

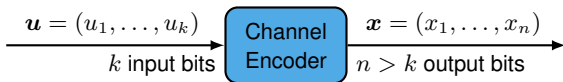
# Spatially Coupled LDPC Codes – Theory and Applications

Laurent Schmalen

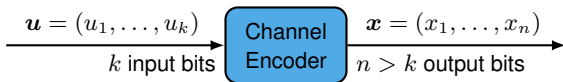
Communications Engineering Lab (CEL)



- 1 Introduction: Channel Coding and LDPC Codes
- 2 Spatially Coupled LDPC Codes
  - Motivation and Definition
  - Performance of SC-LDPC Codes
  - Practical Implementation of SC-LDPC Codes
  - Improvement of SC-LDPC Codes by Non-Uniform Coupling
- 3 Burst Correction Capabilities of Spatially Coupled LDPC Codes
  - Error Probability after Burst Erasures
  - Application Example
- 4 Conclusions



- Channel coding is indispensable in any communication and storage device
- The channel encoder encodes  $k$  information bits into  $n$  output bits, adding  $n - k$  parity bits



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## Channel code

A block code  $\mathcal{C}$  of length  $n$  and cardinality  $M = 2^k$  over a field  $\mathbb{F}$  is a collection of  $M$  elements  $\mathbf{x}^{[i]}$  from  $\mathbb{F}^n$

$$\mathcal{C}(n, M) := \left\{ \mathbf{x}^{[1]}, \mathbf{x}^{[2]}, \dots, \mathbf{x}^{[M]} \right\}, \mathbf{x}^{[m]} \in \mathbb{F}^n, 1 \leq m \leq M$$

The elements  $\mathbf{x}^{[i]}$  are called *codewords*. Here we only consider *binary codes*, where  $\mathbb{F} = \{0, 1\}$ .

- We consider only the sub-class of **linear block codes**

## Linear block code

A linear block code of length  $n$  and dimension  $k$  is a  $k$ -dimensional linear subspace of the vector space  $\mathbb{F}^n$ . As a linear subspace, the code can be represented also as the *span* of  $k$  codewords forming a basis. These basis codewords are the *rows* of a **generator matrix**  $\mathbf{G} \in \mathbb{F}^{k \times n}$ . The code is generated by multiplying all possible information vectors by the generator matrix

$$\mathcal{C}(n, k) = \{\mathbf{x}^{[i]} = \mathbf{u}^{[i]} \mathbf{G} : \forall \mathbf{u}^{[i]} \in \mathbb{F}^k\}$$

Alternative, the linear code can be seen as the *null space* of a **parity-check matrix**  $\mathbf{H}$  with  $\mathbf{G}\mathbf{H}^T = \mathbf{0}$

$$\mathcal{C}(n, k) = \{\mathbf{x} \in \mathbb{F}^n : \mathbf{H}\mathbf{x}^T = \mathbf{0}\}$$

# Linear Block Code – Example

- Consider the following parity-check matrix

$$H = \begin{pmatrix} 0 & 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 \end{pmatrix}$$

- For example:  $x = (1 \ 1 \ 1 \ 0 \ 0 \ 0)$  is a codeword

## Definition: Low-Density Parity-Check (LDPC) Codes

A low-density parity-check (LDPC) code is a *linear block code* with a *sparse* parity-check matrix  $H$ , i.e., the number of “1”s in  $H$  is very small, compared to the number of “0”s

- We consider the sub-class of *regular* LDPC codes

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- We consider the sub-class of *regular* LDPC codes

## Definition: Regular $[d_v, d_c]$ LDPC Codes

A regular  $[d_v, d_c]$  LDPC code is an LDPC code where each *column of  $H$  contains exactly  $d_v$  “1”s* and each *row of  $H$  contains exactly  $d_c$  “1”s*. We assume  $d_v \geq 2$  and  $d_c \geq 2$ .

- As the number of “1”s in  $H$  is small, the memory complexity is only  $O(n)$  as the number of “1”s per column is constant (and independent of  $n$ )





# Graph Representation of LDPC Codes

■ **Example:**  $[2, 4]$  LDPC code with  $n = 6$

$x_1$  ●

$x_2$  ●

$x_3$  ●

$x_4$  ●

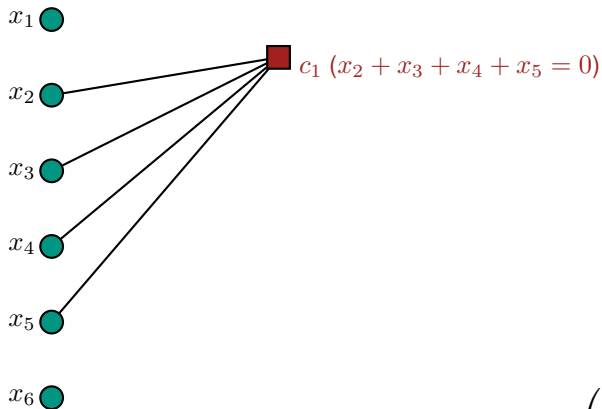
$x_5$  ●

$x_6$  ●

$$H = \begin{pmatrix} 0 & 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 \end{pmatrix}$$

