

Probabilistic Design of Multi-Dimensional Spatially-Coupled Codes

Canberk İrimağzı, **Ata Tanrikulu**, Ahmed Hareedy
Middle East Technical University

ISIT 2024
Athens, Greece

Presentation Outline

➤ Motivation and technical vision

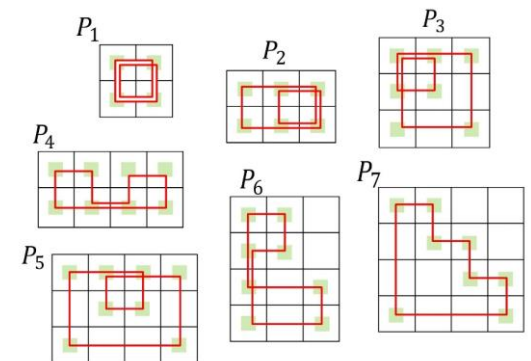
➤ Preliminaries

- ❑ Spatially-Coupled (SC) codes
- ❑ Multi-Dimensional SC (MD-SC) construction

➤ Novel framework for probabilistic MD-SC code design

- ❑ MD Solution form for short cycles
- ❑ Expected number of short cycles
- ❑ Multi-Dimensional gradient-descent distributor (MD-GRADE)
- ❑ Finite-length algorithmic optimizer (FL-AO)
- ❑ Comparison with previous MD-SC schemes

➤ Conclusion and ongoing research



Presentation Outline

➤ Motivation and technical vision

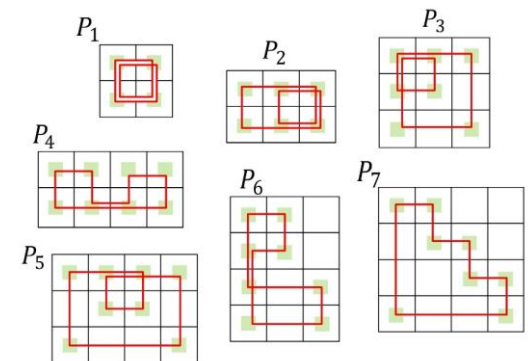
➤ Preliminaries

- ❑ Spatially-Coupled (SC) codes
- ❑ Multi-Dimensional SC (MD-SC) construction

➤ Novel framework for probabilistic MD-SC code design

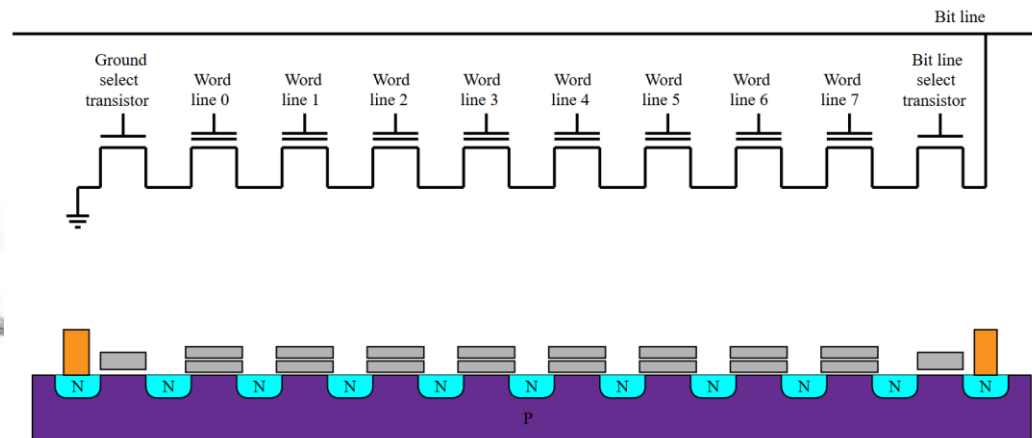
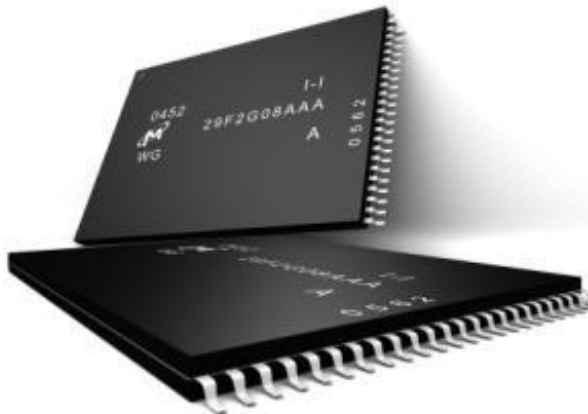
- ❑ MD Solution form for short cycles
- ❑ Expected number of short cycles
- ❑ Multi-Dimensional gradient-descent distributor (MD-GRADE)
- ❑ Finite-length algorithmic optimizer (FL-AO)
- ❑ Comparison with previous MD-SC schemes

