

Spatial Coupling vs. Block Coding: A Comparison



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Coding: From Theory to Practice
UC Berkeley, Feb 9th-13th 2015

Research Collaborators: David Mitchell,
Michael Lentmaier, and Ali Pusane

■ LDPC Block Codes

- ➔ Parity-check matrix and Tanner graph representations, minimum distance bounds, iterative decoding thresholds, regular and irregular code designs, protograph-based constructions

■ Spatially Coupled LDPC Codes

- ➔ Protograph representation, edge-spreading construction, termination
- ➔ Iterative decoding thresholds, threshold saturation, minimum distance

■ Practical Considerations

- ➔ Finite-length scaling, window decoding, performance, latency, and complexity comparisons to LDPC block codes, implementation aspects

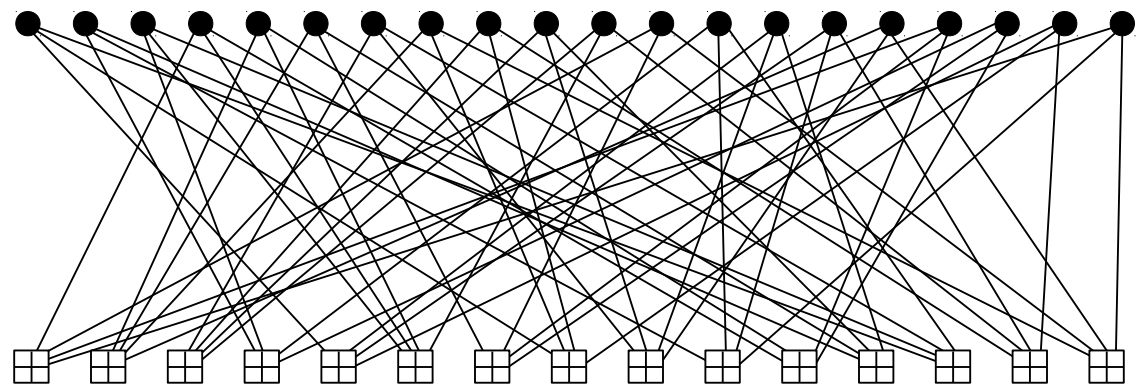
Definition by parity-check matrix: [Gallager, '62]

Bipartite graph representation: [Tanner, '81]

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15 × 20

$n = 20$ variable nodes of degree $J = 3$



$l = 15$ check nodes of degree $K = 4$

Code: $\{ \mathbf{v} \mid \mathbf{v} \mathbf{H}^T = \mathbf{0} \}$

(J,K)-regular LDPC code: $R \geq 1 - \frac{J}{K}$

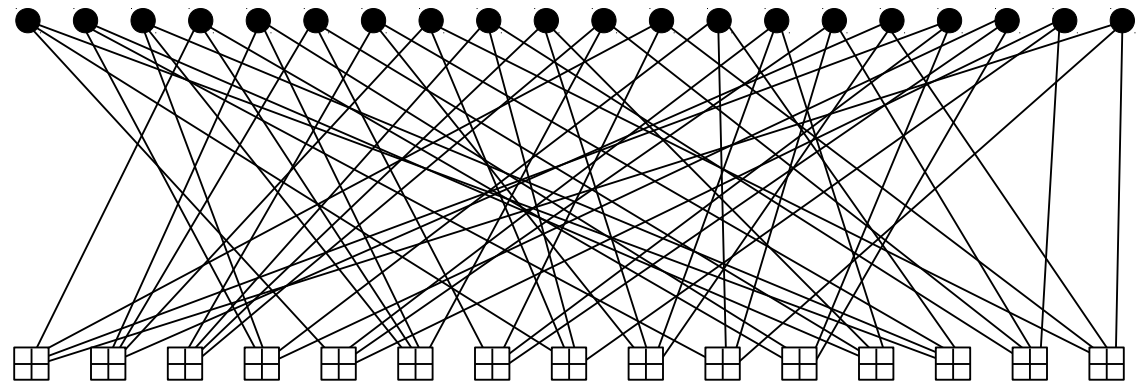
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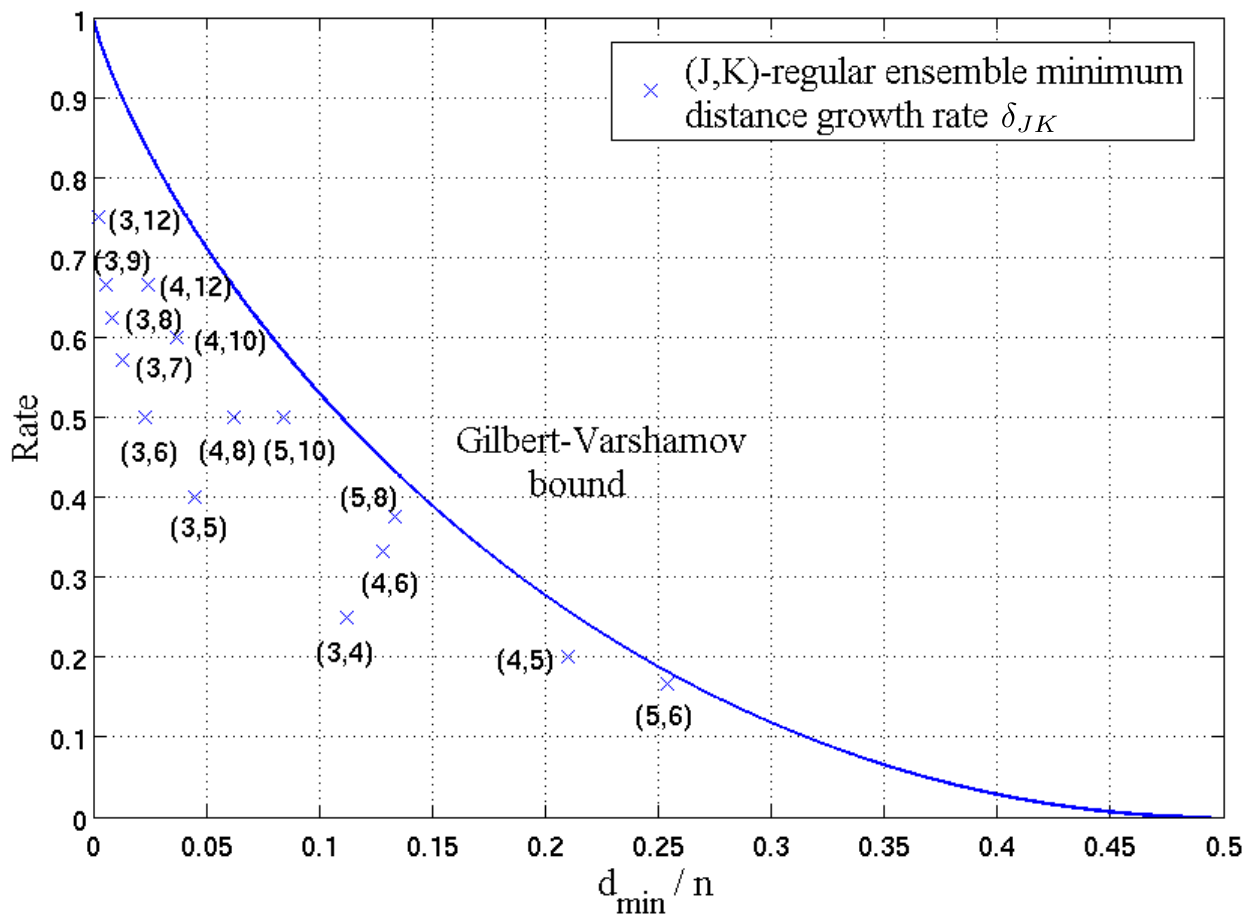
Code: $\{ \mathbf{v} \mid \mathbf{vH}^T = \mathbf{0} \}$

(J,K)-regular LDPC code: $R \geq 1 - \frac{J}{K}$

- Graph-based codes can be decoded **iteratively** with **low-complexity** by exchanging messages in the graph using **Belief Propagation (BP)**.

Code Ensembles – Minimum Distance Growth Rates

- For an **asymptotically good** code ensemble, the minimum distance grows linearly with the block length n



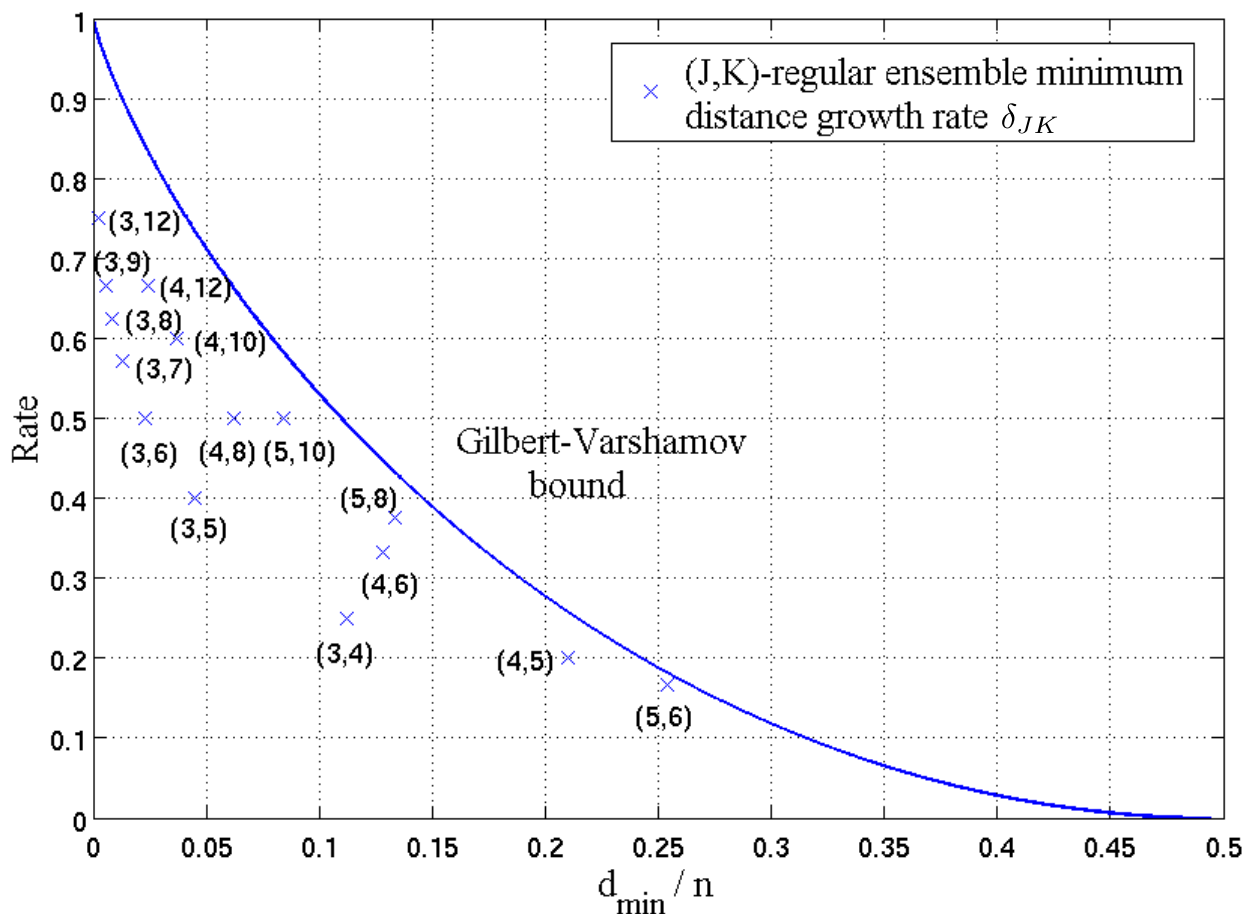
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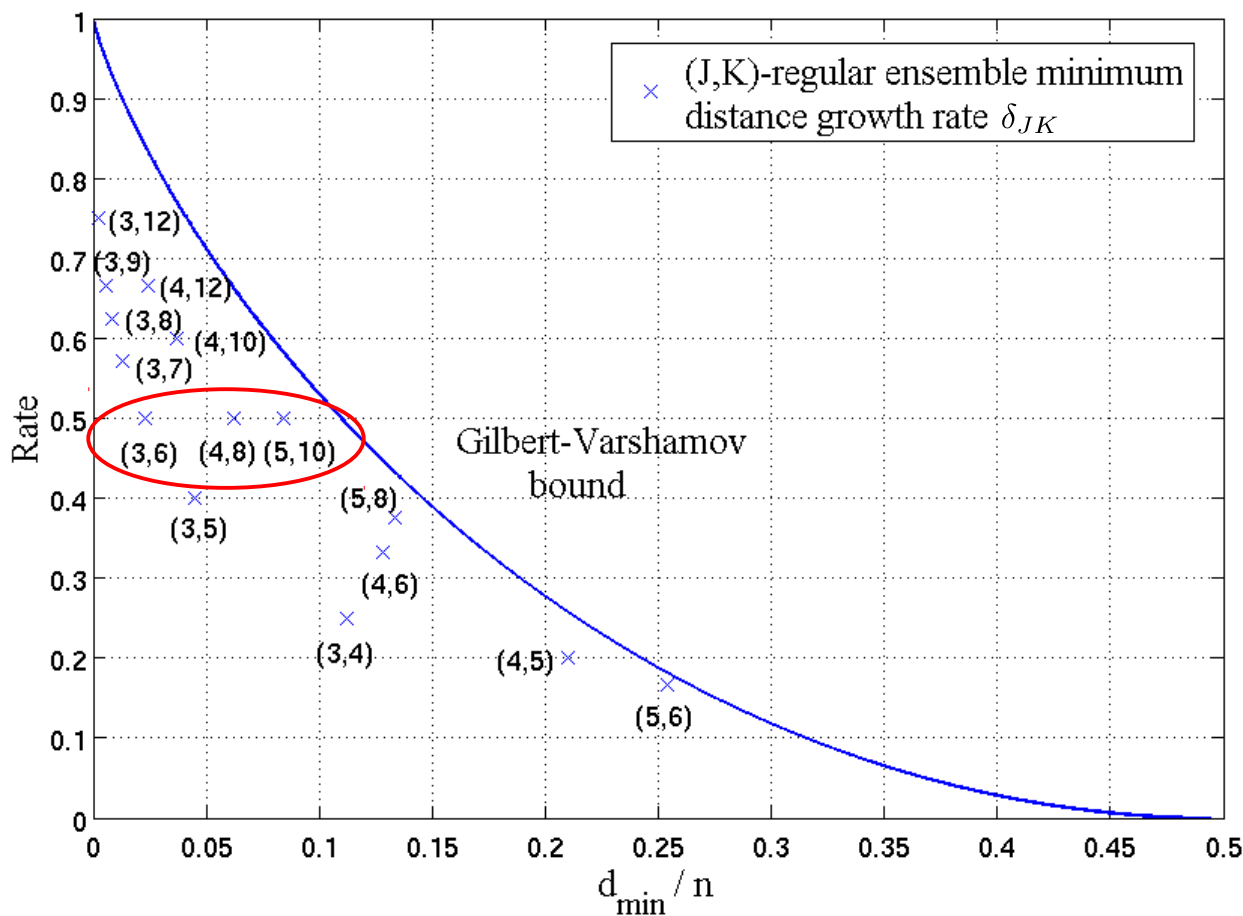
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where δ_{JK} is called the **typical minimum distance ratio**, or **minimum distance growth rate**



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- As the density of (J,K) -regular ensembles increases, δ_{JK} approaches the **Gilbert-Varshamov bound**.

Thresholds of (J,K) -regular LDPC Block Code Ensembles

- Iterative decoding thresholds can be calculated for (J,K) -regular LDPC block code ensembles using density evolution (DE).

BEC thresholds

J	K	Rate	ε^*	ε_{Sh}
3	6	0.5	0.429	0.5
4	8	0.5	0.383	0.5
5	10	0.5	0.341	0.5
3	5	0.4	0.517	0.6
4	6	0.333	0.506	0.667
3	4	0.25	0.647	0.75

AWGNC thresholds

J	K	Rate	$(E_b/N_0)^*$	$(E_b/N_0)_{\text{Sh}}$
3	6	0.5	1.11	0.184
4	8	0.5	1.61	0.184
5	10	0.5	2.04	0.184
3	5	0.4	0.96	-0.229
4	6	0.333	1.67	-0.480
3	4	0.25	1.00	-0.790

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- There exists a relatively **large gap to capacity**.
- Iterative decoding thresholds get **further from capacity** as the graph **density increases**.

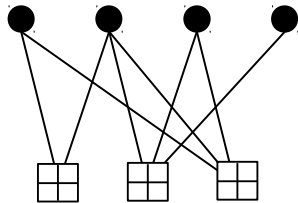
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- **Compact representation** of a structured LDPC block code ensemble with code length $n = Mb_v$ and code design rate $R \geq (b_v - b_c)/b_v$

$$\mathbf{B} = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 0 \end{bmatrix}_{b_c \times b_v}$$

base matrix

b_v variable nodes



b_c check nodes

protograph

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Protograph Representation

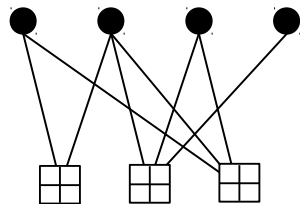
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base matrix

parity-check matrix

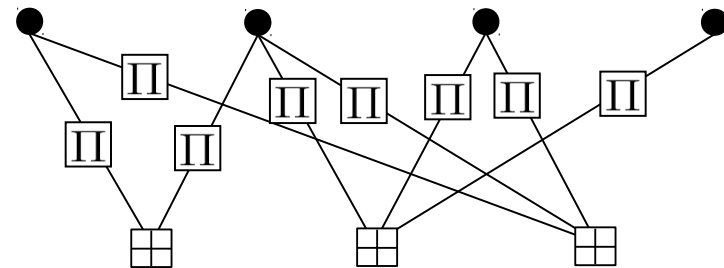
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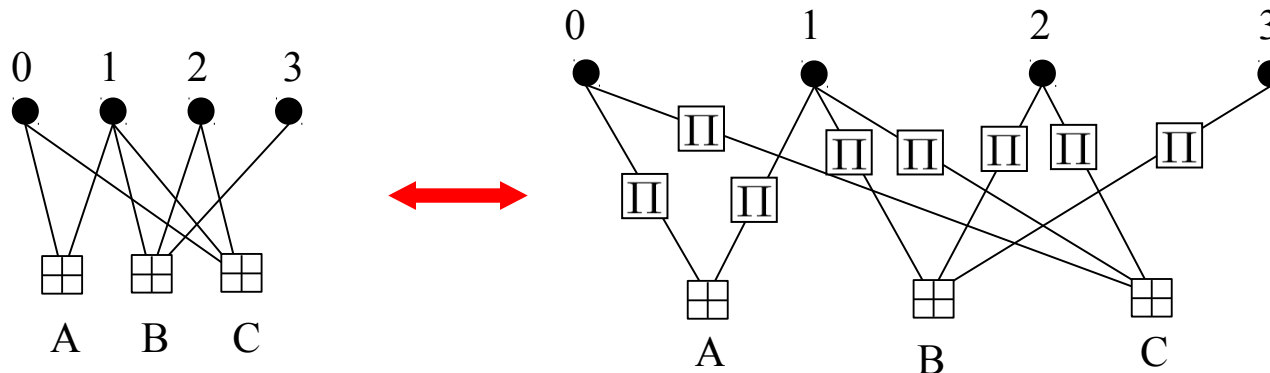
Tanner graph

lifting factor M

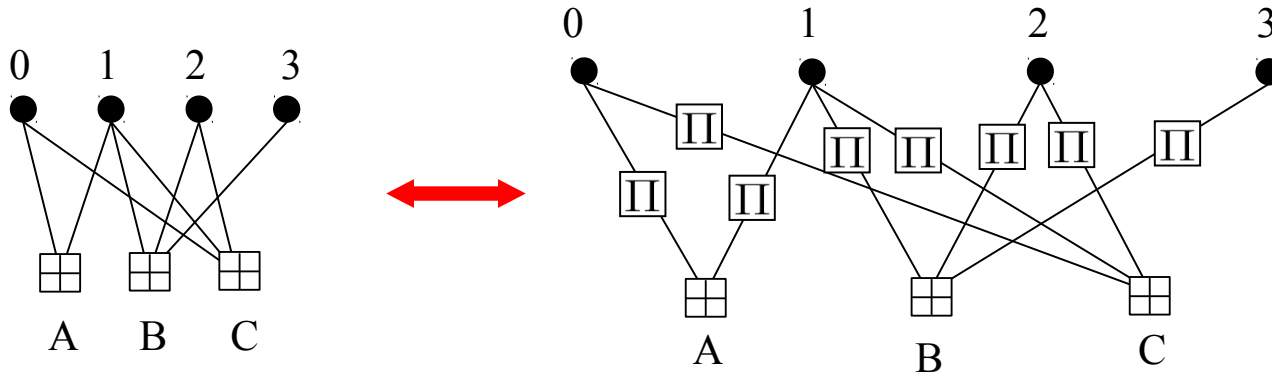
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Protograph Analysis

- By design, every member of a protograph-based ensemble **preserves the structure** of the base protograph.

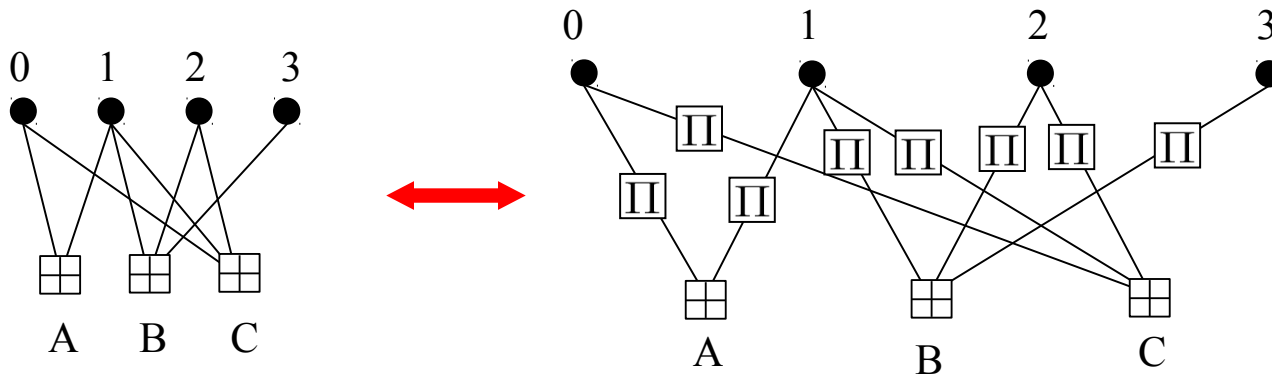


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- ➔ From the protograph, an expression [Divsalar '06] can be obtained for the **ensemble average weight enumerator**,

$$\overline{A}(z) = \sum_{d=0}^n \overline{A}_d z^d, \quad \left(\overline{A}_d = \text{avg. number of codewords of weight } d \right)$$

which can be used to test if the ensemble is **asymptotically good**.

■ LDPC Block Codes

- ➔ Parity-check matrix and Tanner graph representations, minimum distance bounds, iterative decoding thresholds, regular and irregular code designs, protograph-based constructions

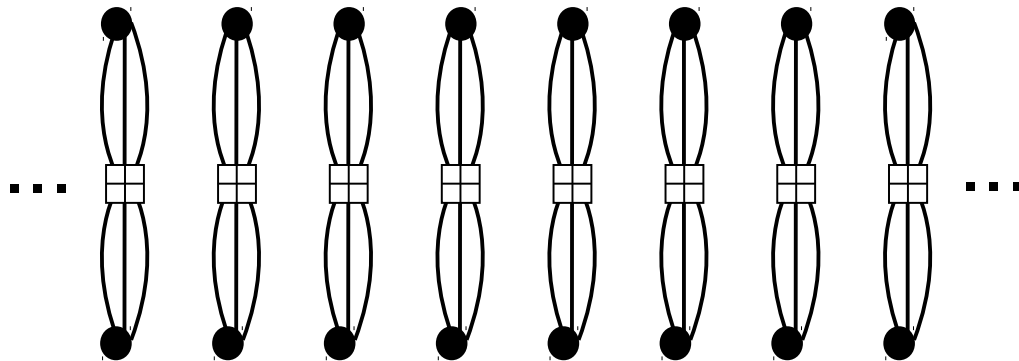
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■ Practical Considerations

- ➔ Finite-length properties, window decoding, comparison to block codes, implementation aspects

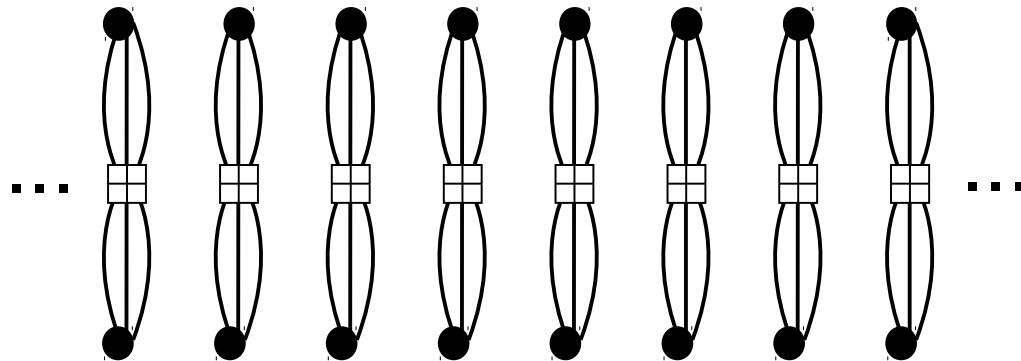
- Consider transmission of consecutive blocks (protograph representation):



$$\mathbf{B} = \begin{bmatrix} 3 & 3 \end{bmatrix}$$

(3,6)-regular
LDPC-BC
base matrix

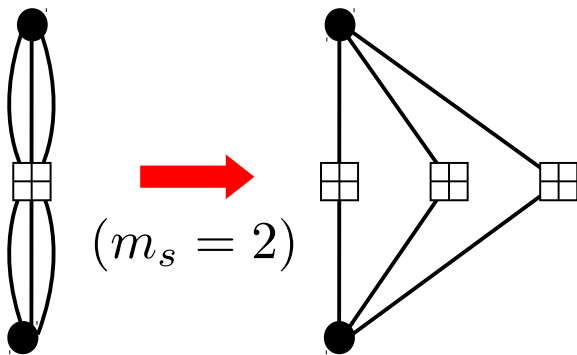
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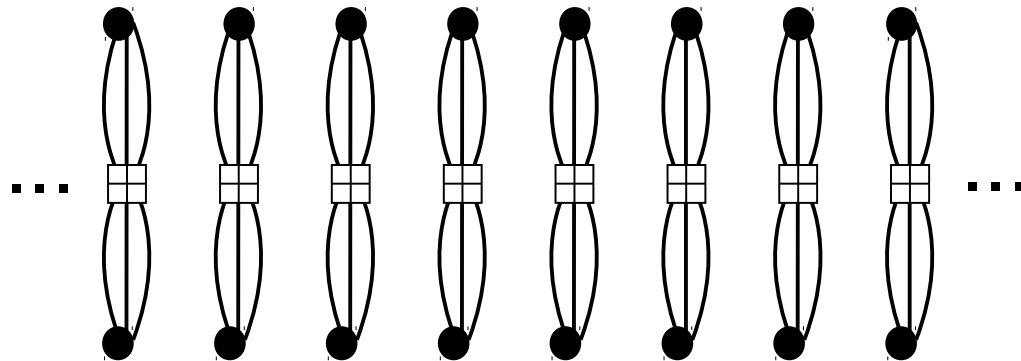
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- Blocks are **spatially coupled** (introducing **memory**) by **spreading edges** over time:



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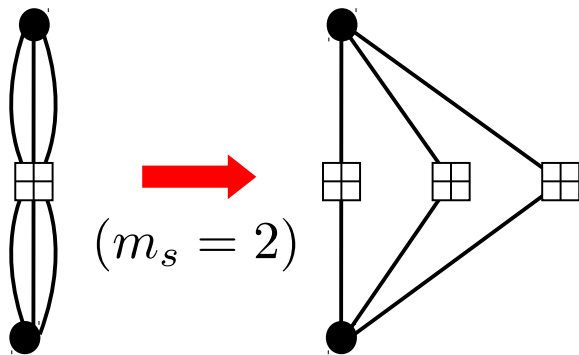
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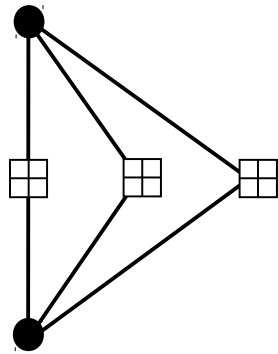