# Graph Representation of LDPC Codes

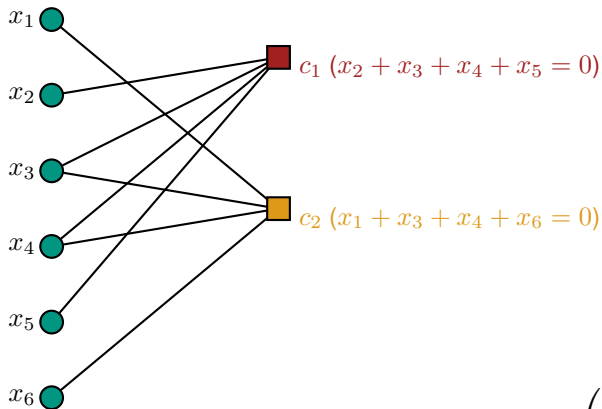
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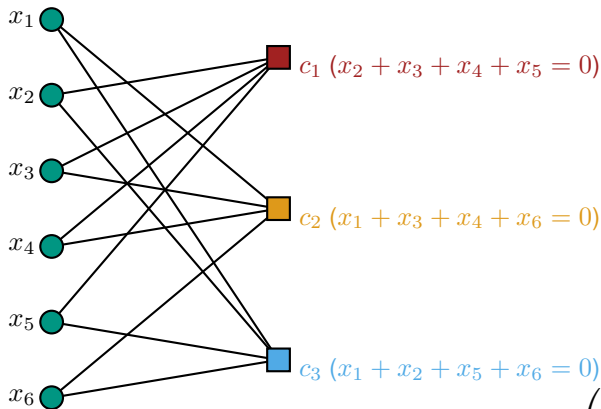
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$$H = \begin{pmatrix} 0 & 1 & 1 & 1 & 1 & 0 \\ 1 & 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 \end{pmatrix}$$

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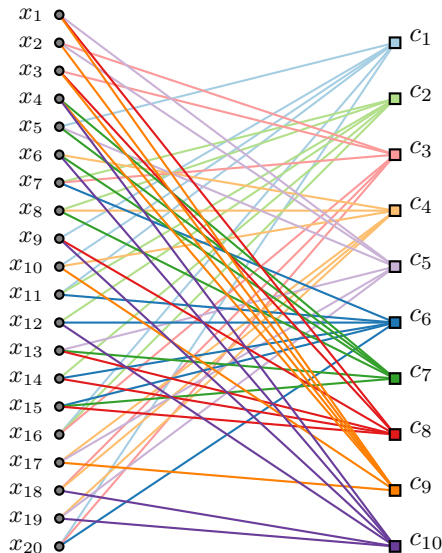
## Example: $[3, 6]$ Regular Code

Consider the following parity-check matrix of the  $d_v = 3, d_c = 6$  code

$$H = \begin{pmatrix} 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 1 & 0 \\ 1 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 \end{pmatrix}$$



# Graph of $[3, 6]$ Regular Code

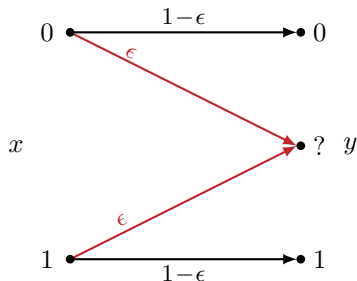


- Every constraint  $c_1, \dots, c_{10}$  corresponds to a row of the parity-check matrix of previous slide
- Every code bit (variable) has  $d_v = 3$  outgoing edges
- Every constraint node  $c_i$  has  $d_c = 6$  connected edges



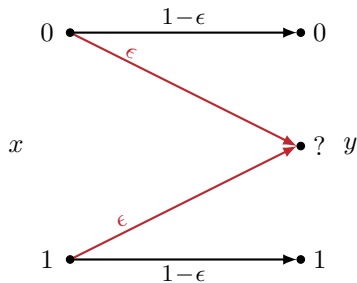
# Binary Erasure Channel (BEC)

- Simple channel model that enables easy analysis but is close to practice
- Channel with 3 output symbols, where “?” denotes an *erasure*, i.e., no information about transmitted symbol
- BEC can be used to model, e.g., packet losses or high SNR channels

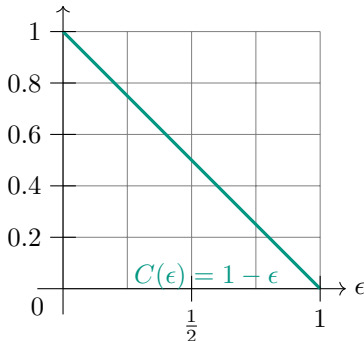


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Capacity  $C(\epsilon)$



- *What is the necessary  $\epsilon$  for iterative decoding to be successful?*

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- Denote the channel erasure probability by  $\epsilon$
- Iterative message passing decoder exchanging *beliefs* or *messages* along the edges of the Tanner graph
- Analyse decoding behavior of the code using a simple update equation

## Update equation for regular $[d_v, d_c]$ LDPC codes on the BEC

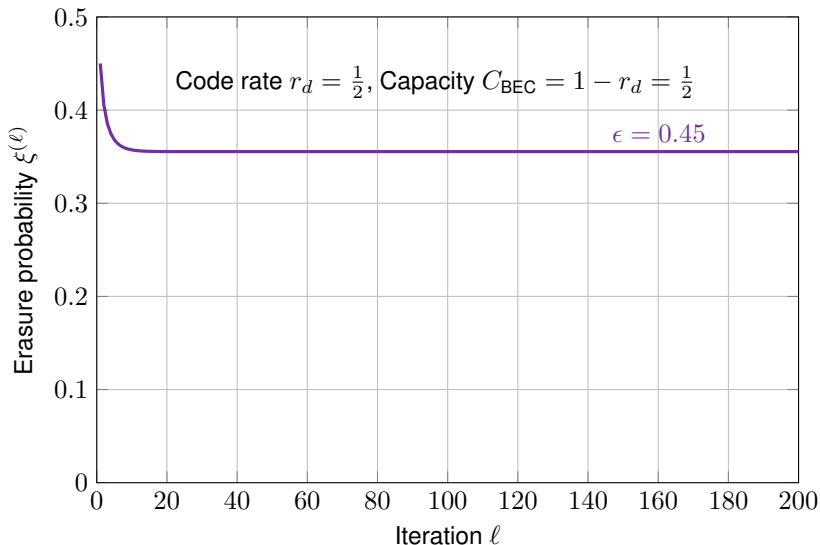
The average message erasure probability  $\xi_\ell$  after  $\ell$  decoding iterations is given by

$$\xi^{(\ell)} = \epsilon \left( 1 - (1 - \xi^{(\ell-1)})^{d_c-1} \right)^{d_v-1}$$

