

# Hey! You Can't Do That With My Code!

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CIMPA-UNESCO-PHILIPPINES Research Summer School  
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# Outline

$(T, M, S)$ -Nets

Resilient Functions

Fuzzy Extractors

## First: The Omissions

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- ▶ the theory of **quantum error-correcting codes**
- ▶ The **PCP Theorem** in computational complexity theory: e.g.  $NP = PCP_{1-\epsilon, \frac{1}{2}}[O(\log n), 3]$  (Håstad, 2001)

## Part I: (T, M, S)-Nets



## Using Codes to Estimate Integrals

If orthogonal arrays can be used to approximate Hamming space, can they also be used to approximate other spaces?

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- ▶ **late 90s+:** Many new constructions (Adams/Edel/Bierbrauer/et al.)
- ▶ **2004+:** Improved bounds (Schmid/Schürer/Bierbrauer/Barg/Purkayastha/Trinker/Visentin)

## What is a $(T, M, S)$ -Net?



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of volume  $q^{T-M}$  (i.e., with  $d_1 + d_2 + \cdots + d_S = M - T$ ) contains exactly  $q^T$  points from  $\mathcal{N}$ .

## Simple Example of a $(T, M, S)$ -Net

- ▶ binary code with minimum distance three
- ▶ four points in  $[0, 1)^2$

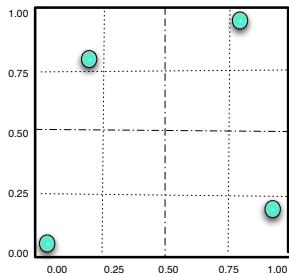
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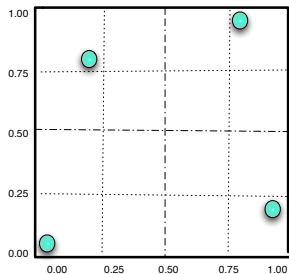


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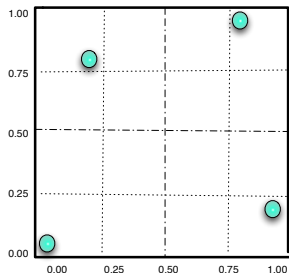


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- ▶ **Lawrence/Mullen/Schmid**:  $\exists(T, M, S)$ -net in base  $q \Leftrightarrow \exists OOA$  over  $\mathbb{F}_q$  with  $q^m$  rows,  $s = S$ ,  $\ell = t = M - T$ .

## The Theorem of Mullen & Schmid and (indep.) Lawrence

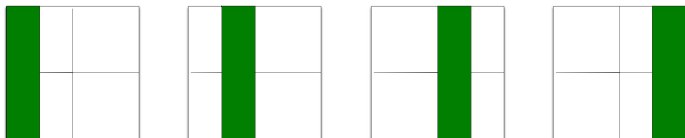


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## Idea of Proof

$$\mathcal{N} = \left\{ \left( \frac{0}{4}, \frac{0}{4} \right), \left( \frac{1}{4}, \frac{3}{4} \right), \left( \frac{2}{4}, \frac{2}{4} \right), \left( \frac{3}{4}, \frac{1}{4} \right) \right\}$$

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$$\mathcal{N} = \{(.00, .00), (.01, .11), (.10, .10), (.11, .01)\}$$

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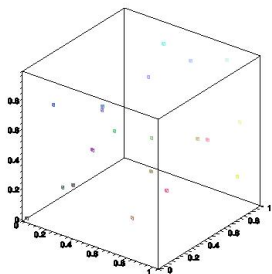
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## Nets from Many Sources



two mutually orthogonal latin squares of order five (color/height)

## Niederreiter/Xing Construction (Simplified)

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- ▶ They show that the same works over algebraic curves (global function fields)

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- ▶ **Research Problem:** Are there any non-trivial perfect codes in the Rosenbloom-Tsfasman metric?



## Part II: Resilient Functions



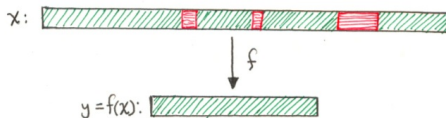
# Resilient Functions

How can a code be used to bolster randomness?

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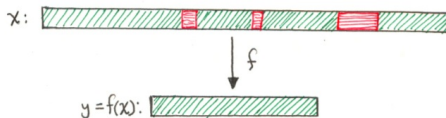
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# Resilient Functions

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We have a secret string  $x$ . An opponent learns  $t$  bits of  $x$ , but we don't know which ones.

After applying function  $f$ , we guarantee that our opponents knows nothing.

