

On the construction of PIR schemes

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Séminaire CASYS

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1. Private information retrieval

2. PIR schemes for common storage systems

- Distributed storage systems

- A PIR scheme on RS-coded databases

- A PIR scheme with regenerating codes

3. PIR schemes with low computation

- Transversal designs and codes

- A PIR scheme with transversal designs

- Instances

4. Conclusion

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Given a **remote** database $F \in \Sigma^M$ and $i \in [1, M]$,
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Trivial solution: full download.

Introduced in:

 *Private Information Retrieval*. Chor, Goldreich, Kushilevitz, Sudan. FOCS. 1995.

Database F stored (in some way) on n servers S_1, \dots, S_n ,
user U wants to recover F_i privately.

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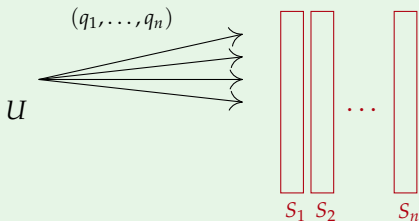
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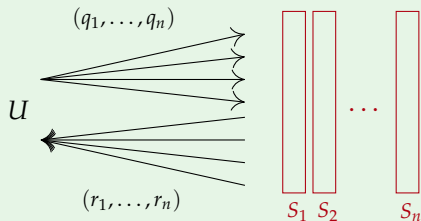
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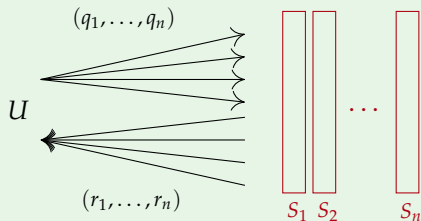
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3. U recovers $F_i = \mathcal{R}(\mathbf{q}, \mathbf{r}, i)$



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

$$t := \max\{|T|, T \subseteq [1, n] \text{ is a collusion}\} \geq 1$$

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- **Information-theoretic privacy:**

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

- **Computational privacy:** by varying the index i , distributions of queries $q_{|T} = \mathcal{Q}(i)_{|T}$ are computationally indistinguishable.

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

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

We focus on **IT-privacy**
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Parameters to be taken into account:

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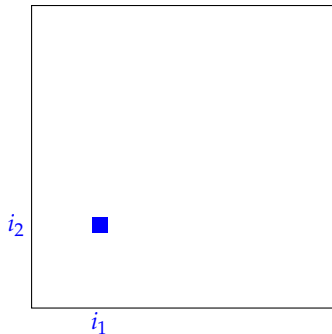
Several possible **settings**:

- bounded vs. unbounded number of entries in the database
- replicated database vs. coded database
- small entries vs. large entries
- dynamic database vs. static database
- unresponsive or byzantine servers

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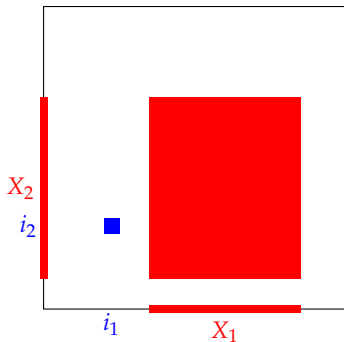
- ▶ $|F| = M$ bits, with $M = L^2$, and $[1, M] \simeq [1, L]^2$.
- ▶ $n = 4$ servers $S_{00}, S_{01}, S_{10}, S_{11}$, each storing a replica of F .
- ▶ **Goal:** retrieve $F_i = F_{(i_1, i_2)}$, for $1 \leq i_1, i_2 \leq L$.



