

# Private information retrieval with a coding-theoretic perspective

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Séminaire de l'équipe ECO

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## 1. Private information retrieval

## 2. PIR schemes with low computation and storage

- Transversal designs and codes

- A PIR scheme with transversal designs

- Collusion-resistant PIR schemes with weighted lifted codes

## 3. PIR schemes for common storage systems

- Distributed storage systems

- A PIR scheme on RS-coded databases

- A PIR scheme with regenerating codes

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**Trivial solution:** full download.

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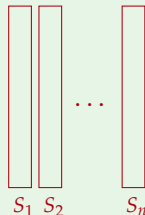
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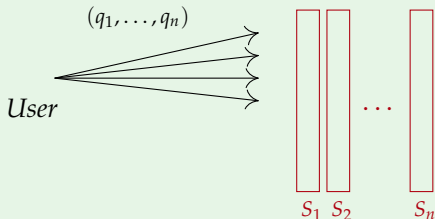
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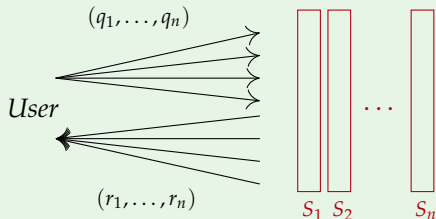
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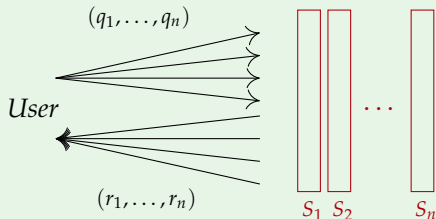
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3.  $U$  recovers the desired entry

$$F_i = \mathcal{R}(\mathbf{q}, \mathbf{r}, i).$$



The adversary: a **collusion of servers** = a subset of servers  $\{S_j : j \in T\}$ , where  $T \subset [1, n]$ , which exchange information about queries, data, etc.

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- **Information-theoretic (IT) privacy:**

$$I(i; q_{|T}) = 0, \quad \forall T \subseteq [1, n], |T| \leq t.$$

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**Theorem [CGKS95, CG97].** If  $t = n$  (in particular if  $n = 1$  server), then:

- for IT privacy, **no better solution than full download**,
- **computational** privacy is possible, but remains **expensive** as of now.

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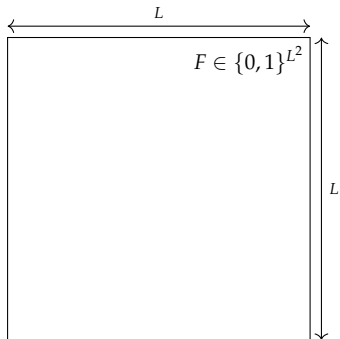
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- small entries vs. large entries
- bounded vs. unbounded number of entries in the database
- dynamic database vs. static database

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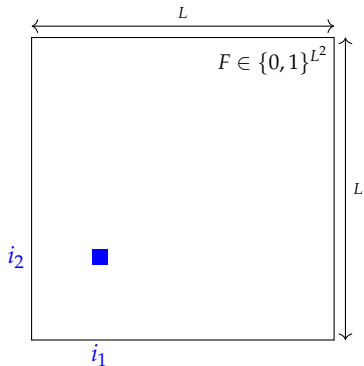


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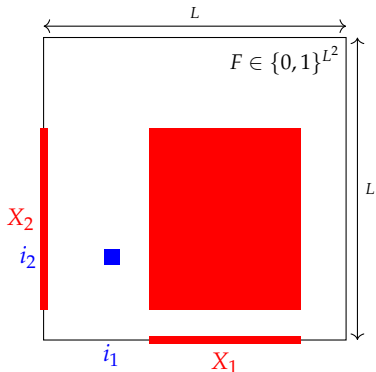


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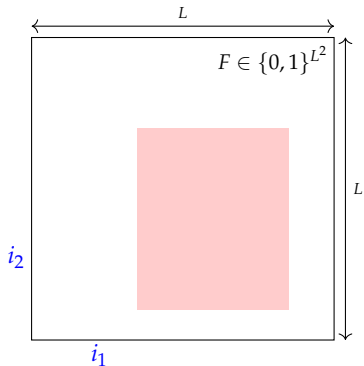
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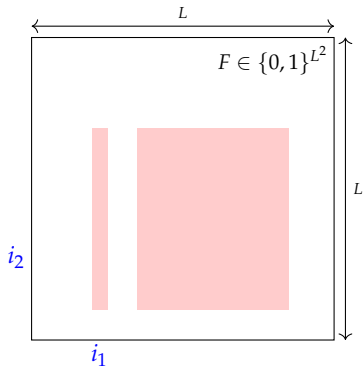
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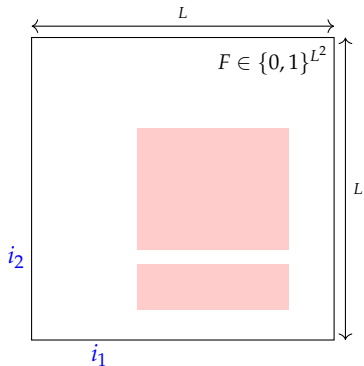
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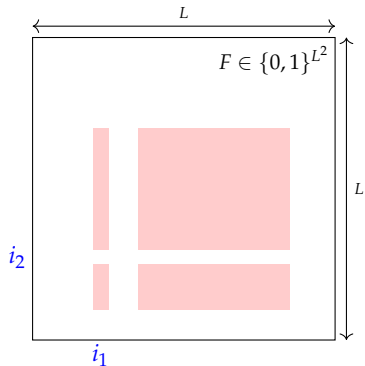
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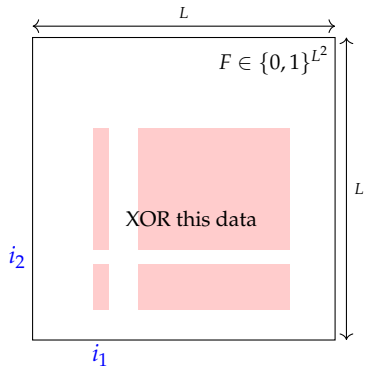
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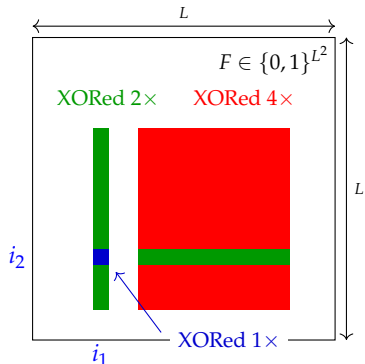
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3. User XORs the 4 bits and retrieves  $F_i$ .