➤ Conclusion and ongoing research



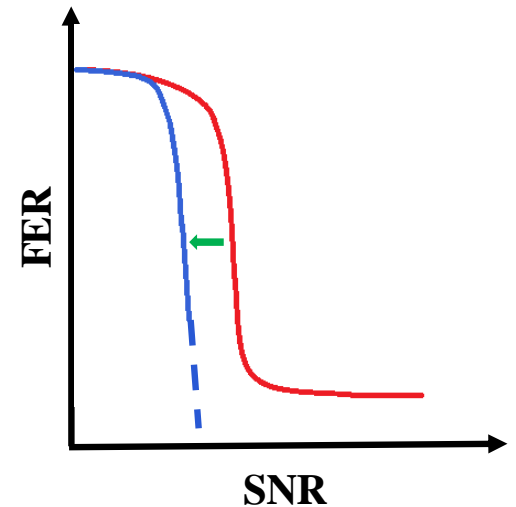
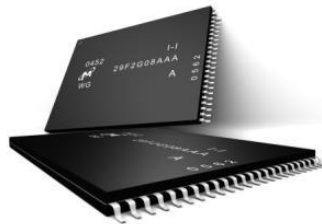
Motivation of the Research

- **We are in the era of big data.**
 - ❑ Modern data centers now have storage capacities in the exabyte range (10^{18} bytes) or higher.
 - ❑ SSDs and HDDs are nearing storage densities of 10 terabits per square inch!
- **These high storage densities increase and intensify error sources in modern storage devices.**
 - ❑ Flash: Inter-cell interference (ICI) and wear-out.



Technical Vision

- **Modern storage devices, including both Flash and magnetic recording devices, must operate at very low error rates.**
 - ❑ Effective ECC techniques are **essential** for allowing storage engineers to confidently use these high-density devices.
 - ❑ Graph-based codes provide excellent performance in this regard!



Our mission is to develop effective ECC techniques that exploit the characteristics of the channels in underlying storage devices.

Presentation Outline

➤ Motivation and technical vision

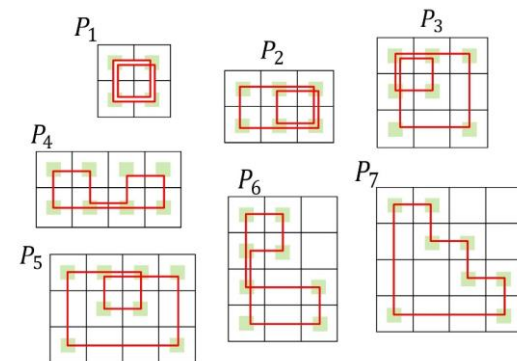
➤ Preliminaries

- ❑ Spatially-Coupled (SC) codes
- ❑ Multi-Dimensional SC (MD-SC) construction

➤ Novel framework for probabilistic MD-SC code design

- ❑ MD Solution form for short cycles
- ❑ Expected number of short cycles
- ❑ Multi-Dimensional gradient-descent distributor (MD-GRADE)
- ❑ Finite-length algorithmic optimizer (FL-AO)
- ❑ Comparison with previous MD-SC schemes

➤ Conclusion and ongoing research



Reminder on SC Codes

➤ **Parameters:** γ, κ, z, m, L m : memory, L : coupling length

➤ **Ingredients:**

❑ A matrix \mathbf{K} of size $\gamma \times \kappa$ whose entries are in $\{0, 1, \dots, m\}$.

