

# Hyper-Invertible Matrices and Applications

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Theory and Practice of MPC, Aarhus, June 2012

# Outline

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## Hyper-Invertible Matrices

- Motivation
- Definition & Properties
- Construction
- Applications
- Conclusions

# How can $n$ parties generate random values?

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## Model

- $n$  parties,  $t$  are bad
- aim for random *shared* values (sharing doesn't matter)

## Approach 1

1. Every  $P_i$  shares random value  $x_i$
2.  $y = \sum_{i=1}^n x_i$       Only one good sharing from  $n$  sharings

## Approach 2

1. Every  $P_i$  shares random value  $x_i$
2.  $y_1 = \sum_i \lambda_{1i} x_i, \quad y_2 = \sum_i \lambda_{2i} x_i, \quad \dots$

How many good sharings from  $n$  sharings?

Best we can hope for:  $n - t$

## More Abstractly ...

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**Given:**  $n$  values

$$x_1 \ x_2 \ x_3 \ x_4 \ x_5 \ \dots \ x_n$$

where

- $n - t$  values are good (e.g. uniformly random),
- $t$  values are bad (e.g. chosen by adversary).

**Goal:** Find (the)  $n - t$  good values

**Goal':** Find  $y_1, \dots, y_{n-t}$  which are “as good as”  $x_2, x_5, \dots, x_n$ .

$$\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_{n-t} \\ y_{n-t+1} \\ \vdots \\ y_n \end{bmatrix} = \begin{bmatrix} \text{Hyper-Invertible} \\ \text{Matrix} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ \vdots \\ x_n \end{bmatrix}$$

## Hyper-Invertible Matrix — The Definition

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**Def:** M is *hyper-invertible* :  $\iff$  every square sub-matrix  $M_R^C$  is invertible.

$$\begin{bmatrix} \lambda_{11} & \lambda_{12} & \lambda_{13} & \cdots & \lambda_{1n} \\ \lambda_{21} & \lambda_{22} & \lambda_{23} & & \lambda_{2n} \\ \vdots & & & & \vdots \\ \lambda_{m1} & \lambda_{m2} & \lambda_{m3} & \cdots & \lambda_{mn} \end{bmatrix}$$

**Note:** Cf. Parity-check matrix of MDS-Codes, Cauchy matrices, ...

## Properties (1/2)

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**Property 1:** Given some  $x_j$ -s and some  $y_i$ -s (in total  $n$  values), one can compute all other  $x_j$ -s and  $y_i$ -s .

$$\begin{bmatrix} y_1 \\ y_2 \\ \cdot \\ y_m \end{bmatrix} = \begin{bmatrix} & & & \\ & & & \\ & & M & \\ & & & \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ \cdot \\ x_n \end{bmatrix}$$

**Lemma 1:** Given HIM  $M$ , index sets  $C \subseteq \{1 \dots n\}$ ,  $R \subseteq \{1 \dots m\}$  with  $|\bar{C}| = |R|$ . Then given  $(\vec{x}_C, \vec{y}_R)$  one can compute  $(\vec{x}_{\bar{C}}, \vec{y}_{\bar{R}})$ .

**Proof:** 1.  $\vec{y}_R = M_R \vec{x} = M_R^C \vec{x}_C + M_R^{\bar{C}} \vec{x}_{\bar{C}}$

2.  $\vec{x}_{\bar{C}} = (M_R^{\bar{C}})^{-1} (\vec{y}_R - M_R^C \vec{x}_C)$







# The Construction

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**Idea:** Construct mapping  $(x_1, \dots, x_n) \mapsto (y_1, \dots, y_m)$  with Property 1.

## Construction

1. fix values  $\alpha_1, \dots, \alpha_n, \beta_1, \dots, \beta_m$  in  $\mathcal{F}$
2. let polynomial  $f(z)$  s.t.  $f(\alpha_j) = x_j \quad \forall j$
3. compute  $y_i = f(\beta_i) \quad \forall i$

## Formally

- $$f(z) = \sum_{j=1}^n \prod_{\substack{k=1 \\ k \neq j}}^n \frac{z - \alpha_k}{\alpha_j - \alpha_k} x_j$$

- $$y_i = f(\beta_i) = \sum_{j=1}^n \underbrace{\prod_{\substack{k=1 \\ k \neq j}}^n \frac{\beta_i - \alpha_k}{\alpha_j - \alpha_k}}_{\lambda_{i,j}} x_j = \sum_{j=1}^n \lambda_{i,j} x_j$$

- $$M := [\lambda_{i,j}]$$

# The Field

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## The Field Size

- Previous construction requires  $|\mathcal{F}| \geq n + m$ .
- Easy patch:  $|\mathcal{F}| = n + m - 1$ .

## Lower Bounds (Conjecture)

- $|\mathcal{F}| = n + m - 1$  is optimal for  $\mathcal{F} \neq \text{GF}(2^k)$
- But:  $\begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 3 \\ 1 & 3 & 2 \end{bmatrix}$  is HIM over  $\text{GF}(4)$  (though  $m + n - 1 = 5$ )

# Randomness Extraction – Passive Security

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## Model

- $n$  parties,  $t$  are bad (**passive only**)
- aim for random *shared* values
- given  $n \times n$  hyper-invertible matrix  $M$

## Protocol

1. Every  $P_i$  shares random value  $x_i \rightarrow [x_i]$
2.  $([y_1], \dots, [y_n]) = M([x_1], \dots, [x_n])$
3. Output  $[y_1], \dots, [y_{n-t}]$

## Analysis

- Adversary  $A \subseteq \{1, \dots, n\}$ ,  $|A| = t$ , hence knows  $\overrightarrow{[x]}_A$ .
- Prop. 2: Fix  $A$ ,  $\overrightarrow{[x]}_A$ , mapping  $\overrightarrow{[x]}_{\overline{A}} \mapsto \overrightarrow{[y]}_{\{1, \dots, n-t\}}$  is bijective.