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Generalisable to  $n = 2^b$  servers:

- ▶ **Communication:**  $b2^b M^{1/b} = n \log(n) M^{1/\log(n)}$  uploaded bits,  $n$  downloaded bits,
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- 1995: first definition [CGKS95]
- 2000: reduction from smooth locally decodable codes [KT00]
- 2000-10's: many improvements
  - ▶ PIR with 3 servers and subpolynomial communication [Yek08, Efr09]
  - ▶ PIR with 2 servers and subpolynomial communication [DG16]
  - ▶ lower storage overhead with *PIR codes* [FVY15]
- 2016-now: capacity-achieving schemes, schemes dedicated to storage systems
  - ▶ capacity of PIR [SJ17, BU18]
  - ▶ (nearly) capacity-achieving schemes [SRR14, CHY15, TR16, ...]

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## Tools: coding theory

- ▶ codes from transversal designs
- ▶ “lifted” codes



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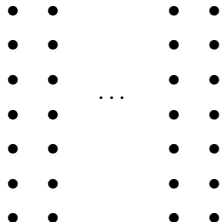
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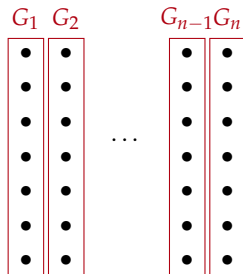
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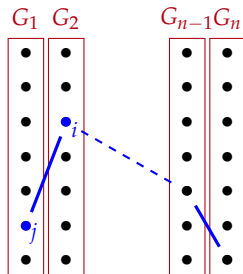


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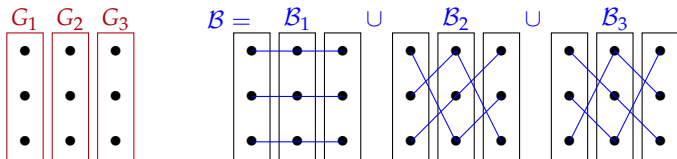
- ▶ blocks  $B \in \mathcal{B}$  satisfying
  - $B \subset X$  and  $|B| = n$ ;
  - for all  $\{i, j\} \subset X$ ,  $\{i, j\}$  lie:
    - either** in a single group  $G \in \mathcal{G}$ ,
    - or** in a unique block  $B \in \mathcal{B}$



# Example: a TD(3,3)

An example for a TD(3,3) :

- $ns = 9$  points
- $s = 3$  groups  $G_1, G_2, G_3$  of size 3
- $ns = 9$  blocks of  $n = 3$  points, partitionned into 3 parallel classes  $B_1, B_2, B_3$



Let  $\mathcal{T}$  be a transversal design  $\text{TD}(n, s) = (X, \mathcal{B}, \mathcal{G})$ .

Its **incidence matrix**  $M$  has size  $|\mathcal{B}| \times |X| = ns \times ns$ , and is defined by:

$$M_{i,j} = \begin{cases} 1 & \text{if } x_j \in B_i \\ 0 & \text{otherwise.} \end{cases}$$

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**Definition.** The **code  $\mathcal{C}$  based on  $\mathcal{T}$  over  $\mathbb{F}_q$**  is the  $\mathbb{F}_q$ -linear code having  $M$  as a parity-check matrix (*i.e.*  $\mathcal{C}^\perp$  is generated by  $M$ ).

- $\text{length}(\mathcal{C}) = |X| = ns$ ,
- $\dim(\mathcal{C}) = \dim(\ker M)$ ,
- every block  $B \in \mathcal{B}$  gives a parity-check equation  $\mathbf{h} \in \mathcal{C}^\perp$ , such that

$$\text{wt}(\mathbf{h}|_{G_j}) = 1, \quad \forall j = 1, \dots, n$$

# Example

The transversal design TD(3,3) represented by:

codeword  $c \in \mathbb{F}_q^9$

$c_1$	$c_2$	$c_3$
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**Dimension of the code?**

- ▶ depends on  $q$
- ▶ for instance, over  $\mathbb{F}_3$ , we have  $\text{rk}(H) = 6 \quad \dim(\mathcal{C}) = 3$ .

## 1. Private information retrieval

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Let  $\mathcal{C} \subseteq \mathbb{F}_q^N$  be a code based on a TD( $n, s$ ), with  $N = ns$ .

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Then  $U$  defines:

$$q_j = \mathcal{Q}(i)_j := \begin{cases} \text{unique } \in B \cap G_j & \text{if } i \notin G_j \\ \text{a random point in } G_j & \text{otherwise.} \end{cases}$$



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3.  $U$  recovers

$$c_i = - \sum_{j: i \notin G_j} c_{q_j} = - \sum_{b \in B \setminus \{i\}} c_b$$

**Theorem.** This PIR protocol is information-theoretically private.

Proof:

- the only server which holds  $F_i$  received a random query;
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**Question:** transversal designs leading to large dimension codes?

An example: the **classical affine transversal design**:

- ▶  $X = \mathbb{F}_q^m$  for  $m \geq 2$ ,
- ▶  $\mathcal{G}$  a partition of  $X$  into  $q$  hyperplanes  $G_1, \dots, G_q$ ,
- ▶  $\mathcal{B} = \{\text{affine lines } L \text{ secant to each } G_j\}$ .

The code has:

- length  $ns = q^m$ ,
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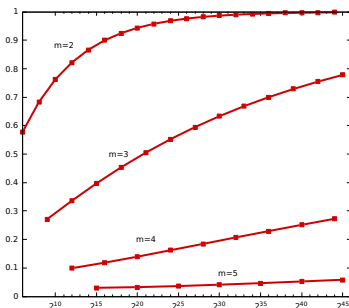
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$$\text{length } N = ns = 2^{em}$$

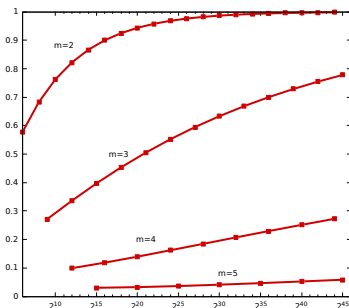
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**Question:** how to deal with collusions and byzantine errors?