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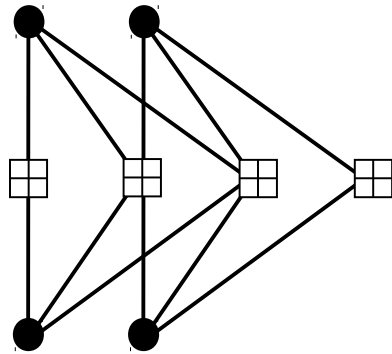
- Spreading constraint:**

$$\sum_{i=0}^{m_s} \mathbf{B}_i = \mathbf{B}$$

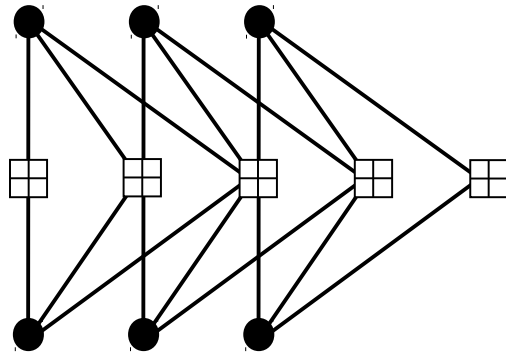
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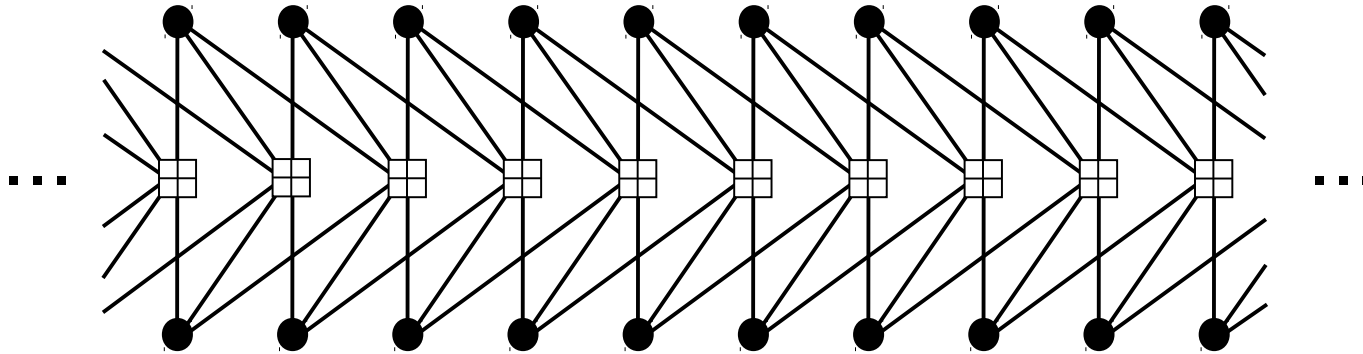
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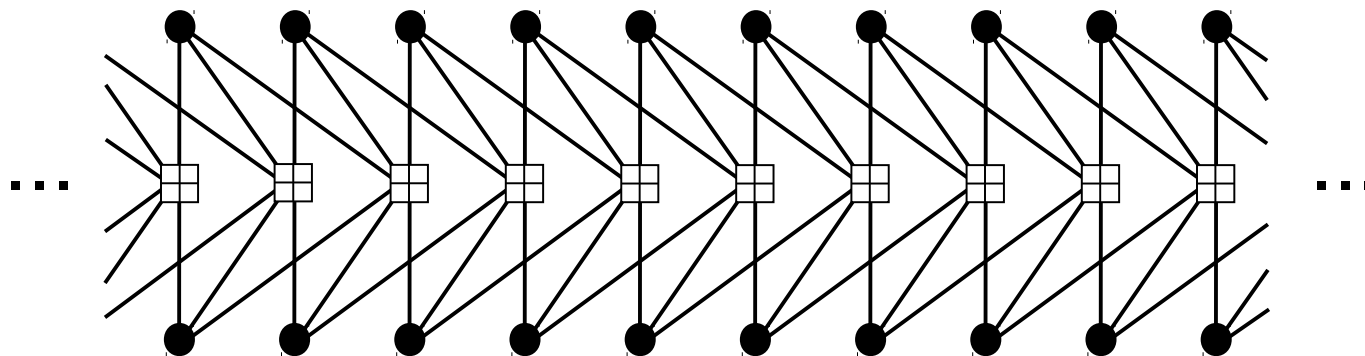
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- The bi-infinite convolutional protograph corresponds to a bi-infinite **convolutional base matrix**:

$$\mathbf{B}_{[-\infty, \infty]} = \left[\begin{array}{ccccccc}
 \ddots & & & & & & \\
 & \mathbf{B}_{m_s} & \cdots & \mathbf{B}_1 & \mathbf{B}_0 & & \\
 & & \ddots & & \ddots & \ddots & \\
 & & & \mathbf{B}_{m_s} & \cdots & \mathbf{B}_1 & \mathbf{B}_0 \\
 & & & & \ddots & \ddots & \ddots
 \end{array} \right]$$

\mathbf{B}_i has size $b_c \times b_v$
Rate:

$$R = \frac{b_v - b_c}{b_v}$$
Constraint length:

$$\nu_s = b_v(m_s + 1)$$

Terminated Spatially Coupled Codes

- Consider **terminating** $\mathbf{B}_{[-\infty, \infty]}$ to a (block code) **base matrix** of length Lb_v :

$$\mathbf{B}_{[0, L-1]} = \begin{bmatrix} \mathbf{B}_0 & & & \\ \vdots & \ddots & & \\ \mathbf{B}_{m_s} & & \mathbf{B}_0 & \\ & & \vdots & \\ & & \mathbf{B}_{m_s} & \end{bmatrix} \quad (L+m_s)b_c \times Lb_v$$

(\mathbf{B}_i is a $b_c \times b_v$ matrix)

Code rate:

$$R_L = \frac{Lb_v - (L + m_s)b_c}{Lb_v}.$$

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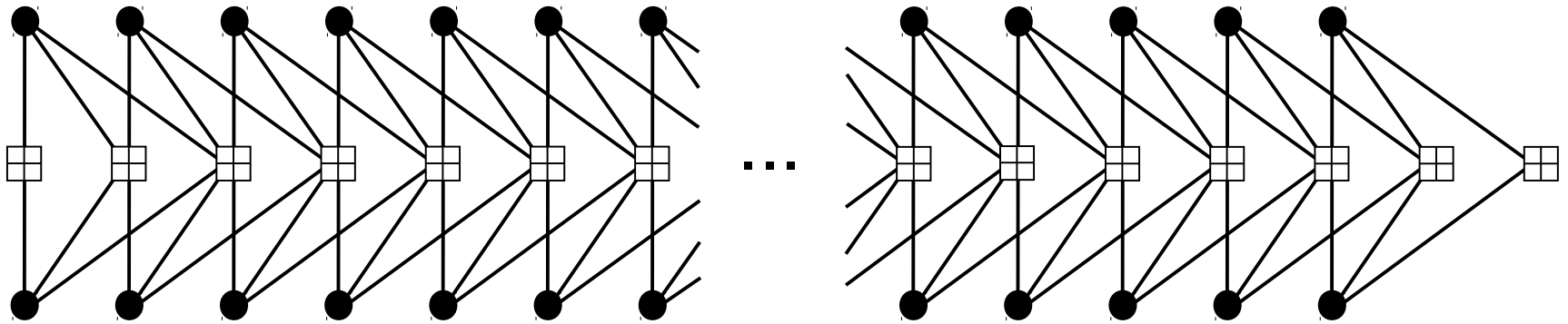
Code rate:

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- For large L , R_L approaches the **unterminated** code rate $R = (b_v - b_c)/b_v$.

Wave-like Decoding of Terminated Spatially Coupled Codes

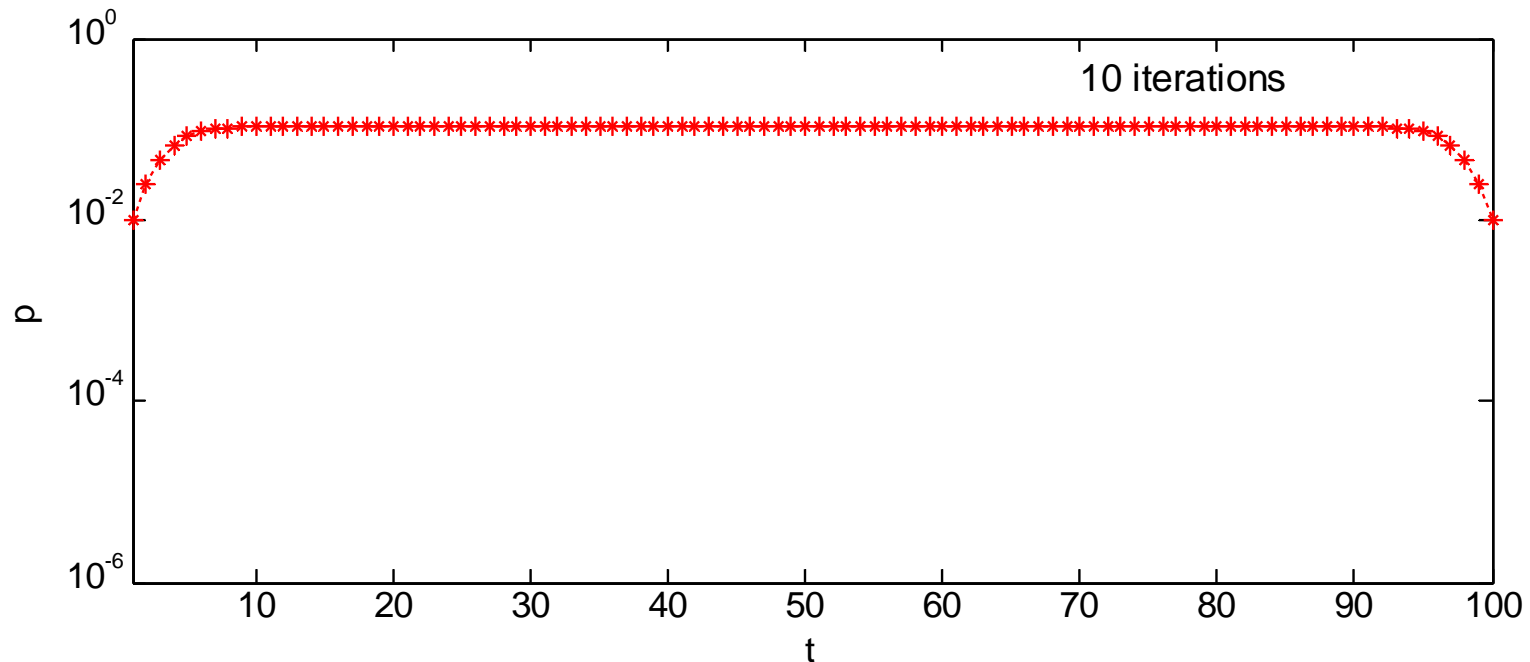
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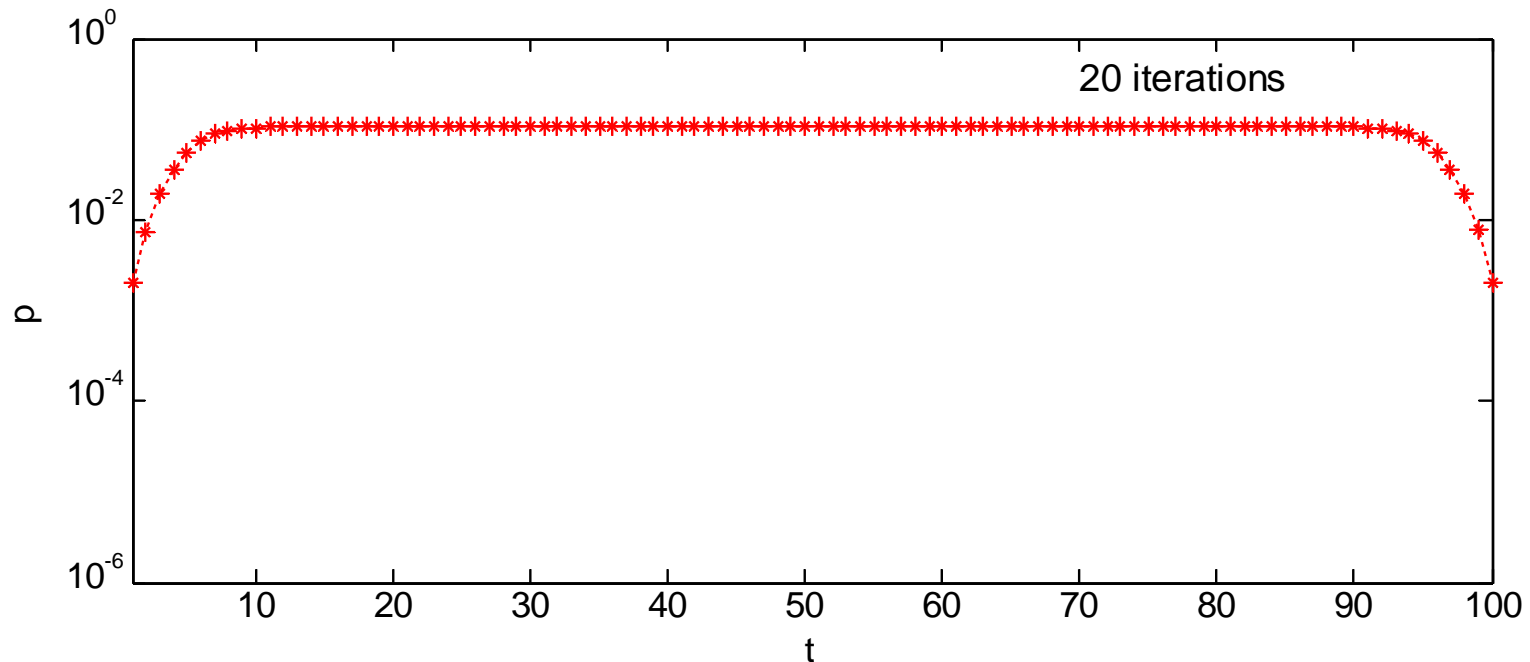
Evolution of message probabilities: (3,6)-regular SC-LDPC code ($L = 100$)



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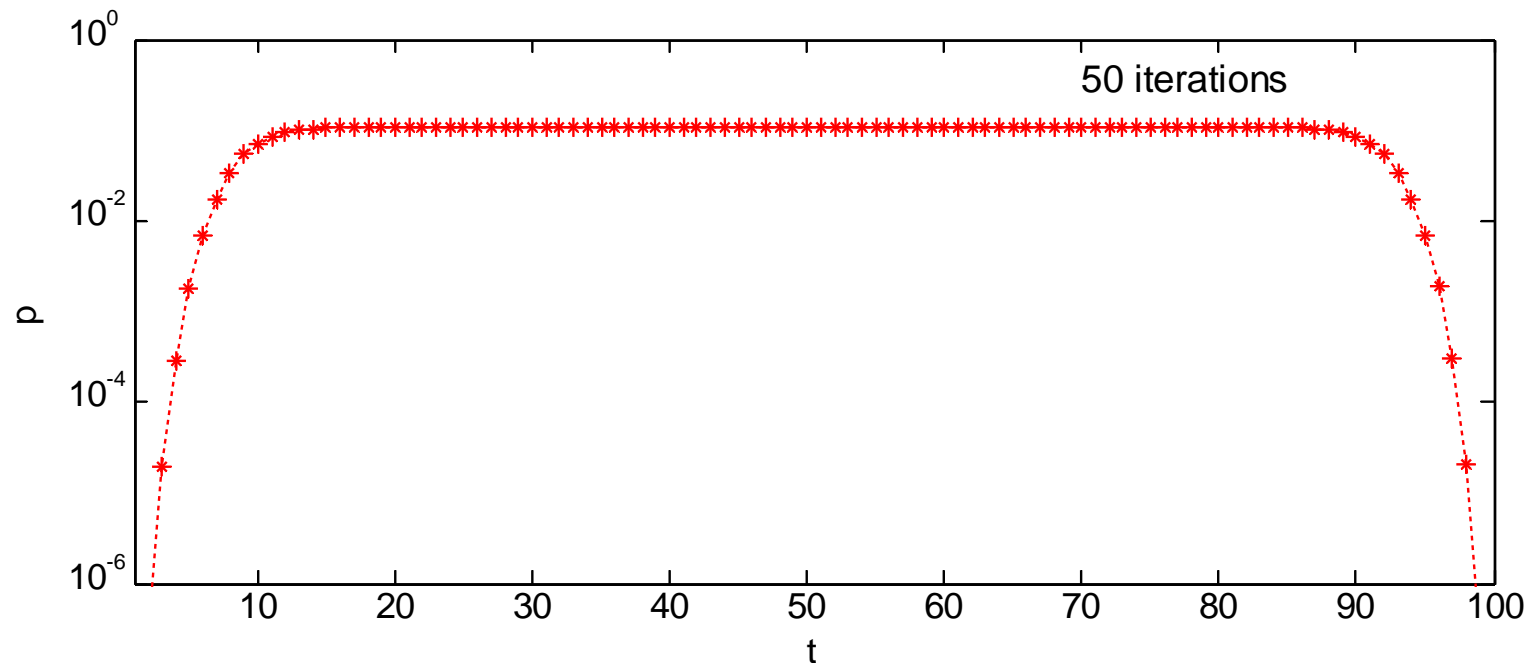
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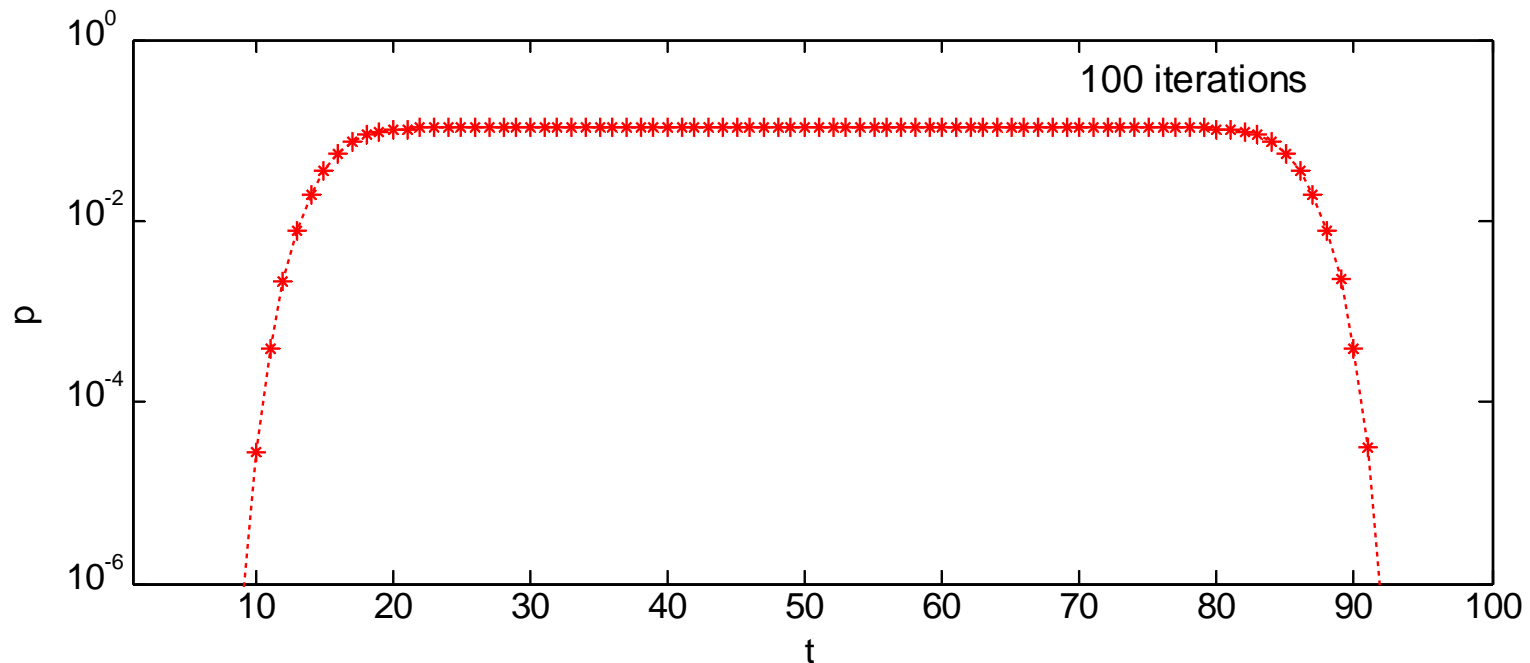
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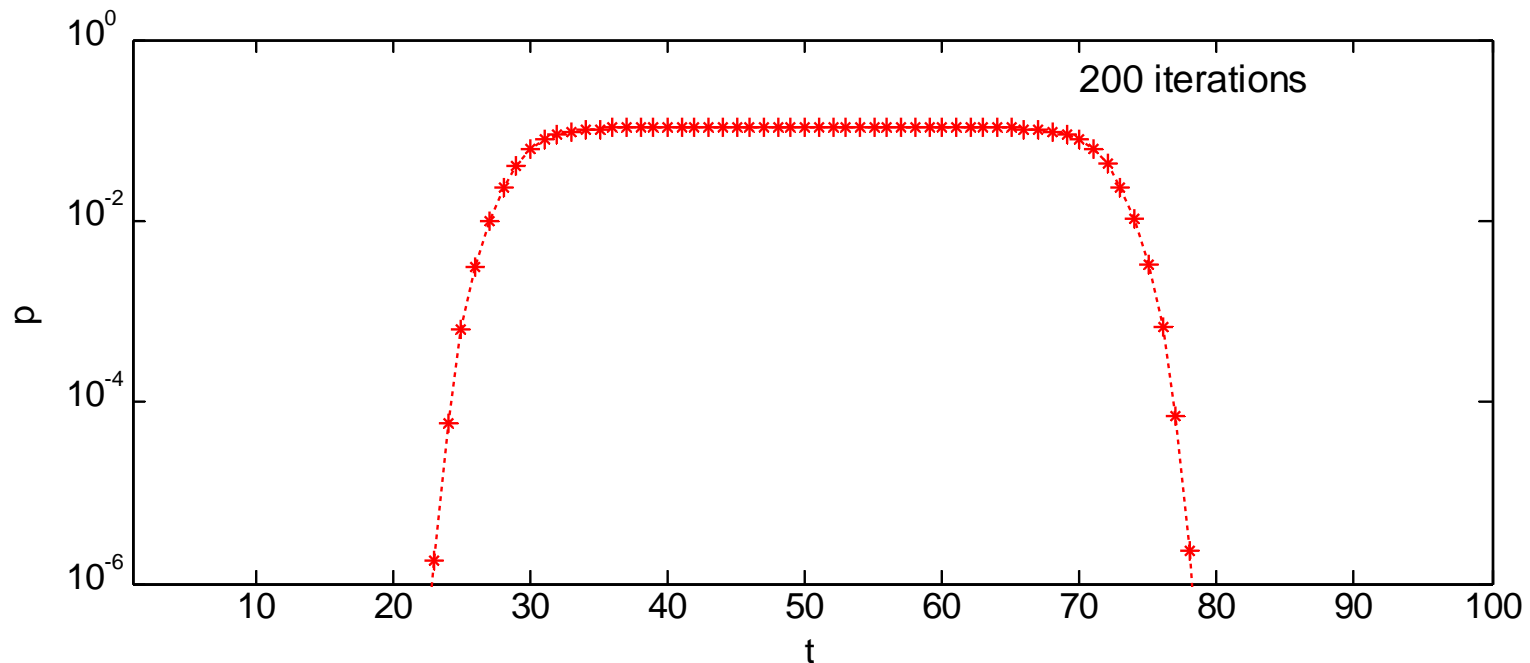
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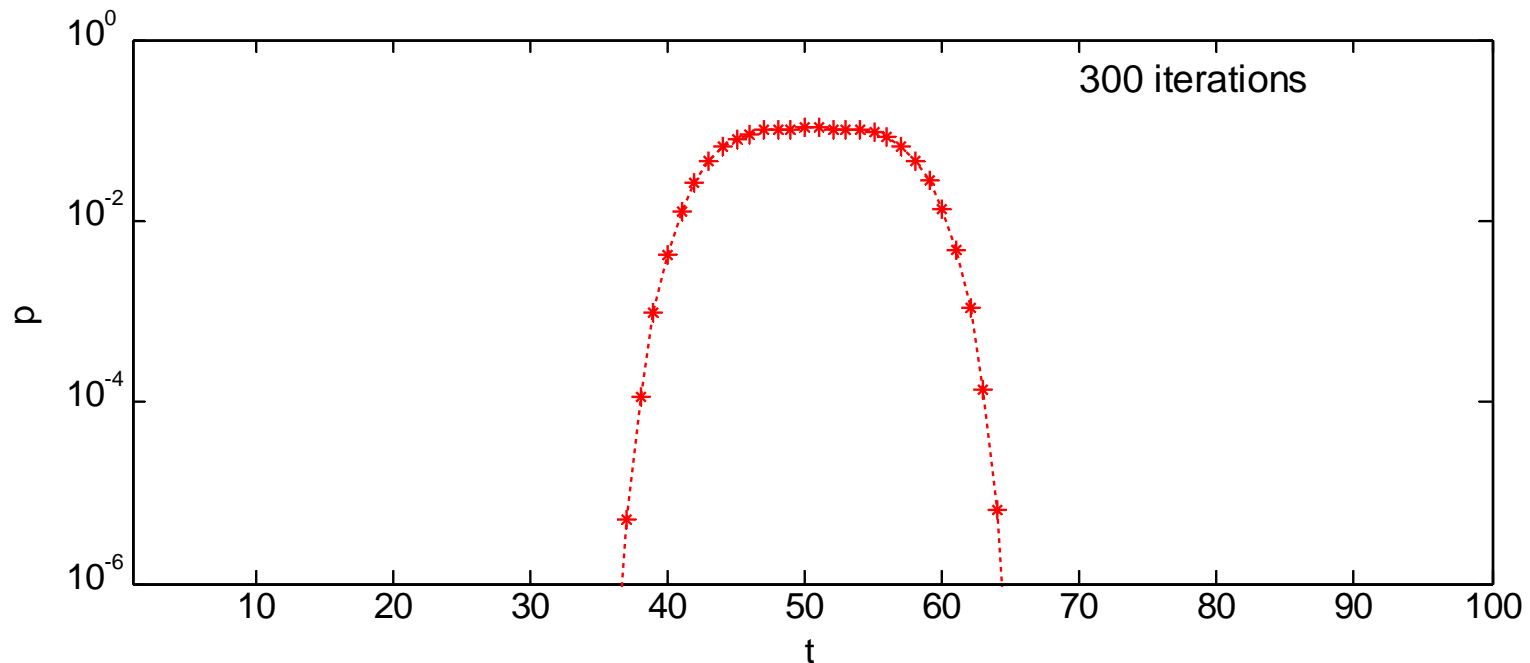
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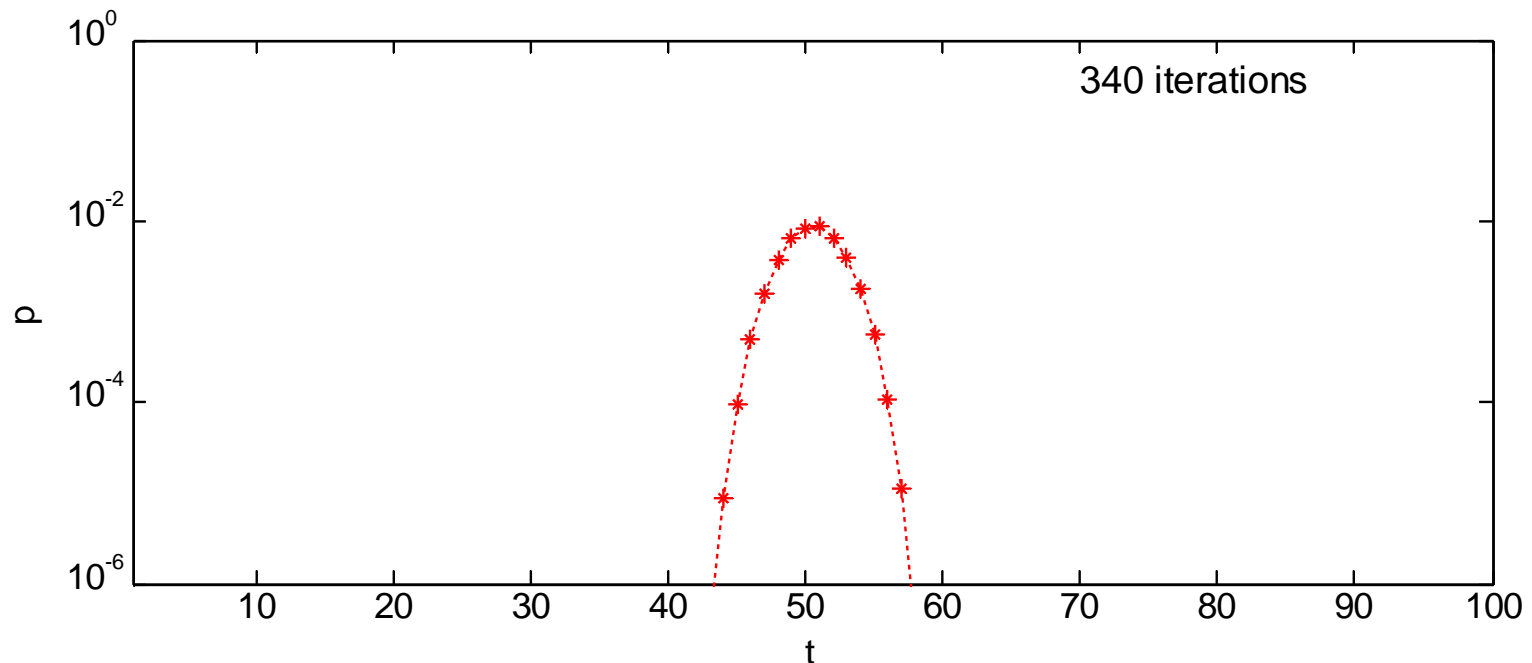
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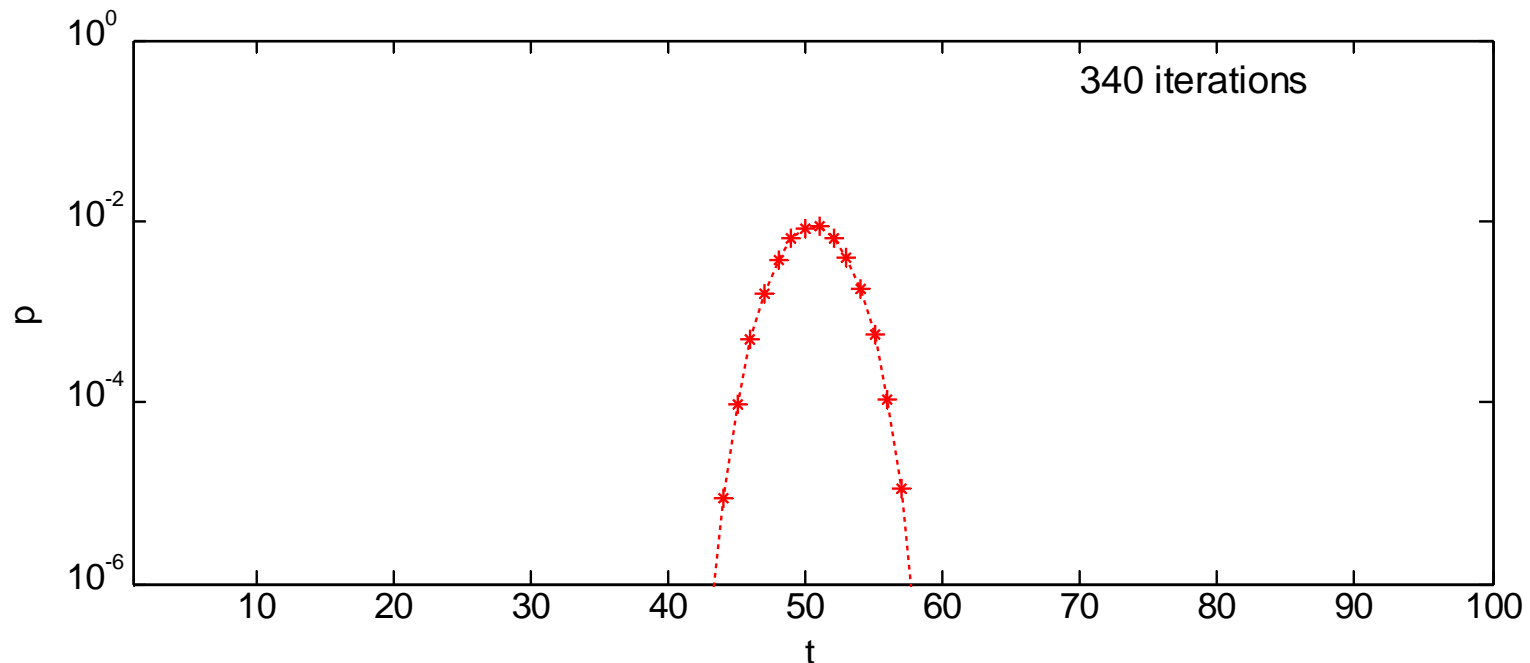
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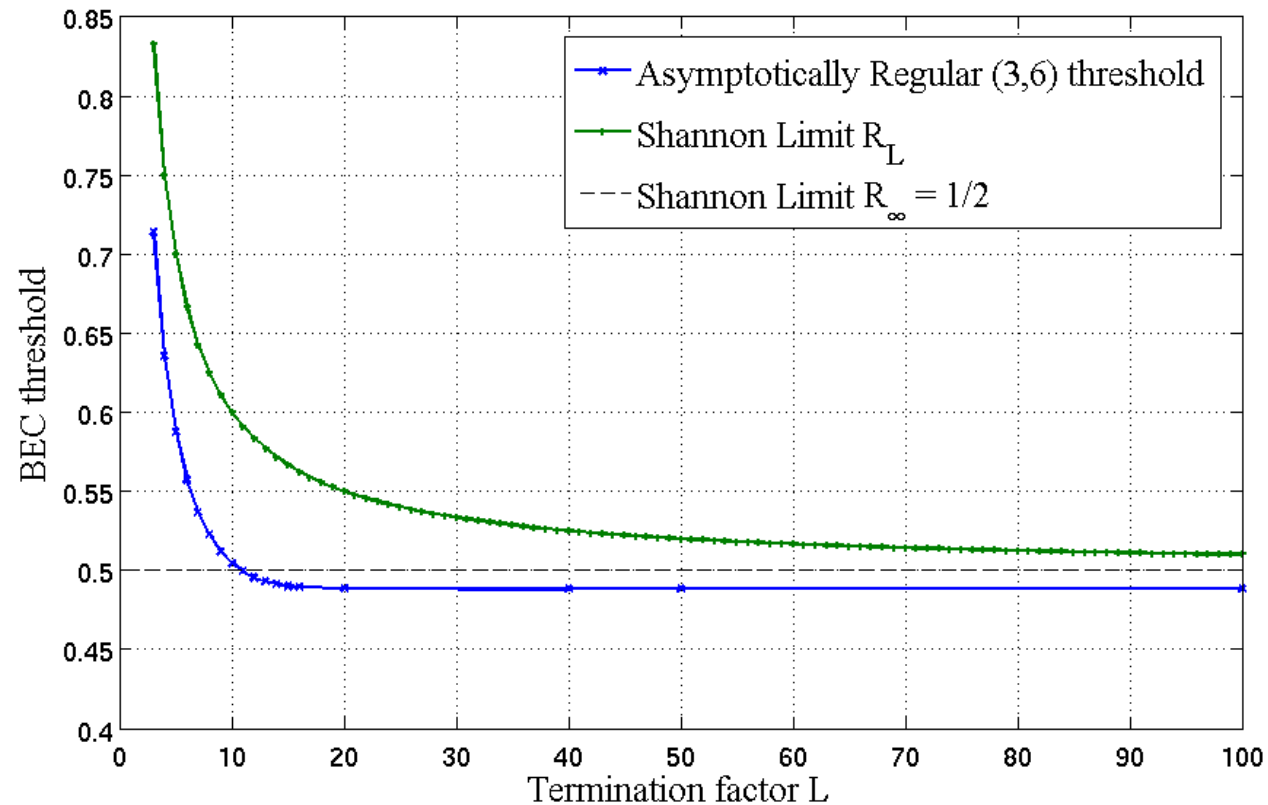


- Note: the **fraction** of **lower degree** nodes tends to zero as $L \rightarrow \infty$, i.e., the codes are **asymptotically regular**.

