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# **Optimized Graph-Based Codes For Modern Flash Memories**

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**10/08/2016**

# Presentation Outline

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- **Background and motivation**
- **Non-binary LDPC code optimization for flash memory**
  - New combinatorial objects
  - The optimization framework
  - Simulation results for practical Flash
- **Analysis and design of spatially-coupled LDPC codes**
  - Block vs spatially-coupled LDPC codes
  - The optimization framework
  - Simulation results for AWGN channel
- **Conclusions and future work**

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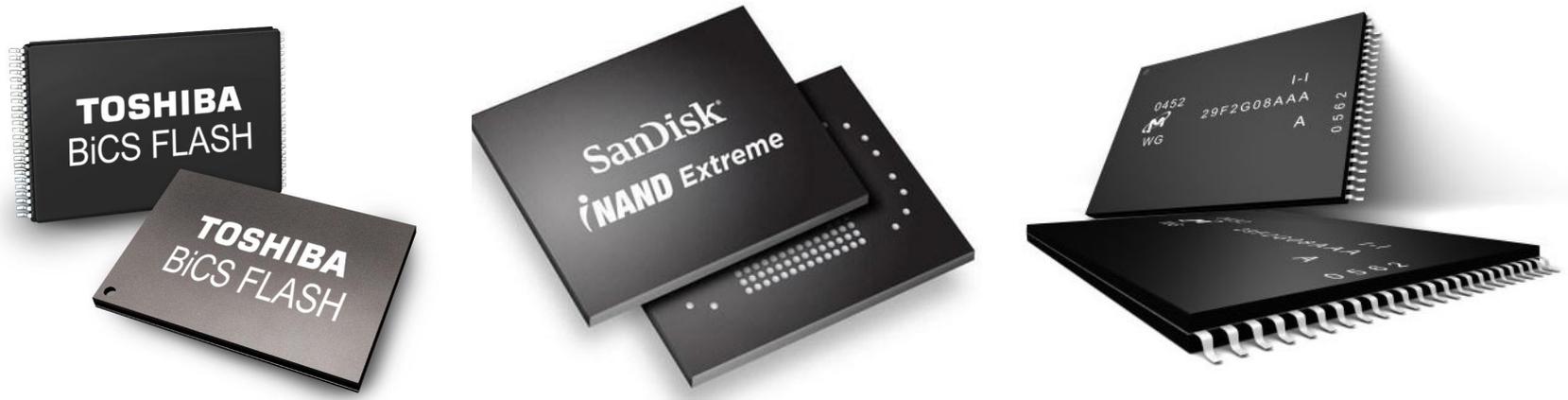
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# Graph-Based Codes with Low Error Rate

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- Modern flash memories operate at very low frame error rates. That is why strong ECC schemes are needed.



- Low-density parity-check (LDPC) codes are graph-based codes that have capacity approaching performance.
  - ❖ Non-binary LDPC (NB-LDPC) codes
  - ❖ Spatially-coupled (SC) LDPC codes

# Non-Binary vs Binary LDPC Codes

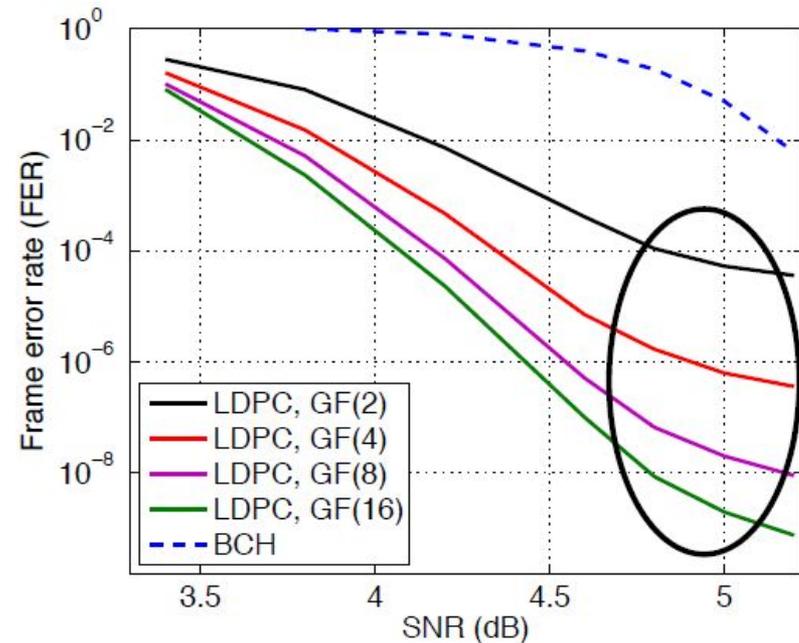
- **Why non-binary?**

- Grouping bits into symbols over  $GF(q)$  decreases the probability of decoding failure.
- Increasing the Galois Field size  $q$  results in better performance.