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## 2. PIR schemes with low computation and storage

- Transversal designs and codes

- A PIR scheme with transversal designs

- Collusion-resistant PIR schemes with weighted lifted codes

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**Definition.** The (full-length) **Reed–Solomon code** of dimension  $k$  over  $\mathbb{F}_q$  is:

$$\text{RS}_q(k) := \{\text{ev}_{\mathbb{A}^1}(f) := (f(x_1), \dots, f(x_q)) \mid \deg(f) \leq k - 1\}.$$

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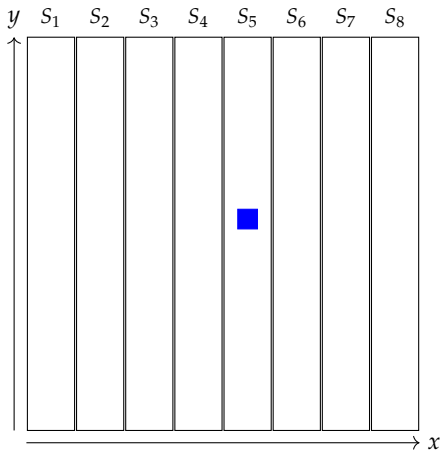
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Reed–Muller codes have the following property:

$$\forall c = \text{ev}_{\mathbb{A}^m}(f) \in \text{RM}_q(m, r) \quad \text{and} \quad \forall \text{ affine line } L \subset \mathbb{A}^m, \\ \text{ev}_{\mathbb{A}^1}(f|_L) \in \text{RS}_q(r + 1).$$

(where  $f|_L$  is the lowest-degree univariate polynomial interpolating  $f$  over  $L$ )

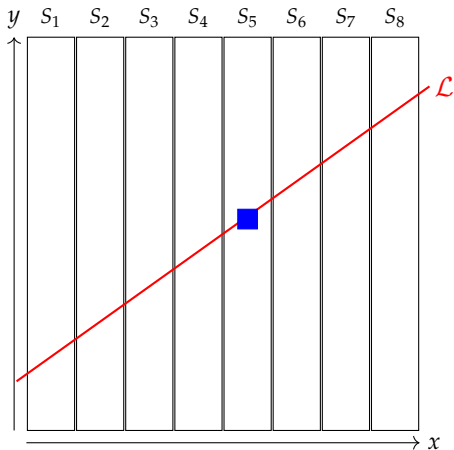
# A PIR scheme based on Reed–Muller codes



Database  $F$  is encoded with  $\text{RM}_q(m, r)$ , then distributed across the servers

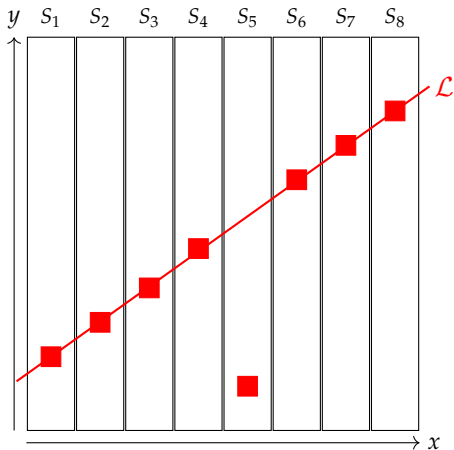


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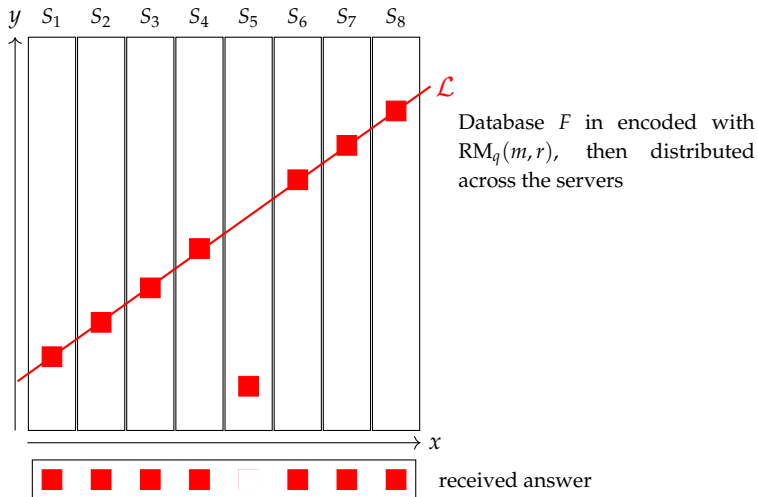
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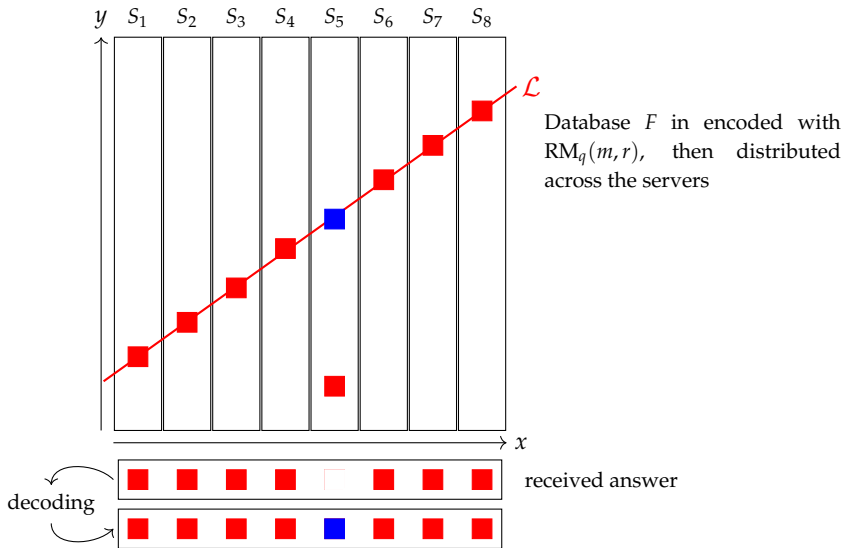