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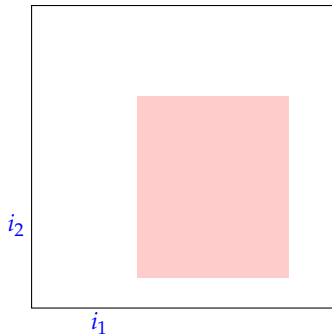


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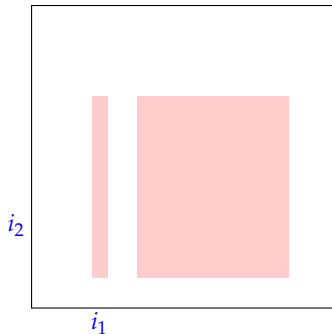


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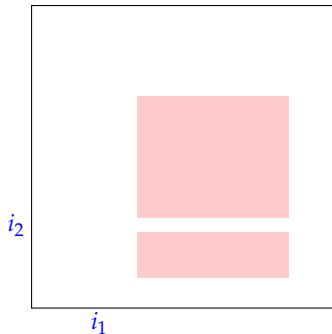


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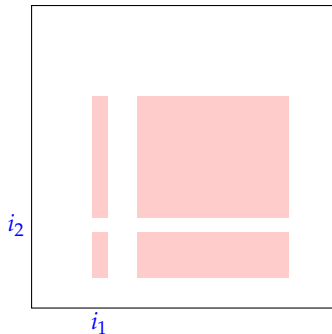


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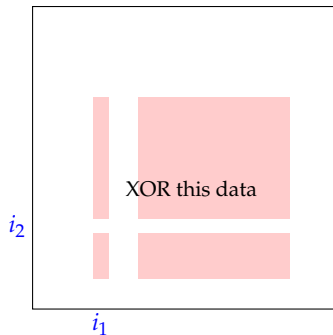


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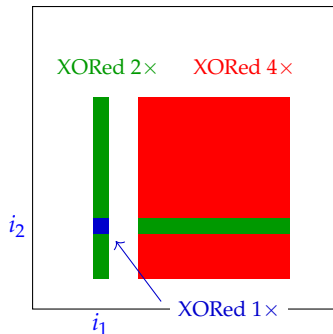


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3. User XORs the 4 bits and retrieves F_i .

Correct, and **secure** if no collusion.

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With $n = 4$ servers:

- ▶ **Communication:** $8\sqrt{M}$ uploaded bits, 4 downloaded bits,
- ▶ **Storage:** replication of F over 4 servers,
- ▶ **Complexity:**
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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,
- ▶ **Storage:** replication of F over n servers,
- ▶ **Complexity:**
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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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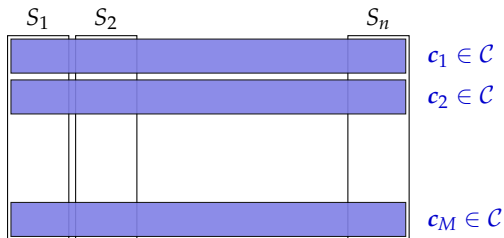
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- ▶ Before 2010: mostly replication or parity-check.
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Given a code \mathcal{C} of length n :



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

$$\text{RS}_q(k, n) := \{(f(x_1), \dots, f(x_n)), f \in \mathbb{F}_q[X], \deg f < k\}$$

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$\mathcal{C} = \text{RS}_q(k, n)$ is **MDS**:

- ▶ every codeword $\mathbf{c} \in \mathcal{C}$ can be reconstructed from any k -subset of coordinates of \mathbf{c} ,
- ▶ any subset of $d^\perp(\mathcal{C}) - 1 = k$ coordinates of \mathbf{c} are independent.

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a file $F_i \in \Sigma \simeq \mathbb{F}_{q^s}^k$ is encoded into $\mathbf{c}_i \in \text{RS}_q(k, n) \otimes \mathbb{F}_{q^s}$

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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].
- ▶ Scalable.
- ▶ 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**.

