. Sun and S. A. Jafar, “Private information retrieval from MDS coded data with colluding servers: Settling a conjecture by Freij-Hollanti et al.”, *IEEE Transactions on Information Theory*, vol. 64, no. 2, pp. 1000–1022, 2018.
- [Sun+18b] —, “The capacity of robust private information retrieval with colluding databases”, *IEEE Transactions on Information Theory*, vol. 64, no. 4, pp. 2361–2370, 2018.
- [Wan+17a] Q. Wang and M. Skoglund, “Linear symmetric private information retrieval for MDS coded distributed storage with colluding servers”, in *2017 IEEE Information Theory Workshop (ITW)*, 2017, pp. 71–75.
- [Wan+17b] —, “Secure symmetric private information retrieval from colluding databases with adversaries”, in *2017 55th Annual Allerton Conference on Communication, Control, and Computing (Allerton)*, 2017, pp. 1083–1090.
- [Wan+17c] —, “Symmetric private information retrieval for MDS coded distributed storage”, in *2017 IEEE International Conference on Communications (ICC)*, 2017, pp. 1–6.

A close-up photograph of a fluffy white rabbit with long, wispy ears, sleeping peacefully. The rabbit is positioned on the left side of the frame, with its head resting against a vibrant red background. Its eyes are closed, and its long whiskers are visible. The lighting is soft, highlighting the texture of its fur. In the bottom right corner, there is a logo for Aalto University School of Science, consisting of a large black 'A' with a red exclamation mark, followed by the text 'Aalto University' and 'School of Science' in a smaller font. To the left of the logo, the words 'THANKS!!' are written vertically in a black, sans-serif font.

T  
H  
A  
N  
K  
S  
!!



Aalto University  
School of Science

# Linear and strongly linear (SL) PIR

---

## Definition

A PIR scheme is *linear* if the responses are given by

$$A_j^i = \langle Q_j^i, Y_j \rangle, \forall j \in [n].$$

## Definition

A linear PIR scheme is *strongly linear* (SL) if each symbol of the desired file  $x^i$  is obtained as a deterministic linear function over  $\mathbb{F}_q$  of the response vector  $(A_1^i, \dots, A_n^i)$ , not depending on the randomness used to produce the queries.

- Strong linearity is important in practice, since it allows for **small field size** and **low sub-packetization level**  $O(k(n - k))$  (in contrast to  $O(n^m)$ ).

# PIR codes from star products

---

- Consider  $m$  files  $x^1, \dots, x^m$ .

$$\begin{matrix} x^1 \\ \\ x^m \end{matrix} \begin{pmatrix} \boxed{x_1^1 \quad \dots \quad x_k^1} \\ \vdots \quad \ddots \quad \vdots \\ \boxed{x_1^m \quad \dots \quad x_k^m} \end{pmatrix}$$

## PIR codes from star products

- Consider  $m$  files  $x^1, \dots, x^m$ .
- We encode this data using an  $[n, k, d_C]$  storage code  $C$  with generator matrix  $G_C$  and store it on  $n$  servers.

$$\begin{matrix} x^1 \\ \vdots \\ x^m \end{matrix} \begin{pmatrix} \boxed{x_1^1 \quad \dots \quad x_k^1} \\ \vdots \quad \ddots \quad \vdots \\ \boxed{x_1^m \quad \dots \quad x_k^m} \end{pmatrix} \cdot G_C = \begin{matrix} \text{Server 1} & \dots & \text{Server } n \end{matrix} \begin{pmatrix} \boxed{y_1^1} & \dots & \boxed{y_n^1} \\ \vdots & \ddots & \vdots \\ \boxed{y_1^m} & \dots & \boxed{y_n^m} \end{pmatrix}$$

## PIR codes from star products

---

- Consider  $m$  files  $x^1, \dots, x^m$ .
- We encode this data using an  $[n, k, d_C]$  storage code  $C$  with generator matrix  $G_C$  and store it on  $n$  servers.

$$\begin{matrix} x^1 \\ \vdots \\ x^m \end{matrix} \begin{pmatrix} \boxed{x_1^1 \quad \dots \quad x_k^1} \\ \vdots \quad \ddots \quad \vdots \\ \boxed{x_1^m \quad \dots \quad x_k^m} \end{pmatrix} \cdot G_C = \begin{matrix} \text{Server 1} & \dots & \text{Server } n \end{matrix} \begin{pmatrix} \boxed{y_1^1} & \dots & \boxed{y_n^1} \\ \vdots & \ddots & \vdots \\ \boxed{y_1^m} & \dots & \boxed{y_n^m} \end{pmatrix}$$

- Protects against failure of up to  $d_C - 1$  servers.