Thresholds of Terminated Spatially Coupled Codes

- **Density evolution** can be applied to the protograph-based ensembles with $M \rightarrow \infty$ [Sridharan et al. '04]:

Example: BEC



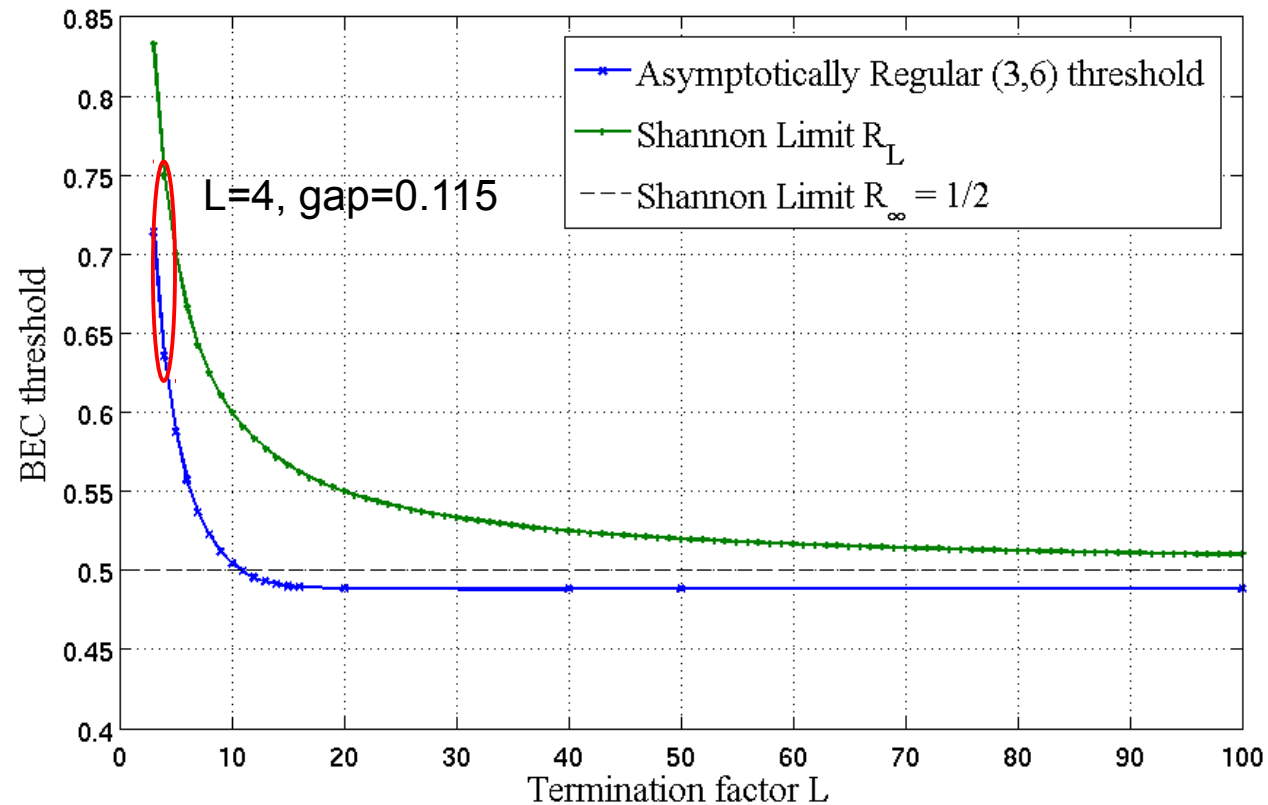
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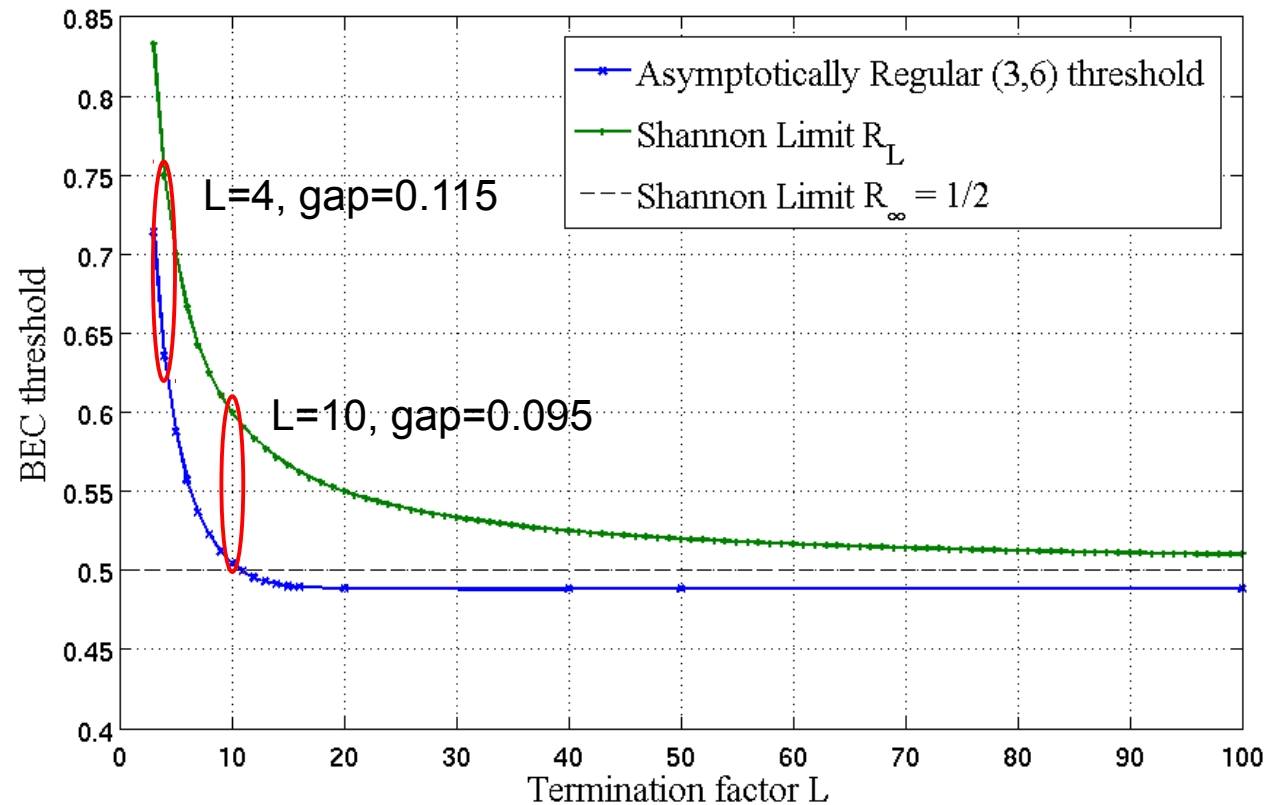
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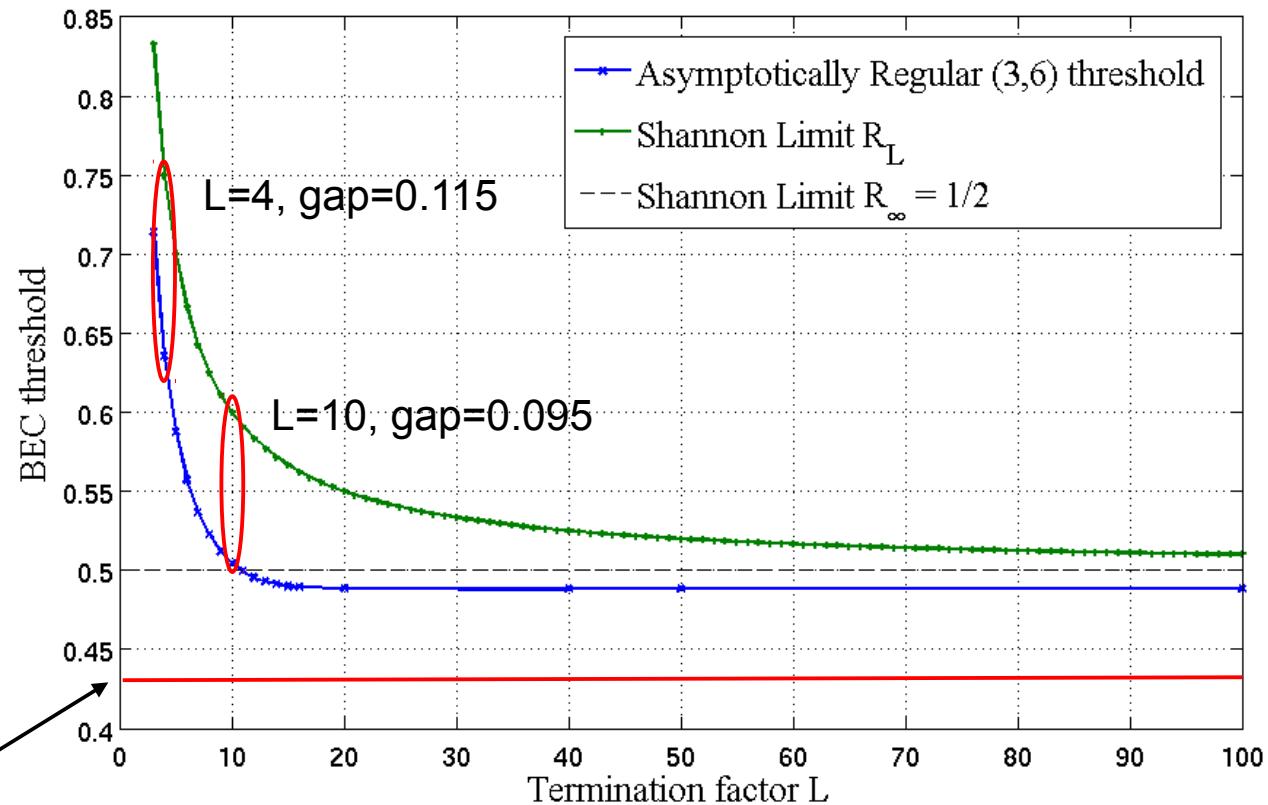
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⋮

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(3,6)-regular block code:

$$\varepsilon^* = 0.429$$



Thresholds of Terminated Spatially Coupled Codes

Iterative decoding thresholds (structured protograph-based ensembles)

BEC

(J, K)	ϵ_{SC}^*	ϵ_{blk}^*
(3,6)	0.488	0.429
(4,8)	0.497	0.383
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AWGN

(J, K)	$E_b/N_{o_{SC}}$	$E_b/N_{o_{blk}}$
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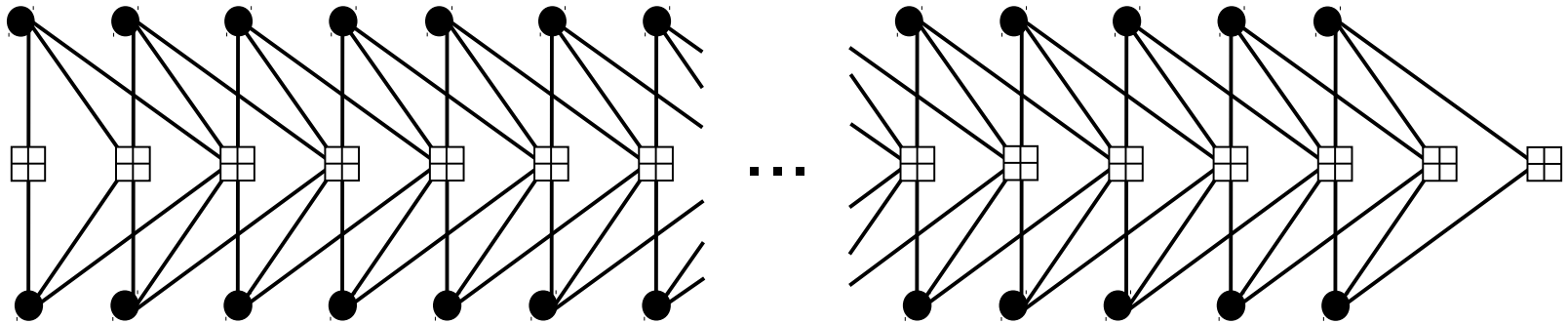
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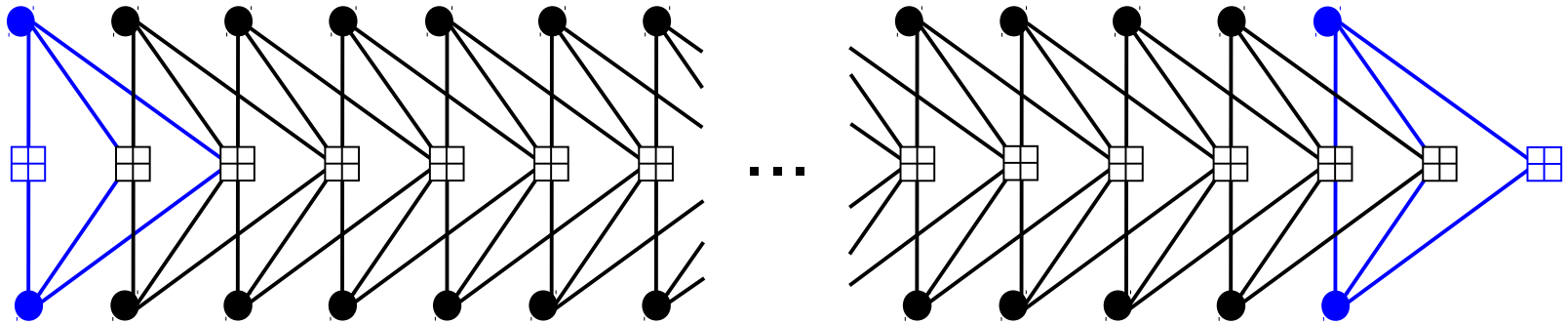
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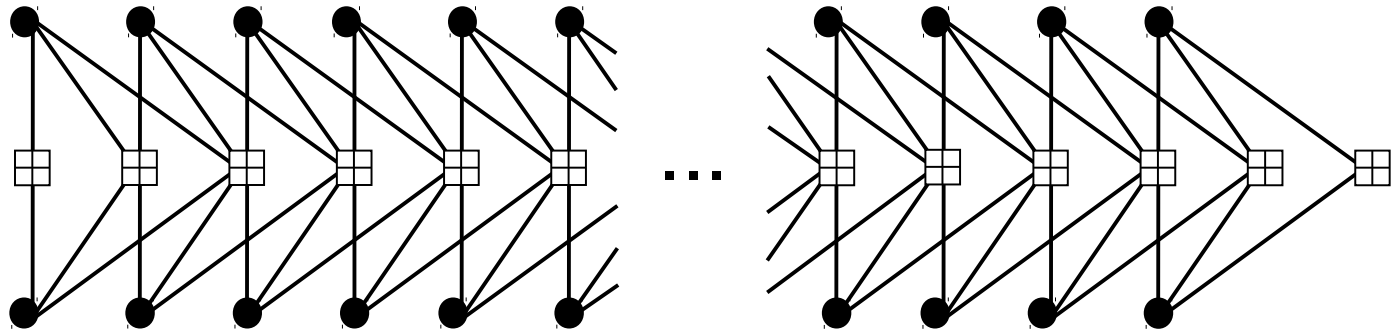
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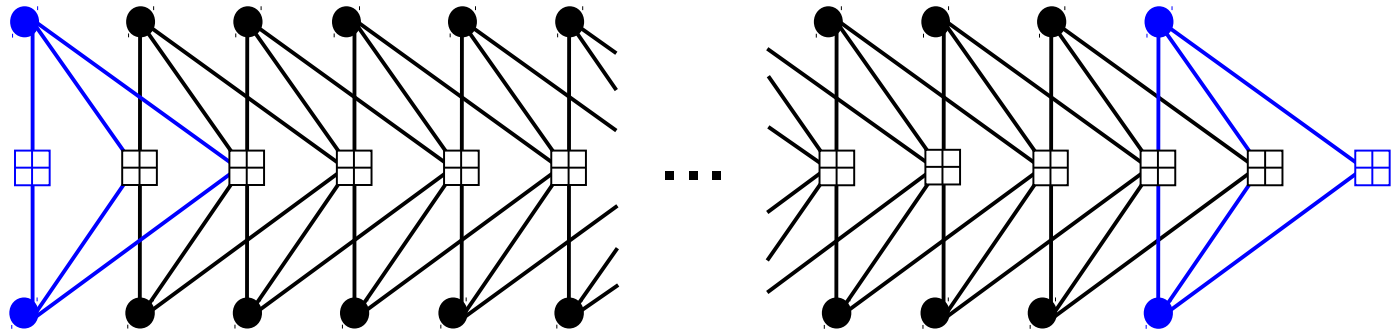
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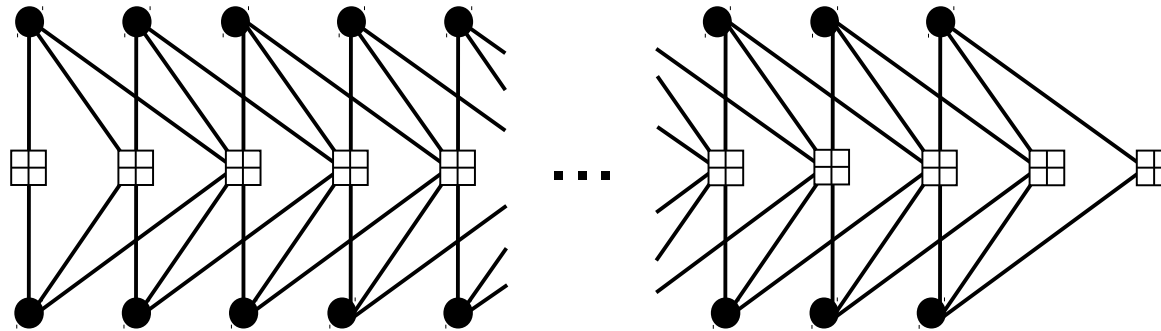
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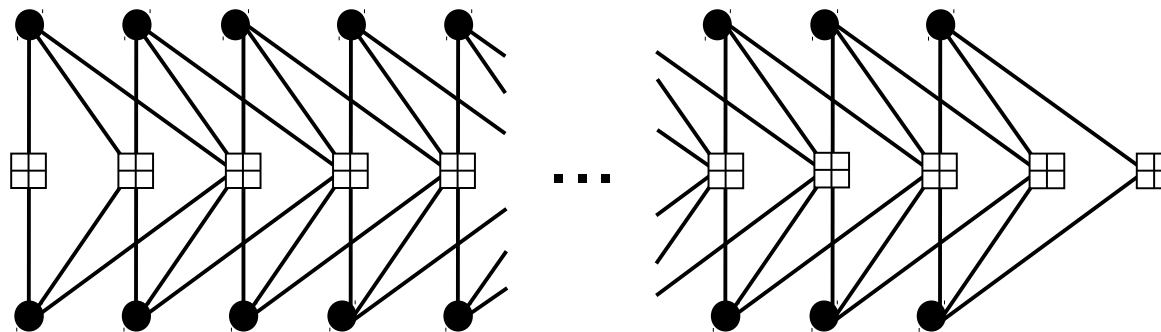
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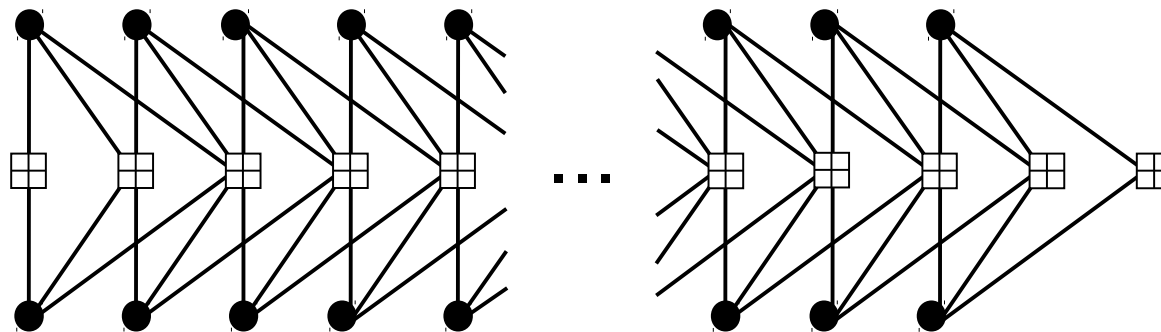


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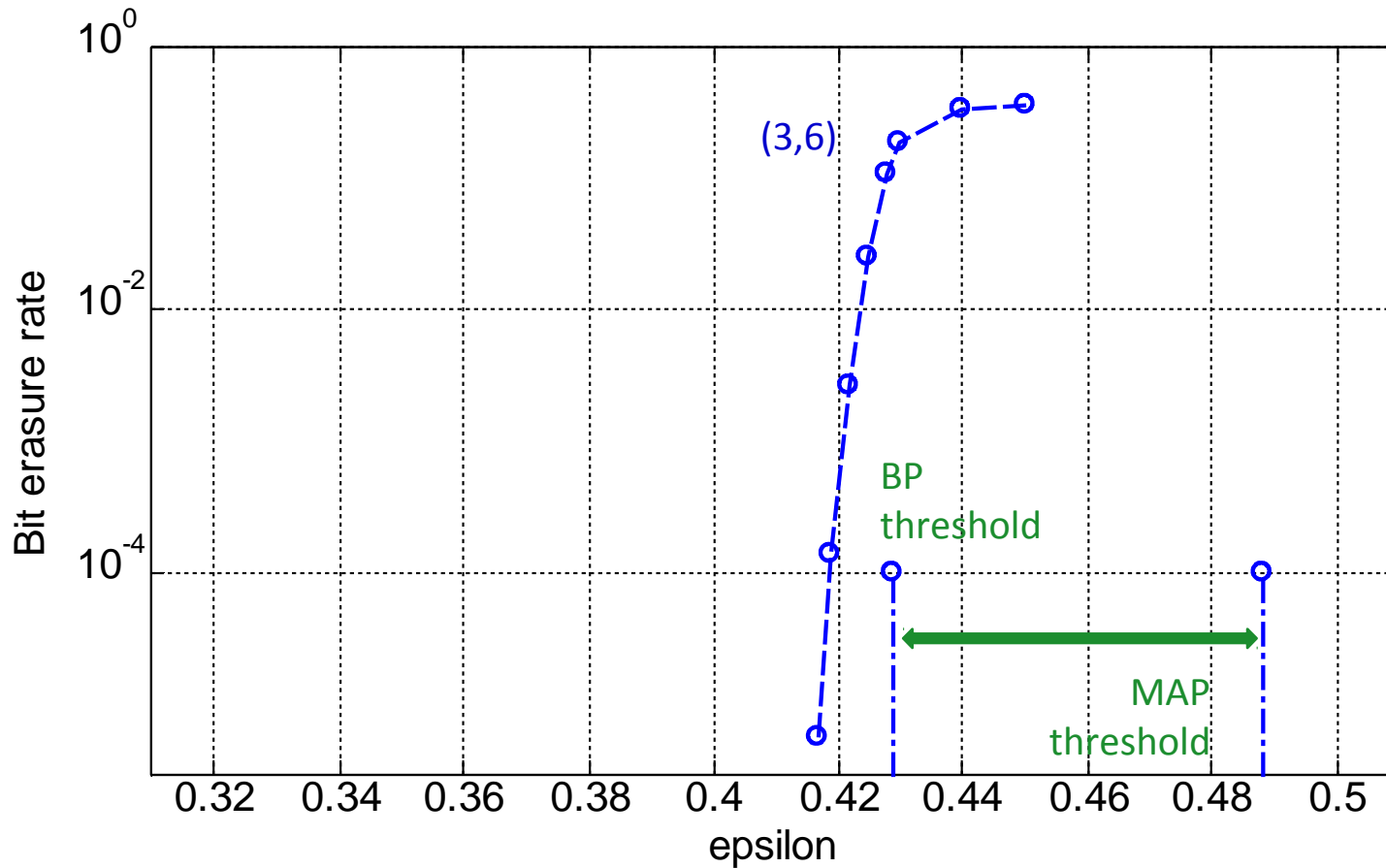
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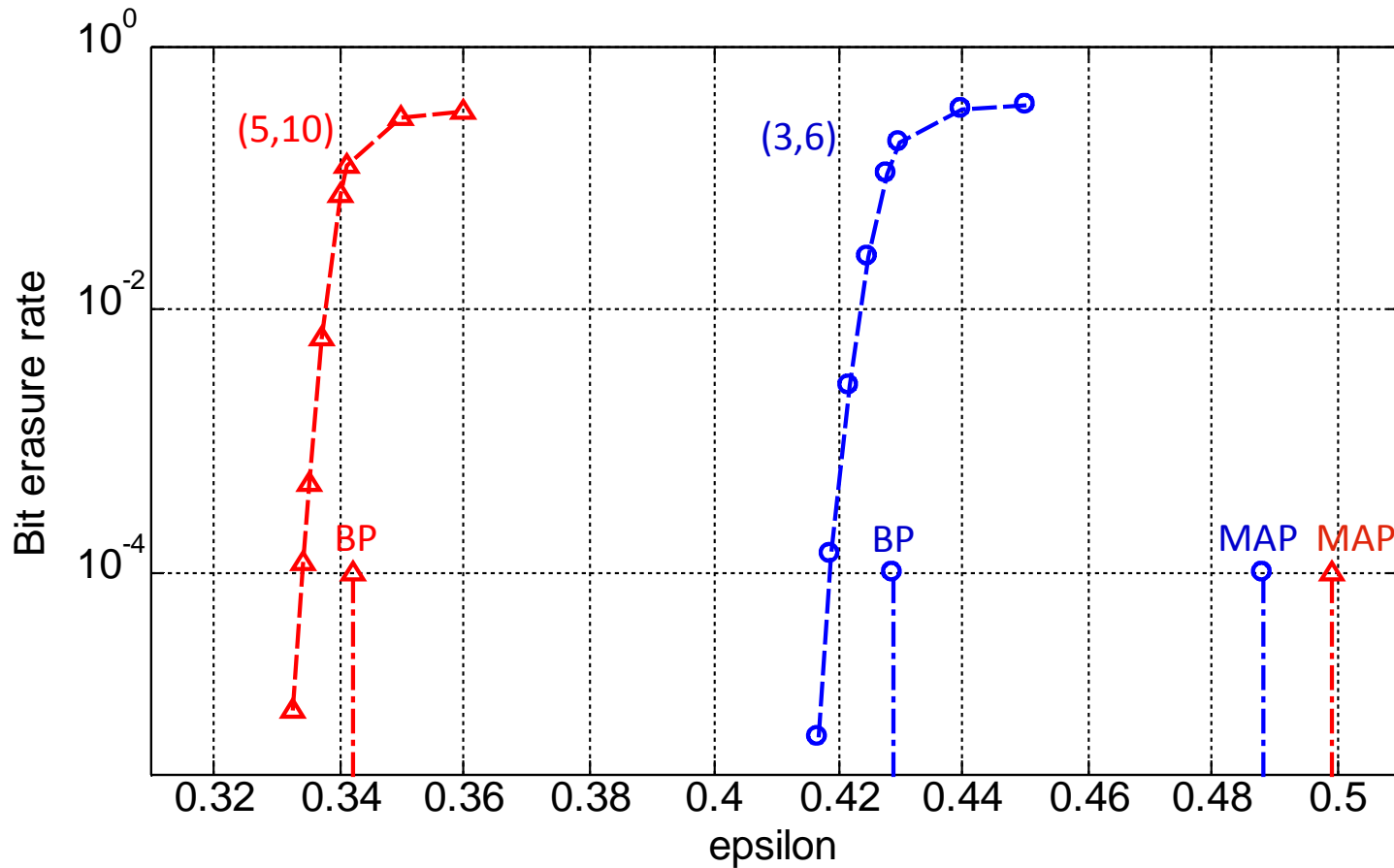
[KRU11] S. Kudekar, T. J. Richardson and R. Urbanke, “Threshold saturation via spatial coupling: why convolutional LDPC ensembles perform so well over the BEC”, *IEEE Trans. on Inf. Theory*, 57:2, 2011