Notation:

$$a \star b := (a_1 b_1, \dots, a_n b_n)$$
$$\mathcal{C} \star \mathcal{C}' := \langle \{c \star c' \mid c \in \mathcal{C}, c' \in \mathcal{C}'\} \rangle$$

The protocol: query generation

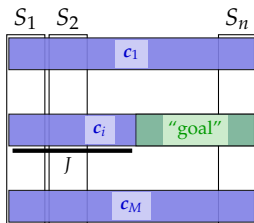
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System parameters:

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

$\mathcal{D} \subseteq \mathbb{F}_q^n$ a query code of dual distance $d^\perp(\mathcal{D}) = t + 1$

$J \subseteq [1, n]$ an information set for $\mathcal{C} \star \mathcal{D}$, and $\bar{J} := [1, n] \setminus J$



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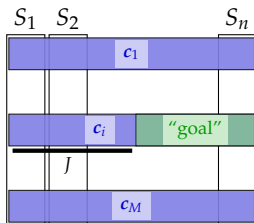
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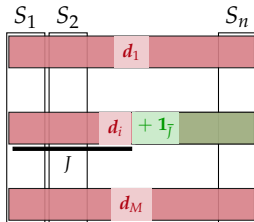
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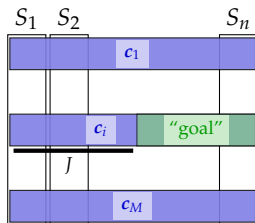
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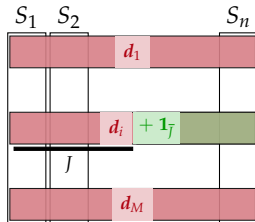
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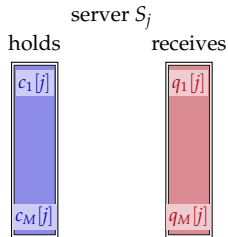
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Remark: queries remain private against collusions of servers of size $\leq t$.

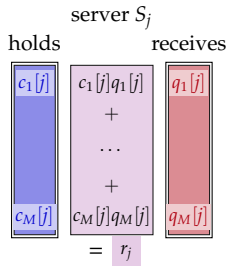


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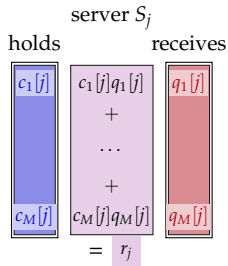
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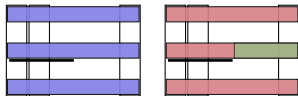


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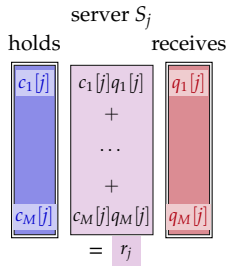


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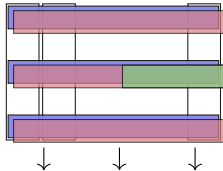
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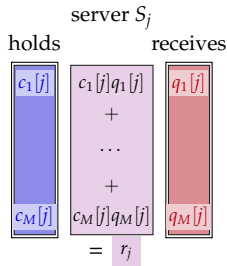
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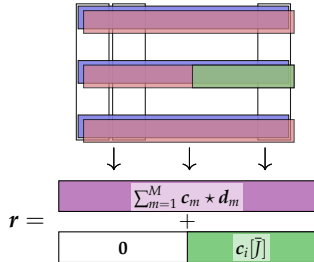


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and interpolates on J to recover

- $\sum_{m=1}^M d_m \star c_m$,
- 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 $s \gg M$)
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4. Conclusion

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- ▶ \mathcal{C} is a linear space of dimension B , consisting in $(\alpha \times n)$ -matrices over \mathbb{F}_q ,
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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**.