❑ A matrix \mathbf{L} of size $\gamma \times \kappa$ whose entries are in $\{0, 1, \dots, z - 1\}$.

➤ **Output:** A parity-check matrix \mathbf{H}_{SC} of size $(\gamma z(m + L)) \times (\kappa z L)$

➤ An underlying block code is partitioned into a number of component matrices:

$$\mathbf{H}_0, \mathbf{H}_1, \dots, \mathbf{H}_m$$

such that

$$\mathbf{H} = \sum_{i=0}^m \mathbf{H}_i,$$

and L copies of **replicas** are coupled together to make a chain of coupled block codes.

$$\mathbf{H}_{\text{SC}} \triangleq \begin{bmatrix} \mathbf{H}_0 & \mathbf{0} & & & & & \mathbf{0} \\ \mathbf{H}_1 & \mathbf{H}_0 & & & & & \vdots \\ \vdots & \mathbf{H}_1 & \ddots & & & & \vdots \\ \mathbf{H}_m & \vdots & \ddots & \ddots & & & \mathbf{0} \\ \mathbf{0} & \mathbf{H}_m & \ddots & \ddots & \ddots & & \mathbf{H}_0 \\ \vdots & \mathbf{0} & & \ddots & \ddots & \ddots & \mathbf{H}_1 \\ \vdots & \vdots & & & \ddots & \ddots & \vdots \\ \mathbf{0} & \mathbf{0} & & & & \ddots & \mathbf{H}_m \end{bmatrix}$$

What is MD Construction?

- **Coupling copies of an SC code notably mitigates channel non-uniformity.**
- **Parameters:** $\gamma, \kappa, z, m, L, M$ M : number of auxiliary matrices
- **The MD-SC code is obtained from M copies of \mathbf{H}_{SC} on the diagonal by**
 - ❑ relocating some of its circulants from each replica of every copy of \mathbf{H}_{SC} to the corresponding locations in \mathbf{X}_l copies, and then by
 - ❑ coupling them in a sliding manner as shown below.
- **For convention, $\mathbf{X}_0 \triangleq \mathbf{H}'_{\text{SC}}$.**

$$\mathbf{H}_{\text{MD}} \triangleq \begin{bmatrix} \mathbf{H}'_{\text{SC}} & \mathbf{X}_{M-1} & \mathbf{X}_{M-2} & \cdots & \mathbf{X}_2 & \mathbf{X}_1 \\ \mathbf{X}_1 & \mathbf{H}'_{\text{SC}} & \mathbf{X}_{M-1} & \cdots & \mathbf{X}_3 & \mathbf{X}_2 \\ \mathbf{X}_2 & \mathbf{X}_1 & \mathbf{H}'_{\text{SC}} & \cdots & \mathbf{X}_4 & \mathbf{X}_3 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \mathbf{X}_{M-2} & \mathbf{X}_{M-3} & \mathbf{X}_{M-4} & \cdots & \mathbf{H}'_{\text{SC}} & \mathbf{X}_{M-1} \\ \mathbf{X}_{M-1} & \mathbf{X}_{M-2} & \mathbf{X}_{M-3} & \cdots & \mathbf{X}_1 & \mathbf{H}'_{\text{SC}} \end{bmatrix},$$