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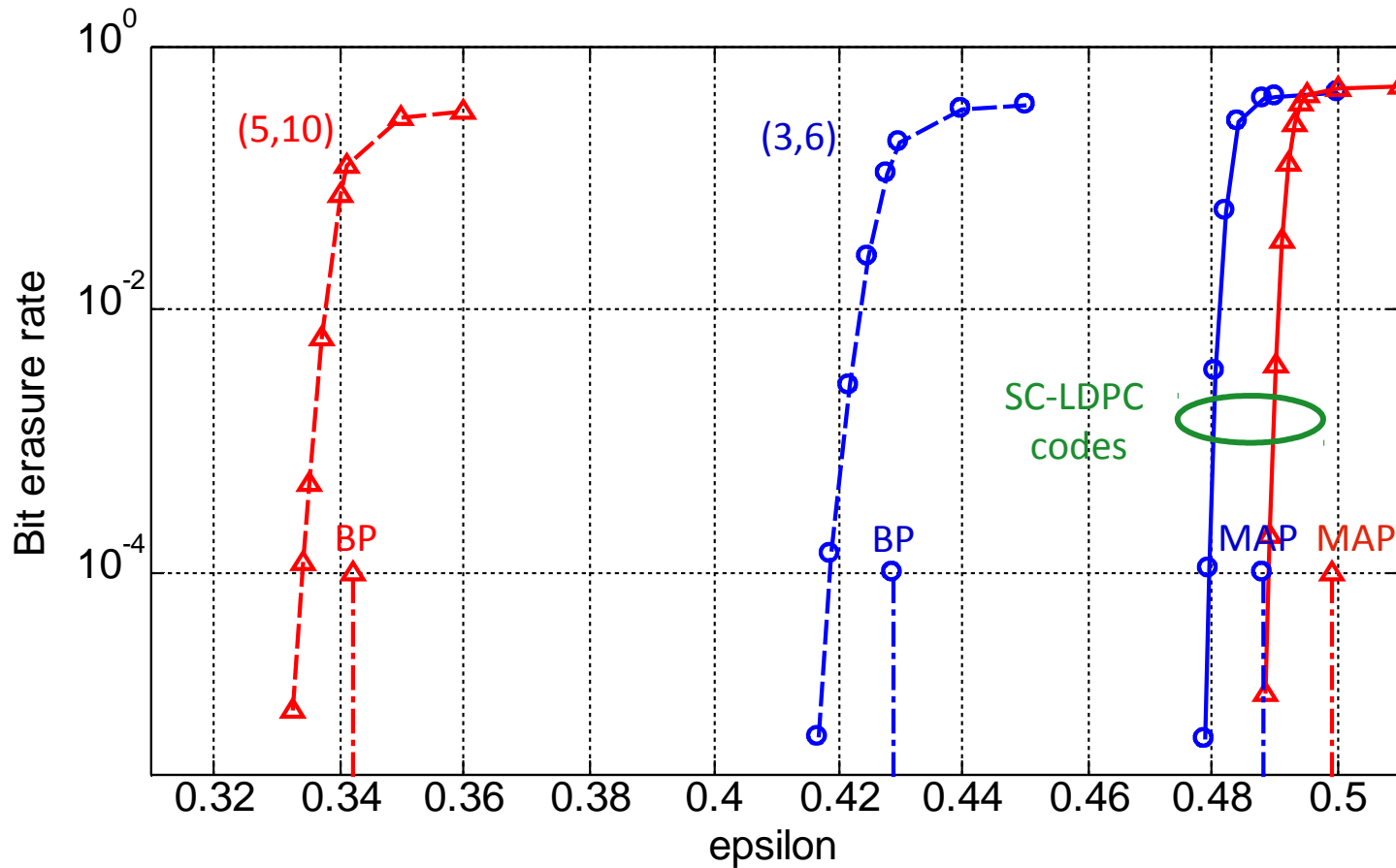
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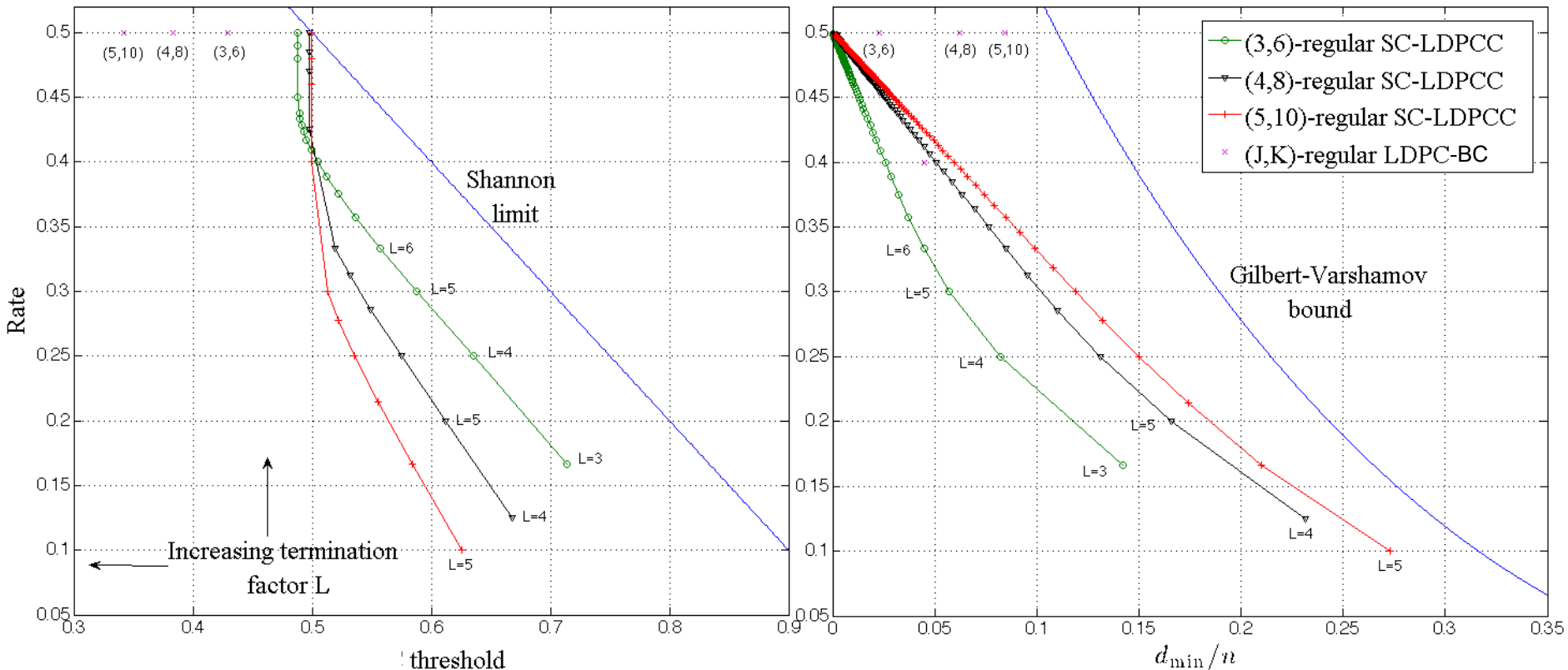


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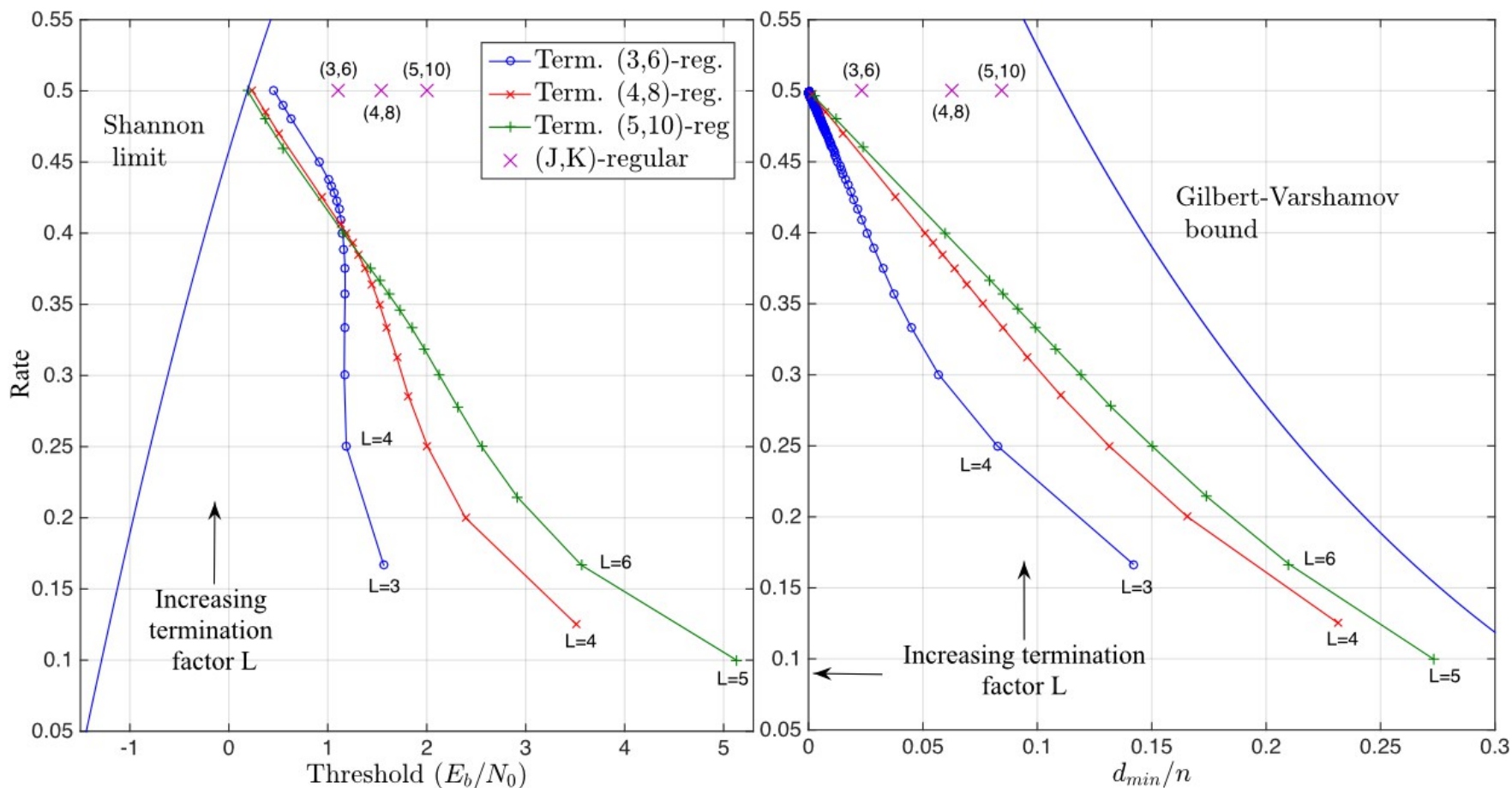


BEC Thresholds vs Distance Growth

- By increasing J and K , we obtain **capacity achieving** (J,K) -regular SC-LDPC code ensembles with linear minimum distance growth.

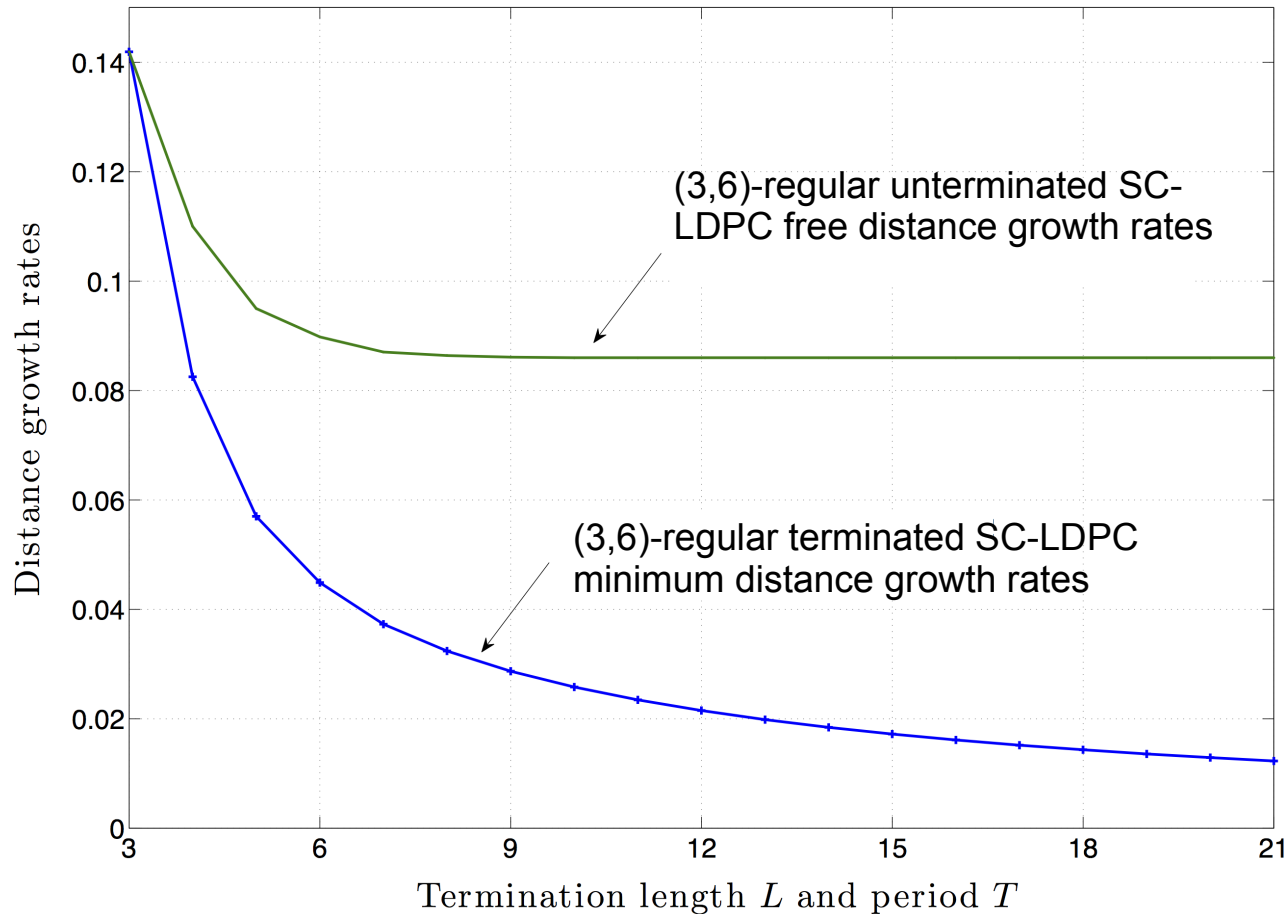


■ Similar results are obtained for the AWGNC

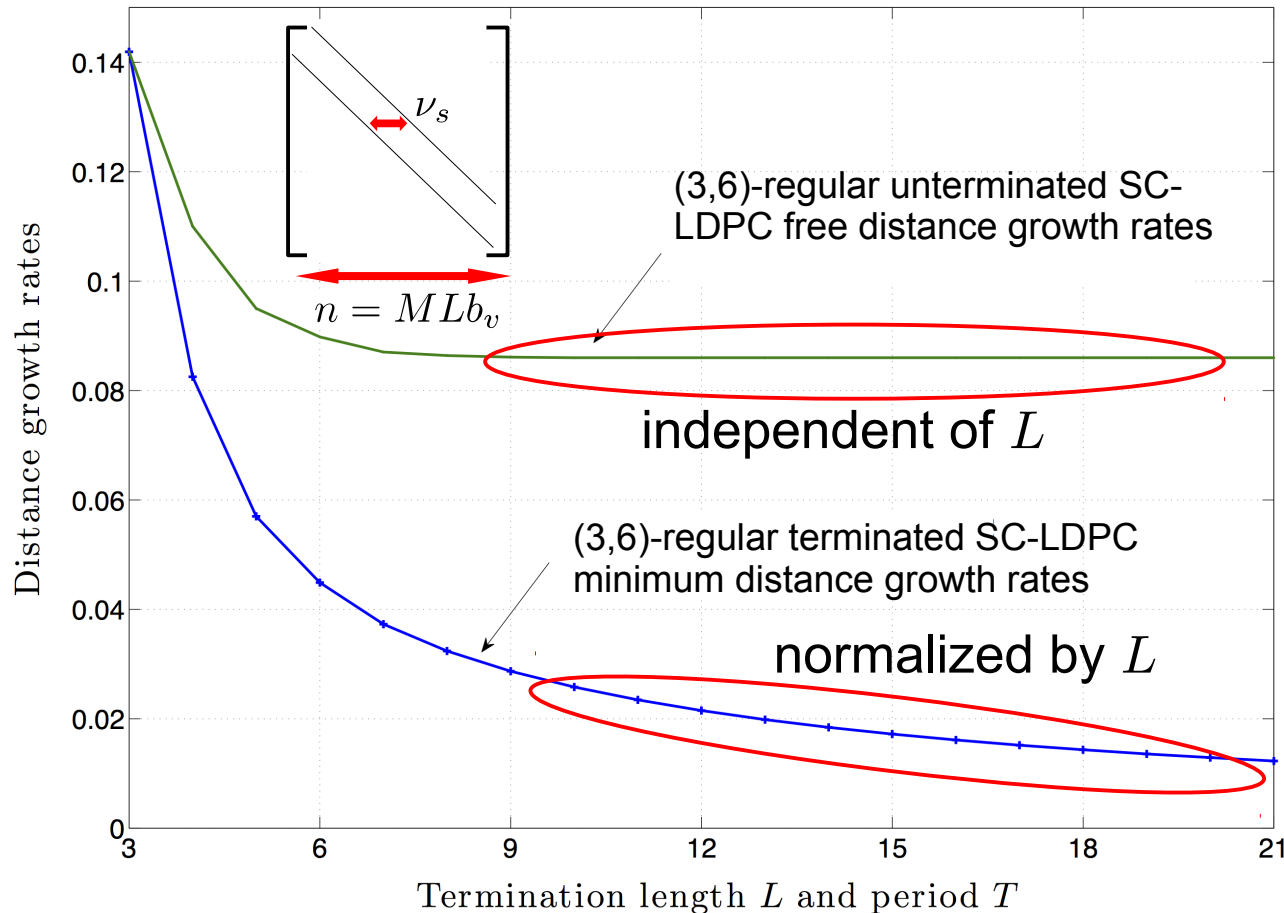


[MLC10] D. G. M. Mitchell, M. Lentmaier and D. J. Costello, Jr., "AWGN Channel Analysis of Terminated LDPC Convolutional Codes", *Proc. Information Theory and Applications Workshop*, San Diego, Feb. 2011.

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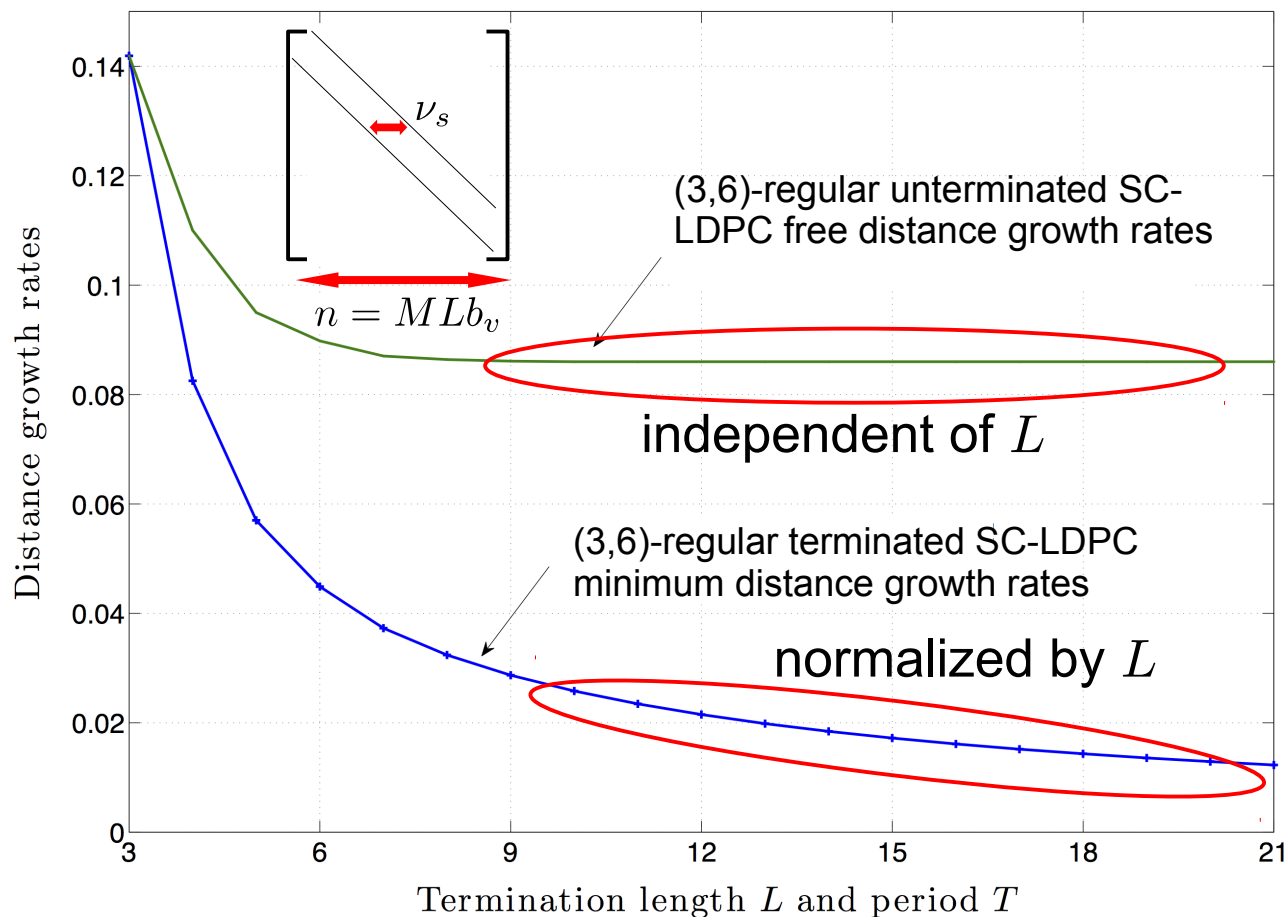


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- An appropriate distance measure for 'convolutional-like' terminated ensembles should be independent of L .

■ LDPC Block Codes

- ➔ Parity-check matrix and Tanner graph representations, minimum distance bounds, iterative decoding thresholds, regular and irregular code designs, protograph-based constructions

■ Spatially Coupled LDPC Codes

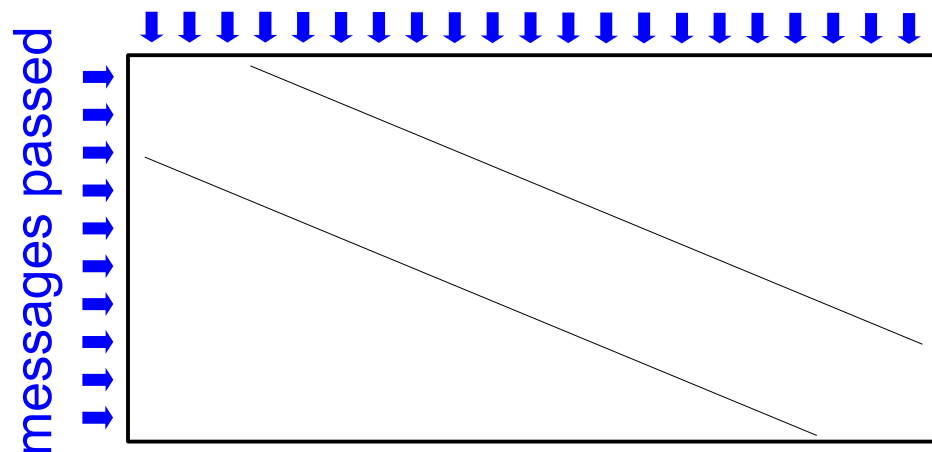
- ➔ Protograph representation, edge-spreading construction, termination
- ➔ Iterative decoding thresholds, threshold saturation, minimum distance

■ Practical Considerations

- ➔ Finite-length scaling, window decoding, performance, latency, and complexity comparisons to LDPC block codes, implementation aspects

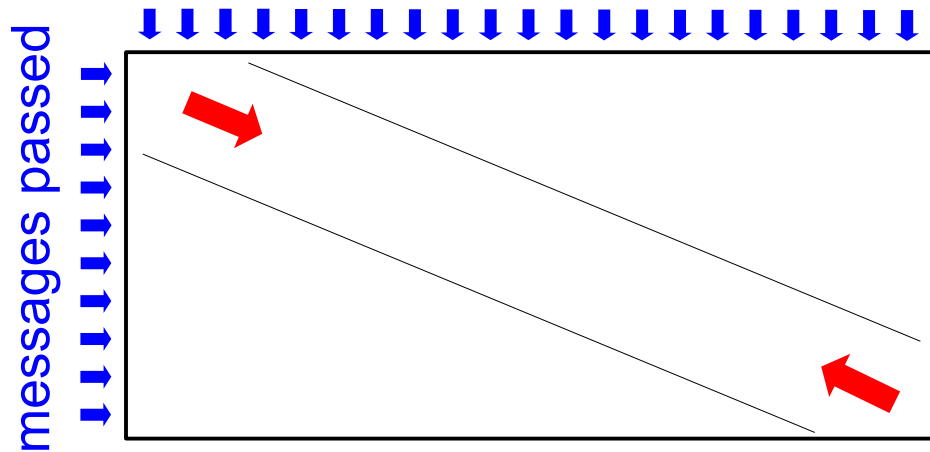
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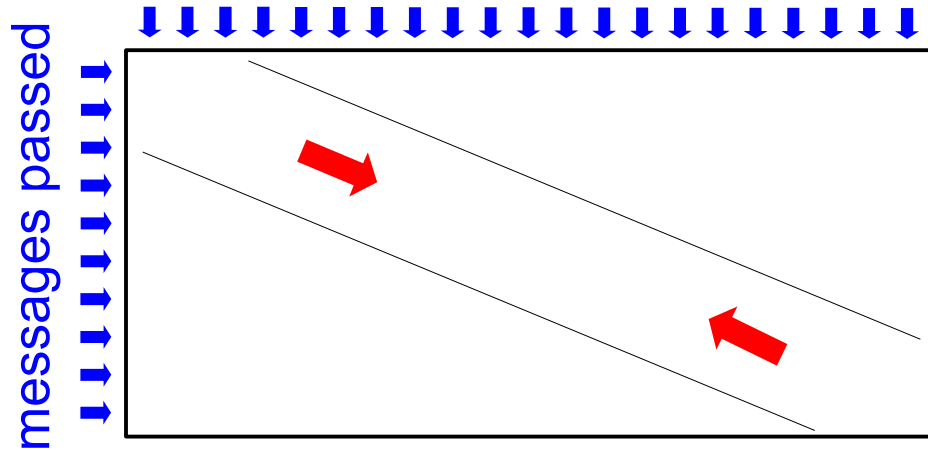
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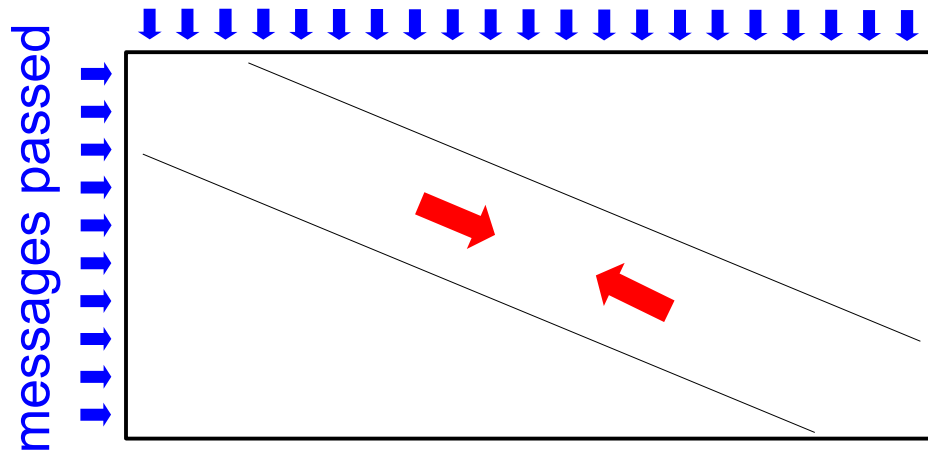
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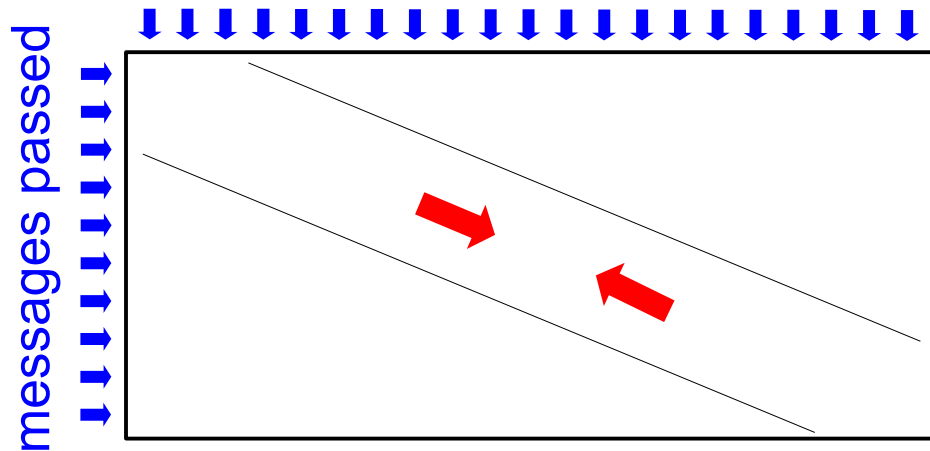
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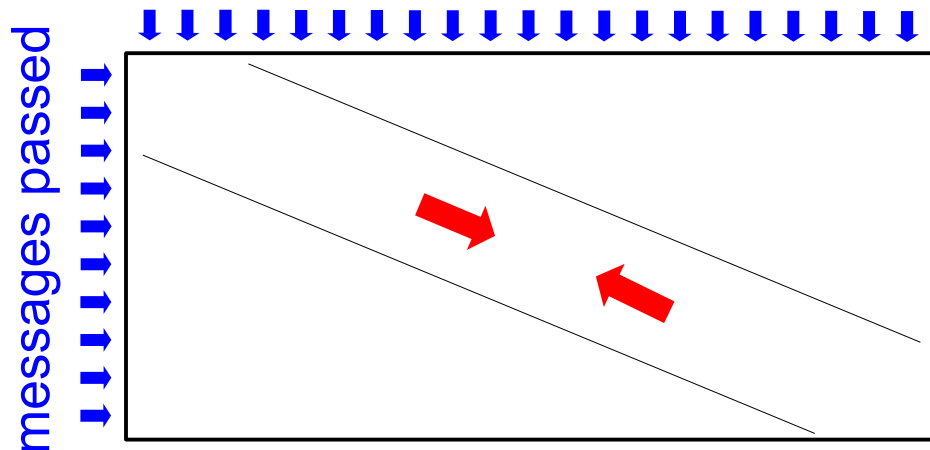