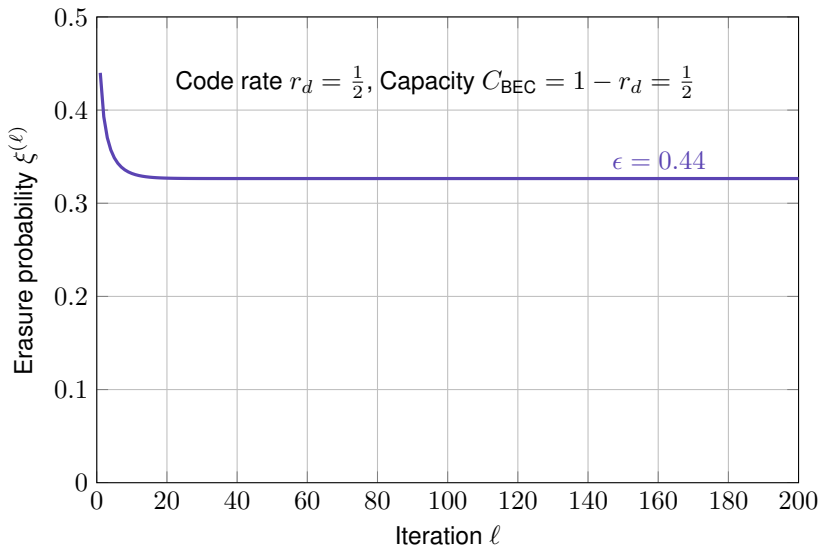
with  $\xi_0 = 1$ .

- We can decode successfully (if  $n \rightarrow \infty$ ) if  $\xi^{(\ell)} \rightarrow 0$
- The **threshold**  $\epsilon^*$  is the largest  $\epsilon$  for which  $\xi^{(\ell)} \rightarrow 0$

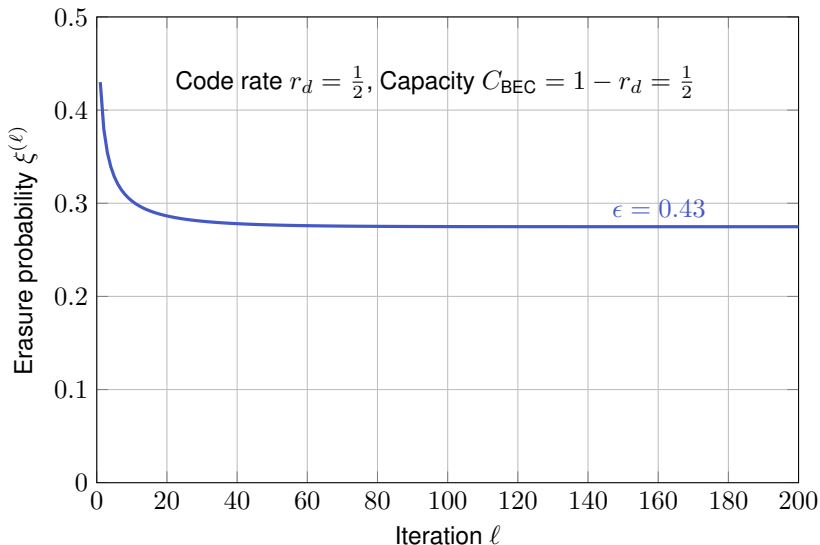
# Density Evolution for the Regular [3,6] LDPC Code