where

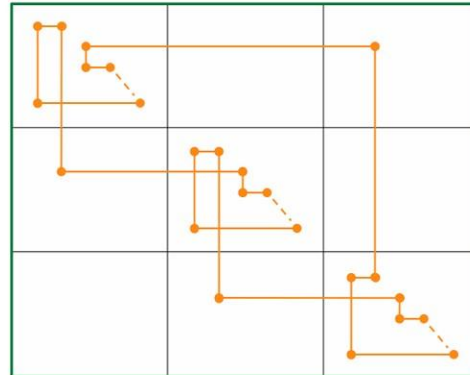
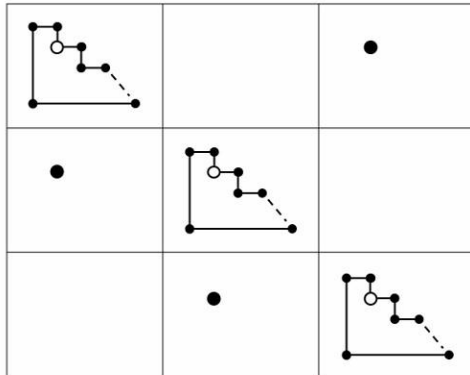
$$\mathbf{H}_{\text{SC}} = \mathbf{H}'_{\text{SC}} + \sum_{\ell=1}^{M-1} \mathbf{X}_\ell.$$

What is an MD Protograph?

- **DEFINITION:** The matrix \mathbf{H}_{MD}^g obtained by replacing \mathbf{H}'_{SC} with $\mathbf{H}'_{\text{SC}}{}^g$ and \mathbf{X}_l 's with \mathbf{X}_l^g 's is called the **MD protograph**.
 - ❑ \mathbf{X}_l 's are obtained by replacing each nonzero (zero) entry in \mathbf{X}_l^g with the $z \times z$ circulant $\sigma^{f_{i,j}}$ (the zero matrix $\mathbf{0}_{z \times z}$) that has the appropriate power $f_{i,j}$ from the lifting matrix \mathbf{L} .
 - ❑ The sum $\mathbf{H}'_{\text{SC}}{}^g + \sum_{l=1}^{M-1} \mathbf{X}_l^g$ gives the SC protograph.
- **Relocations are represented by an MD mapping as follows:**
$$F : \{C_{i,j} | 1 \leq i \leq \gamma, 1 \leq j \leq \kappa\} \rightarrow \{0, 1, \dots, M - 1\},$$
 where
 - ❑ $C_{i,j}$ is the circulant corresponding to 1 at entry (i, j) of the all-one base matrix, and
 - ❑ $F(C_{i,j})$ is the index of the auxiliary matrix to which $C_{i,j}$ is located.
- The percentage of relocated circulants is called the **MD density**.

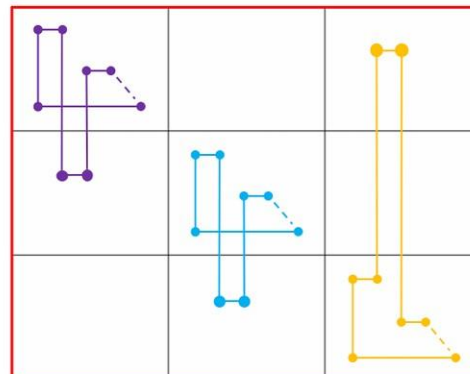
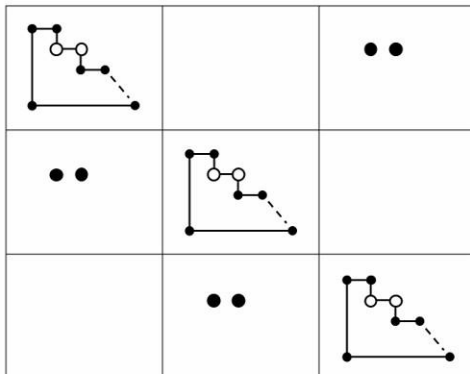
Relocations Eliminate Short Cycles!

➤ EXAMPLE 1 (effective relocation arrangement):



H'_{SC}	X_2	X_1
X_1	H'_{SC}	X_2
X_2	X_1	H'_{SC}

➤ EXAMPLE 2 (ineffective relocation arrangement):



Design complexity increases as we have more degrees of freedom.