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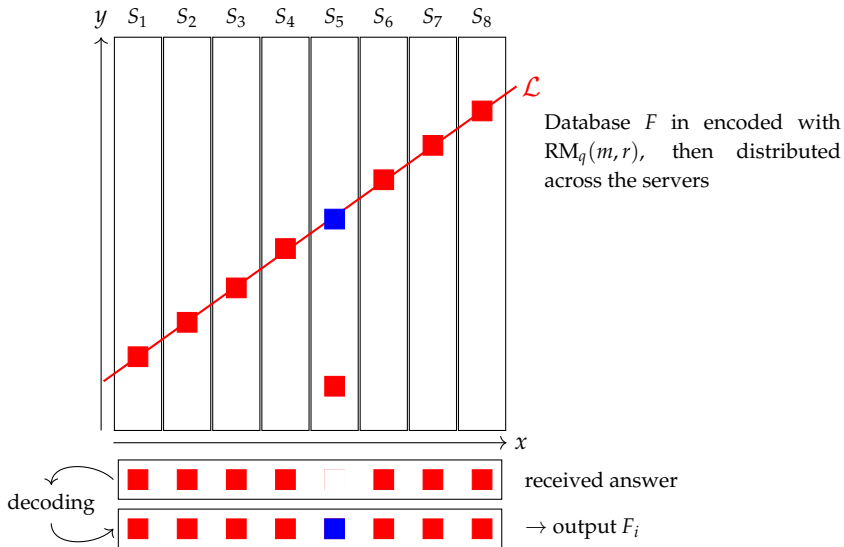
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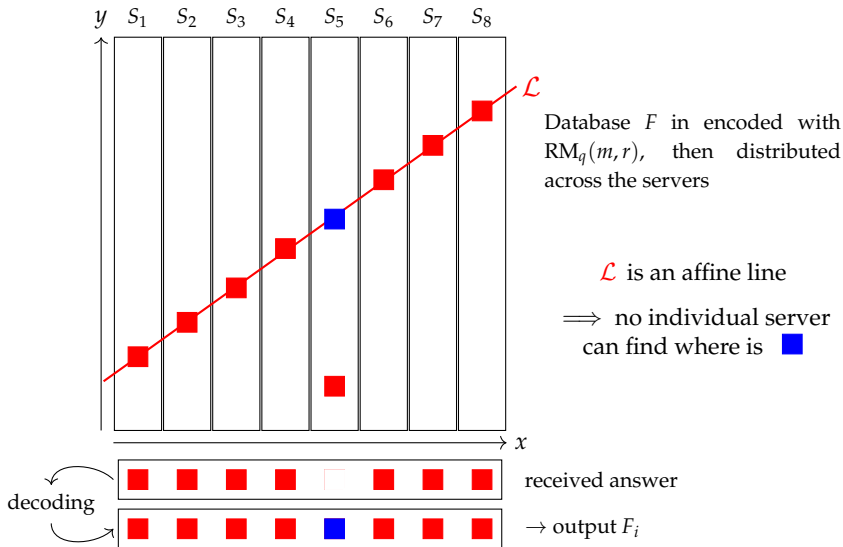
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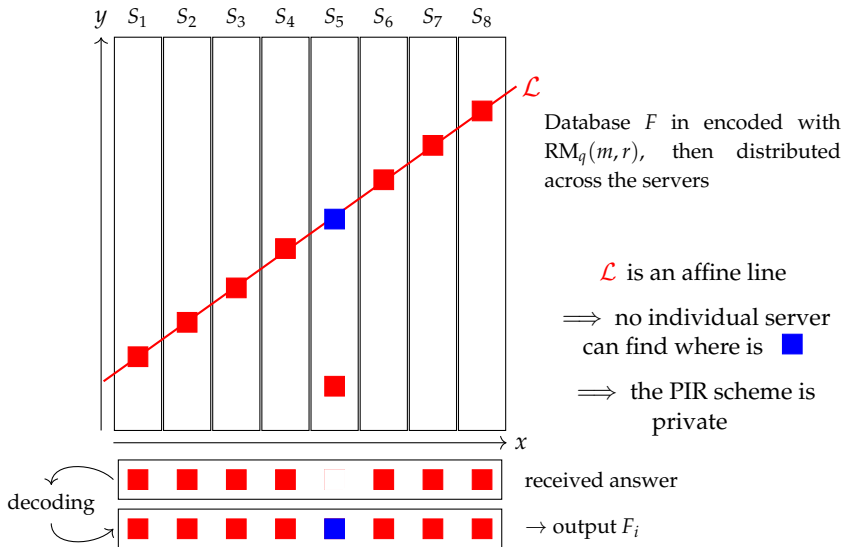
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**Features** with  $\text{RM}_q(m, r)$  of length  $q^m$ .


- ▶ communication complexity:  $(m - 1)q \log q$  uploaded bits,  $q \log q$  downloaded bits
- ▶ computational complexity:
  - ▶ **only 1 read for each server (optimal)**
  - ▶ a decoding procedure for  $\text{RS}_q(r)$  for the user
- ▶ storage overhead: the rate of  $\text{RM}_q(m, r)$  with  $r \leq q - 1$  is

$$\simeq \frac{(r/q)^m}{m!} \dots$$

⇒ We need codes with the same RM properties, but larger dimension.



“Lifted” codes are the largest codes having the same property as Reed–Muller codes.

 *New affine-invariant codes from lifting*. Guo, Kopparty, Sudan. ITCS. 2013.

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**Definition.** The  $m$ -th lifted Reed-Solomon code of degree  $r$  over  $\mathbb{F}_q$  is:

$$\text{Lift}_q(m, r) := \{ \text{ev}_{\mathbb{A}^m}(f) \mid f \in \mathbb{F}_q[\mathbf{X}] \text{ and } \forall \text{ affine line } L \subset \mathbb{A}^m, \deg(f|_L) \leq r \}.$$

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**Definition.** The  $m$ -th lifted Reed-Solomon code of degree  $r$  over  $\mathbb{F}_q$  is:

$$\text{Lift}_q(m, r) := \{\text{ev}_{\mathbb{A}^m}(f) \mid f \in \mathbb{F}_q[\mathbf{X}] \text{ and } \forall \text{ affine line } L \subset \mathbb{A}^m, \deg(f|_L) \leq r\}.$$

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Hence,

$$\text{ev}(X^2Y^2) \in \text{Lift}_4(2, 2) \quad \text{but} \quad \text{ev}(X^2Y^2) \notin \text{RM}_4(2, 2).$$

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**Fact.** For every  $m$ , lifted codes reach arbitrarily large information rates.