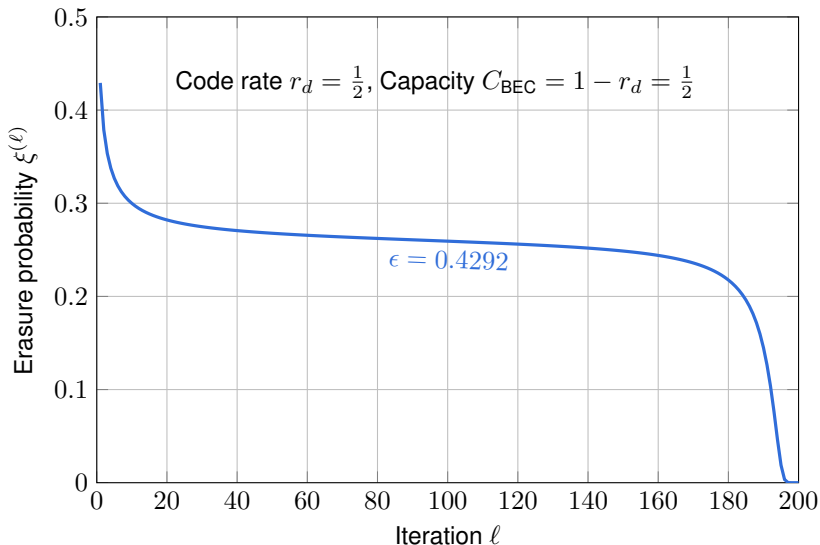
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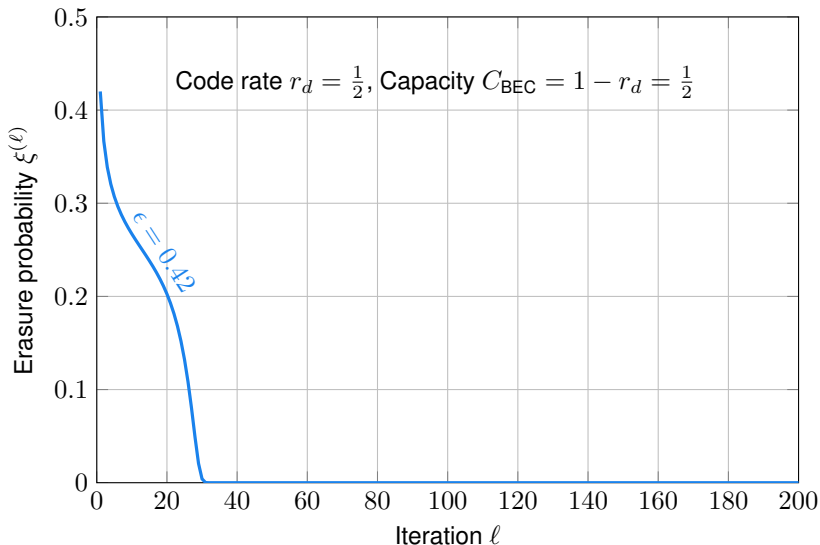


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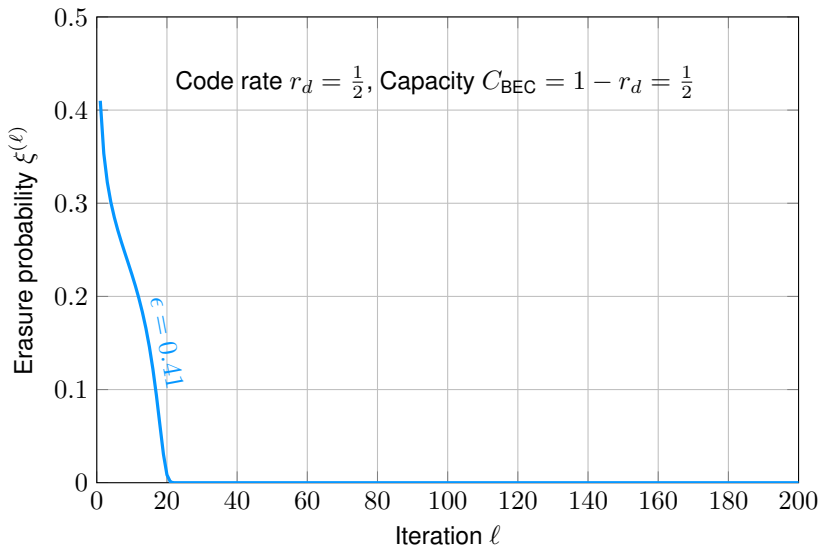




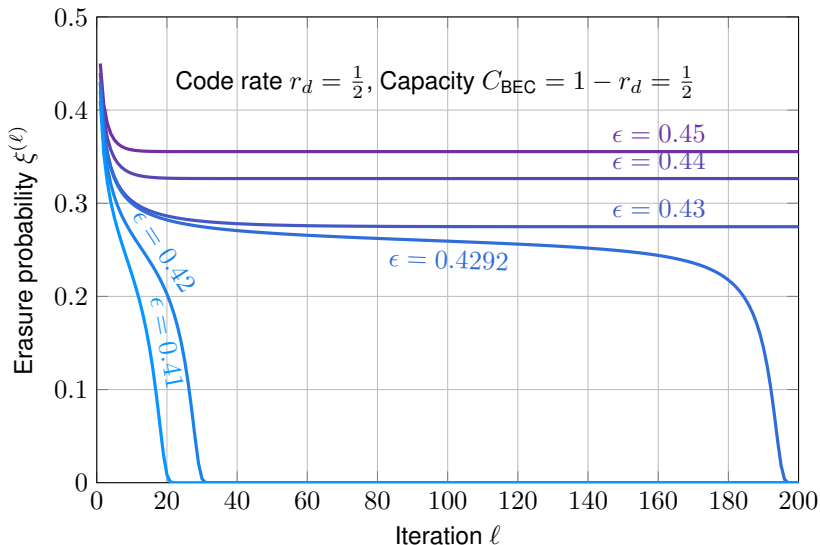
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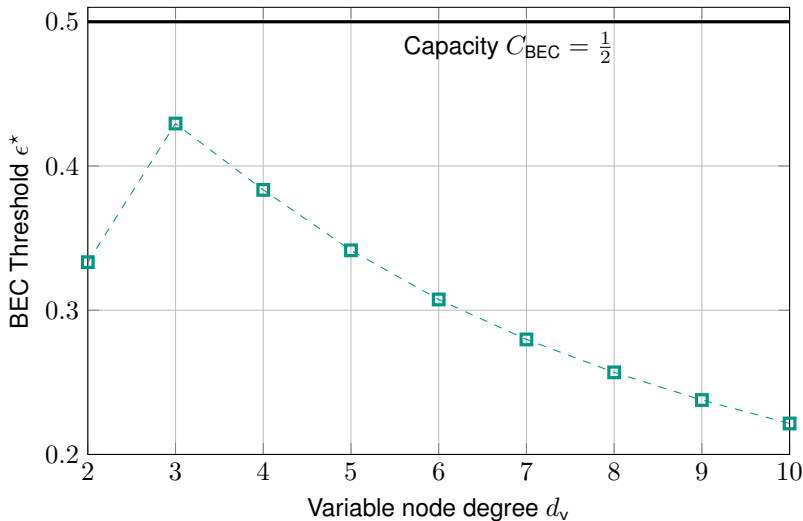
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# Thresholds of Rate- $\frac{1}{2}$ $[d_v, d_c = 2d_v]$ Codes