Solution: Probabilistic design!

Set-up for Probabilistic Framework

- For $m \geq 1$ and $M \geq 2$, consider the matrix

$$\mathbf{P} = \begin{bmatrix} p_{0,0} & p_{0,1} & \cdots & p_{0,M-1} \\ p_{1,0} & p_{1,1} & \cdots & p_{1,M-1} \\ \vdots & \vdots & \ddots & \vdots \\ p_{m,0} & p_{m,1} & \cdots & p_{m,M-1} \end{bmatrix}_{(m+1) \times M}$$

where

probability of m^{th} component of the 1^{st} auxiliary matrix

$$\sum_{i=0}^m \sum_{j=0}^{M-1} p_{i,j} = 1$$

and each $p_{i,j} \in [0,1]$ specifies the probability that a non-zero circulant in the MD protograph is assigned to the i^{th} component of the j^{th} auxiliary matrix.

- \mathbf{P} is referred as the **(joint) probability-distribution matrix**.

Set-up for Probabilistic Framework

- The coupling polynomial of an MD-SC code associated with the probability-distribution matrix \mathbf{P} is

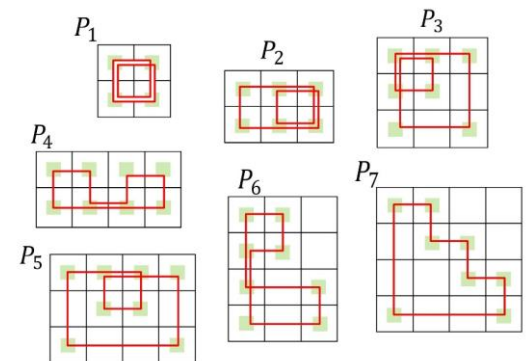
$$f(X, Y, \mathbf{P}) = \sum_{i=0}^m \sum_{j=0}^{M-1} p_{i,j} X^i Y^j$$

which is abbreviated as $f(X, Y)$.

- The vector \mathbf{p}^{con} obtained by concatenating the rows of \mathbf{P} from top to bottom is called the **probability-distribution vector**.
- We express the expected number of cycles-6 and cycles-8 candidates as a function of the probabilities $p_{i,j}$'s after partitioning and relocations.

Presentation Outline

- **Motivation and technical vision**
- **Preliminaries**
 - ❑ Spatially-Coupled (SC) codes
 - ❑ Multi-Dimensional SC (MD-SC) construction
- **Novel framework for probabilistic MD-SC code design**
 - ❑ MD Solution form for short cycles
 - ❑ Expected number of short cycles
 - ❑ Multi-Dimensional gradient-descent distributor (MD-GRADE)
 - ❑ Finite-length algorithmic optimizer (FL-AO)
 - ❑ Comparison with previous MD-SC schemes
- **Conclusion and ongoing research**



MD-SC Solution Form for Cycles-6

- **THEOREM 1:** The expected number $N_6(\mathbf{p}^{\text{con}})$ of cycles-6 in the MD protograph is given by

$$6 \binom{Y}{3} \binom{K}{3} \sum_{M|b} [f^3(X, Y) f^3(X^{-1}, Y^{-1})]_{0,b}$$

where $[\cdot]_{i,j}$ denotes the coefficient of $X^i Y^j$ in a two-variable polynomial.

- **LEMMA 1:** For $N_6(\mathbf{p}^{\text{con}})$ to be locally minimized subject to the constraints

$$p_{i,0} + p_{i,1} + \dots + p_{i,M-1} = p_i^*$$

for all $i \in \{0, 1, \dots, m\}$ and $j \in \{0, 1, \dots, M-1\}$, it is necessary that the following equations hold for some $c_i \in R$:

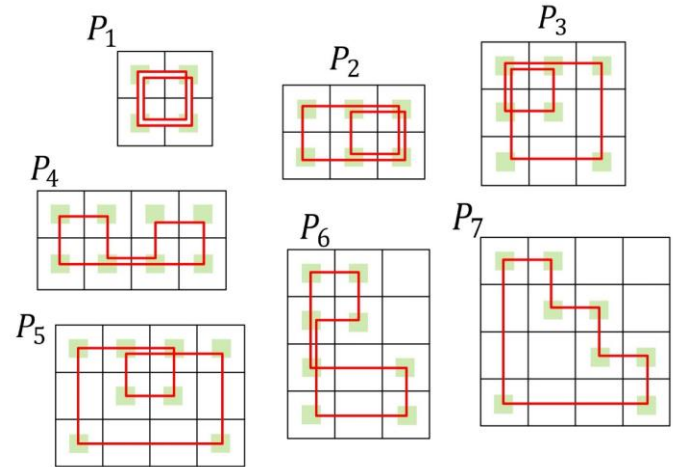
$$\sum_{M|b} [f^3(X, Y) f^2(X^{-1}, Y^{-1})]_{i,b+j} = c_i.$$

- Here, p_i^* 's are obtained by GRADE-AO Algorithm in [R1].

Cycles-8 Candidates

➤ **THEOREM 2:** The expected number $N_8(\mathbf{p}^{\text{con}})$ of cycles-8 candidates in the MD protograph is given by

$$\begin{aligned}
 N_8(\mathbf{p}^{\text{con}}) &= \sum_{M|b} \left\{ w_1 [f^2(X, Y) f^2(X^{-1}, Y^{-1})]_{0,b} \right. \\
 &+ w_2 [f(X^2, Y^2) f(X^{-2}, Y^{-2}) f^2(X, Y) f^2(X^{-1}, Y^{-1})]_{0,b} \\
 &+ w_3 [f(X^2, Y^2) f^2(X, Y) f^4(X^{-1}, Y^{-1})]_{0,b} \\
 &\left. + w_4 [f^4(X, Y) f^4(X^{-1}, Y^{-1})]_{0,b} \right\},
 \end{aligned}$$



where $w_1 = \binom{\gamma}{2} \binom{\kappa}{2}$, $w_2 = 3 \binom{\gamma}{2} \binom{\kappa}{3} + 3 \binom{\gamma}{3} \binom{\kappa}{2}$, $w_3 = 18 \binom{\gamma}{3} \binom{\kappa}{3}$, $w_4 = 6 \binom{\gamma}{2} \binom{\kappa}{4} + 6 \binom{\gamma}{4} \binom{\kappa}{2} + 36 \binom{\gamma}{3} \binom{\kappa}{4} + 36 \binom{\gamma}{4} \binom{\kappa}{3} + 24 \binom{\gamma}{4} \binom{\kappa}{4}$ if $\gamma \geq 4$, and $w_4 = 6 \binom{\gamma}{2} \binom{\kappa}{4} + 36 \binom{\gamma}{3} \binom{\kappa}{4}$ if $\gamma = 3$, where $\kappa \geq 4$.

Expected Number of Cycles

- We compute the expected number of cycles- k , $k \in \{6,8\}$, under a specific MD probability distribution.
 - ❑ This gives an estimate of what the finite-length algorithmic optimizer (FL-AO) algorithm can produce.
 - ❑ These numbers inform us what to expect from incorporating the probability-distribution matrix in designing gradient-descent MD-SC (GD-MD) codes under **random** partitioning and lifting.
- **THEOREM 3:** After random partitioning, relocations, and lifting based on \mathbf{p}^{con} , the **expected number of cycles-6** in the Tanner graph of \mathbf{H}_{MD} is

$$\approx N_6(\mathbf{p}^{\text{con}}) * \frac{2L - m}{2} * M$$

and the **expected number of cycles-8** in the Tanner graph of \mathbf{H}_{MD} is

$$\approx N_8(\mathbf{p}^{\text{con}}) * (L - m) * M.$$

MD-GRADE Algorithm

- **The algorithm is designed to find**
 - ❑ the probability distribution matrix of an MD-SC code with arbitrary number of auxiliary matrices and memory,
 - ❑ to estimate the expected number of short cycles along with their upper and lower bounds.