Block length  $\approx 1000$  bits

Rate  $\approx 0.9$

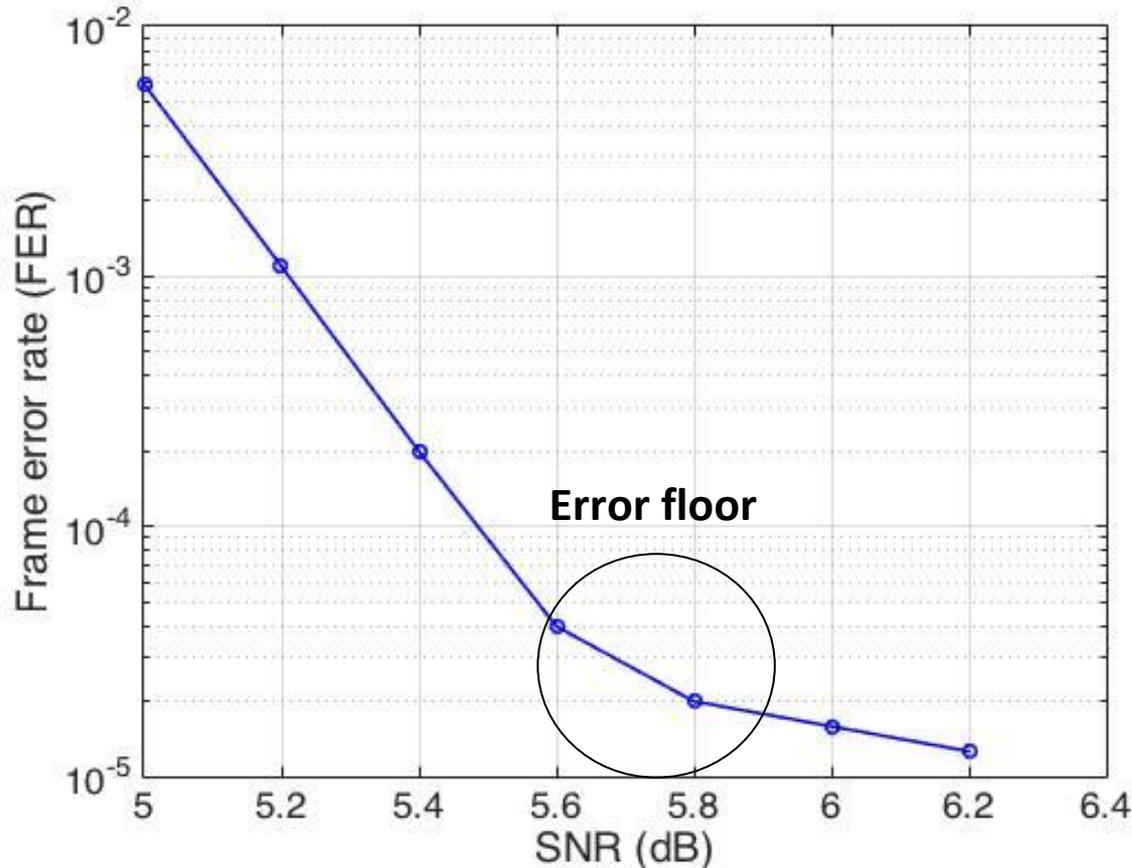
Column weight = 4



- **Disadvantage: Decoding complexity increases.**

# Error Floor of LDPC Codes: Absorbing Sets

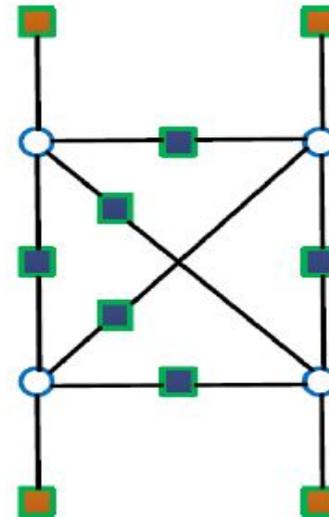
- For the AWGN channels, absorbing sets (ASs) [Dolecek 10] are the reason behind error floor.



# Error Floor of LDPC Codes: Absorbing Sets

- An  $(a, b)$  absorbing set:
  - is a subgraph of the Tanner graph.
  - $a$  is the number of variable nodes in the configuration.
  - $b$  is the number of unsatisfied check nodes connected to the configuration.
  - each variable node is connected to more satisfied than unsatisfied check nodes.

- Example:  $(4, 4)$  absorbing set:

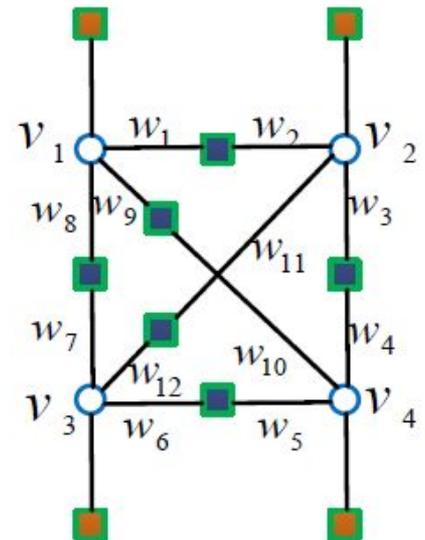




# Binary vs. Non-Binary Absorbing Sets

- Binary AS --> Binary LDPC codes
- Non-binary AS --> Non-binary LDPC codes
- Binary absorbing sets are described in terms of topological conditions only. For non-binary absorbing sets,
  - The values of the variable nodes matter.
  - The **topological conditions** alone are not enough; **weight conditions** have to be added [Amiri 14].