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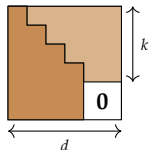
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$$A = \begin{pmatrix} S & T^\top \\ T & \mathbf{0} \end{pmatrix}$$

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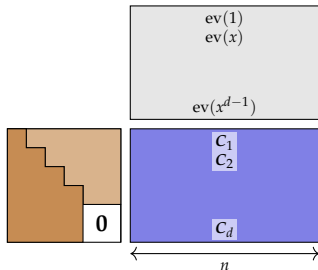
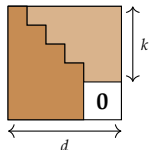
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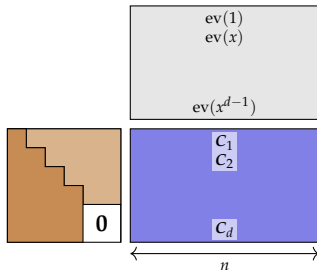
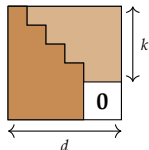
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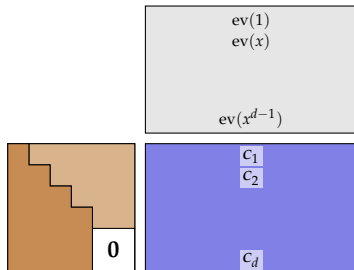
Remark: row C_j of C is a word of a RS code

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



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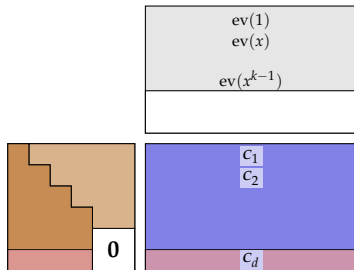
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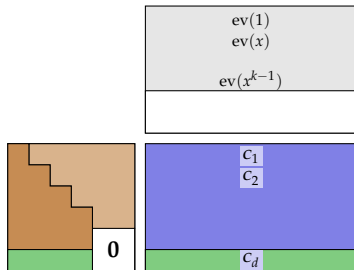
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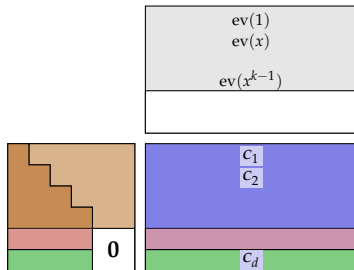


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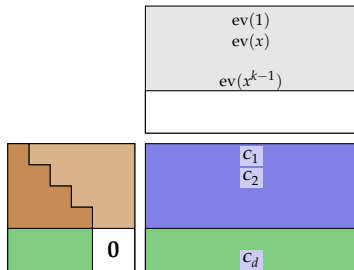
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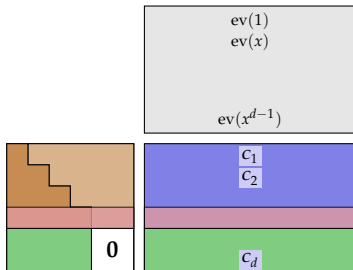


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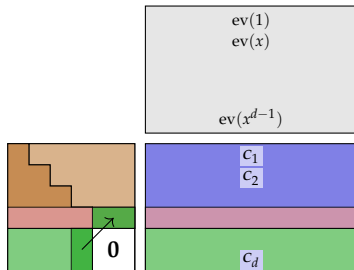
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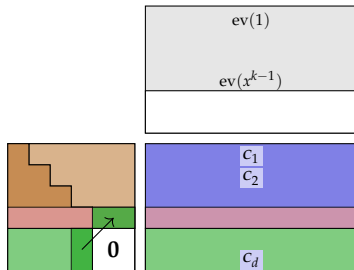
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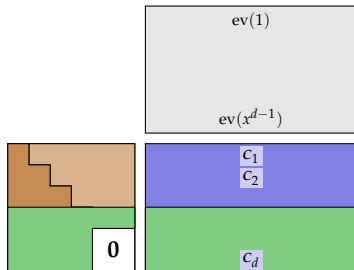
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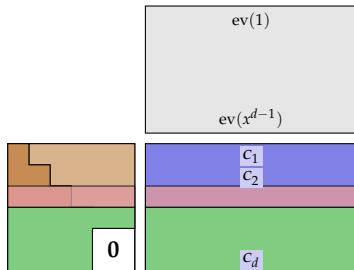