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**Definition.** The **weighted lifted Reed-Solomon code** of degree  $r$  and weight  $t$  over  $\mathbb{F}_q$  is:

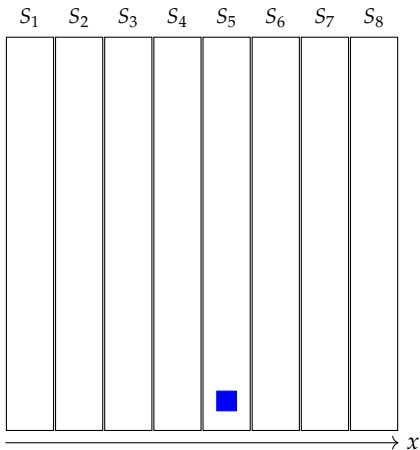
$$\text{WLift}_q(t, r) := \{\text{ev}_{\mathbb{A}^2}(f) \mid f \in \mathbb{F}_q[X, Y] \text{ and } \forall t\text{-curve } \mathcal{L} \subset \mathbb{A}^2, \deg(f|_{\mathcal{L}}) \leq r\}$$

**Consequence:** for every codeword  $c \in \text{WLift}_q(t, r)$  and every  $t$ -curve  $\mathcal{L}$ , we have:

$$c|_{\mathcal{L}} \in \text{RS}_q(r + 1).$$

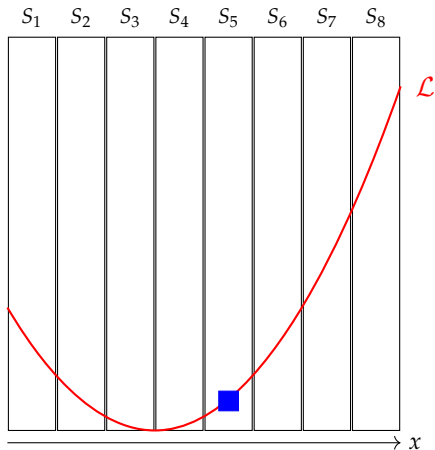


# A PIR scheme based on weighted lifted codes



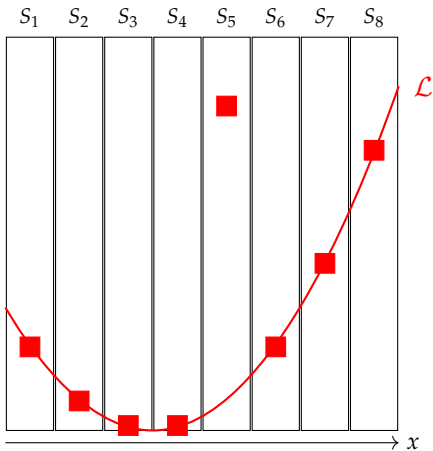
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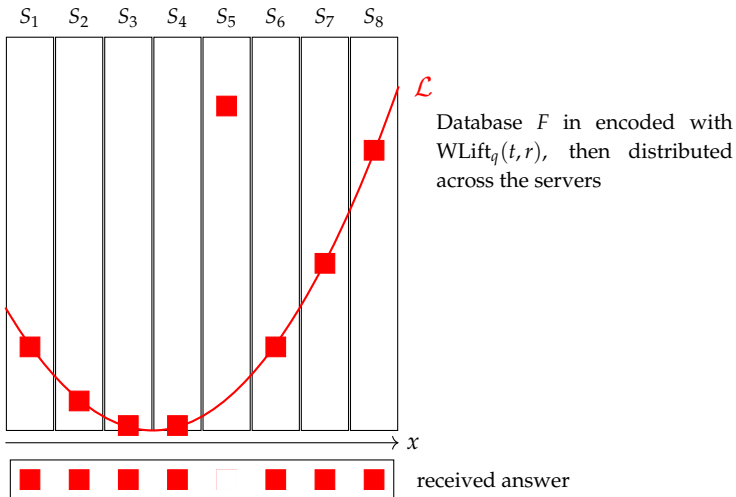
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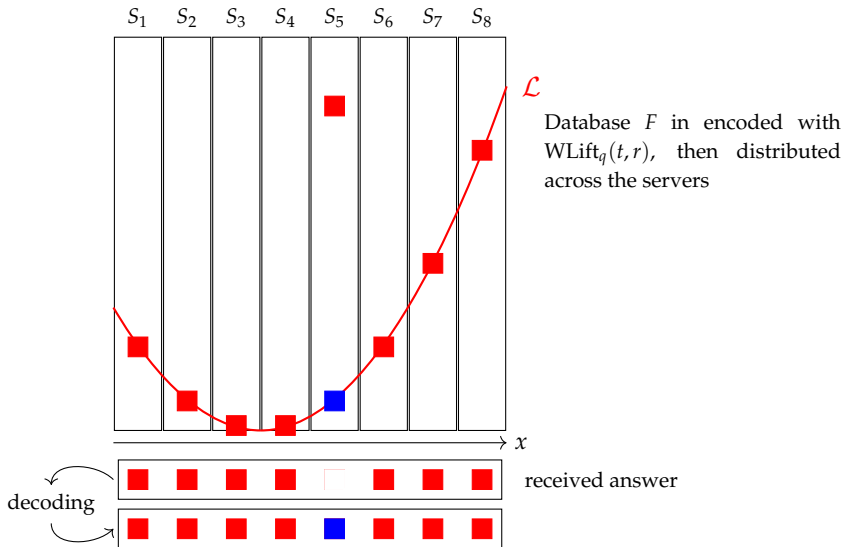