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- ▶ **1999+:** Applications to fault-tolerant distributed computing,  
key distribution, quantum cryptography, etc.

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- ▶ **Why?** Any linear combination of entries of  $f(x)$  is a dot product of  $x$  with some codeword
- ▶ So any non-trivial linear function of entries involves at least one random input position



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- ▶ Device must tap some physical source of known behavior
- ▶ Even the best sources of randomness have “quiet” periods
- ▶ Assuming 80% of input bits are random samples and 20% are from quiet periods
- ▶ Resilient function collapses samples to strings one-tenth the size
- ▶ What if quiet period is longer than expected?

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- ▶ **Lemma** (Sunar/WJM): Let  $X$  be a random variable taking values in  $\{0, 1\}^n$  according to a probability distribution  $\mathcal{D}_{T,A}$ . Then

$$\text{Prob}[H_{\text{out}} = k - r \mid |T| = i] = B_{i,r} \binom{n}{i}^{-1}.$$

## A Research Problem

Higher weight enumerators are known only for very few codes:

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- ▶ MDS codes: partial information only (Dougherty, et al.)
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- ▶ Hamming codes

Can we work out these statistics for the other standard families of codes?

## Part III: Fuzzy Extractors



## Codes for Biometrics

How can we eliminate noise if we are not permitted to choose our codewords?



## Selected References

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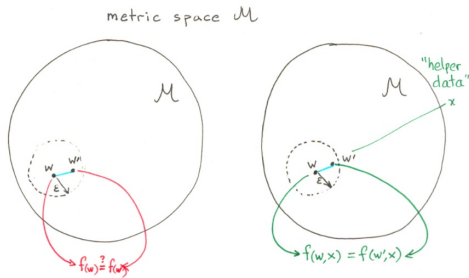
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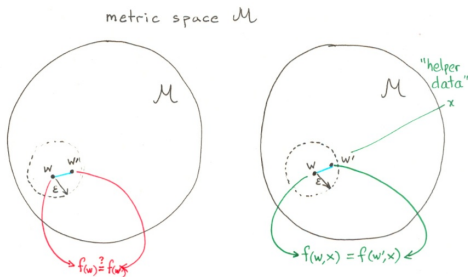
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# Fuzzy Extractors

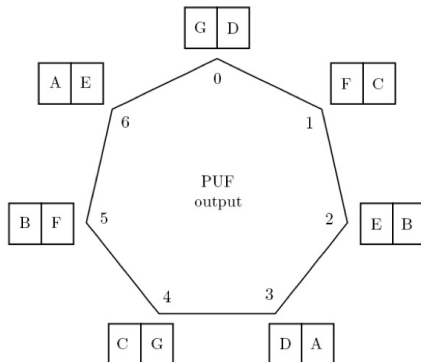


# Fuzzy Extractors



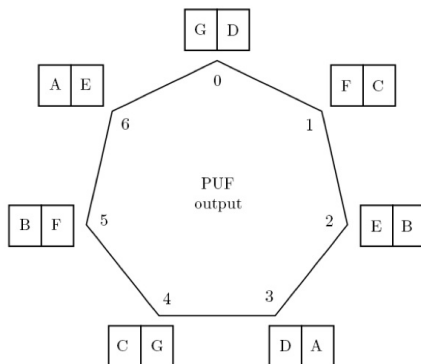
Metric space  $\mathcal{M}$  and function  $f : \mathcal{M} \times \{0, 1\}^* \rightarrow \{0, 1\}^*$  such that  $f(w', x) = f(w, x)$  provided  $x$  valid for  $w$  and  $d(w', w) < \epsilon$ .

## Fuzzy Extractor: Toy Example





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Baseline reading  $w = 3$  is obtained from temporal reading  $w' = 2$   
and hint  $x = D$ .

But  $w$  is not recoverable from either  $w'$  or  $x$  alone.

## Code-Offset Construction (Dodis, et al.)

Fuzzy extractor for Hamming metric:

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- ▶ To recover  $m$  from  $x$  and  $w'$ , decode  $w' + x = mG + (w - w')$
- ▶ Provided  $k$  and  $d$  are both linear in  $n$ , recovery of  $m$  from just  $x$  or  $w'$  is hard

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Fuzzy extractors are known for several metrics:

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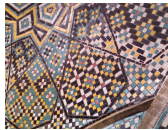
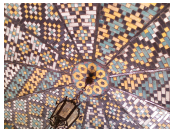
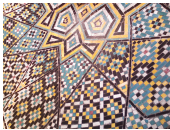
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Can we build efficient fuzzy extractors for the Euclidean metric?

# The End



Thank you all!