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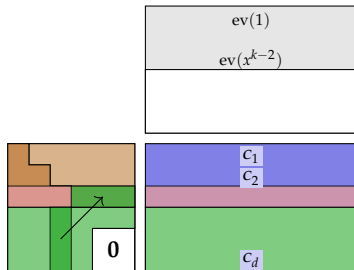
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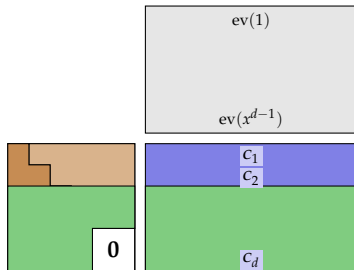
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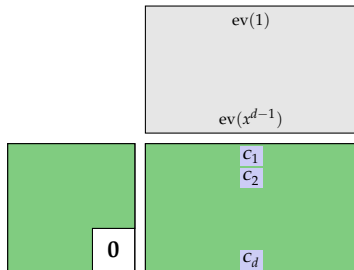


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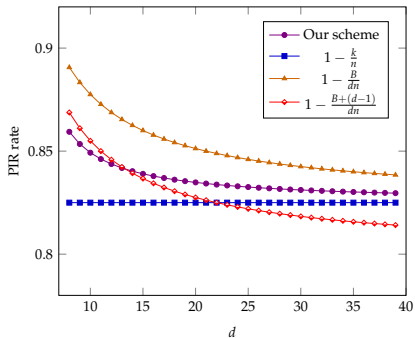
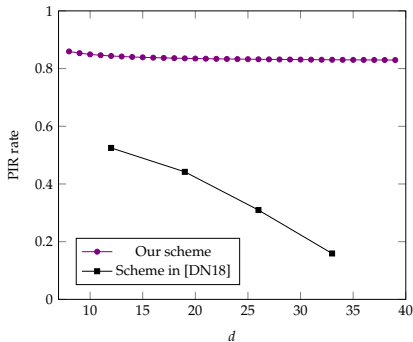
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Comparison of PIR rates for $n = 40$ and $k = 7$.

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4. Conclusion

Previous schemes:

- ▶ low communication complexity
- ▶ computationally inefficient (linear in $|F| = \sum_{m=1}^M |F_m|$)

Our goal:

- ▶ optimal computation ($|r_j|$ for each server S_j)
- ▶ remove the assumption $s \gg M$
- ▶ moderate communication complexity

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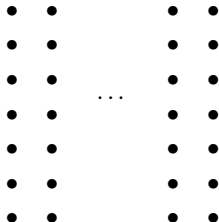
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A **transversal design** $\text{TD}(n, s) = (X, \mathcal{B}, \mathcal{G})$ is given by:

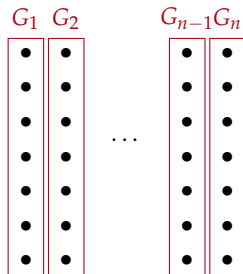
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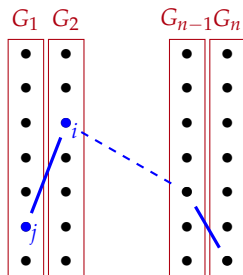


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- ▶ blocks $B \in \mathcal{B}$ satisfying
 - $B \subset X$ and $|B| = n$;
 - for all $\{i, j\} \subset X$, $\{i, j\}$ lie:
 - either** in a single group $G \in \mathcal{G}$,
 - or** in a unique block $B \in \mathcal{B}$



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

Its **incidence matrix** M has size $|\mathcal{B}| \times |X|$ 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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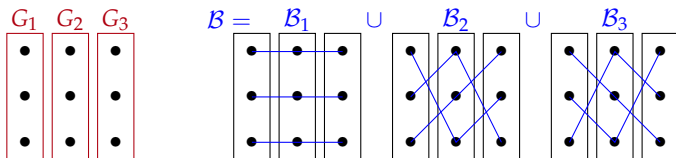
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The **code** \mathcal{C} based on \mathcal{T} over \mathbb{F}_q is the \mathbb{F}_q -linear code admitting M as a parity-check matrix (\mathcal{C}^\perp is generated by M).