# Randomness Extraction – Active Security – Attempt #1

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## Model

- $n$  parties,  $t$  are bad (**active**)

## Protocol

- Every  $P_i$  **VSSes** random value  $x_i \rightarrow [x_i]$
- ...

## Analysis

- works, but complicated & inefficient

# Randomness Extraction – Active Security – Attempt #2

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## Model

- $n$  parties,  $t$  are bad (**active**)
- **detectable security** (cf player elimination / dispute control)

## Protocol

1. Every  $P_i$  **passively** shares random  $x_i \rightarrow [x_i]$
2.  $([y_1], \dots, [y_n]) = M([x_1], \dots, [x_n])$
3. Reconstruct and **check degree** of  $[y_1], \dots, [y_t]$
4. Output  $[y_{t+1}], \dots, [y_{n-t}]$

## Analysis

- Adversary  $A \subseteq \{1, \dots, n\}$ ,  $|A| = t$ ;  $H \subseteq \bar{A}$ ,  $|H| = n - 2t$ .
- Prop. 1: Degrees of  $\vec{[x]}_{\bar{A}}$  and  $\vec{[y]}_{\{1, \dots, t\}}$  ok  $\rightarrow$  all degrees ok.
- Prop. 2: Fix  $A, \vec{[x]}_A, \vec{y}_{\{1, \dots, t\}}$ , bij. mapping  $\vec{[x]}_H \mapsto \vec{[y]}_{\{t+1, \dots, n-t\}}$ .

# Randomness Extraction – Active Security – Attempt #3

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## Protocol

1. Every  $P_i$  **passively** shares random  $x_i \rightarrow [x_i]$
2.  $([y_1], \dots, [y_n]) = M([x_1], \dots, [x_n])$
3. For  $i = 1, \dots, 2t$ , have  $P_i$  check degree of  $[y_i]$
4. Output  $[y_{2t+1}], \dots, [y_n]$

## Analysis

- Adversary  $A \subseteq \{1, \dots, n\}$ ,  $|A| = t$ ;  $H \subseteq \bar{A}$ ,  $|H| = n - 2t$ .
- Prop. 1: Degrees of  $\vec{[x]}_{\bar{A}}$  and  $\vec{[y]}_{\{1, \dots, 2t\} \cap \bar{A}}$  ok  $\rightarrow$  all degrees ok.
- Prop. 2: Fix  $A, \vec{[x]}_A, \vec{[y]}_{\{1, \dots, 2t\} \cap A}$ ,  
mapping  $\vec{[x]}_H \mapsto \vec{[y]}_{\{2t+1, \dots, n\}}$  is bijective.

## Efficiency

- $n$  **passive** sharings  $\rightarrow n - 2t$  good random sharings

# Enhanced Checks

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## Example: Random Zero-Sharings [0]

1. Every  $P_i$  **passively** shares  $x_i = 0 \rightarrow [x_i]$
2.  $([y_1], \dots, [y_n]) = M([x_1], \dots, [x_n])$
3. For  $i = 1, \dots, 2t$ , have  $P_i$  check degree of  $[y_i]$  and  $y_i \stackrel{?}{=} 0$ .
4. Output  $[y_{2t+1}], \dots, [y_n]$

## Analysis

- Adversary  $A \subseteq \{1, \dots, n\}$ ,  $|A| = t$
- Prop. 1: If  $\vec{[x]}_{\bar{A}}$  and  $\vec{[y]}_{\{1, \dots, 2t\} \cap \bar{A}}$  have right degree and share 0  
 $\Rightarrow$  all sharings have right degree and share 0.

# Enhanced Checks – More Abstractly

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## Requirements

- “Goodness” must be linear:  $x_1$  and  $x_2$  good  $\Rightarrow x_1 + x_2$  good.
- Remember:  $\left( \vec{[x]}_A, \vec{[y]}_{\{t+1, \dots, n\}} \right) = \mathcal{L} \left( \vec{[x]}_{\bar{A}}, \vec{[y]}_{\{1, \dots, t\}} \right)$
- “Badness” does not need to be linear.

## Examples

- Sharings  $[x_i]$  of degree  $\leq t$
- Sharings  $[x_i]$  of degree  $\leq t$  and  $x_i = 0$
- Shared random bits  $[b_i]$  over  $\text{GF}(2^k)$ .
- Double-sharings  $[x_i], [y_i]$  of degrees  $\leq t, \leq 2t$ , resp., and  $x_i = y_i$ .
- ...



# Perfect MPC with Active Security

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## Model

- $n$  parties,  $t < n/3$  actively corrupted
- secure channels model (w/o broadcast)

## Achievements

- $\mathcal{O}(n\kappa)$  bits for multiplying two  $\kappa$ -bit values

## Tools

- Use HIM to generate random  $[x], [y]$  of degree  $t, 2t$  and  $x = y$ .
- Mult.:  $\forall P_i$  compute  $v_i = a_i b_i - y_i$ , reconstruct  $v$ , use  $[x] - v$  for  $[ab]$ .
- Beaver's circuit randomization + Player Elimination

# Conclusions

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## Hyper-Invertible Matrices

- easy to construct
- very good diffusing properties
- perfect security, no probabilities

## Applications

- extract randomness (propagate good properties)
- check consistency (concentrate bad properties)
- linear-complexity perfectly-secure MPC, very small overhead
- many more?