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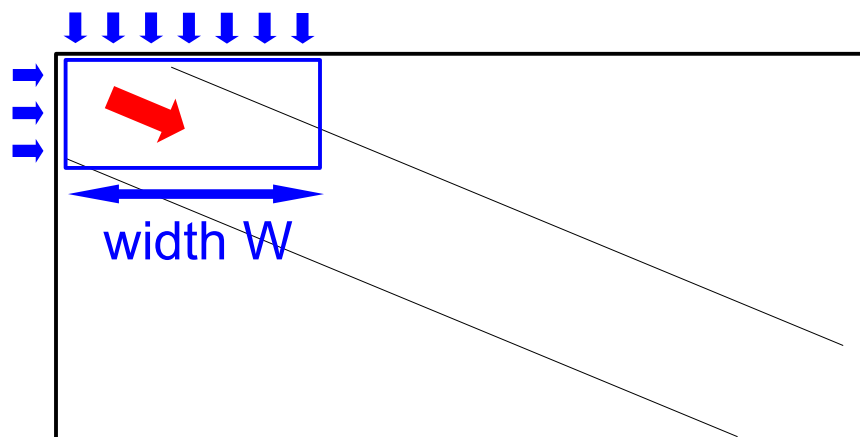
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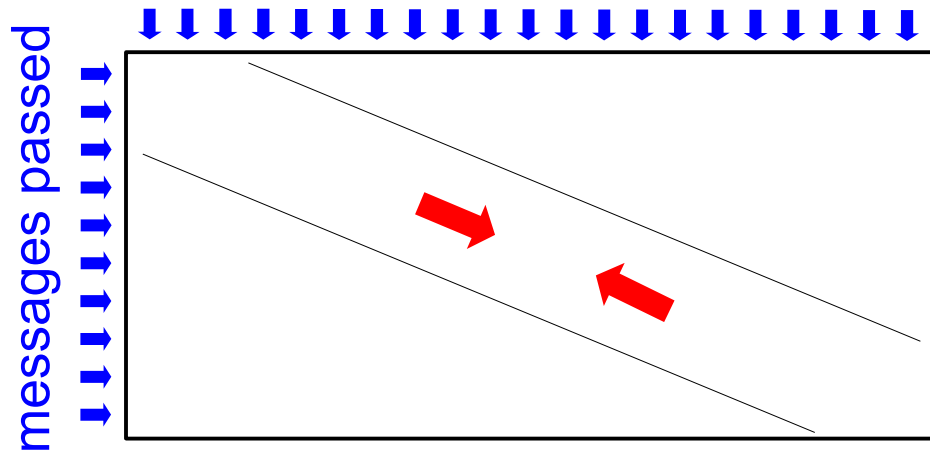
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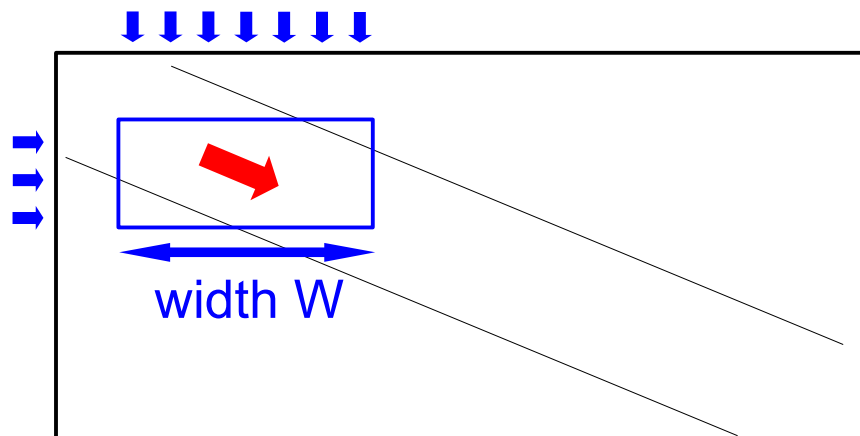
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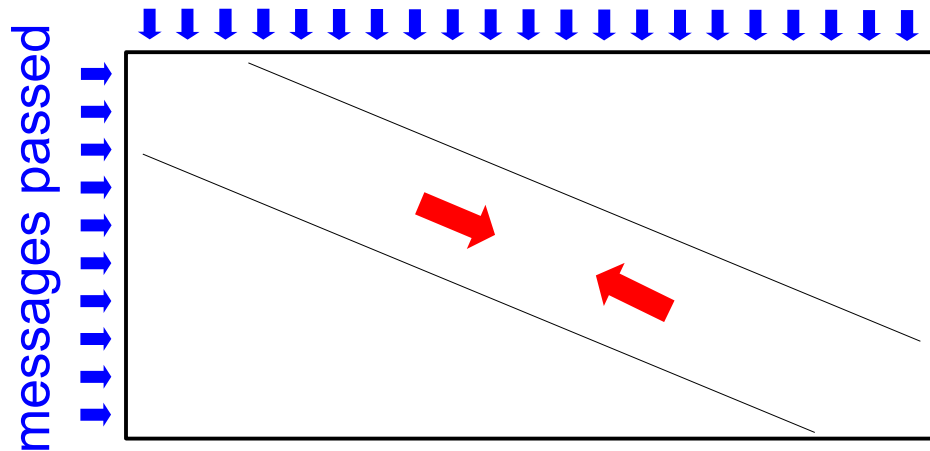
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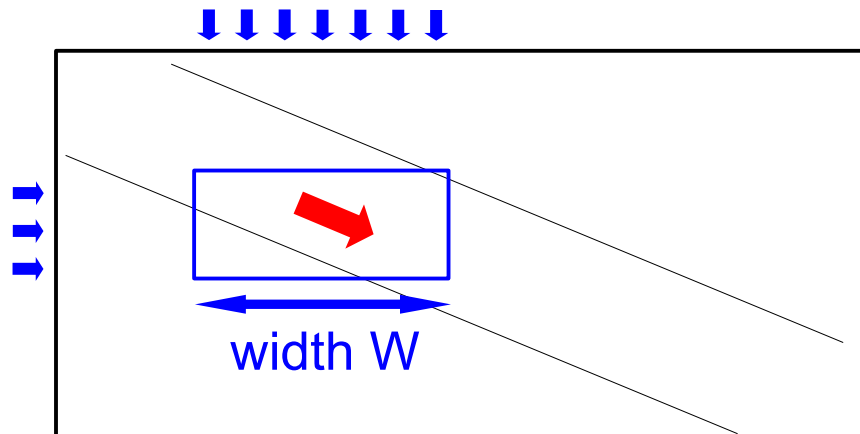
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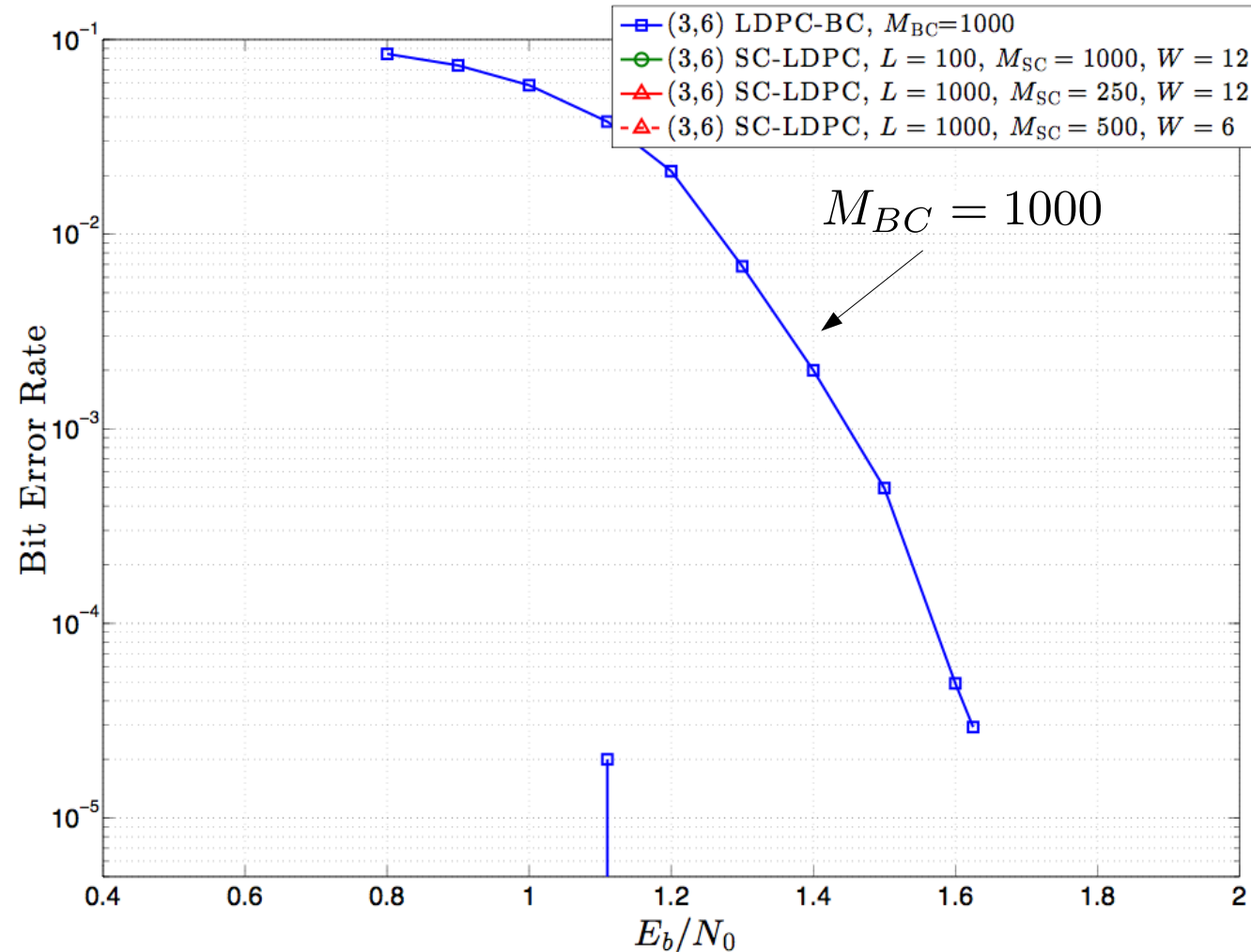
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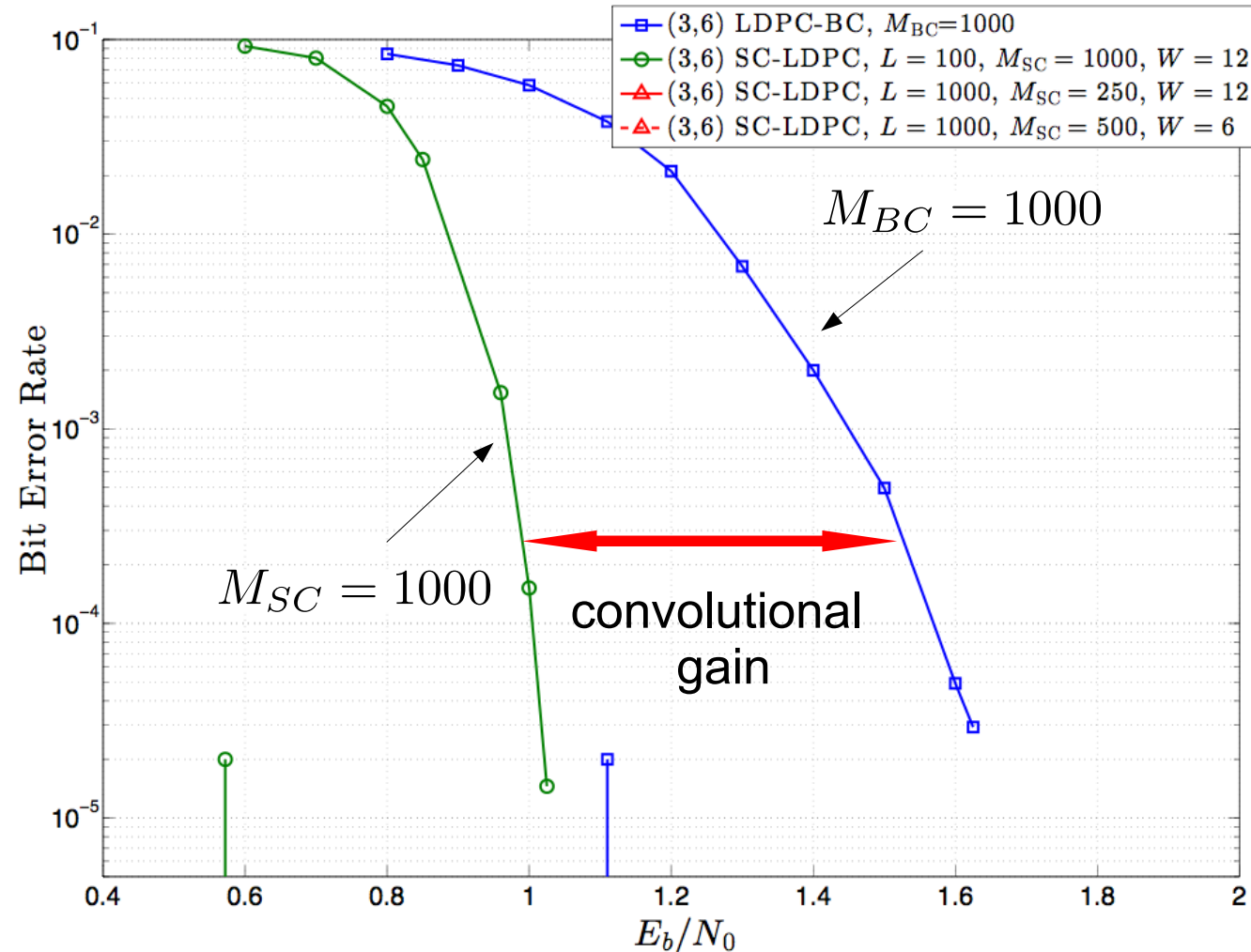
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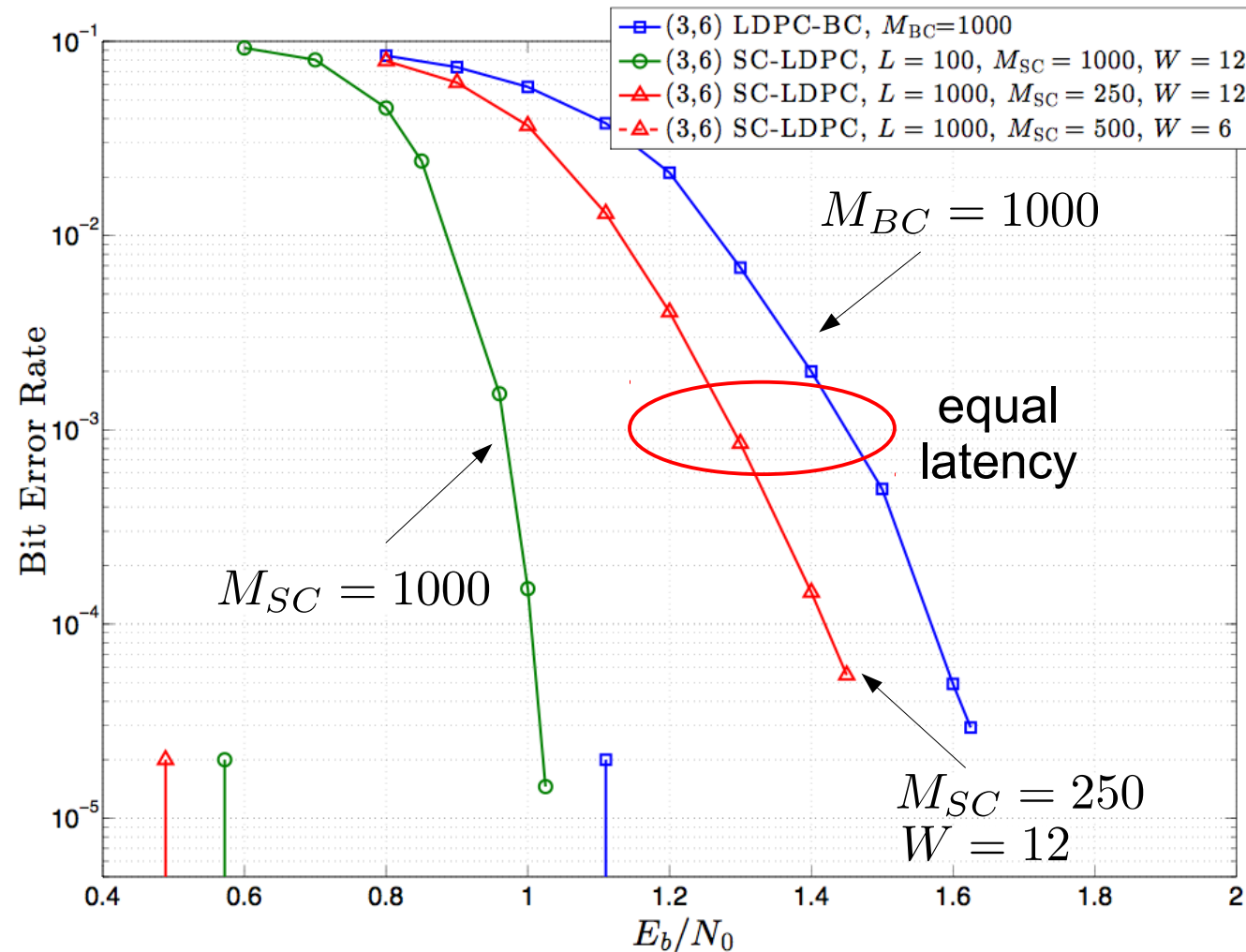
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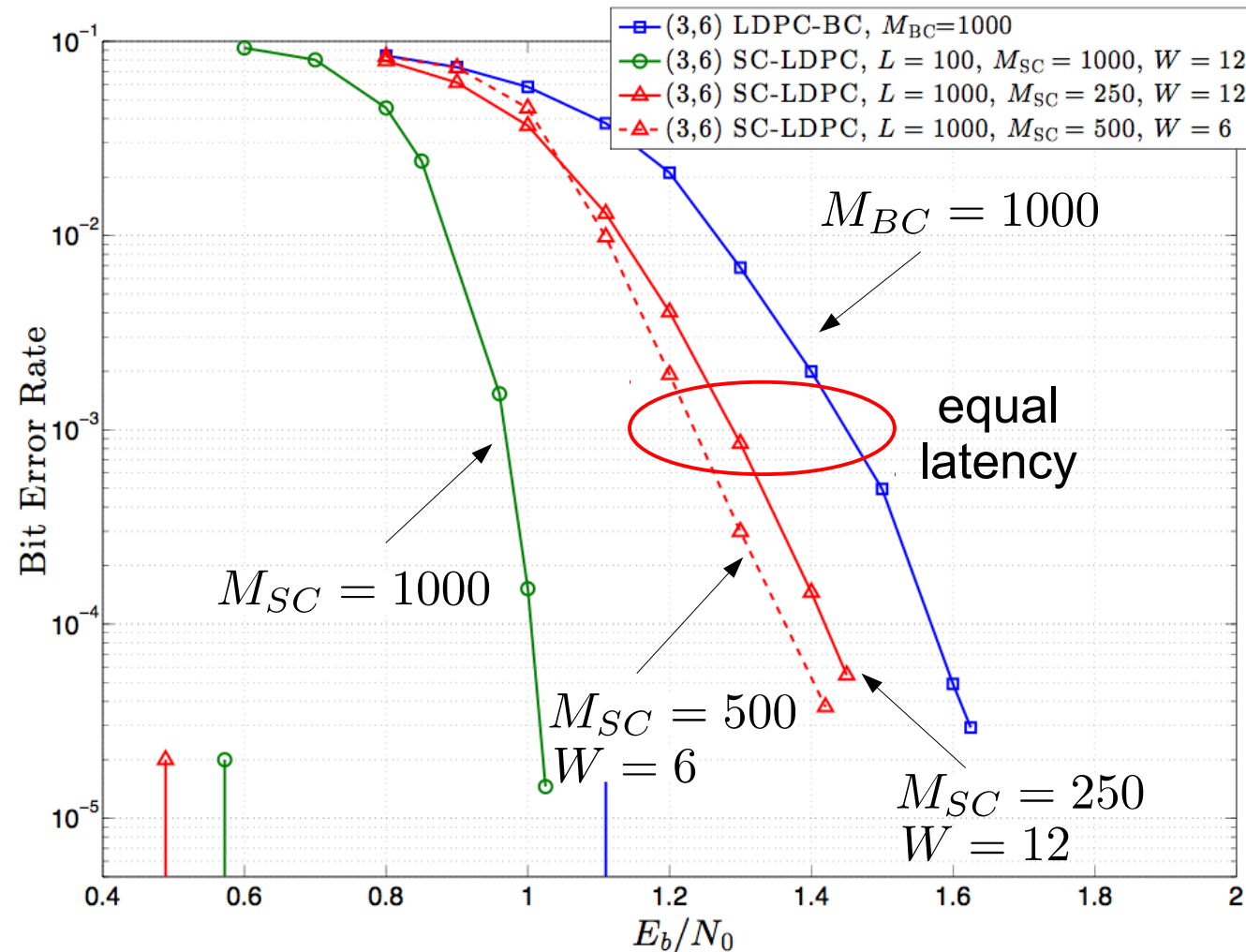
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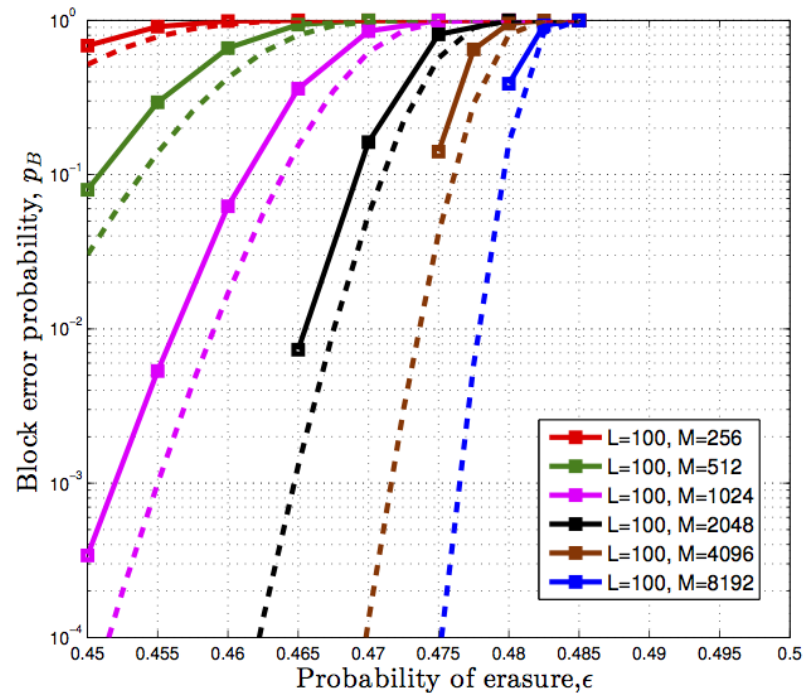
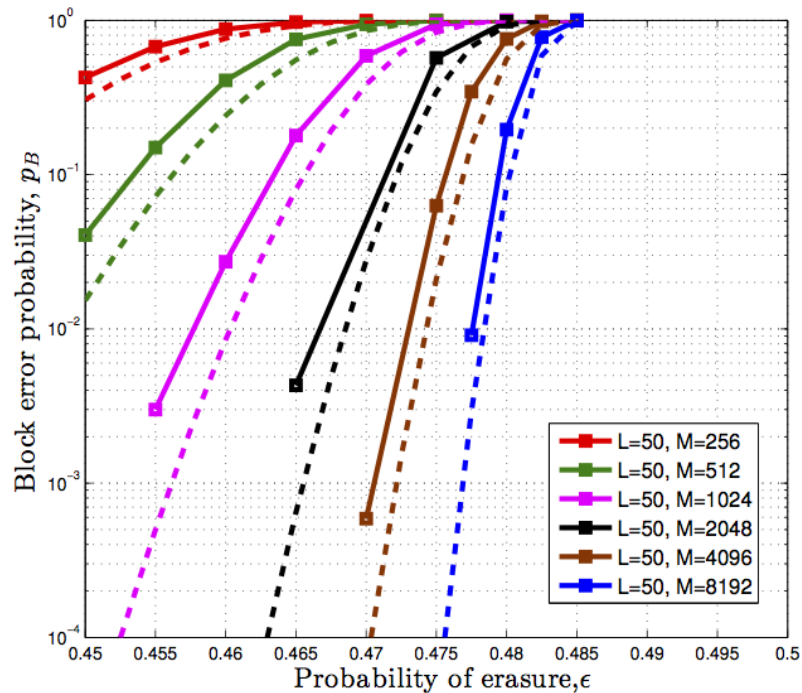
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Finite-Length Scaling for SC-LDPC Codes