- **A modified version of the gradient-descent algorithm is employed to handle the constraints outlined in Lemma 1, initiated with the locally-optimal edge distribution of the underlying SC code, obtained as in [R1].**

- **MD-GRADE provides a **non-trivial** solution where**
 - ❑ relocation percentages of component matrices are not necessarily the same;
 - ❑ typically, more circulants need to be relocated from the middle component matrices of the SC code than from the side ones.

Finite-length Algorithmic Optimizer (FL-AO)

- **The algorithm to construct an MD-SC code starting from an underlying SC code through performing relocations is introduced in [R2] by Esfahanizadeh et al. and modified in [R3] to achieve more reduction in cycle counts.**
- **Our FL-AO initially starts with**
 - ❑ a random relocation based on a locally-optimal GD distribution,
 - ❑ performs relocations based on the majority rule, and makes random decisions between the best options if there are multiple options.
- **The GD-MD distribution results in a significant reduction in the search space of all possible relocation arrangements that the FL-AO operates on.**
 - ❑ This remarkably reduces the framework complexity and latency.
- **FL-AO converges on excellent finite-length MD-SC designs in few iterations.**

Parameters of Codes

- **SC Code 1** is an SC code with parameters $(\gamma, \kappa, z, L, m) = (4, 17, 17, 10, 1)$ and girth 6 [R3].
- **SC Code 1.1** is similar to SC Code 1, but with $L = 30$.
- **GD-MD Code 1.1** with $M = 3$ has final MD density = 33.82% and is obtained after 12 iterations.
- **SC Code 3** is an SC code with parameters $(\gamma, \kappa, z, L, m) = (3, 19, 23, 10, 2)$ and girth 8 [R3].
- **SC Code 3.1** is similar to SC Code 3, but with $L = 40$.
- **GD-MD Code 3.1** with $M = 4$ has final MD density = 31.58% after 11 iterations.

Code name	Length	Rate
GD-MD Code 1.1	8,670	0.74
SC Code 1.1	8,670	0.76
GD-MD Code 2.1-2.2	210,250	0.81
MD-SC (NR)	210,250	0.81
GD-MD Code 3.1	17,480	0.81
SC Code 3.1	17,480	0.83
GD-MD Code 4.1	107,100	0.81
SC Code 4.1	107,100	0.82
GD-MD Code 4.2	154,700	0.81
SC Code 4.2	154,700	0.82
GD-MD Code 5	43,350	0.75
TC Code 1.1	43,350	0.76

Remarkable Reduction in Cycle Counts!

- The following table compares the population of cycles of interest in our proposed GD-MD codes with their 1D-SC counterparts and/or those in the literature [R3].

Code name	Cycle-6 count
GD-MD Code 1.1	6,375
[R3]	9,078
SC Code 1.1	79,917
GD-MD Code 5	0
TC Code 1.1	47,736

Code name	Cycle-8 count
GD-MD Code 2.2	2,768,485
MD-SC (NR)	16,809,705
GD-MD Code 3.1	239,752
[R3]	249,320
SC Code 3.1	1,397,319
GD-MD Code 4.1	112,959
SC Code 4.1	3,819,480
GD-MD Code 4.2	92,001
SC Code 4.2	5,530,280

Better Cycle Counts with Higher Lengths!

- **Significant reduction in**
 - ❑ the number of cycles-6 that ranges between **92% and 100%** and
 - ❑ the number of cycles-8 that ranges between **83% and 98%** compared with 1D SC/TC codes of the same lengths.
- **More intriguingly, our GD MD codes offer lower cycle counts compared with their underlying SC/TC codes, which have notably lower lengths.**
- **Despite that the final MD densities are in the range **25% – 35%**, the number of iterations is always **below 20**.**
- **GD-MD codes are constructed in a computationally fast manner. In fact,**
 - ❑ [R3, Algorithm 2] (with no relocations initially) terminates at MD density 21% in ≥ 21 iterations, whereas
 - ❑ our FL-AO terminates after **11 iterations** only, to design our GD-MD Code 3.1, yielding fewer cycles-8.