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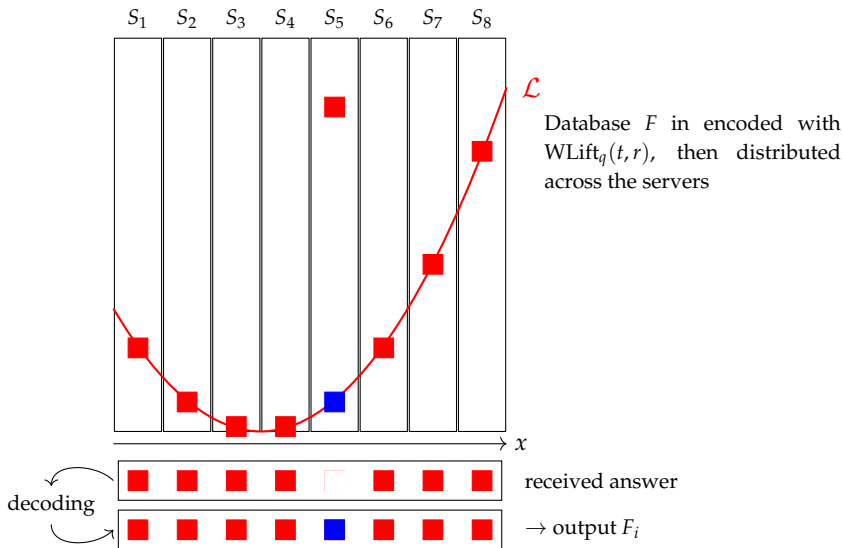
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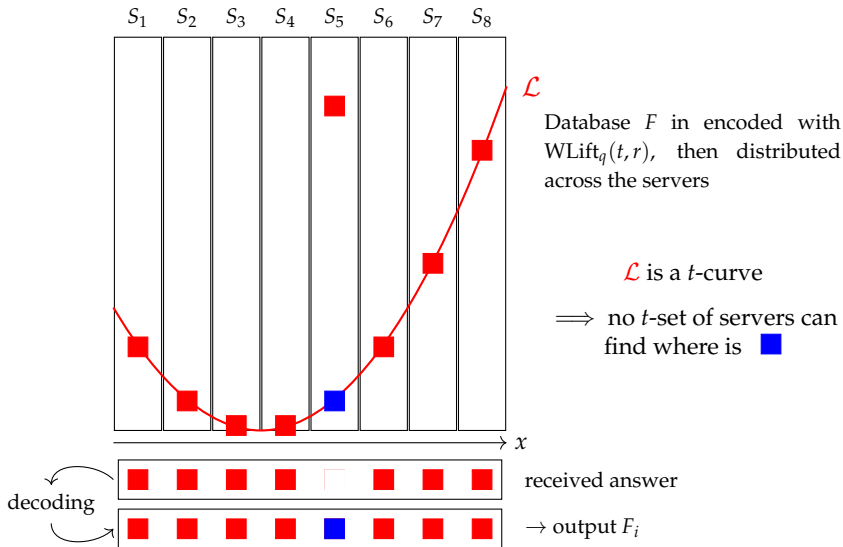
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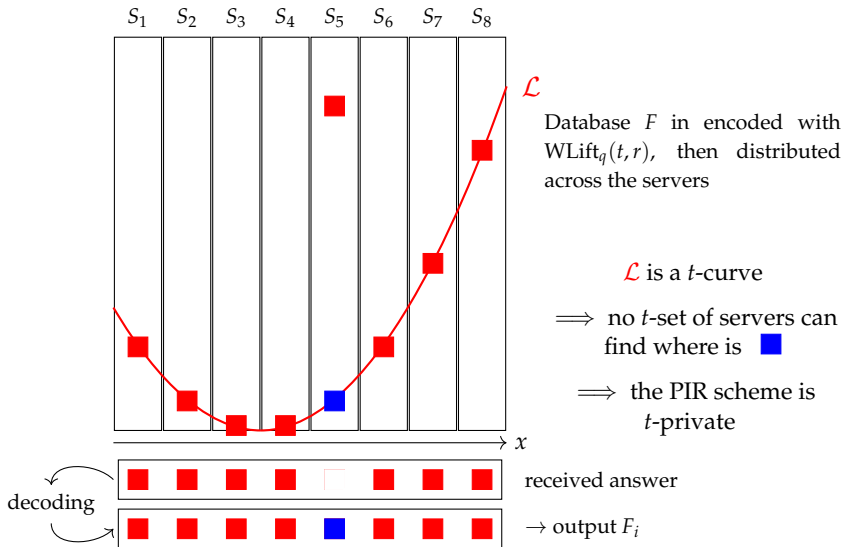
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📄 *Weighted Lifted Codes: Local Correctabilities and Application to Robust Private Information Retrieval.* L., Nardi. IEEE TIT. 2021.

**Theorem.** Let  $p$  be a prime number,  $t \geq 1$  and  $\alpha \geq 2$  be fixed integers. Then, the information rate  $\text{WLift}_{p^e}(t, p^e - \alpha)$  grows to 1 when  $e \rightarrow \infty$ .

**Corollary:** we get PIR schemes with relative storage overhead  $\rightarrow 0$ , for a constant number of adversaries.

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**Theorem.** Let  $p$  be a prime number,  $t \geq 1$  and  $c \geq 1$  be fixed integers. Let  $\gamma = 1 - p^{-c}$  and  $\mathcal{C}_e = \text{WLift}_{p^e}(t, \gamma p^e)$ . Then, the information rate  $R_e$  of  $\mathcal{C}_e$  satisfies:

$$\lim_{e \rightarrow \infty} R_e = K_{t,p,c} > 0$$

**Corollary:** we get PIR schemes with **constant relative storage overhead**, for a **constant number of collusions** and a **constant fraction of errors**.

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## 2. PIR schemes with low computation and storage

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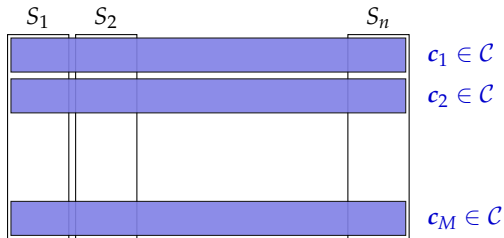
**Storage systems** use codes to cope with node failures.

- ▶ Before 2010: mostly replication or parity-check.
- ▶ 2010's: MDS storage (*e.g.*  $[14, 10]$  Reed-Solomon code for Facebook).
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Given a code  $\mathcal{C}$  of length  $n$ :



**Definition** (Reed-Solomon code). Let  $\mathbf{x} = (x_1, \dots, x_n) \in \mathbb{F}_q^n$ , pairwise distinct.

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**Main assumption** (can be discussed):

$$s \gg M$$

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Next, we present a PIR scheme for RS-coded databases.