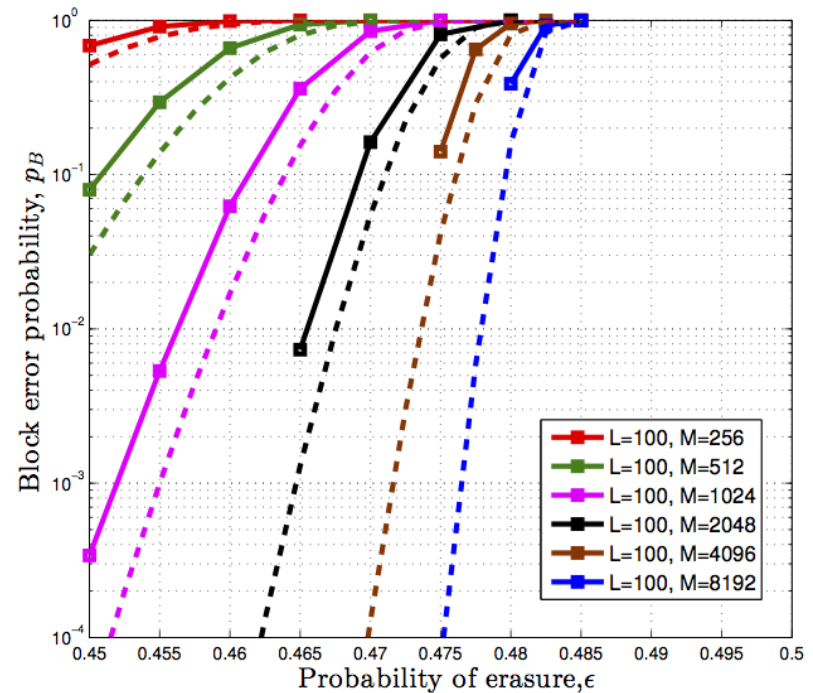
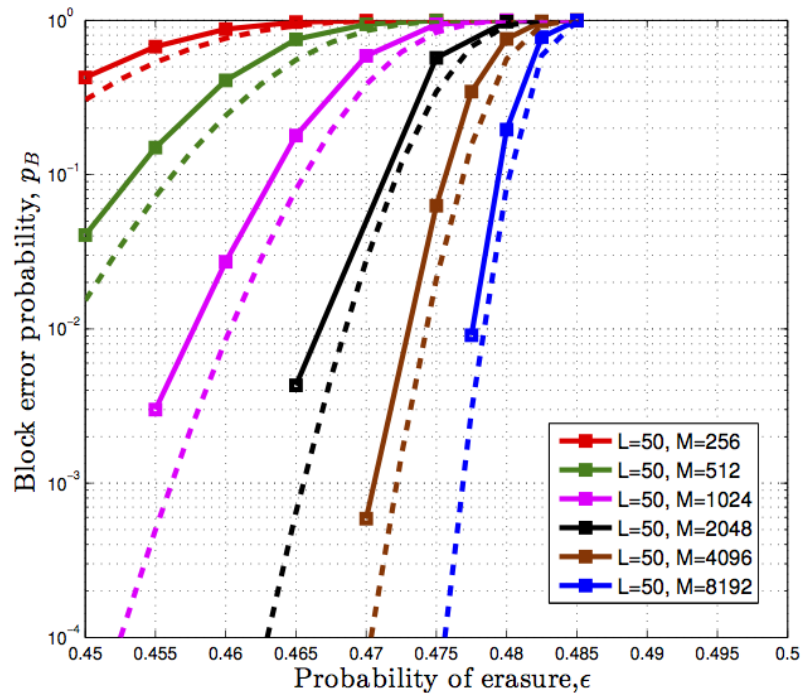
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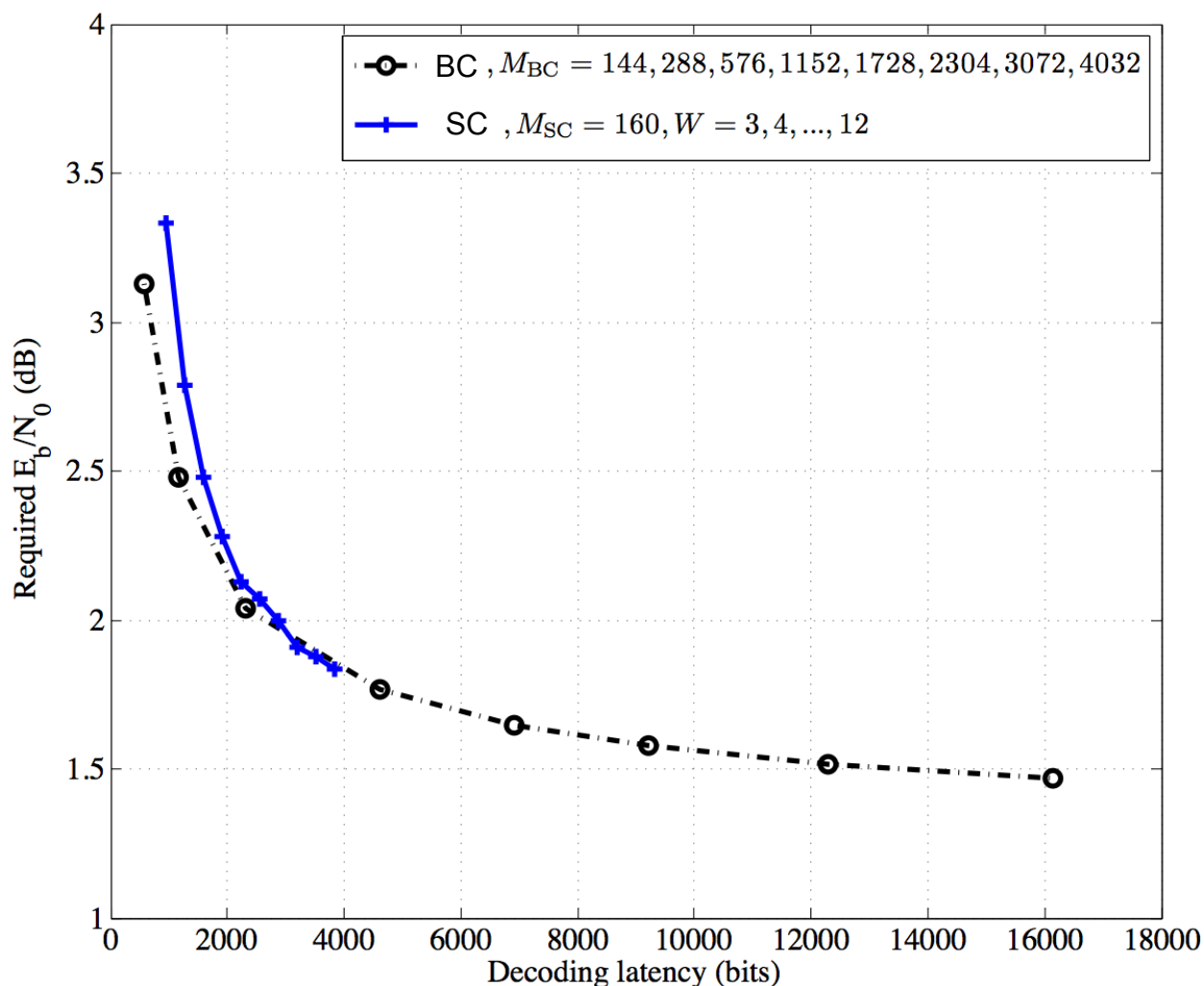


- The scaling law is a useful engineering tool to gain insight into the design of SC-LDPC codes.

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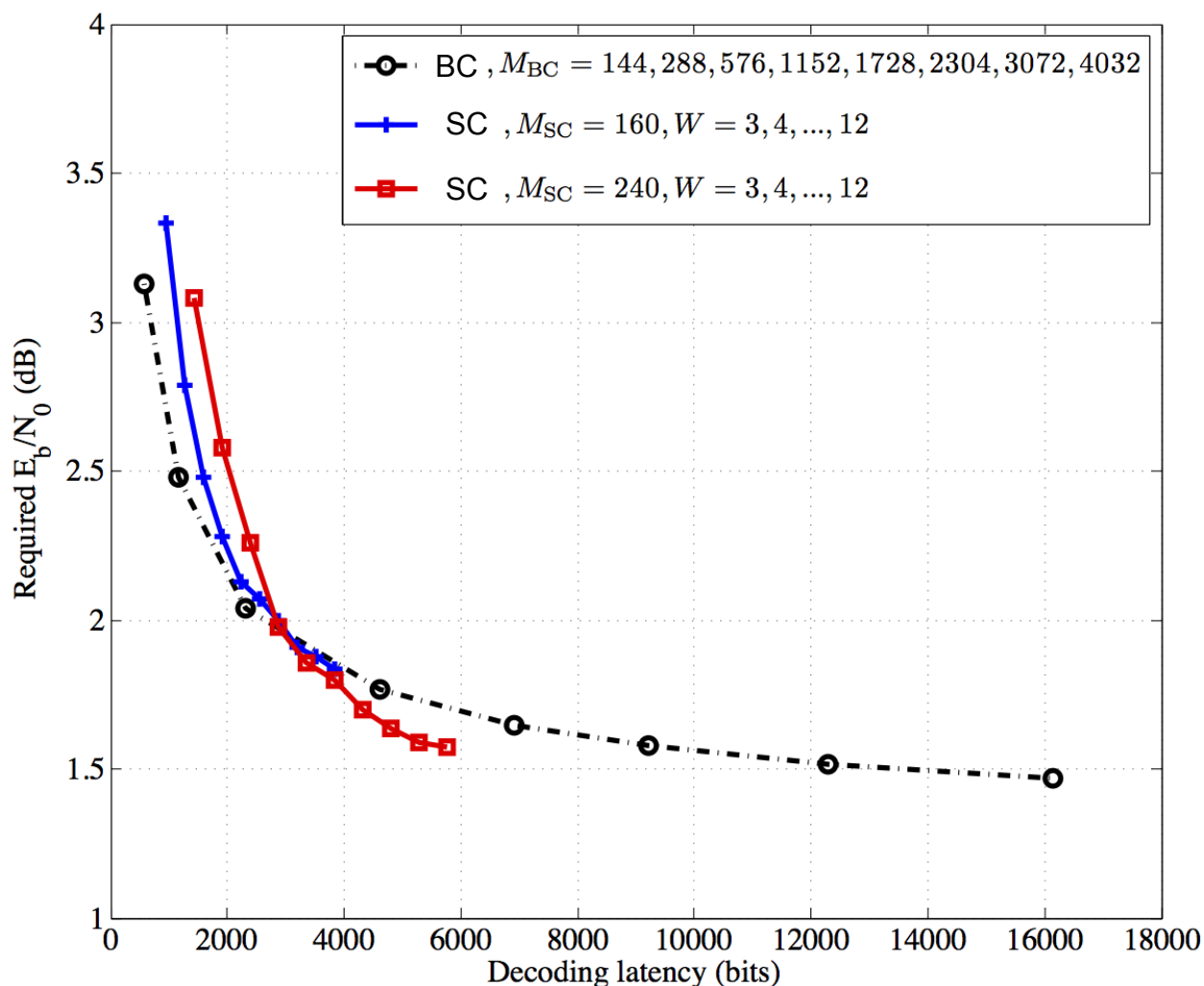
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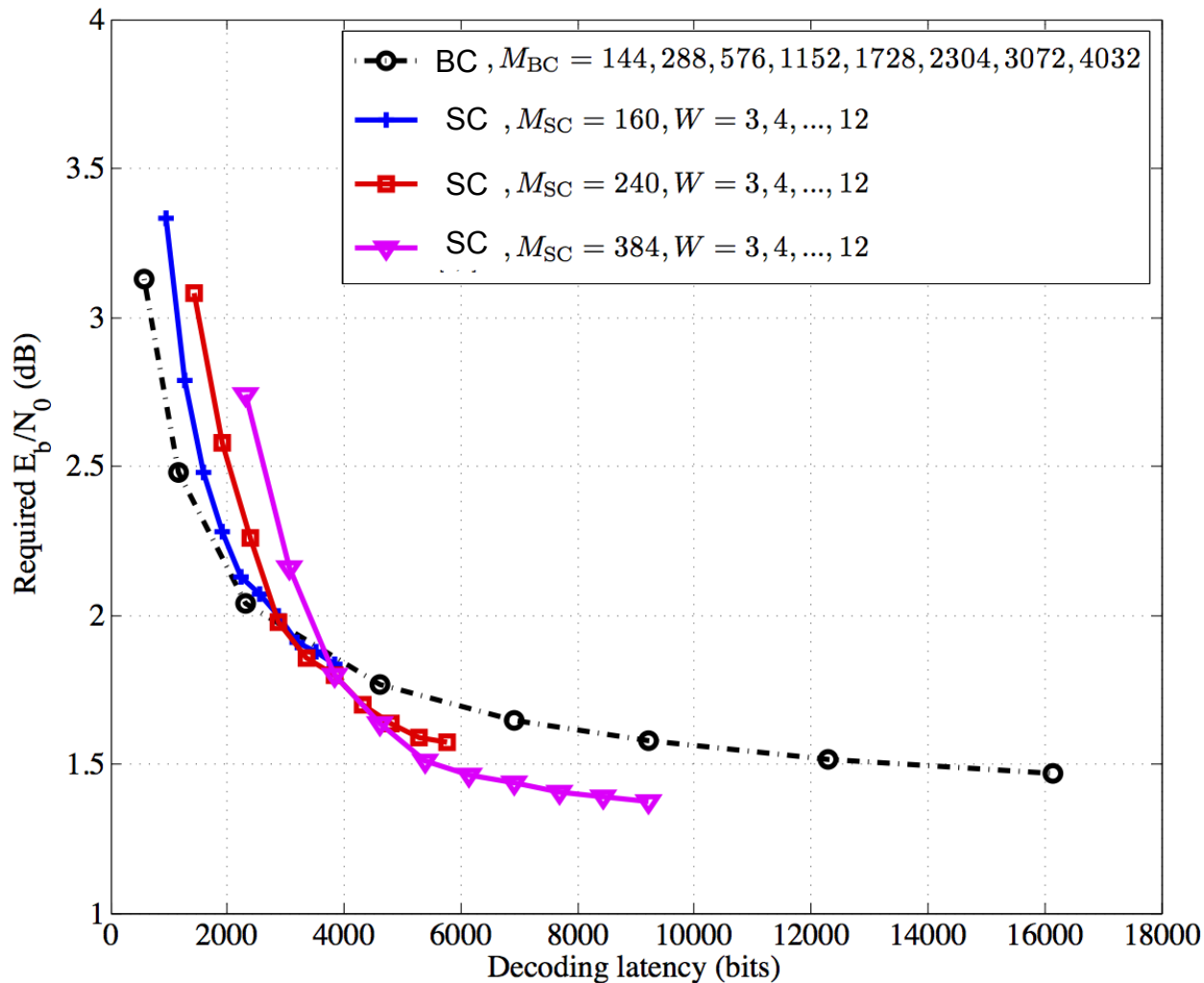
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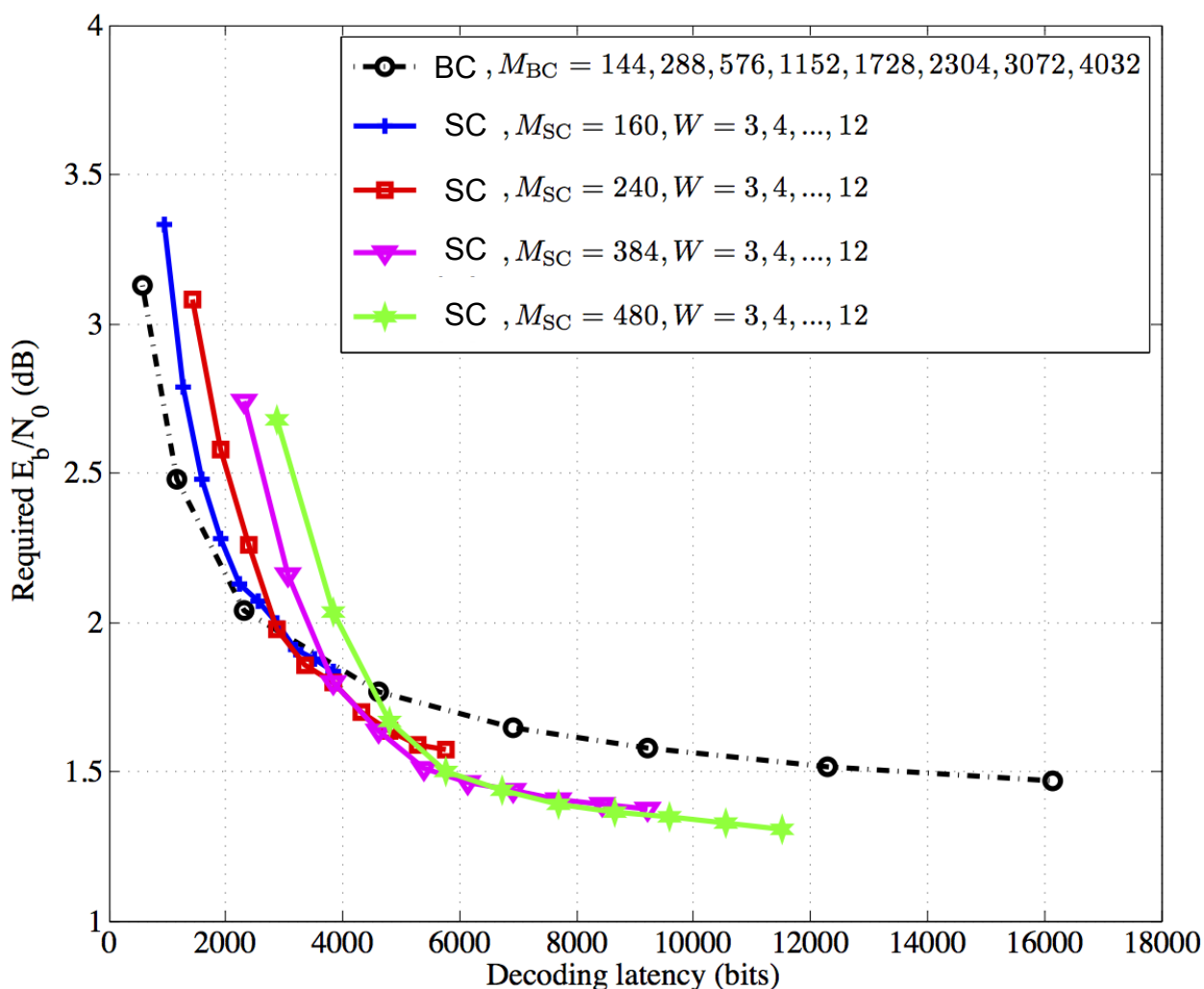
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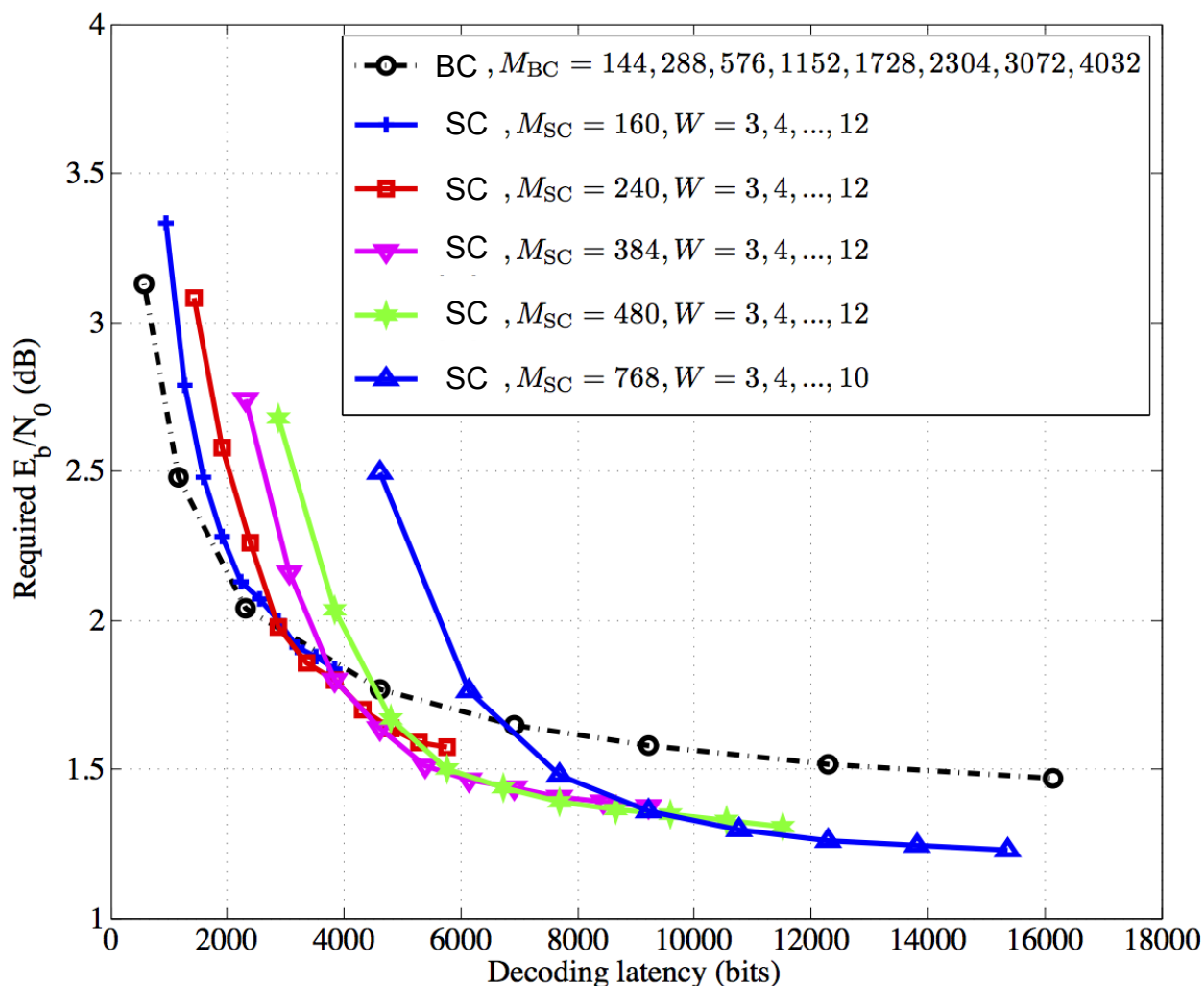
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SC-LDPC: $2M_{SC}W$

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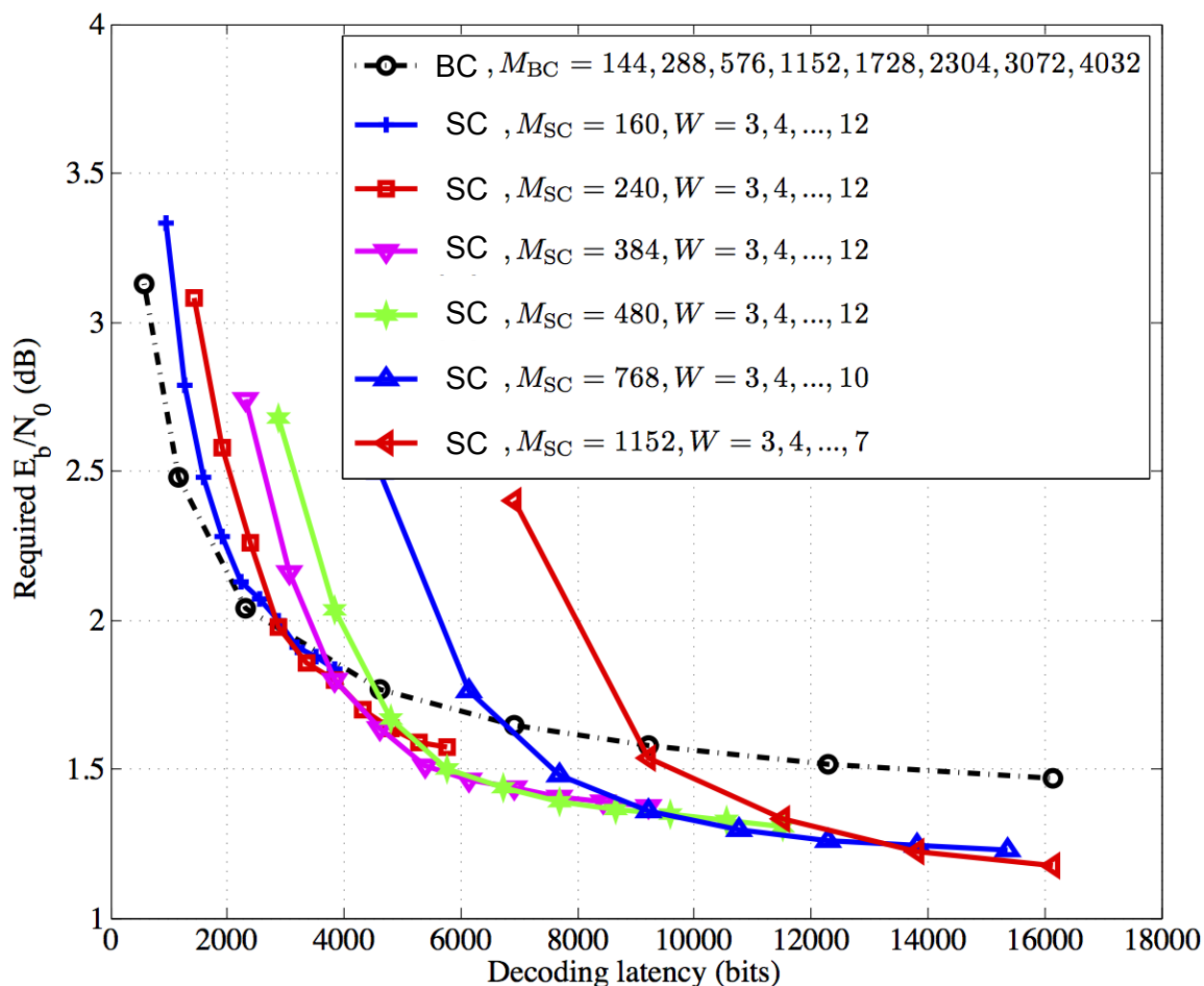
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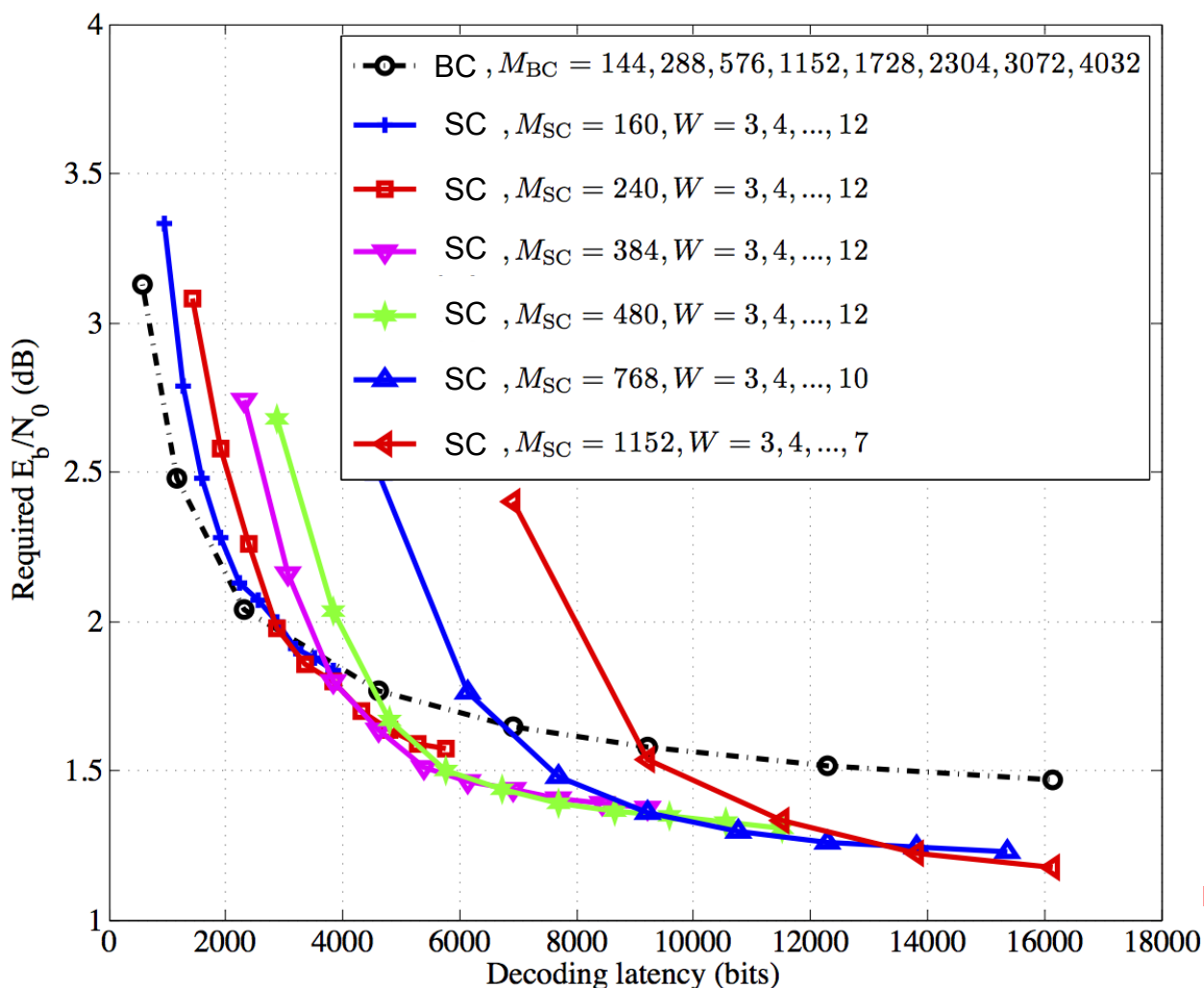
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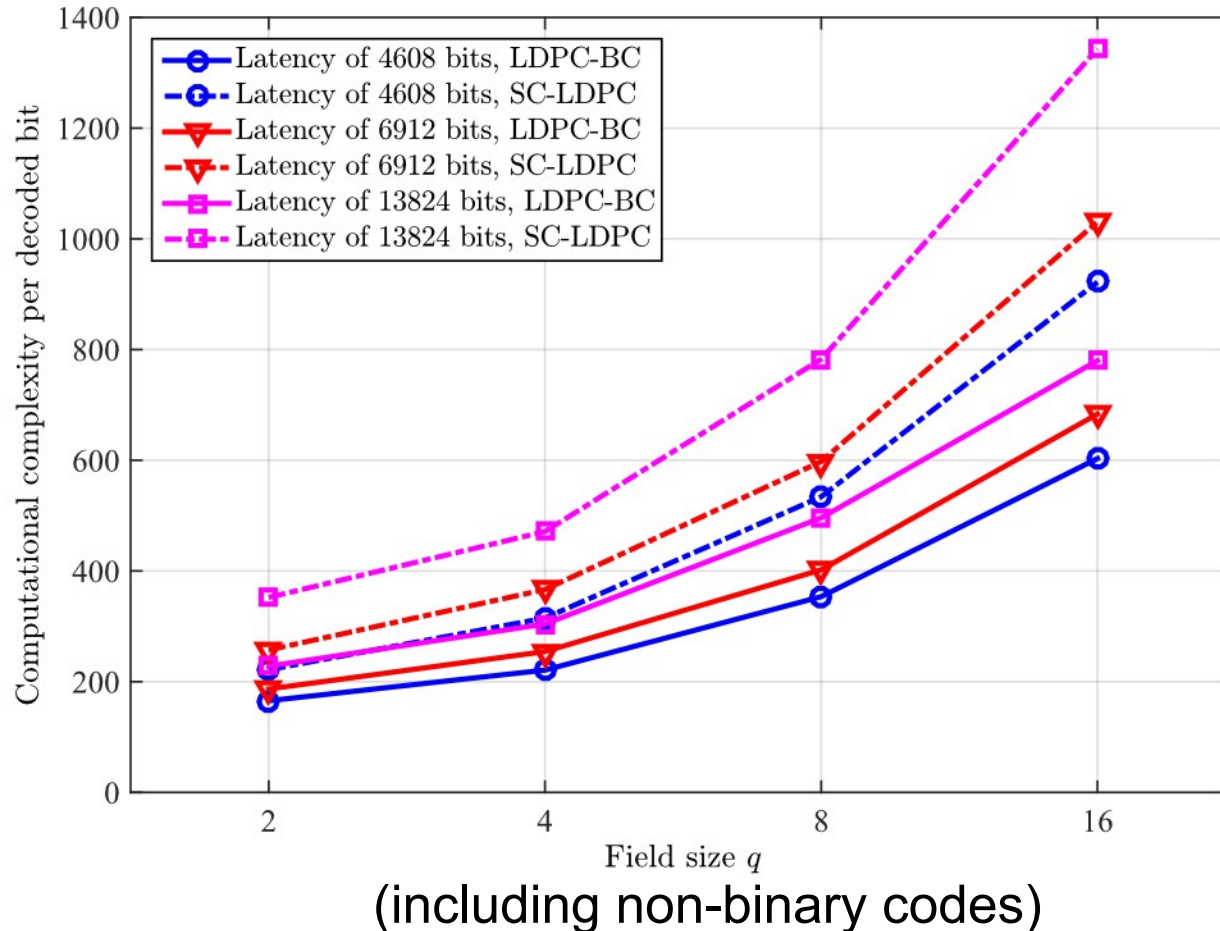
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- When choosing parameters:
 - large M_{SC} improves code performance.
 - large W improves decoder performance.