Strength of Theorem 3

- The final number of cycles we obtain after FL-AO Algorithm is lower than the upper bound on the expected number of cycles obtained in Theorem 3.
- For GD-MD Code 3.1, however, Theorem 3 gives a good indicator of the outcome of FL AO Algorithm, where the (rounded) expected number of cycles-8 is **228,070** and the actual final number of cycles-8 is **239,752**.
- With Theorem 3 in hand, we are able to answer two major questions:
 - ❑ What is the estimated percentage of relocations required to remove all instances of a cycle, assuming it can be removed entirely?
 - ❑ What is the optimal approach to handling a cycle given a pre-specified maximum relocation percentage, imposed by decoding latency requirements?

Presentation Outline

➤ Motivation and technical vision

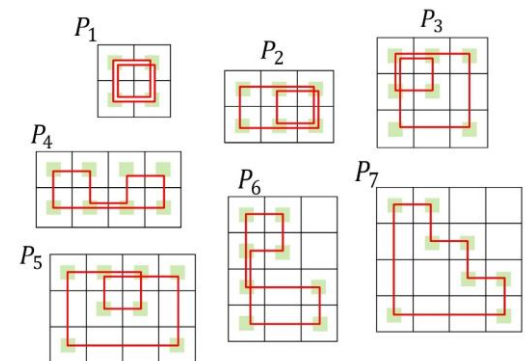
➤ Preliminaries

- ❑ Spatially-Coupled (SC) codes
- ❑ Multi-Dimensional SC (MD-SC) construction

➤ Novel framework for probabilistic MD-SC code design

- ❑ MD Solution form for short cycles
- ❑ Expected number of short cycles
- ❑ Multi-Dimensional gradient-descent distributor (MD-GRADE)
- ❑ Finite-length algorithmic optimizer (FL-AO)
- ❑ Comparison with previous MD-SC schemes

➤ Conclusion and ongoing research



Conclusion and Ongoing Research

- **Our MD-GRADE algorithm provides us with a locally-optimal distribution matrix that guides the FL-AO by reducing its search space.**
- **When fed with a locally-optimal probability-distribution matrix, the FL-AO can converge on excellent finite-length MD-SC designs in a notably fast manner.**
- **Our GD-MD codes achieve notable reduction in the number of short cycles.**
- **Future work includes**
 - ❑ **simulating the codes' performance/density/lifetime gains and**
 - ❑ **extending the analysis to objects that are more advanced than short cycles.**

References

- [R1] S. Yang, A. Hareedy, R. Calderbank, and L. Dolecek, “Breaking the computational bottleneck: Probabilistic optimization of high-memory spatially-coupled codes,” *IEEE Trans. Inf. Theory*, 2023.
- [R2] H. Esfahanizadeh, A. Hareedy, and L. Dolecek, “Multi-dimensional spatially-coupled code design through informed relocation of circulants,” in *Proc. 56th Annual Allerton Conf. Commun., Control, and Computing*, 2018.
- [R3] H. Esfahanizadeh, L. Taus, and L. Dolecek, “Multi-dimensional spatially-coupled code design: Enhancing the cycle properties,” *IEEE Trans. Commun.*, 2020.
- [R4] H. Esfahanizadeh, A. Hareedy, and L. Dolecek, “Finite-length construction of high performance spatially-coupled codes via optimized partitioning and lifting,” *IEEE Trans. Commun.*, 2019.
- [R5] D. Truhachev, D. G. M. Mitchell, M. Lentmaier, and D. J. Costello, “New codes on graphs constructed by connecting spatially coupled chains,” in *Proc. ITA*, 2012.
- [R6] C. İrimağzı, A. Tanrıkulu, and A. Hareedy, “Probabilistic design of multi-dimensional spatially-coupled codes,” 2024. [Online]. Available: <https://arxiv.org/abs/2401.15166>

Thank You!