- ▶ Originally [TR16], then extended and reformulated [TGKFH18, TGR18].
- ▶ Optimal PIR rate for  $t = 1$  and  $M \rightarrow \infty$ .
- ▶ PIR rate conjectured optimal for  $M \rightarrow \infty$ .

 [TR16] *PIR from MDS Coded Data in Distributed Storage Systems*. Tajeddine, El Rouayheb. ISIT. **2016**.

 [TGKFH18] *Robust PIR from Coded Systems with Byzantine and Colluding Servers*. Tajeddine, Gnilke, Karpuk, Freij-Hollanti, Hollanti. ISIT. **2018**.

 [TGR18] *PIR from MDS Coded Data in Distributed Storage Systems*. Tajeddine, Gnilke, El Rouayheb. IEEE-TIT. **2018**.

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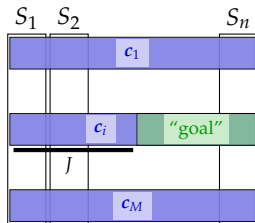
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$\mathcal{C} \subseteq \mathbb{F}_q^n$  the storage code,  $\mathcal{C} \in \mathcal{C}^M$  the coded database

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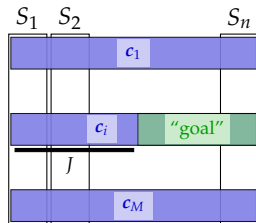
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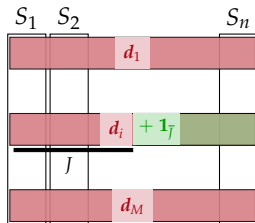
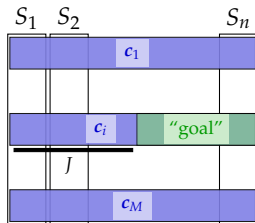
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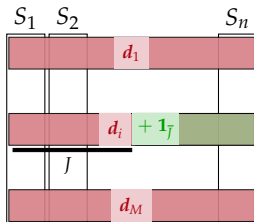
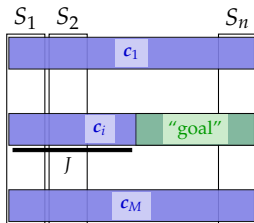
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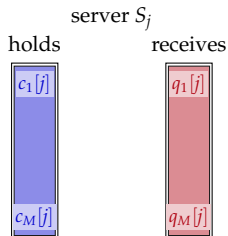
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**Remark:** queries remain private against collusions of servers of size  $\leq t$ .

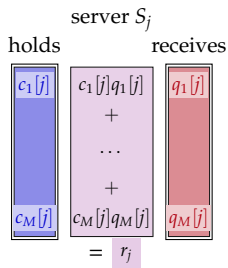


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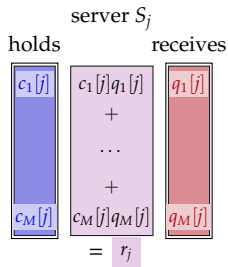
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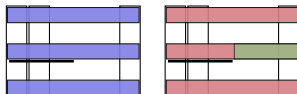


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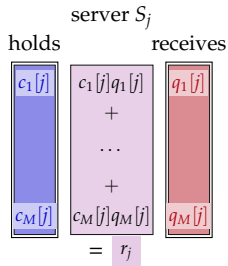


**Reconstruction:**



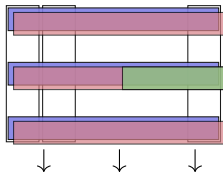
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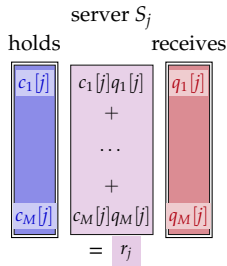


$$r = \text{[purple bar] [grey bar]}$$

# The protocol: server answers and reconstruction

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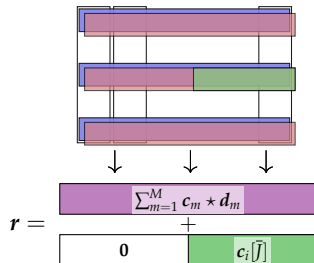


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and interpolates on  $J$  to recover

- $\sum_{m=1}^M d_m \star c_m$ ,
- then  $c_i[[\bar{J}]]$ .



**Features** for 1 run of the protocol.

- ▶ download cost:  $n$  symbols over  $\mathbb{F}_{q^s}$
- ▶ upload cost: an  $(M \times n)$ -matrix over  $\mathbb{F}_q$  (negligible if  $s \gg M$ )
- ▶ retrieval of  $|\bar{J}| = n - \dim(\mathcal{C} \star \mathcal{D})$  symbols of the desired file
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Otherwise, **striping** methods allow to achieve the same PIR rate.

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**Definition:**  $\mathcal{C}$  is an  $(n, k, d, \alpha, \beta, B)$ -regenerating code if:

- ▶  $\mathcal{C}$  is a linear space of dimension  $B$ , consisting in  $(\alpha \times n)$ -matrices over  $\mathbb{F}_q$ ,
- ▶ every  $c \in \mathcal{C}$  is fully determined by any  $k$ -subset of columns,
- ▶ every column of  $c$  can be “repaired”, by downloading  $\beta \leq \alpha$  symbols from any  $d$ -subset of columns (hence  $d\beta \geq \alpha$ ).

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A particular optimal point (minimum-bandwidth repair, MBR):  $d\beta = \alpha$ .  
Then,

$$B = \left( kd - \frac{k(k-1)}{2} \right) \beta.$$

 *Optimal Exact-Regenerating Codes for Distributed Storage at the MSR and MBR Points via a Product-Matrix Construction.* Rashmi, Shah, Kumar. IEEE-TIT. **2011.**

We set  $\beta = 1$ , hence  $\alpha = d$ .