$$\prod_{i=1}^t w_{2i-1} = \prod_{i=1}^t w_{2i} \text{ over } \text{GF}(q)$$

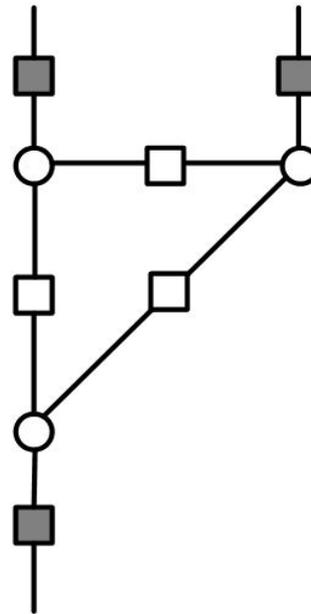


# Objects of Interest for the AWGN Channel

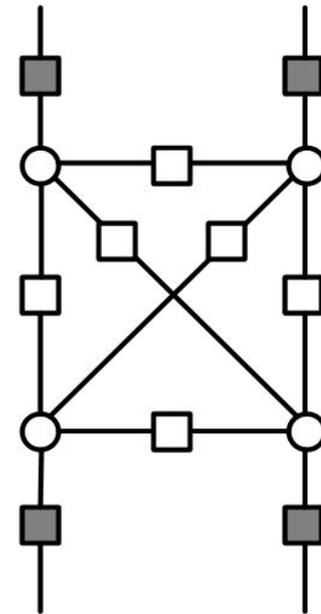
- For symmetric channels (like AWGN), the dominant objects have been the elementary ASs.
- **Elementary AS:** Each satisfied check node is of degree 2 and each unsatisfied check node is of degree 1.

- **Examples:**

- (3, 3) AS,  $\gamma=3$ .
- (4, 4) AS,  $\gamma=4$ .
- Where  $\gamma$  is the column weight of the code.



(3, 3) AS



(4, 4) AS

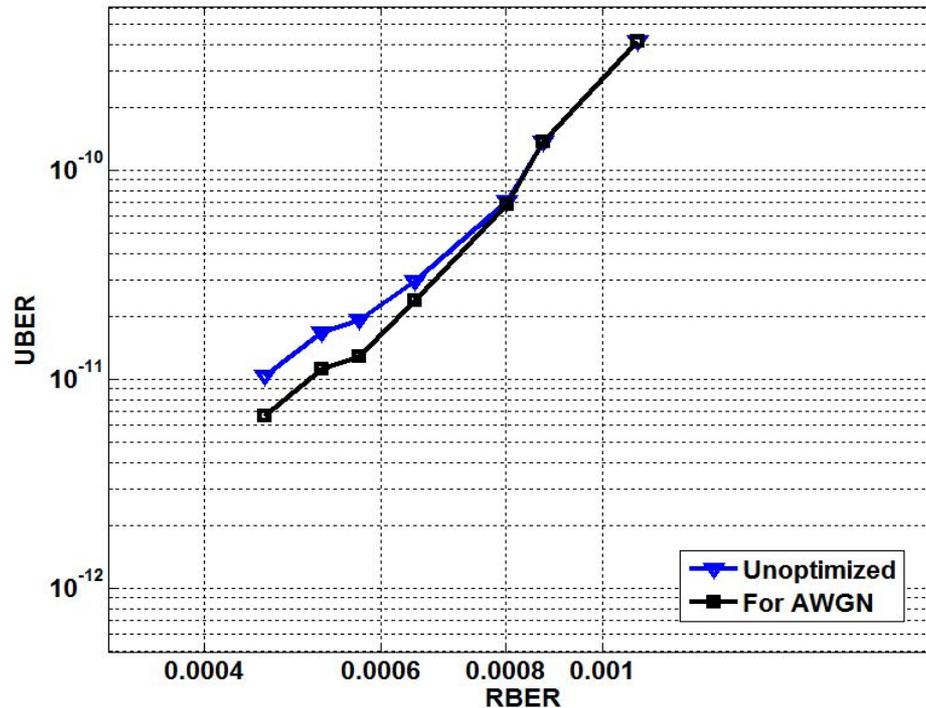
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# AWGN Techniques Don't Work for Flash Memories