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- **Gap to capacity** with regular LDPC codes
- Remedy A: **Irregular LDPC Codes** [RSU01]
  - Can close the gap to capacity
  - Drawback: Increased error floor, not suitable for high-reliability applications
- Remedy B: **Protograph LDPC Codes** [Tho03]
  - Generalizes the construction of LDPC Codes
  - Optimization of very good codes with low error floors still difficult

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- New approach needed that extends protograph LDPC codes
- **Spatially coupled LDPC codes**

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[Tho03] J. Thorpe, "Low-density parity-check (LDPC) codes constructed from protographs," *JPL IPN progress report 42.154*, 2003

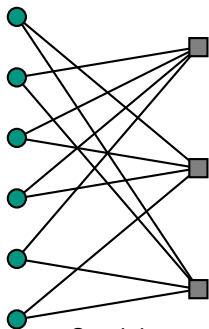


- Block codes:
  - encode a group of  $k$  input bits  $\mathbf{u}$  into a codeword  $\mathbf{x}$
  - transmit over channel and receive  $\mathbf{y}$
  - decode  $\mathbf{y}$  and proceed to next block
- Block processing independent of previous and next block

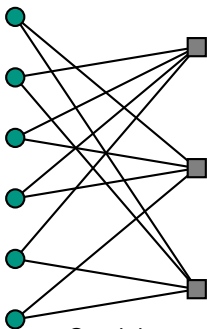
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- **Spatial coupling:** introduce dependencies between the neighboring blocks so that they can help each other during decoding.
- We introduce a “spatial” dimension  $t$  to describe blocks. It is called “spatial” dimension as the temporal dimension is reserved for the bits **inside** a codeword.
- Let  $w$  denote the coupling width.

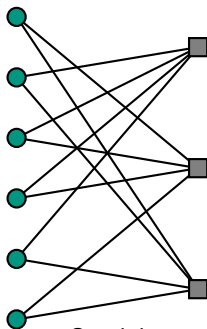
# Example: $L = 3$ Codes Along Spatial Dimension



Spatial position  
 $t = 1$



Spatial position  
 $t = 2$



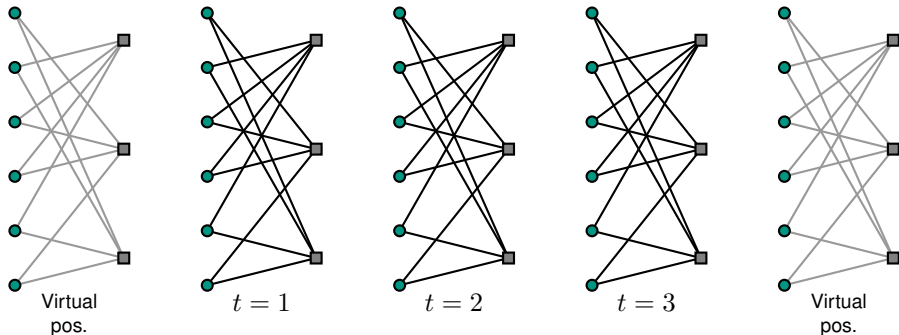
Spatial position  
 $t = 3$

- No coupling, independent codes ( $w = 1$ )

## Spatial Coupling

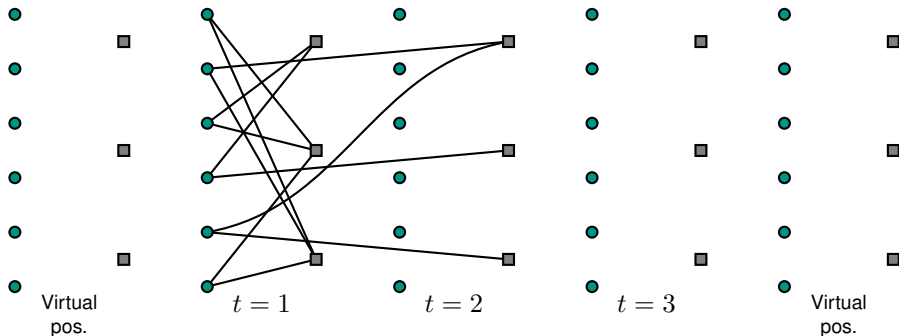
- Consider a spatial chain of  $L$  codes ( $t = 1, \dots, L$ ).
- Consider each edge of the Tanner graph at spatial position  $t$  and connect it **randomly** to a check node of spatial positions  $t, t + 1, \dots, t + w - 1$  with probability  $1/w$  such that the **local degree distributions** are **preserved**
- At each boundary insert  $w - 1$  **virtual** spatial positions at  $-w + 2, \dots, 0$  (left boundary) and  $L + 1, \dots, L + w - 1$  (right boundary) to fulfill the degree distributions at the boundaries.
- The code bits of these virtual positions are fixed to “0” and need not be transmitted. As the code bits are “0”, the edges can be removed in a later step