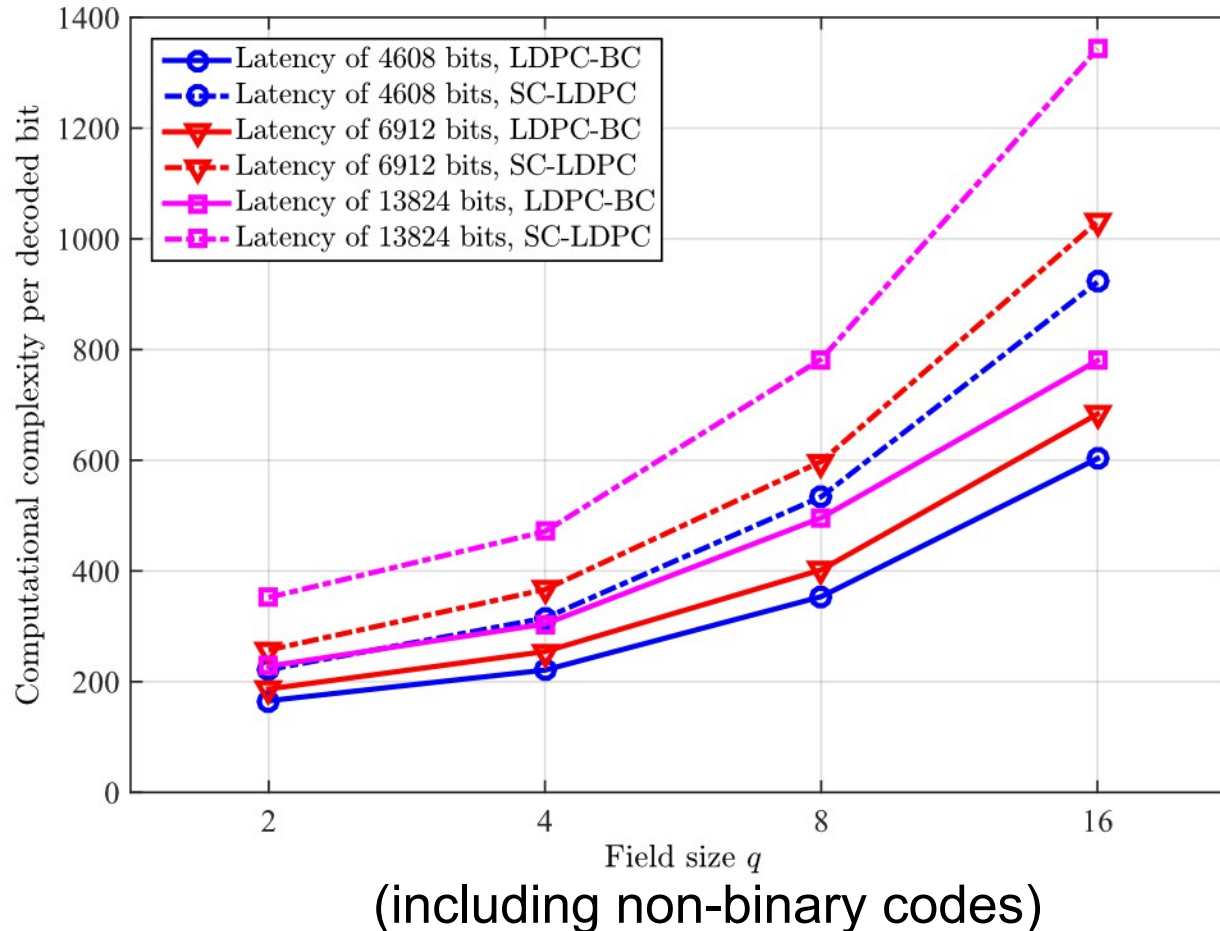
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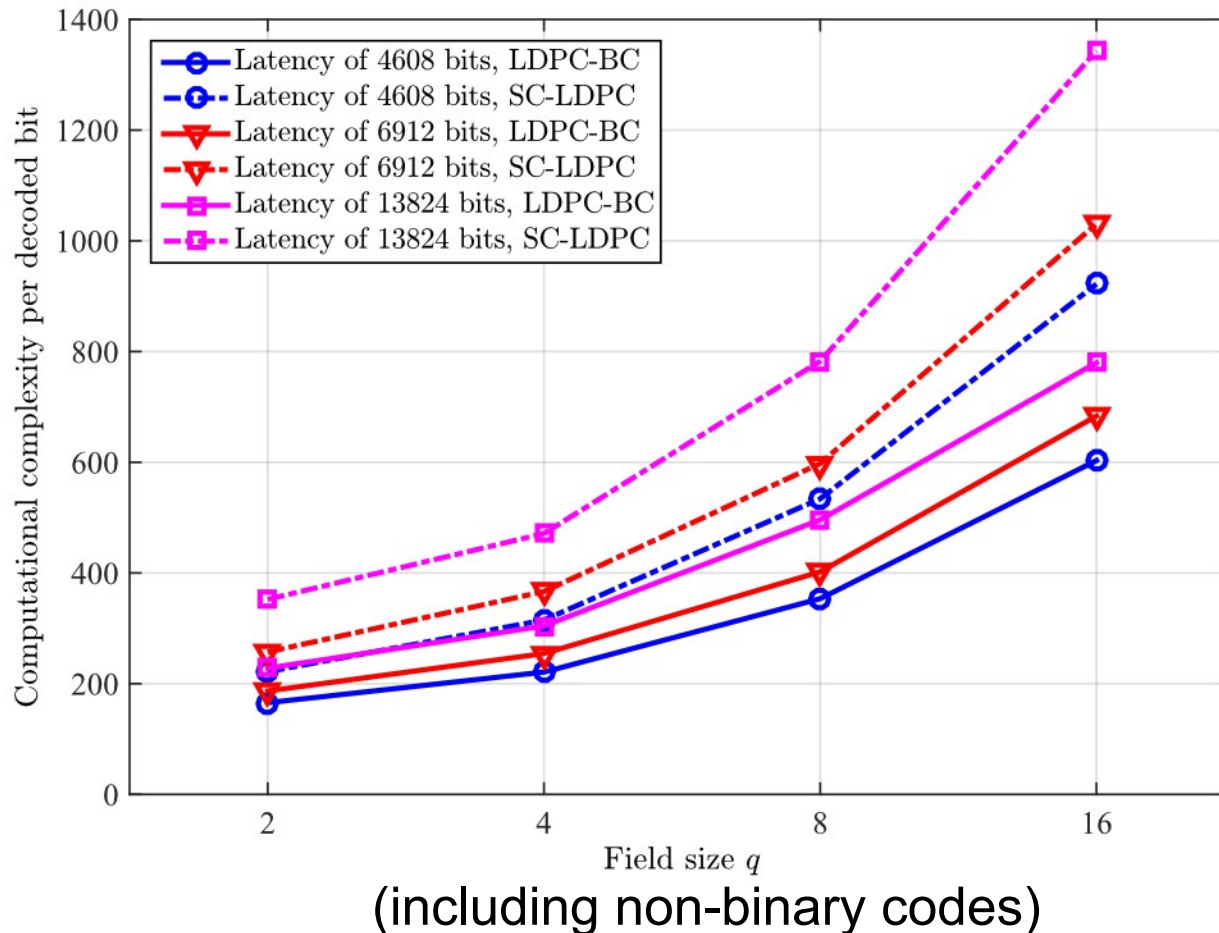
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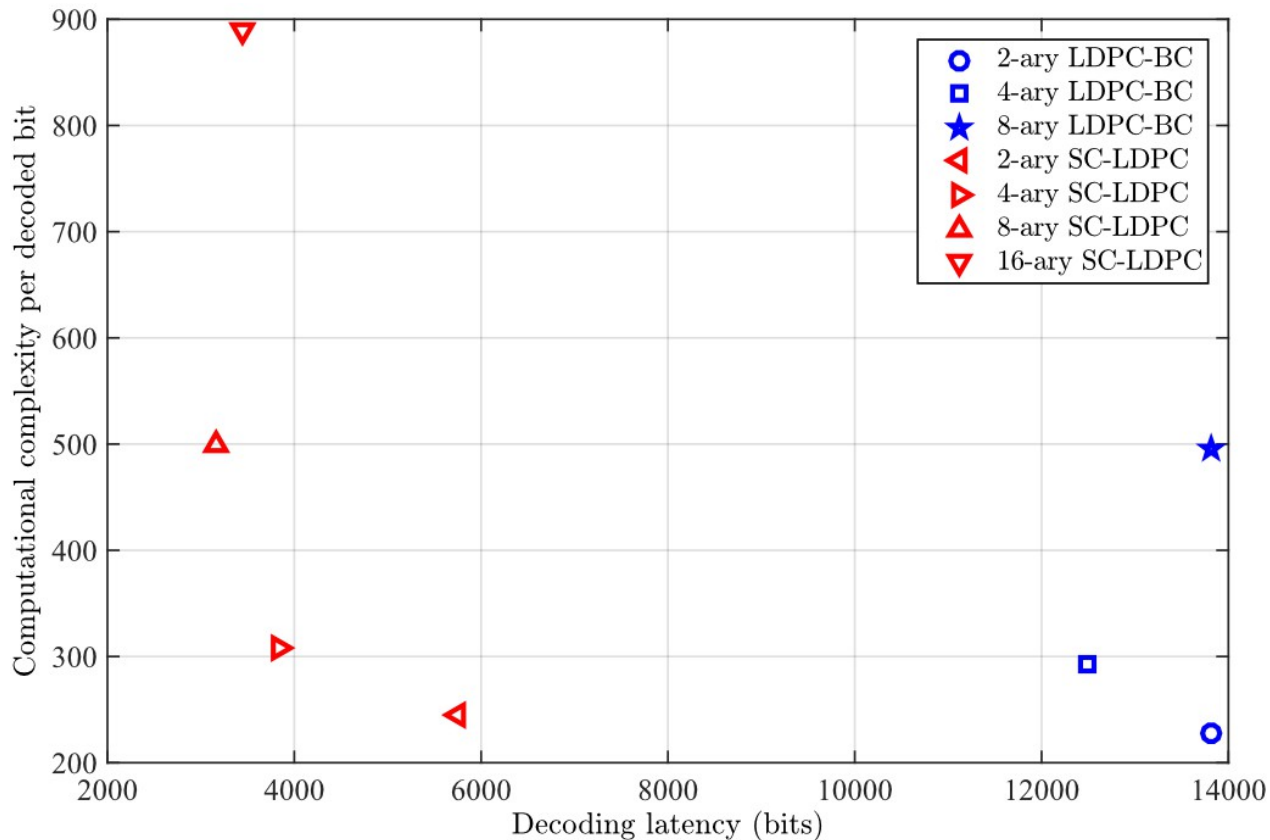
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Complexity/Latency Tradeoffs

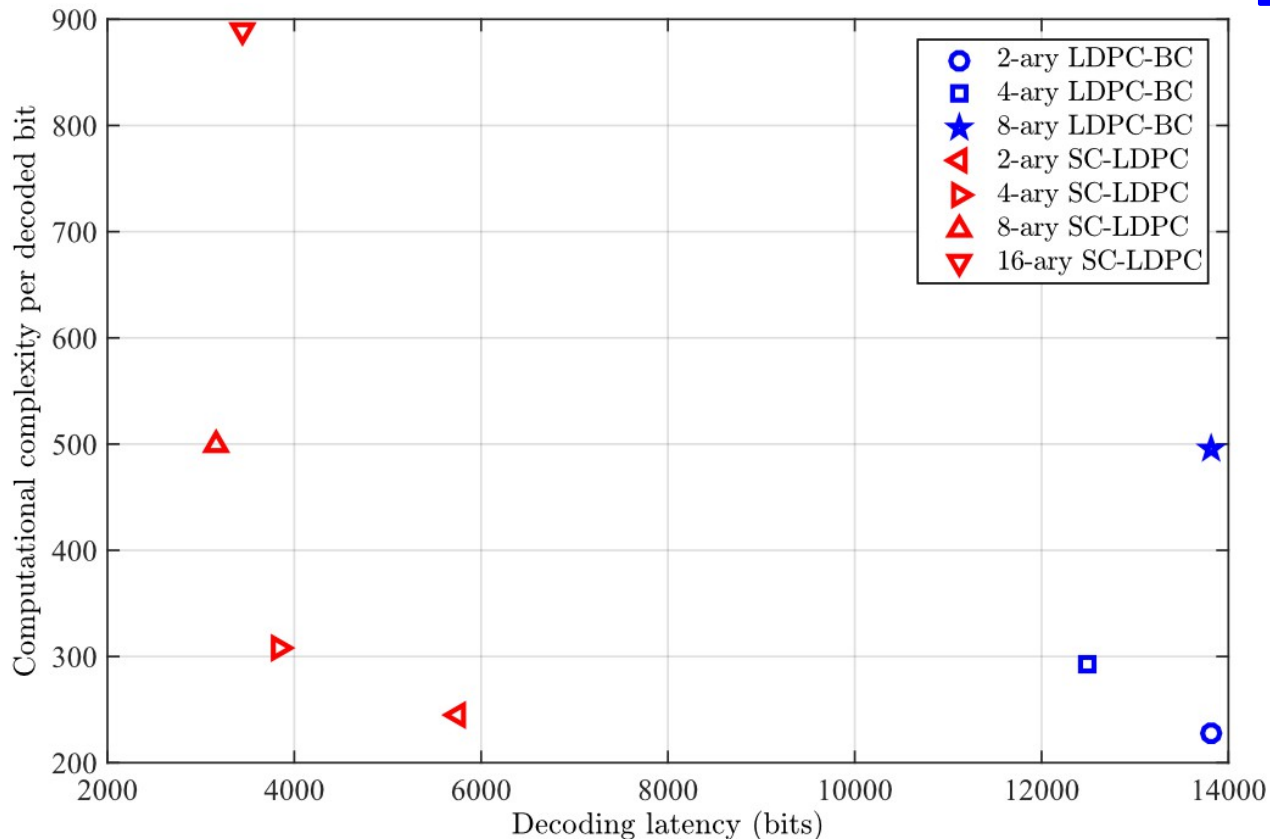
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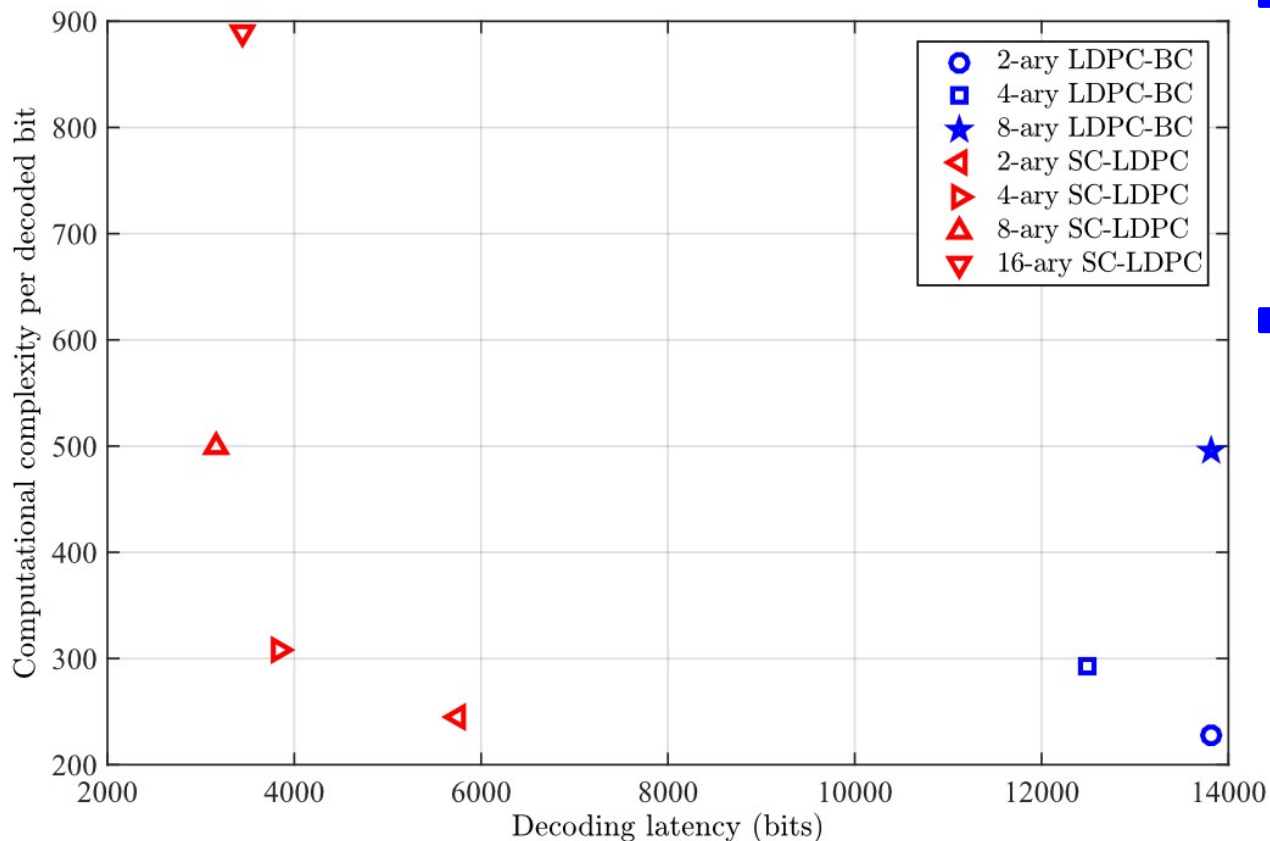


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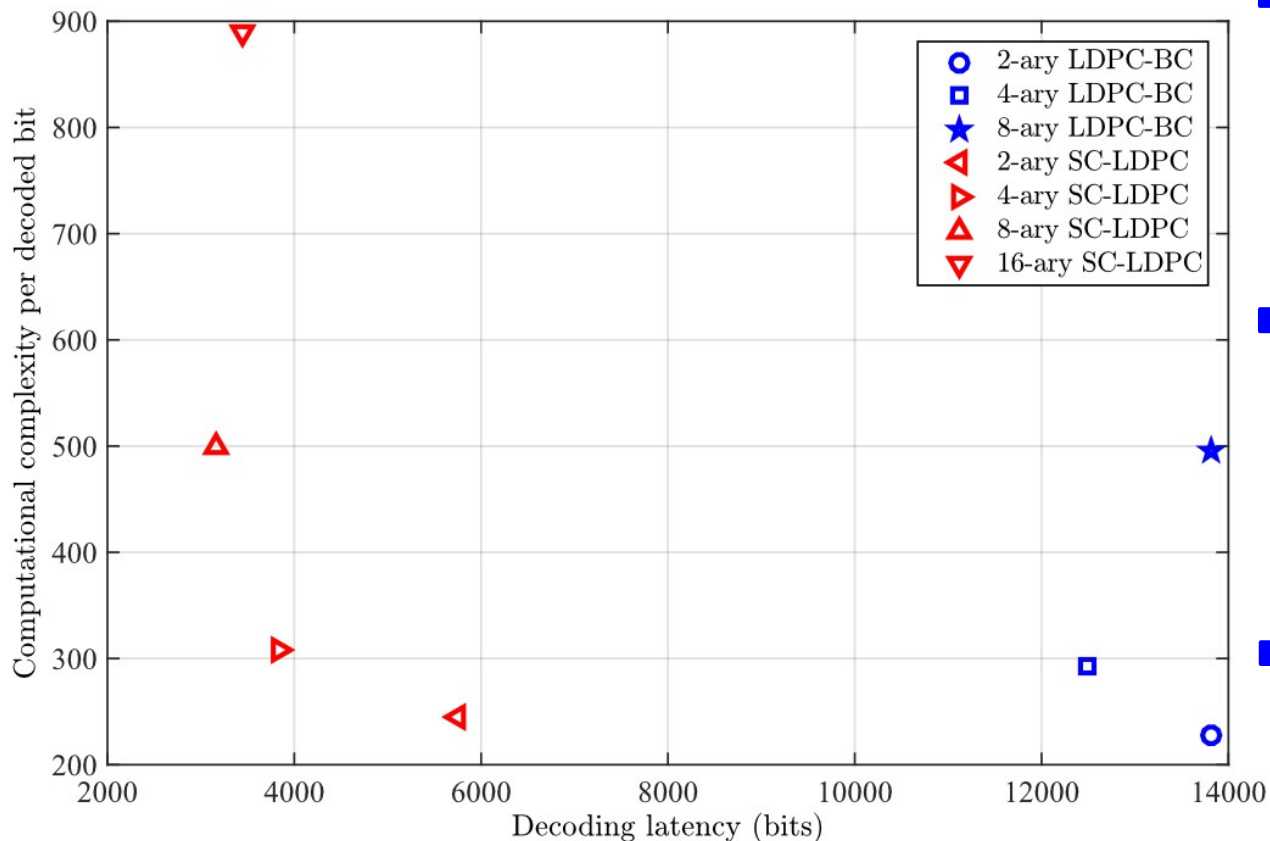


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- With increasing (small) field sizes q , latency decreases for increasing complexity
- For larger q , both latency **and** complexity increase (for both SC-LDPC codes and LDPC-BCs)!
- SC-LDPC codes over GF(4) offer a good balance between complexity and latency

Regular SC-LDPC Codes vs. Irregular LDPC-BCs

- Consider a comparison of a (3,6)-regular SC-LDPC code vs. an optimized irregular LDPC code with degree distribution