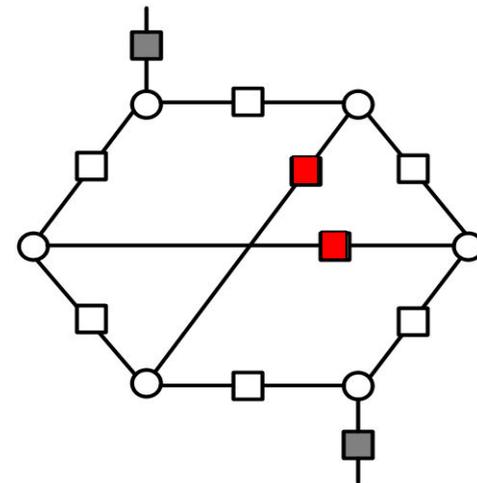
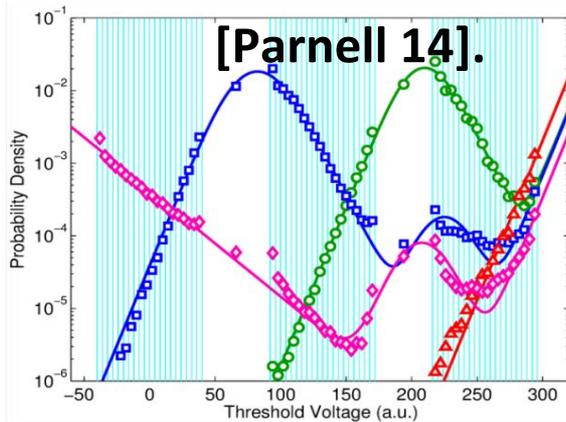
- Using optimization with respect to *elementary objects*, this is the gain we can reach on an asymmetric Flash channel (NLM Flash channel [Parnell 14]):



- Can we do better than techniques that are developed for the AWGN channel?

# The Answer is Yes!

- **Asymmetry in the channel (e.g., in Flash) can result in:**
  - NB ASs with unsatisfied check nodes having degree  $> 1$ .
  - NB ASs with satisfied check nodes having degree  $> 2$ .
- Such dominant objects are **non-elementary!**
- This is mainly because of the **high VN error magnitudes.**
- Example: (6, 4) non-elementary NB AS ( $\gamma=3$ ).



- (6, 2), (6, 3), and (6, 4) are **all problematic** because of asymmetry.

# We Need New Definitions/Objects

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- **General absorbing set (GAS)**
  - We can have unsatisfied check nodes of degree  $> 1$ .
  - We can have satisfied check nodes of degree  $> 2$ .
  
- **Because unsatisfied check nodes of degree  $> 2$  are less likely to happen, we introduce GAST.**
  
- **A GAS of type two (GAST) is a GAS that:**
  - The number of degree 2 check nodes is higher than the number of degree  $> 2$  check nodes.
  - Unsatisfied check nodes are of degree 1 or degree 2.

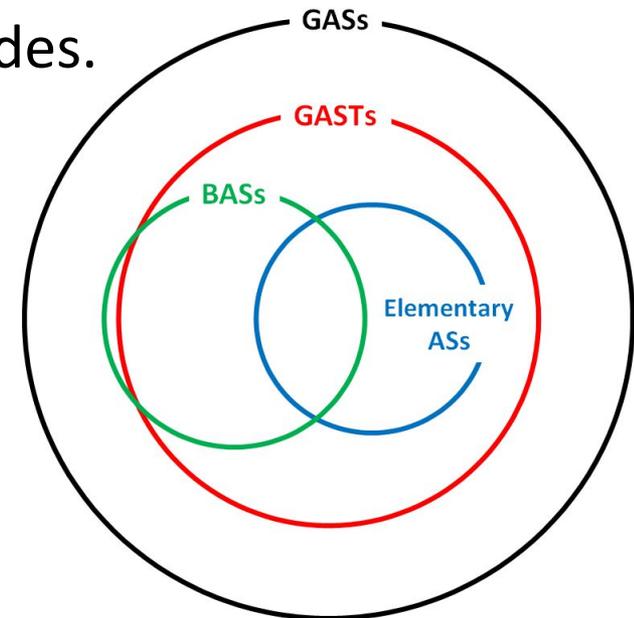
# We Need New Definitions/Objects

- Topological description of the new objects (GASTs):

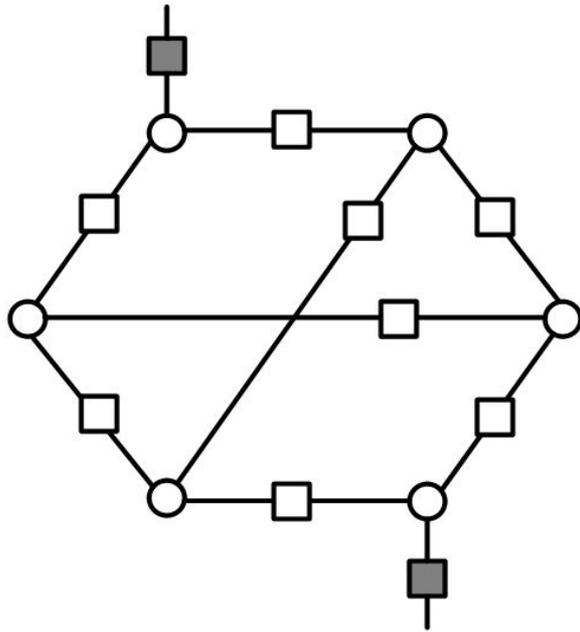
The tuple  $(a, b, d_1, d_2, d_3)$  is an NB AS that:

- $a$  is the number of variable nodes in the set.
- $b$  is the number of unsatisfied check nodes.
- $d_1$  is the number of degree 1 check nodes.
- $d_2$  is the number of degree 2 check nodes.
- $d_3$  is the number of degree  $> 2$  check nodes.