# Spatial Coupling of $L = 3$ Codes ( $w = 2$ )



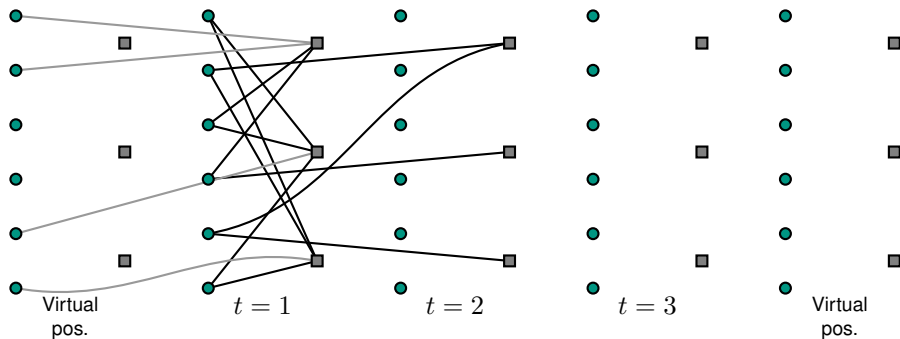
- Starting position, no coupling yet
- Two ( $2 \cdot (w - 1)$ ) virtual positions at boundaries
- Next remove all edges and start coupling for  $t = 1$

## Spatial Coupling of $L = 3$ Codes ( $w = 2$ ) (2)



- As check nodes in  $t = 1$  cannot achieve their degree distribution anymore from variable nodes at  $t = 1$ , we need to fill them using edges from the virtual position

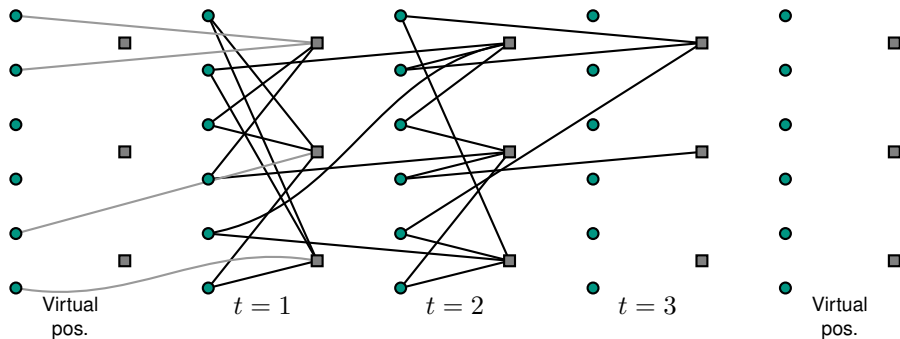
# Spatial Coupling of $L = 3$ Codes ( $w = 2$ ) (3)



- We can now proceed to couple position  $t = 2$ .

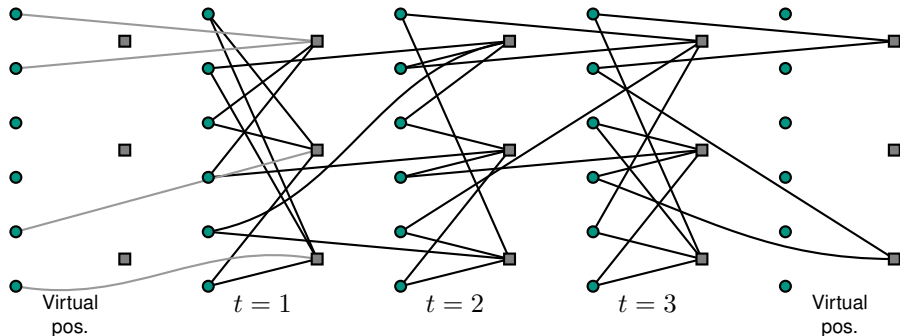


# Spatial Coupling of $L = 3$ Codes ( $w = 2$ ) (4)



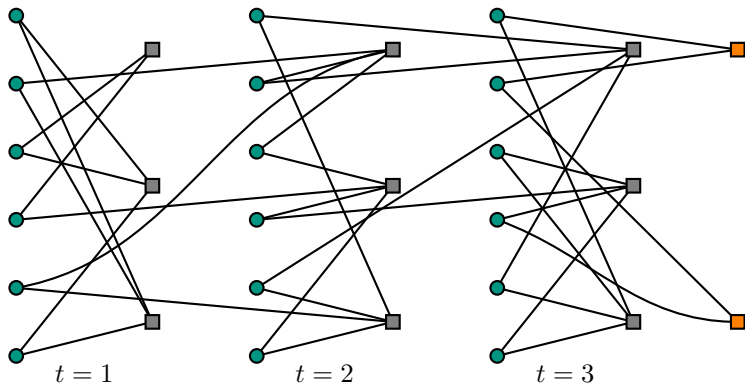
- Finally, we couple the edges of position  $t = 3$ .

# Spatial Coupling of $L = 3$ Codes ( $w = 2$ ) (5)



- We next remove the unconnected nodes and set the variable nodes at the virtual positions to be "0", so we can remove these edges

# Spatial Coupling of $L = 3$ Codes ( $w = 2$ ) (6)



- We have created a new code of  $n' = Ln = 21$  code bits
- Due to coupling we have created *two extra check nodes*
- These extra check nodes decrease the rate of the code

# The Regular SC-LDPC $[d_v, d_c, w, L]$ Code Ensemble

- For simplicity, we consider again regular  $[d_v, d_c]$  LDPC codes and couple them
- We denote the spatially coupled code by the 4-tuple  $[d_v, d_c, w, L]$  where
  - $d_v$  denotes the variable node degree
  - $d_c$  denotes the check node degree
  - $w$  denotes the coupling width
  - $L$  denotes the **replication factor**, i.e., the number of spatial positions