$$\lambda(x) = 0.409x + 0.202x^2 + 0.0768x^3 + 0.1971x^6 + 0.1151x^7$$
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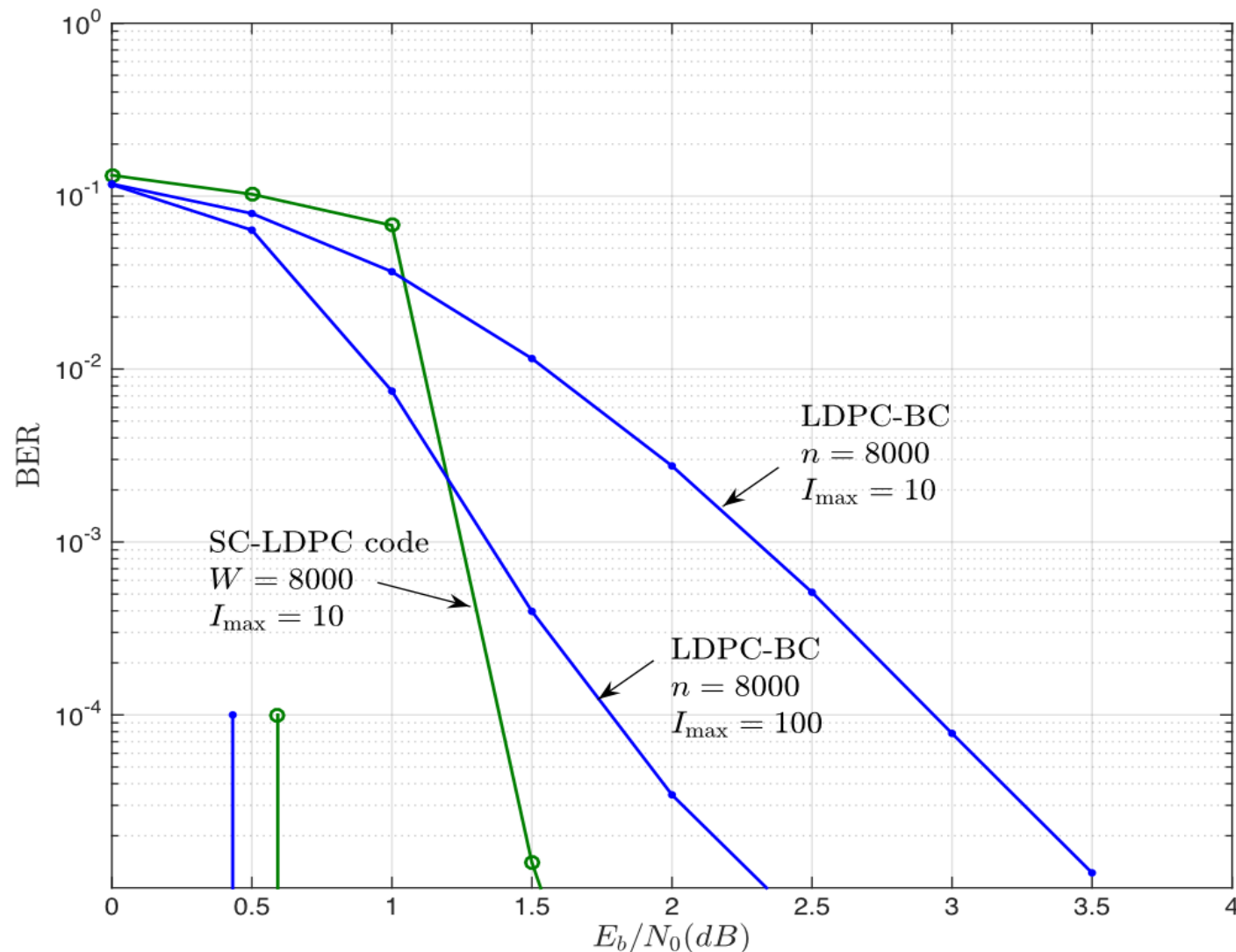
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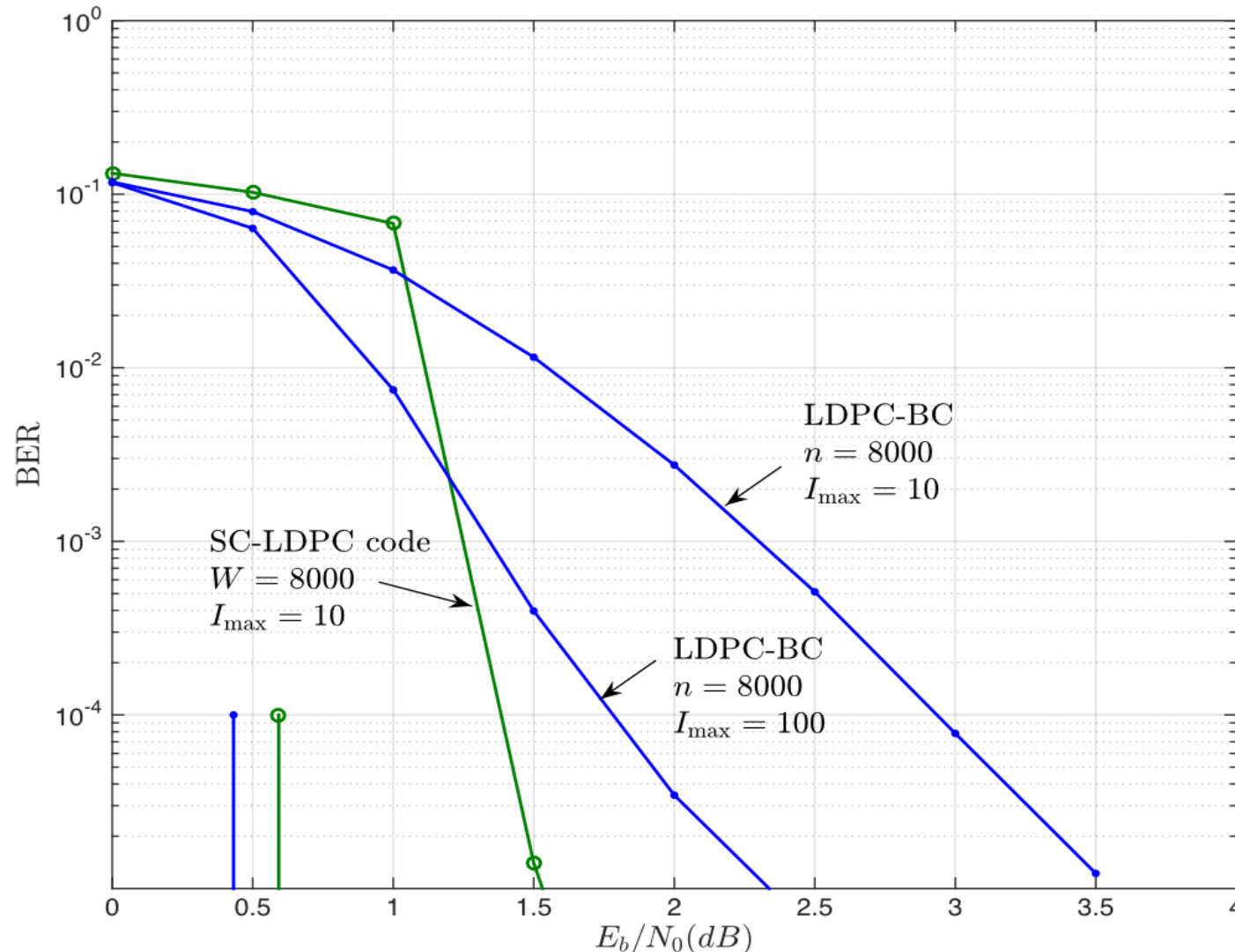
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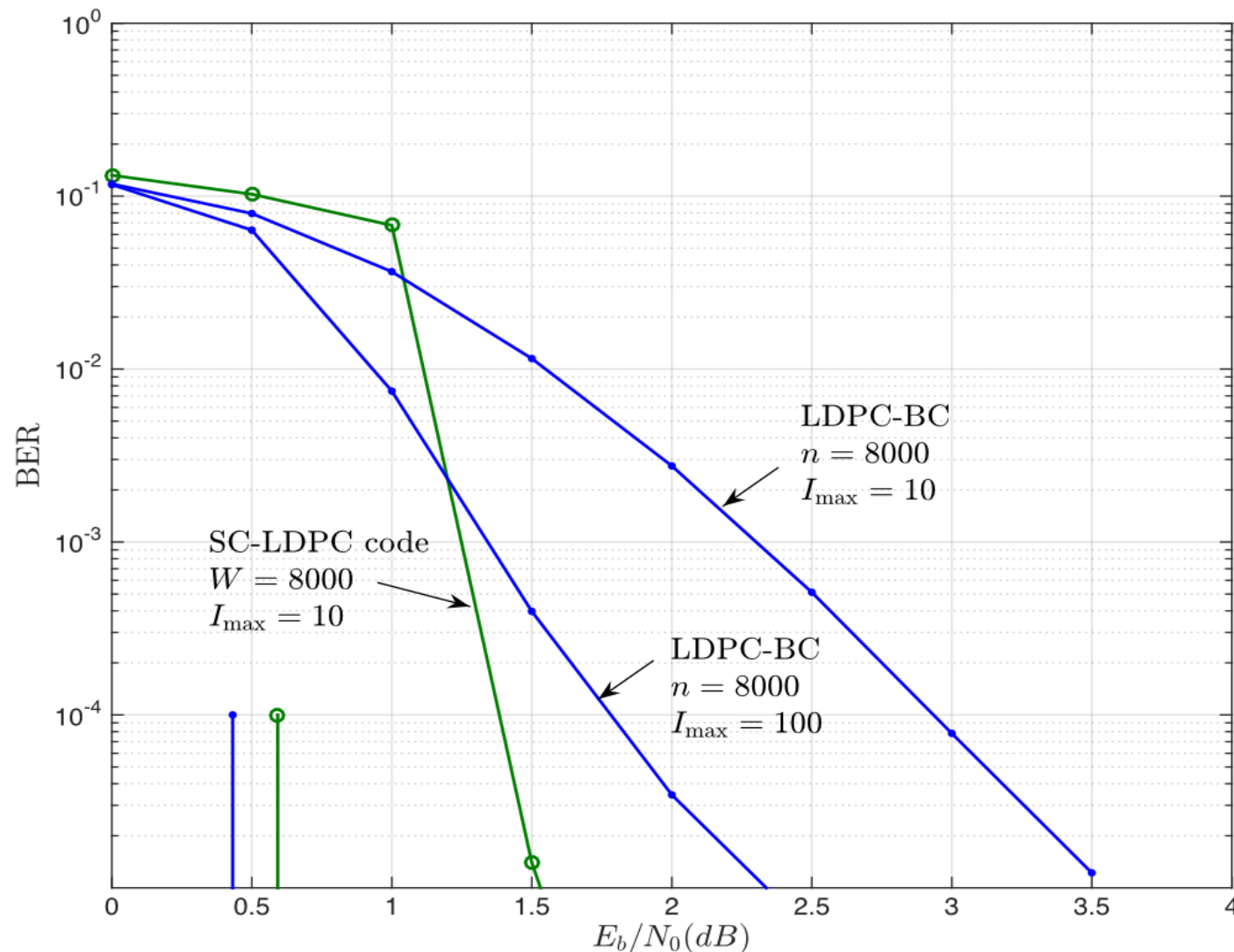
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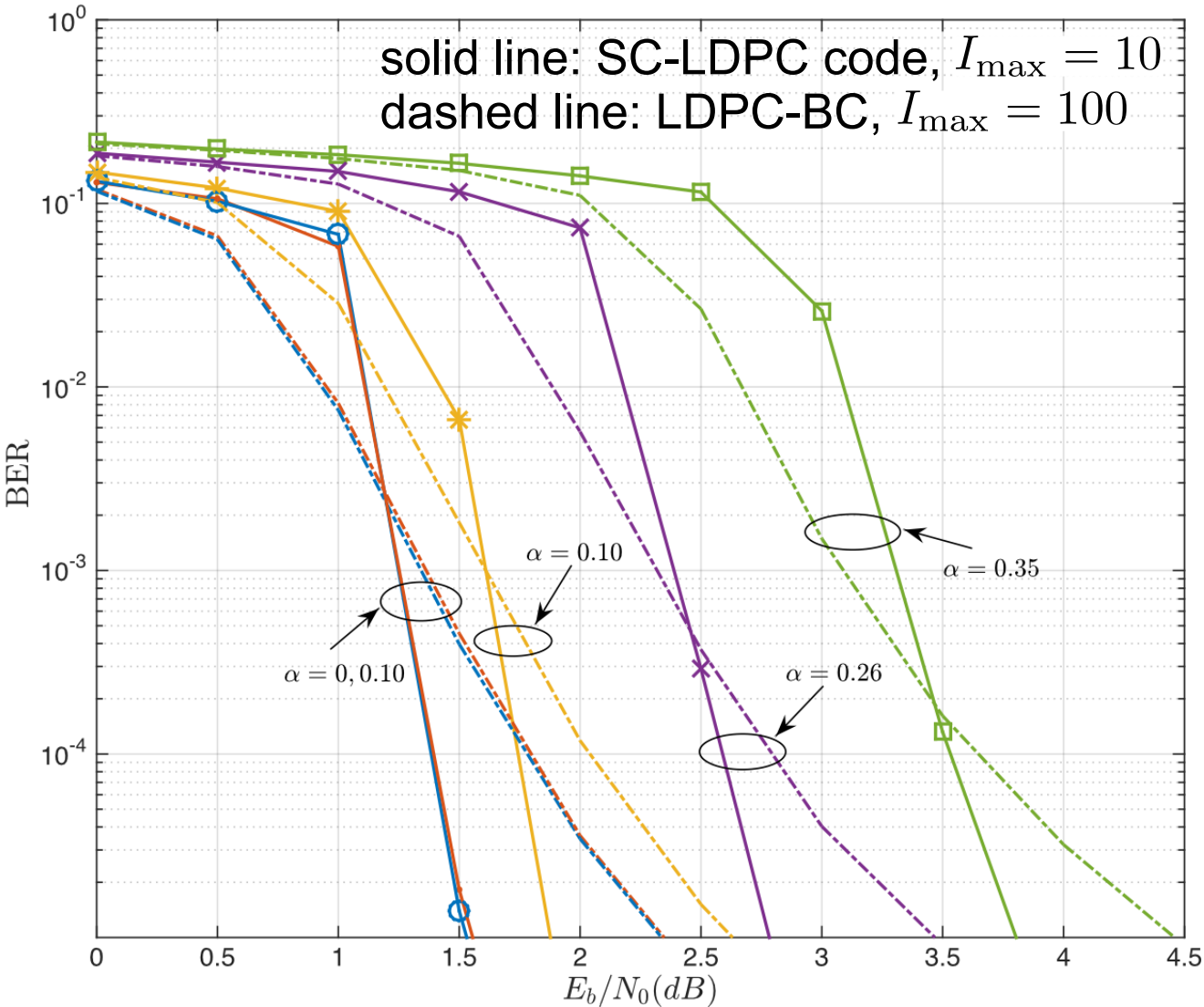
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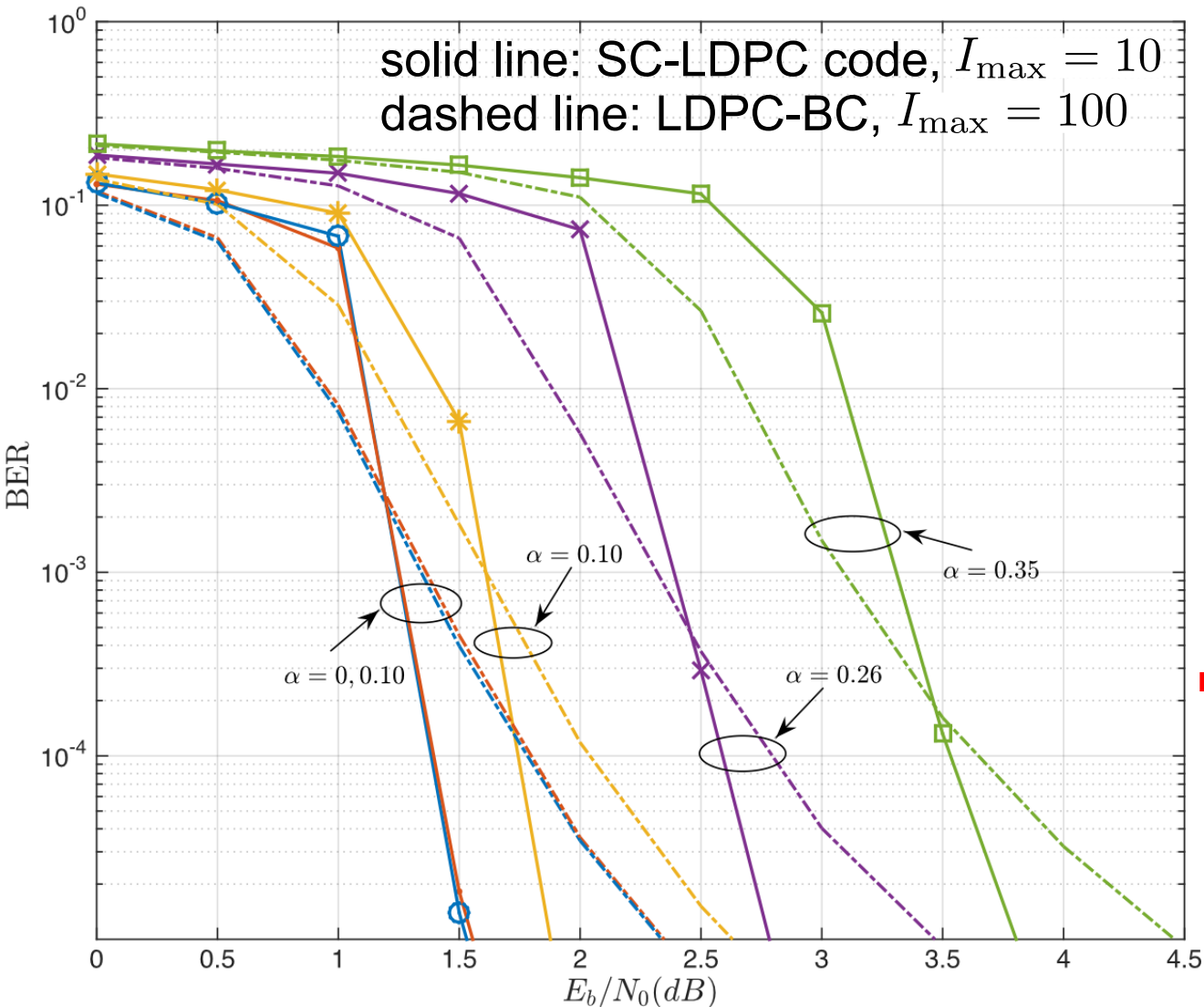
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- The regular structure has implementation advantages

Randomly Punctured LDPC Codes



- Random puncturing can be applied to LDPC code ensembles to increase the rate

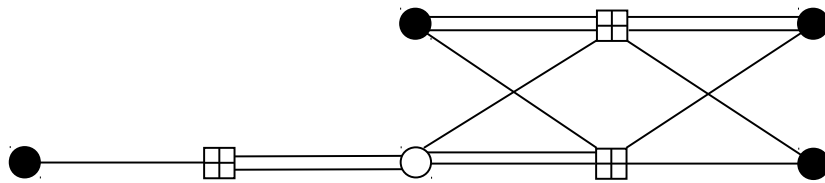
Randomly Punctured LDPC Codes



- Random puncturing can be applied to LDPC code ensembles to increase the rate
- Equal latency performance comparisons are consistent for higher rate ensembles
- ➔ Regular SC-LDPC codes display robust decoding performance compared to irregular LDPC-BCs

An Irregular Example

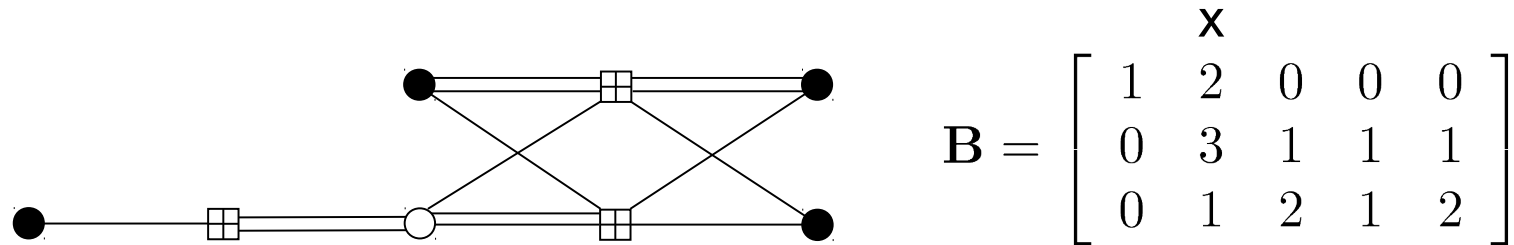
- Alternatively, we can couple irregular codes to construct an irregular SC-LDPC code ensemble. Consider the ARJA LDPC-BC protograph:



$$\mathbf{B} = \begin{matrix} & \mathbf{x} & & & \\ \begin{bmatrix} 1 & 2 & 0 & 0 & 0 \\ 0 & 3 & 1 & 1 & 1 \\ 0 & 1 & 2 & 1 & 2 \end{bmatrix} \end{matrix}$$

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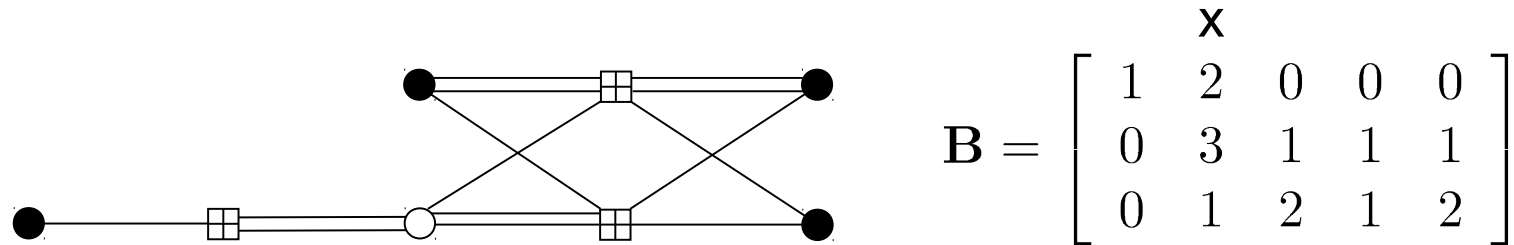


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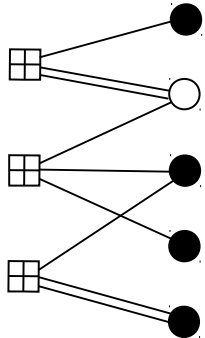
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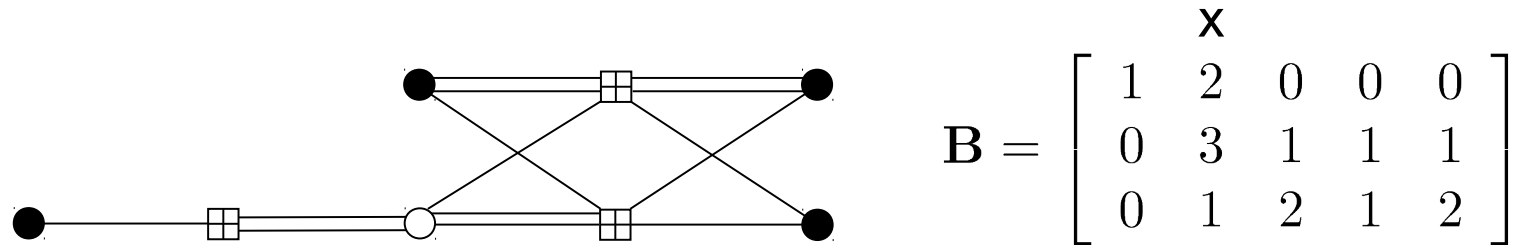


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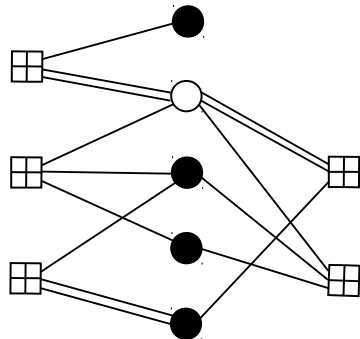


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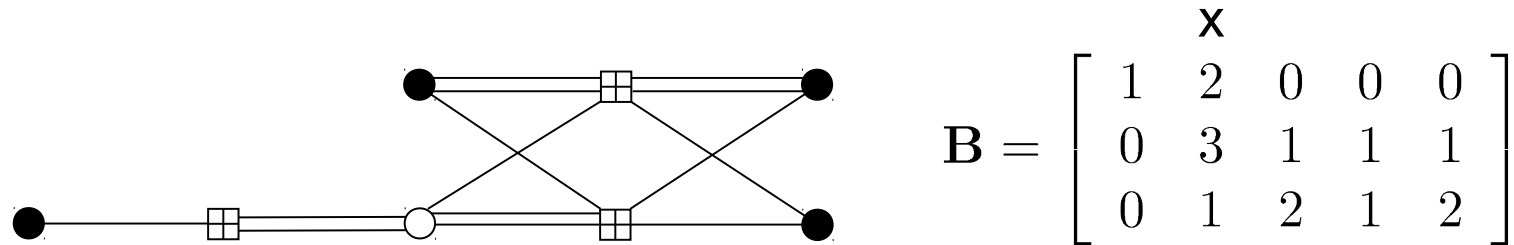


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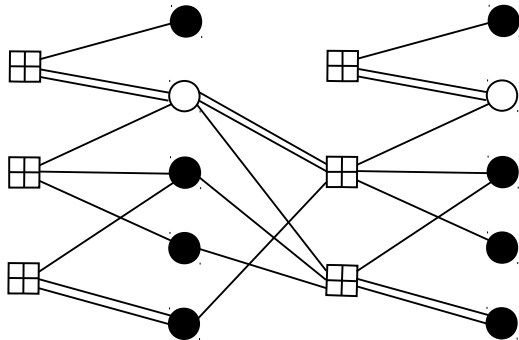
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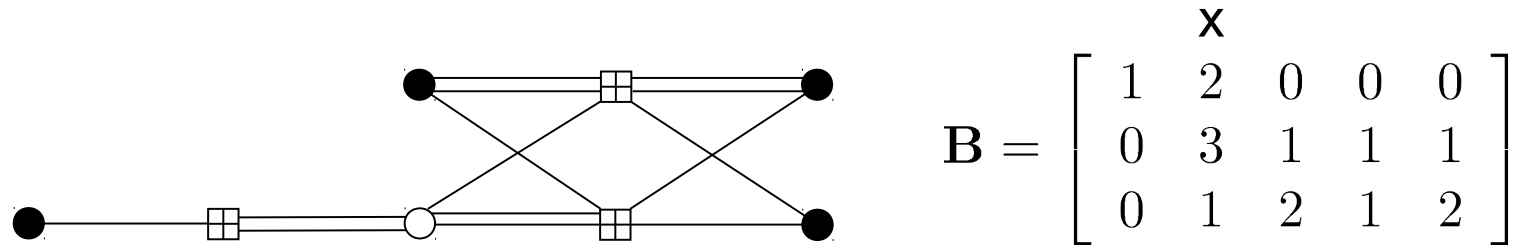
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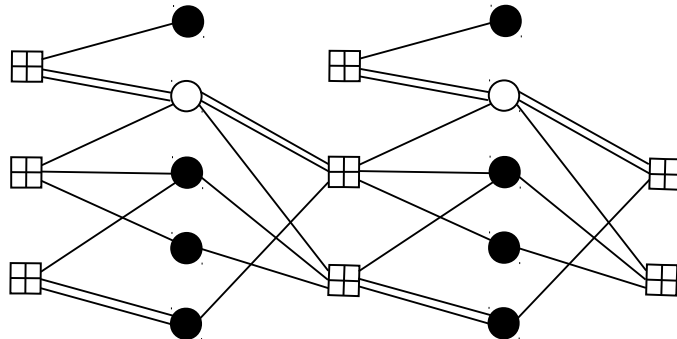


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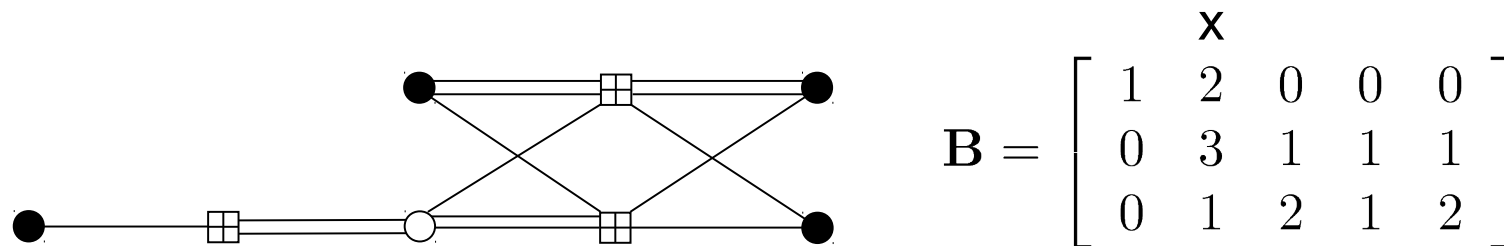


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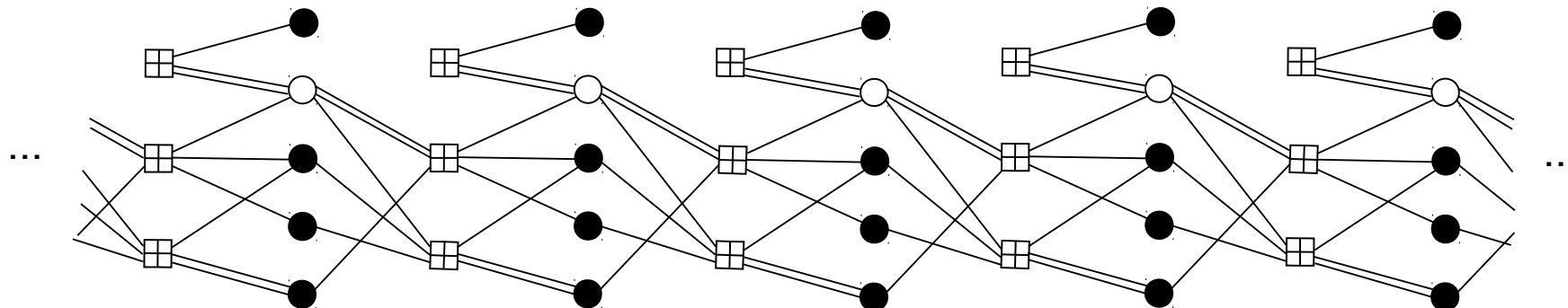


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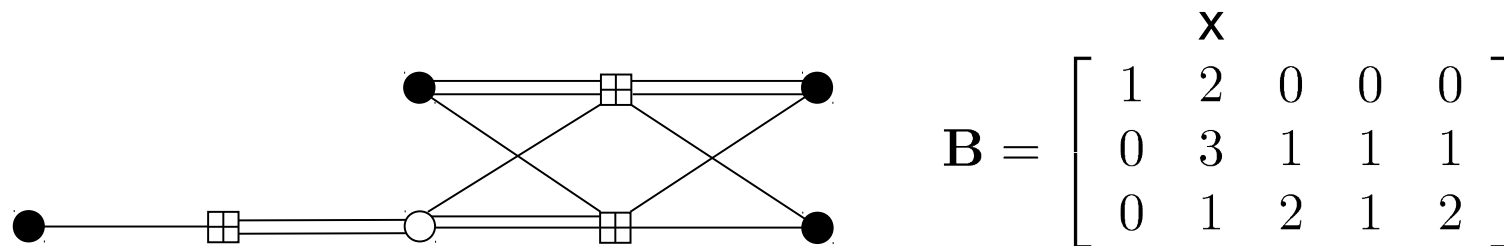


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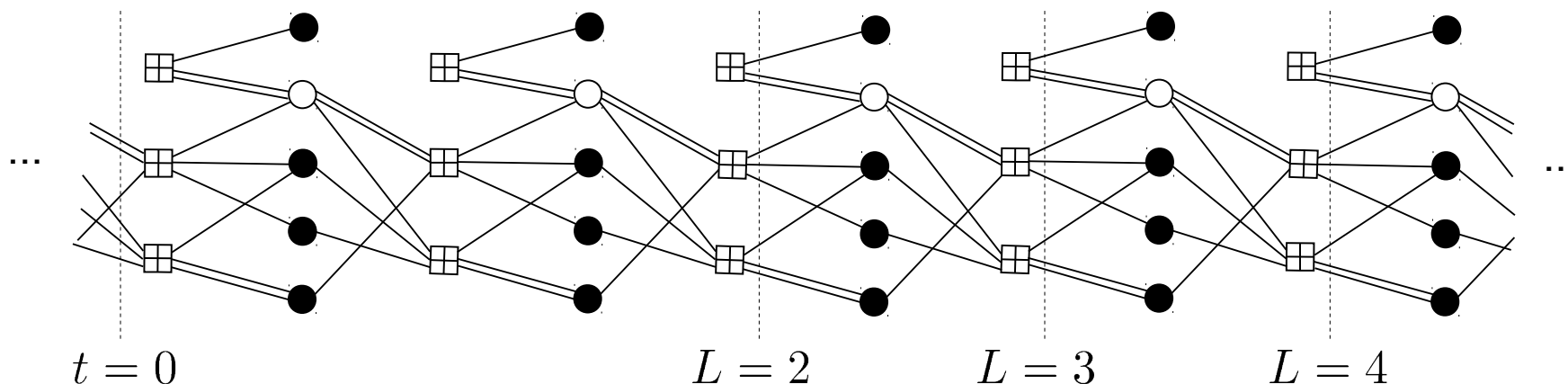


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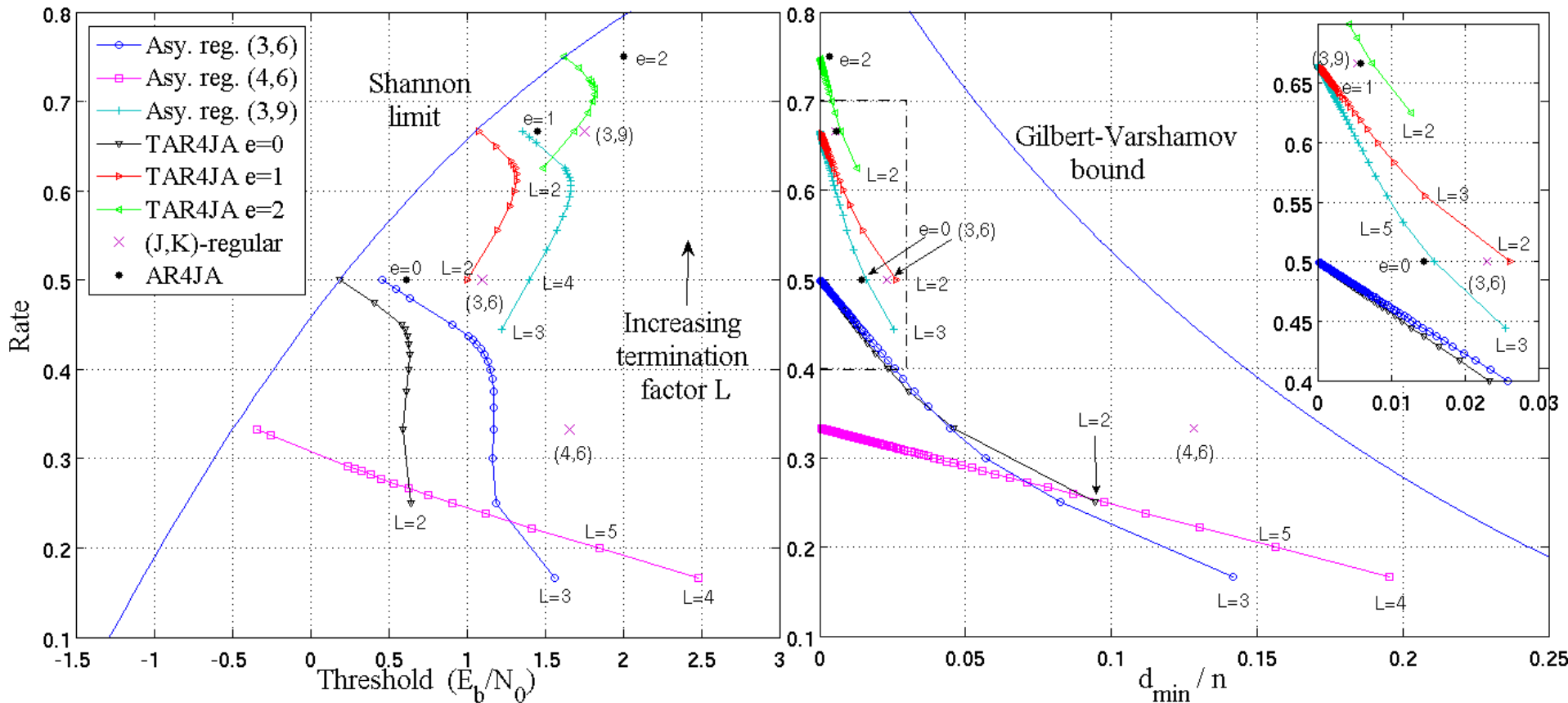
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■ Irregular SC-LDPC code ensembles also display excellent properties



[MLC10] D. G. M. Mitchell, M. Lentmaier and D. J. Costello, Jr., "AWGN Channel Analysis of Terminated LDPC Convolutional Codes", *Proc. Information Theory and Applications Workshop*, San Diego, Feb. 2011.

- As a result of their excellent performance and simple structure, regular SC-LDPC codes may be attractive for future coding standards. Several key features will require further investigation:
 - ➔ Hardware advantages of QC designs obtained by circulant liftings
 - ➔ Hardware advantages of the 'asymptotically-regular' structure
 - ➔ Design advantages of the flexible frame length feature obtained by varying L
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 - ➔ Flexible rate feature obtained by puncturing
- Of particular importance for applications requiring extremely low decoded bit error rates (e.g., optical communication, data storage) is an investigation of error floor issues related to **stopping sets**, **trapping sets**, and **absorbing sets**.

- Spatially coupled LDPC code ensembles achieve **threshold saturation**, i.e., their iterative decoding thresholds (for large L and M) approach the MAP decoding thresholds of the underlying LDPC block code ensembles.
- The threshold saturation and linear minimum distance growth properties of (J,K) -regular SC-LDPC codes combine the best asymptotic features of both regular and irregular LDPC-BCs.
- With window decoding, SC-LDPC codes also compare favorably to LDPC-BCs in the finite-length regime, providing flexible tradeoffs between BER performance, decoding latency, and decoder complexity.