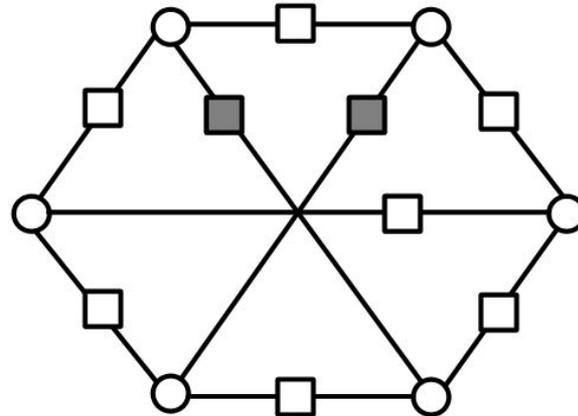
- GASTs are more general than any previously introduced type of absorbing sets



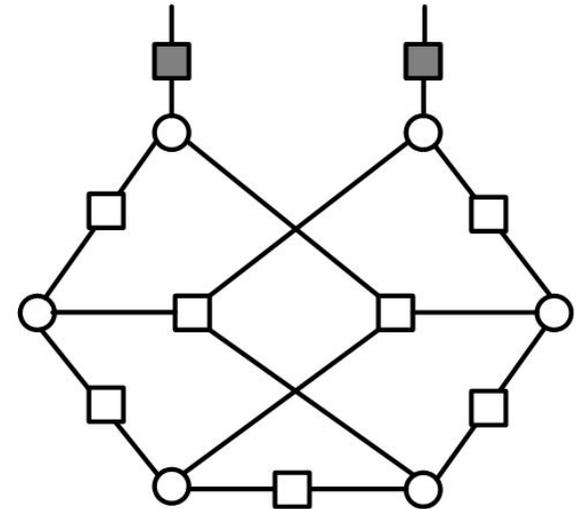
# Examples of GASTs



Config (a)



Config (b)



Config (c)

- Configuration (a) is now a **(6, 2, 2, 8, 0) GAST**.
- Configuration (b) is now a **(6, 2, 0, 9, 0) GAST**.
- Configuration (c) is now a **(6, 2, 2, 5, 2) GAST**.

# How to Remove a GAST

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- **Objective of removal:**
  - The code structure and properties are preserved.
  - Manipulating the edge weights such that problematic GASTs are completely removed (not converted into other GASTs).
  
- **We established a set of necessary algebraic conditions such that when we break one of them, the GAST is removed.**  
Our WCM framework [Hareedy 16]

# Algorithm: NB-LDPC Code Optimization

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- **Input: Tanner graph. Output: Optimized Tanner graph.**
  1. Identify the set ( $G$ ) of problematic GASTs.
  2. For each candidate, extract its subgraph from the Tanner graph of the code.
  3. **Determine the set of necessary conditions of that GAST.**
  4. For each condition in that set:
    - Break the condition via the edge weight manipulation.**
  5. If the GAST removal is successful, reflect the edge weight changes in the Tanner graph of the code.
  6. This process continues until all GASTs in  $G$  are removed or no more GASTs can be removed.

# Applications of Our Framework

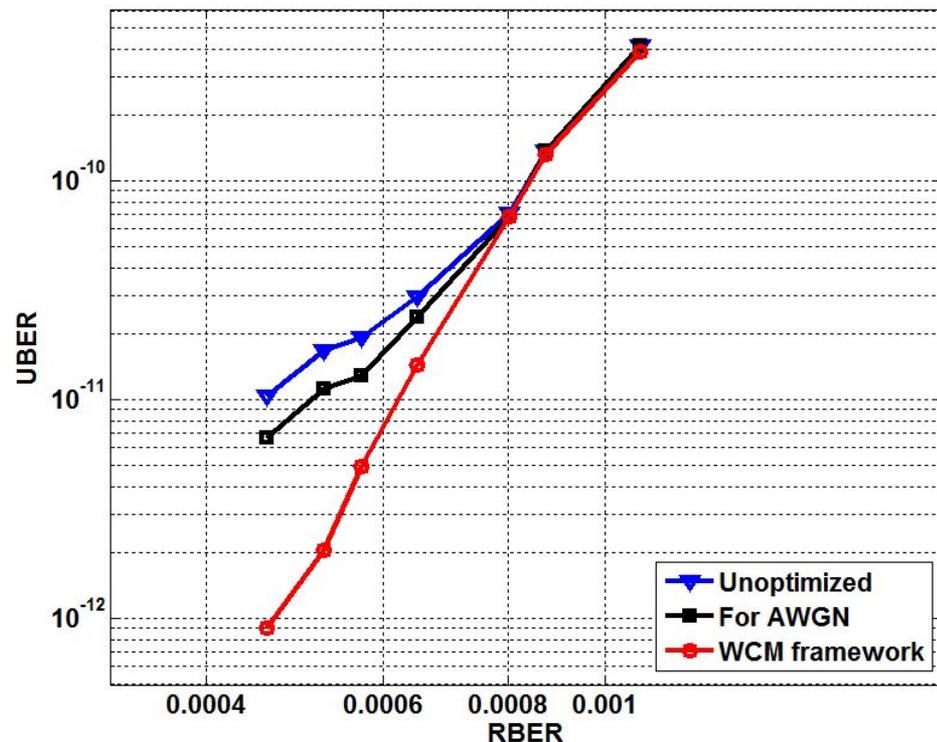
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- **The normal-Laplace mixture (NLM) Flash channel [Parnell 14]:**
  - Accurately models the voltage threshold distribution of sub-20nm MLC (2-bit) NAND Flash memory.
  - Takes into account various sources of error due to wear-out effects (e.g., programming errors).
  - We are using 3 reads (hard decision).
- **Moreover, we test our framework on Cai Flash channel [Cai 13].**
- **Unoptimized codes are designed according to [Bazarsky 13].**