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  - $L$  denotes the **replication factor**, i.e., the number of spatial positions
- Note that the check nodes have only regular degree  $d_c$  at spatial positions  $w, \dots, L$ 
  - check nodes at positions  $1, \dots, w - 1$  have **lower degree**  $\leq d_c$
  - extra check nodes due to right virtual positions also have **lower degree**  $\leq d_c$
- The extra check nodes induce a rate loss

## Theorem (Rate of regular $[d_v, d_c, w, L]$ SC-LDPC codes)

Consider the regular  $[d_v, d_c, w, L]$  SC-LDPC code ensemble and assume  $w < L$ . The design rate of this code ensemble is given by

$$\begin{aligned} r_{d,SC} &= 1 - \frac{d_v}{d_c} - \frac{d_v}{d_c} \frac{w + 1 - 2 \sum_{i=0}^w \left(\frac{i}{w}\right)^{d_c}}{L} \\ &= r_d - \frac{d_v}{d_c} \frac{w + 1 - 2 \sum_{i=0}^w \left(\frac{i}{w}\right)^{d_c}}{L} \end{aligned}$$

where  $r_d := 1 - \frac{d_v}{d_c}$  is the design rate of the regular LDPC code ensemble.

Proof:

- We obtain the result by counting the number of variable nodes  $V$  and the average number of check nodes  $C$ . The rate is obtained as  $r = 1 - \frac{C}{V}$

[KRU11] S. Kudekar, T. Richardson and R. Urbanke, "Threshold Saturation via Spatial Coupling: Why Convolutional LDPC Ensembles Perform So Well over the BEC," *IEEE Trans. Inform. Theory*, vol. 57, no. 2, Feb. 2011

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# Density Evolution Analysis of $[d_v, d_c, w, L]$ SC-LDPC Codes

- We can track the asymptotic behavior using a similar technique as for LDPC codes
- For simplicity, we restrict ourselves to the BEC case
- Contrary to LDPC codes, we need to track  $L$  distinct erasure probabilities  $\xi_t^{(\ell)}$ ,  $t \in \{1, \dots, L\}$ , as each spatial position must be treated differently
- We need to take care of the boundary conditions
- As the code bits at virtual positions are fixed to be “0” and not transmitted, we can fix

$$\xi_t^{(\ell)} = 0 \quad \forall t \in \{-w + 2, -w + 3, \dots, -1, 0, L + 1, L + 2, \dots, L + w - 1\}$$



# BEC Density Evolution of $[d_v, d_c, w, L]$ SC-LDPC Codes

## BEC Density Evolution of $[d_v, d_c, w, L]$ SC-LDPC Codes

- 1 Initialize

$$\xi_t^{(0)} = \begin{cases} 1 & \text{if } t \in \{1, 2, \dots, L\} \\ 0 & \text{otherwise} \end{cases}$$

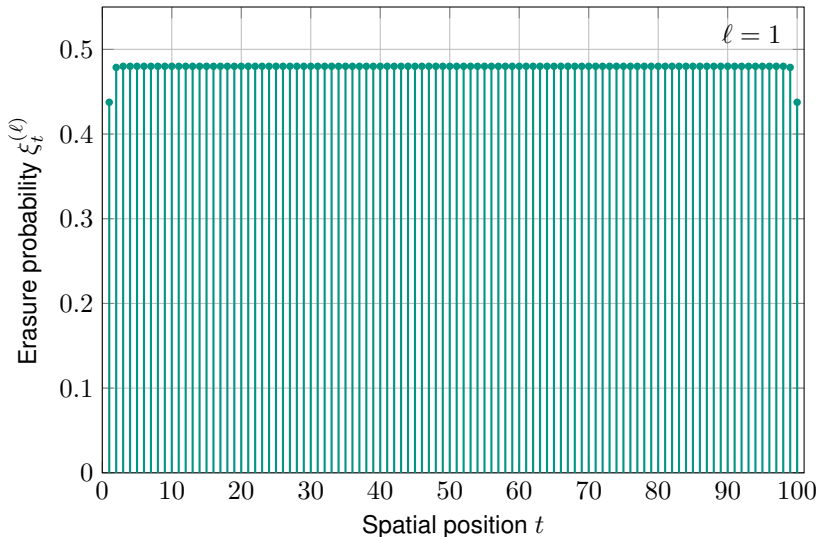
- 2 Update

$$\xi_t^{(\ell)} = \epsilon \left( 1 - \frac{1}{w} \sum_{i=0}^{w-1} \left( 1 - \frac{1}{w} \sum_{j=0}^{w-1} \xi_{t+i-j}^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1} \quad \forall t \in \{1, \dots, L\}$$

- 3 If all  $\xi_t^{(\ell)} = 0, \forall t$ , declare **success** and abort
- 4 If all  $\xi_t^{(\ell)} = \xi_t^{(\ell-1)} > 0, \forall t \in \{1, \dots, L\}$ , we have reached a fixed point. Declare **failure** and abort
- 5 Otherwise, go to step 2)