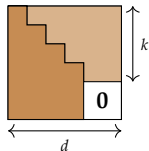
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$$A = \begin{pmatrix} S & T^\top \\ T & \mathbf{0} \end{pmatrix}$$

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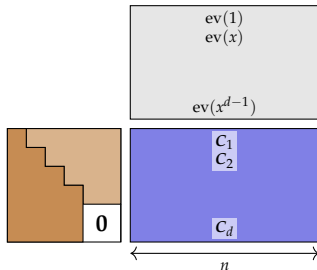
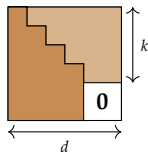
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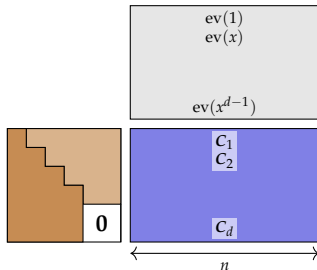
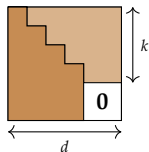
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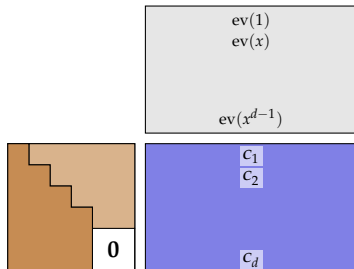
**Remark:** row  $C_j$  of  $C$  is a word of a RS code

- of dimension  $k$ , if  $j > k$ ,
- of dimension  $d > k$  otherwise.



📄 *Private Information Retrieval Schemes With Product-Matrix MBR Codes*. L., Tajeddine, Freij-Hollanti, Hollanti. IEEE IFS. 2021.

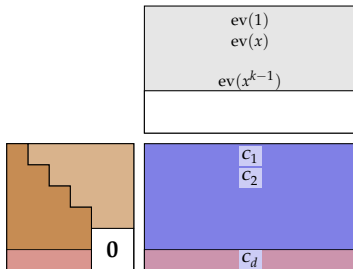
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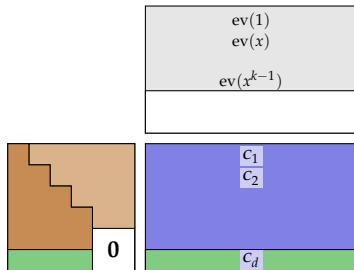
- For row  $j = d$  down to  $k + 1$ :
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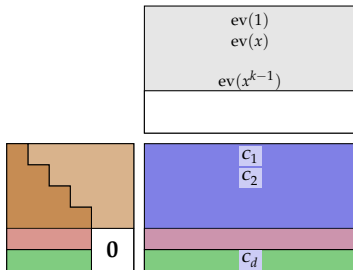


Retrieval rate:  $1 - \frac{k}{n}$

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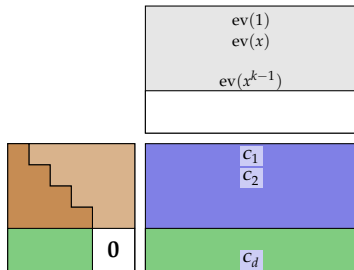
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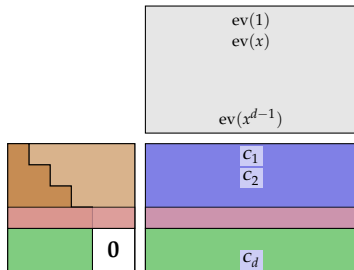
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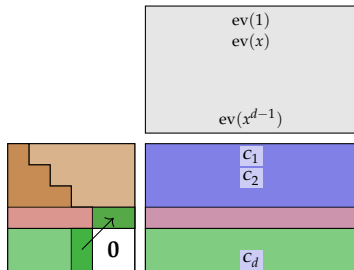
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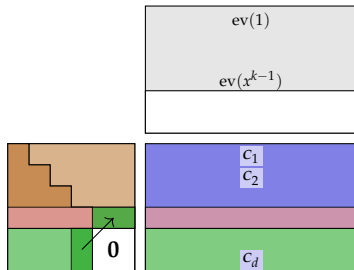
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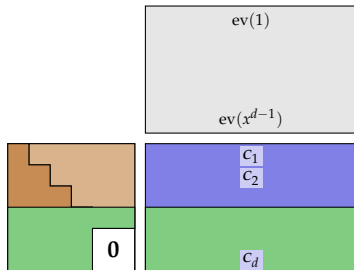
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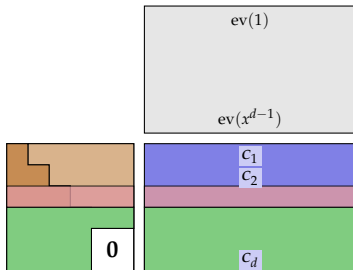



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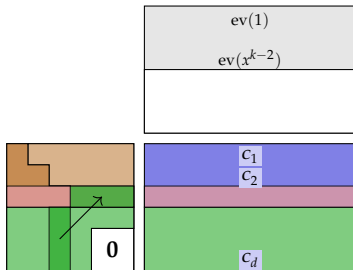
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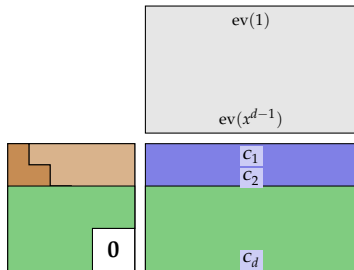
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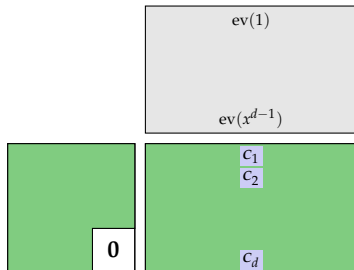


Retrieval rate:  $1 - \frac{k-1}{n}$

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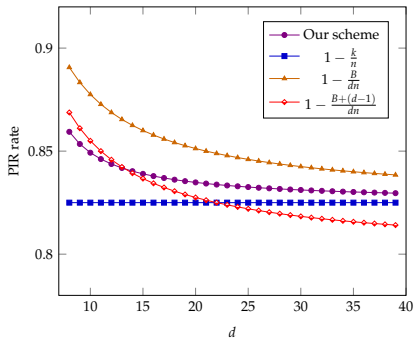
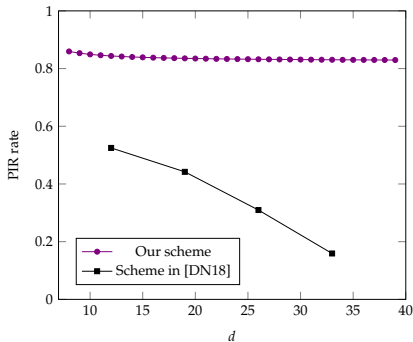


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Comparison of PIR rates for  $n = 40$  and  $k = 7$ .