# NB-QC-LDPC Codes with Column Weight 3

- NB-QC-LDPC code: Block-length = 3996 bits, GF(4), rate  $\approx 0.89$ .

Type	Unopt	For AWGN	WCM Framework
(4, 2, 2, 5, 0)	45	0	0
(4, 3, 2, 5, 0)	15	23	0
(4, 4, 4, 4, 0)	3	0	0
(6, 0, 0, 9, 0)	12	1	0
(6, 1, 0, 9, 0)	7	13	0
(6, 2, 0, 9, 0)	3	5	1
(6, 2, 2, 5, 2)	2	2	0
(7, 1, 0, 10, 1)	4	4	0
Other	9	13	8

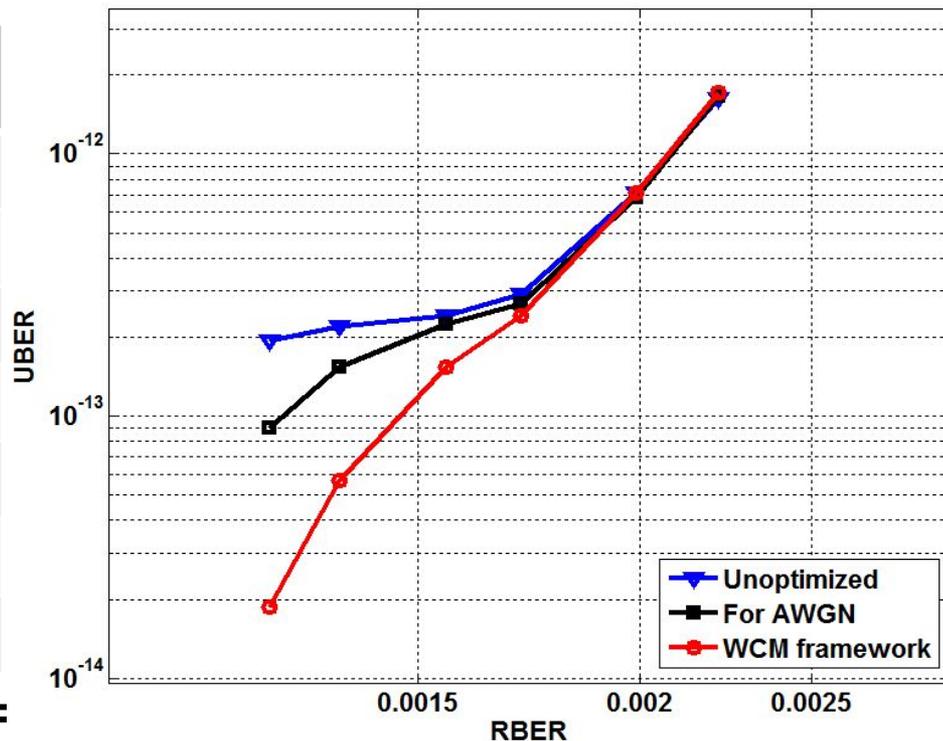


- Tables are extracted at  $\text{RBER} = 4.60 \times 10^{-4}$ 
  - $\text{UBER}(\text{unoptimized}) = 1.04 \times 10^{-11}$ ,  $\text{UBER}(\text{optimized}) = 9.04 \times 10^{-13}$ .

# NB-QC-LDPC Codes with Column Weight 4

- NB-QC-LDPC code: Block-length = 3280 bits, GF(4), rate  $\approx 0.80$ .

Type	Unopt	For AWGN	WCM Framework
(4, 4, 4, 6, 0)	47	2	0
(6, 2, 2, 11, 0)	11	0	0
(6, 4, 2, 11, 0)	14	14	0
(6, 4, 4, 7, 2)	6	6	0
(8, 3, 2, 15, 0)	6	6	1
(8, 4, 4, 14, 0)	5	1	0
Other	11	18	9



- Tables are extracted at RBER =
  - UBER (unoptimized) =  $1.93e-13$ , UBER (optimized) =  $1.86e-14$ .
- Optimizing using WCM gives  $> 1$  order of magnitude gain.
- Optimizing for AWGN (elementary) does not help.