- ▶ $\text{length}(\mathcal{C}) = |X|$,
- ▶ $\dim(\mathcal{C}) = \dim(\ker M)$,
- ▶ every $B \in \mathcal{B}$ gives an $\mathbf{h} \in \mathcal{C}^\perp$ such that $\text{wt}(\mathbf{h}|_{\mathcal{G}_j}) = 1, \forall j = 1, \dots, n$.

Example

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



gives an incidence matrix

$$M = \begin{pmatrix} 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}$$

Its rank over \mathbb{F}_3 is 6 \implies the associated code \mathcal{C} is a $[9,3]_3$ code.

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- **Initialisation.** User U encodes $F \mapsto c \in \mathcal{C}$, and gives $c|_{G_j}$ to server S_j .
- **To recover** $F_i = c_i$, with $i \in X$:
 1. User U randomly picks a block $B \in \mathcal{B}$ containing i .
Then U defines:

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

2. Each server S_j sends back c_{q_j}
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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- ▶ communication complexity: $n \log s$ uploaded bits, $n \log q$ downloaded bits
- ▶ computational complexity:
 - ▶ **only 1 read for each server** (somewhat optimal)
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Question: transversal designs with good $\dim(\mathcal{C})$ depending on (n, s) ?

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\mathcal{T}_A , the **classical affine transversal design**:

- ▶ $X = \mathbb{F}_q^m$, $m \geq 2$,
- ▶ \mathcal{G} a set of q disjoint hyperplanes partitionning X ,
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- length $ns = q^m$,
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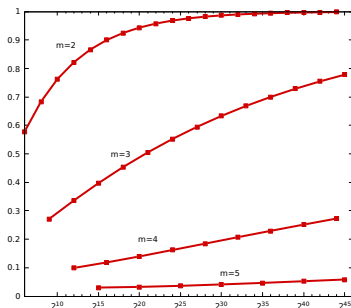
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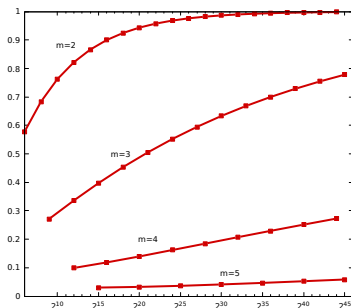
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Question: better instances?

An **orthogonal array** $\text{OA}(t, n, s)$ of strength t is a list A of words

- of length n ,
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- such that, for every $I \subset [1, n]$ of size t , $A|_I = S^t$.

Equivalently, an $\text{OA}(t, n, s)$ is a code $A \subset S^n$ with dual distance $t + 1$.

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$$\text{OA}(2, 3, 2) = \begin{bmatrix} a & b & b \\ b & b & a \\ b & a & b \\ a & a & a \end{bmatrix}$$

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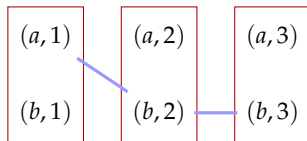
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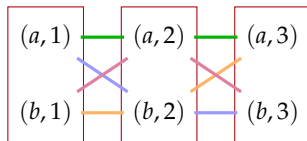
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⇒ The PIR protocol resists $t - 1$ colluding servers.

- ▶ OAs with $t > 2$ exist (e.g. from Reed-Solomon codes)
- ▶ But associated TDs lead to codes with poor rates (except for $t \ll n$)

 *Private Information Retrieval from Transversal Designs*. L.. IEEE-TIT. 2019.

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