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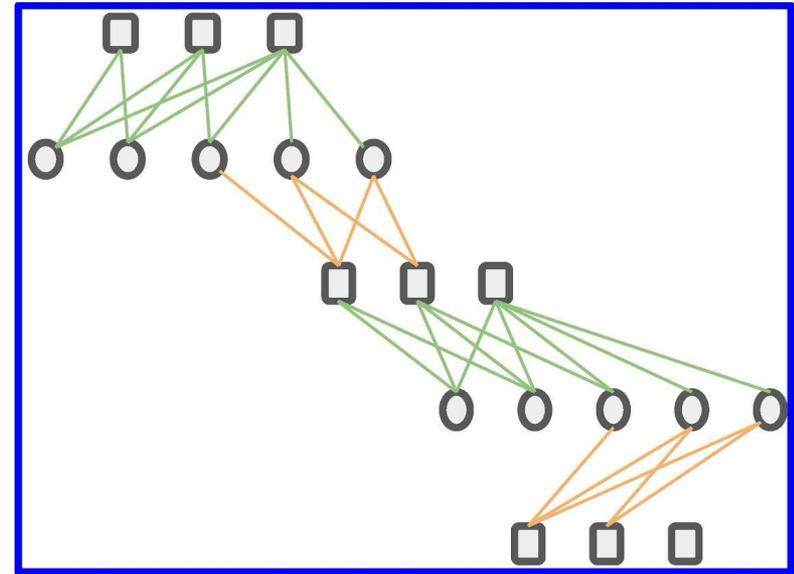
# SC Code Construction: Partitioning and Concatenation

- An spatially-coupled code is formed by coupling replicas of the partitioned sub-matrices together.

$H_{sc} =$

H0	0	0	0	0	0
H1	H0	0	0	0	0
0	H1	H0	0	0	0
0	0	H1	▪	0	0
0	0	0	▪	H0	0
0	0	0	0	H1	H0
0	0	0	0	0	H1

Parity-check matrix of an spatially-coupled (SC) code



Tanner graph corresponding to the sub-matrix with blue borders

# Importance of Finite length Analysis of SC Codes

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- Shown to have excellent performance in the regime of extremely long block lengths when averaged over many codes.
- Many recent papers are on this asymptotic limit:
- [Costello 14] [Urbanke 13] [Lentmaier 15], among others.
- Our research:  
Finite-length performance of spatially-coupled codes

# SC-LDPC Code Optimization

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- We derived a compact mathematical technique for the enumeration of problematic objects of spatially-coupled codes.
- This technique uses the special structure of SC codes to simplify the enumeration of detrimental objects.
- We proposed an optimization framework to find the **optimal cutting vector** that results minimum number of dominant absorbing sets [Amiri 14].

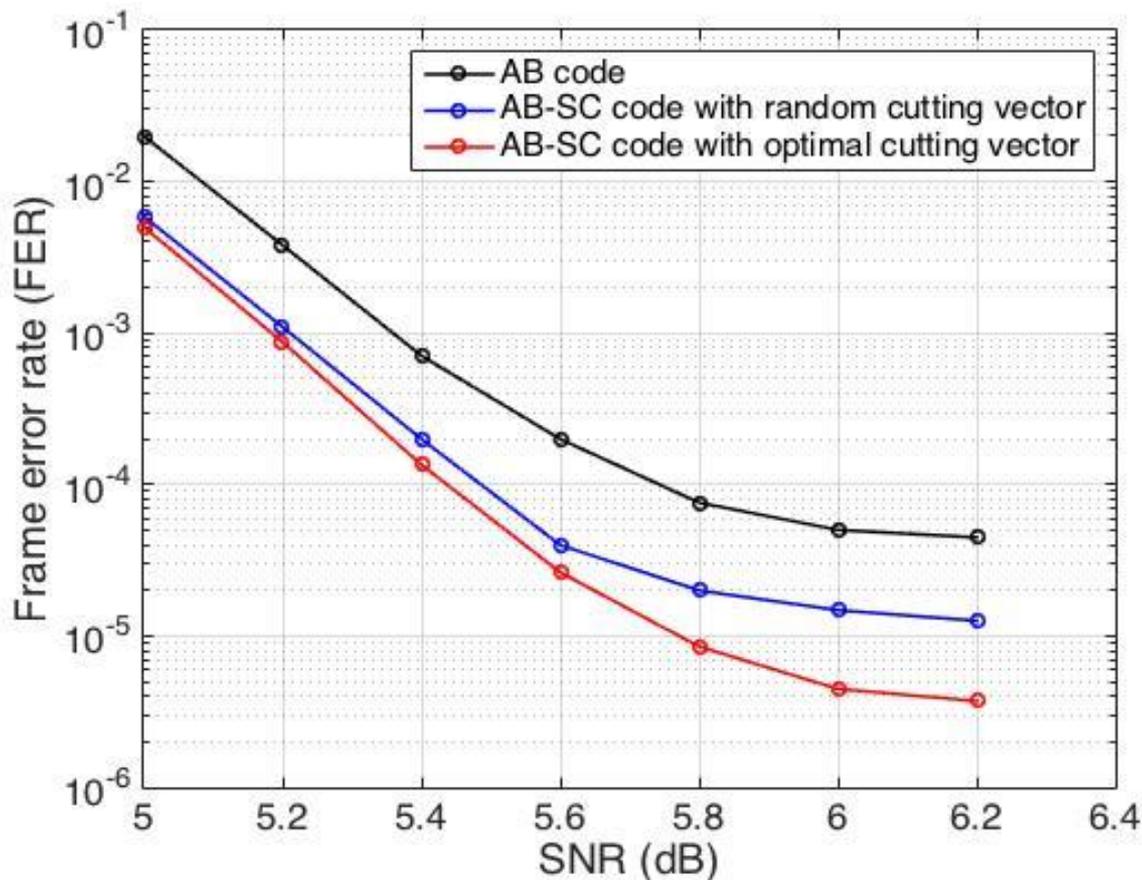
# The Best and Worst Cutting Vector Analysis

Circulant size	Coupling length	The best cutting vector	The worst cutting vector
17	50	[4,8,13] (Equi-partition) number of (3,3) = 99144	[17,17,17] (No coupling) number of (3,3) = 231200
23	100	[5,11,17] (Equi-partition) number of (3,3) = 512302	[23,23,23] (No coupling) number of (3,3) = 1163800
31	100	[7,15,23] (Equi-partition) number of (3,3) = 1288608	[31,31,31] (No coupling) number of (3,3) = 2883000

- “Equi-partition” minimizes the number of (3,3) ASs.
- Due to partitioning the block code, SC codes have fewer detrimental ASs than block-based counterparts.

# Spatially-coupled codes outperform block codes

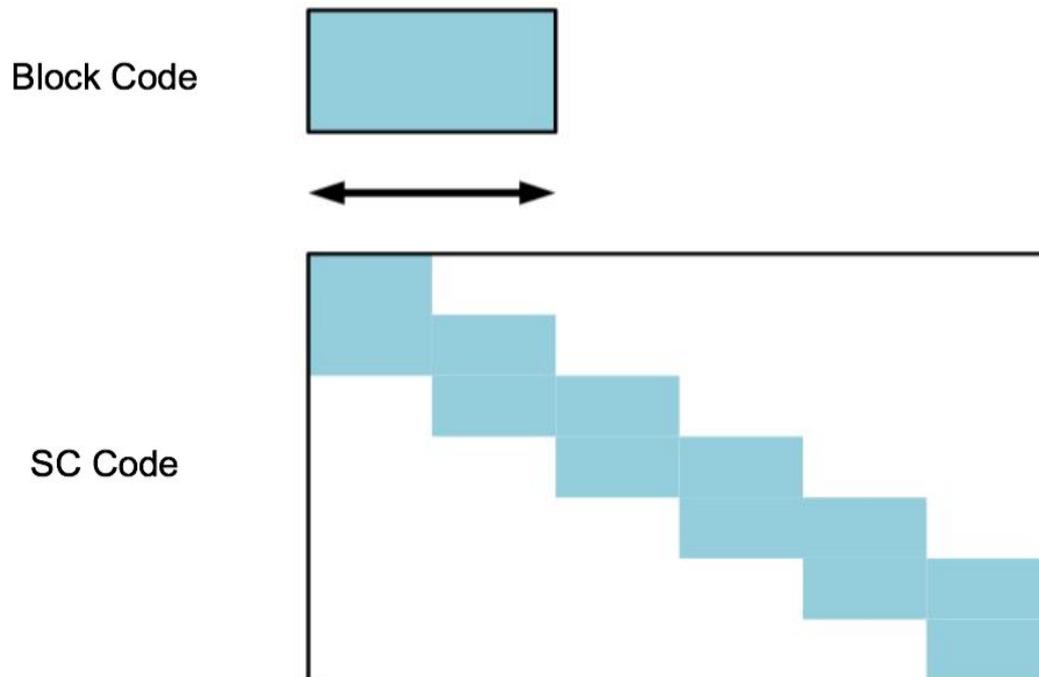
- Binary array-based code with circulant size  $p=67$  and *column weight* 3, and its derived SC codes with coupling length  $L = 50$ .



# Block Codes vs. Spatially-Coupled Codes

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- The research of SC codes often motivated by their “superior” performance (threshold saturation phenomenon, etc.)
- Block codes and SC codes are compared for fixed constraint lengths (equal decoding latency)?



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# Conclusions and Future work

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- **The nature of the errors which dominate the error floor region of NB-LDPC codes in flash memories is different from the AWGN channel.**
  - GASTs are the objects of interest in practical Flash channels.
- **Our framework results at least one order of magnitude performance gain.**
- **We demonstrated that SC codes always have fewer problematic absorbing sets than block codes and that the choice of the cutting vector matters.**
- **Future work includes:**
  - Developing SC code design techniques for flash memories.

# References

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# Thank You