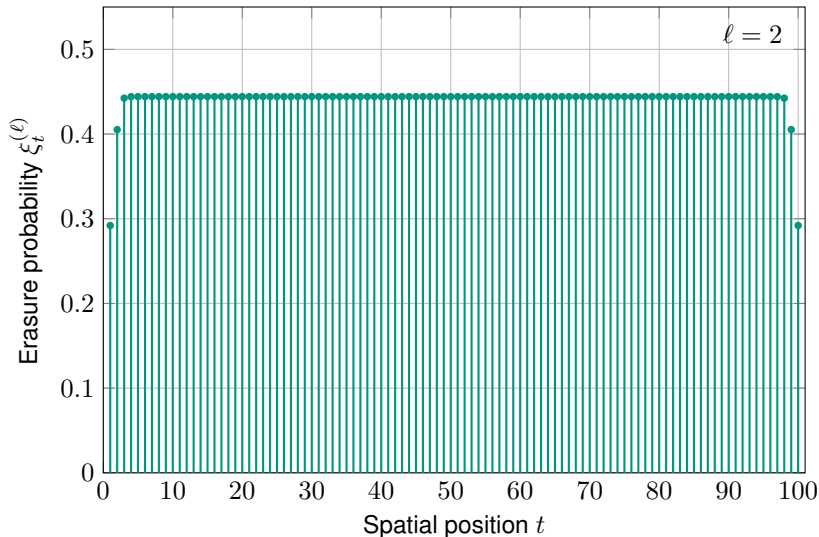
**Example:**  $[d_v = 3, d_c = 6, w = 3, L = 100]$ ,

$\epsilon = 0.48$



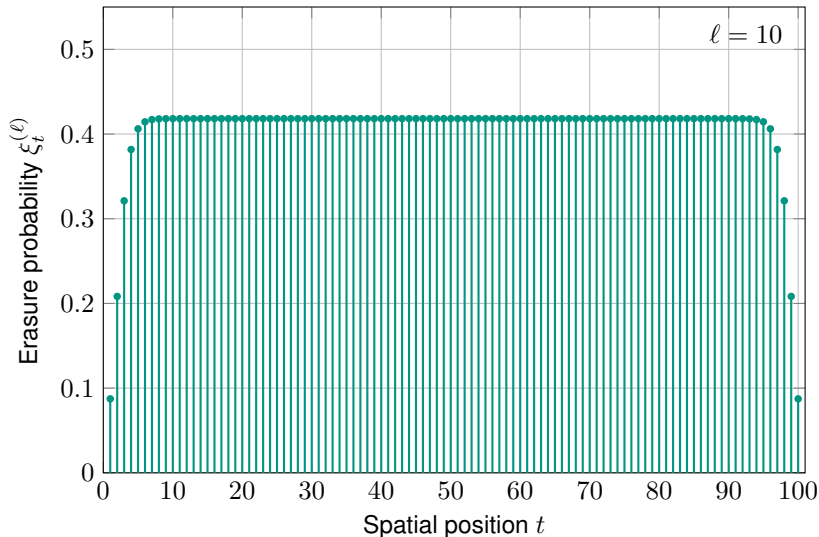
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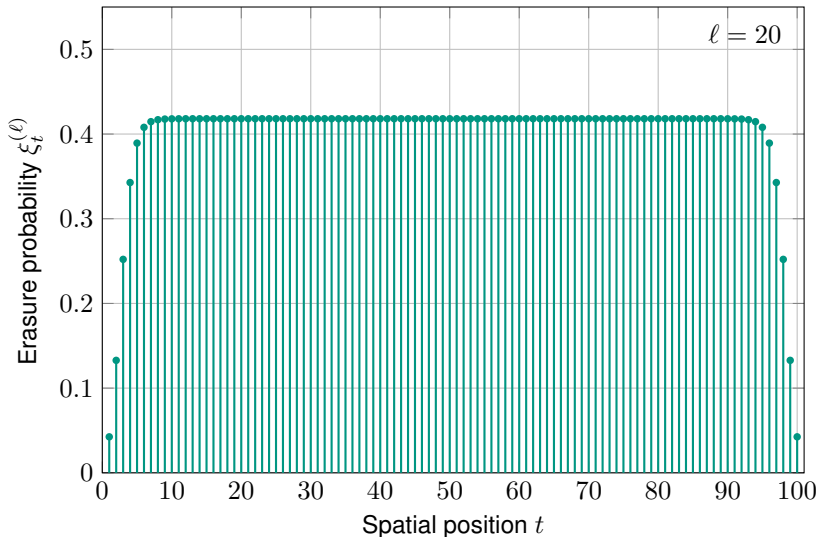
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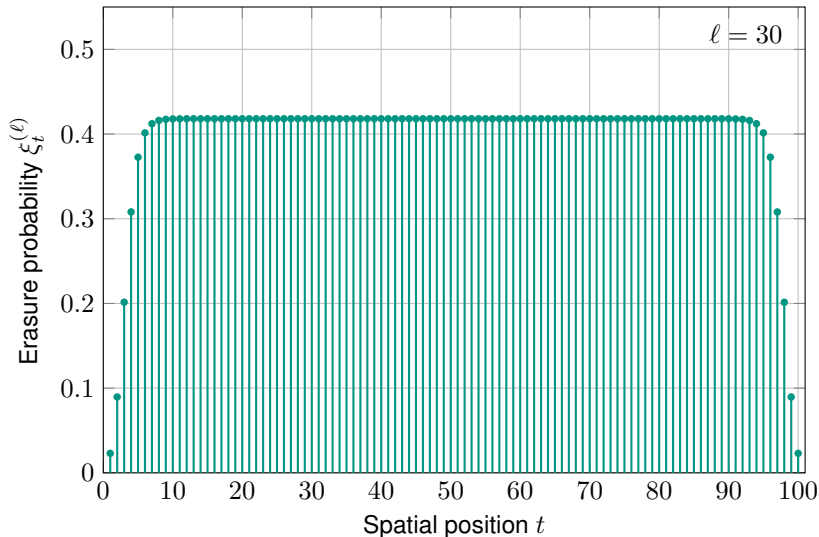
**Example:**  $[d_v = 3, d_c = 6, w = 3, L = 100]$ ,

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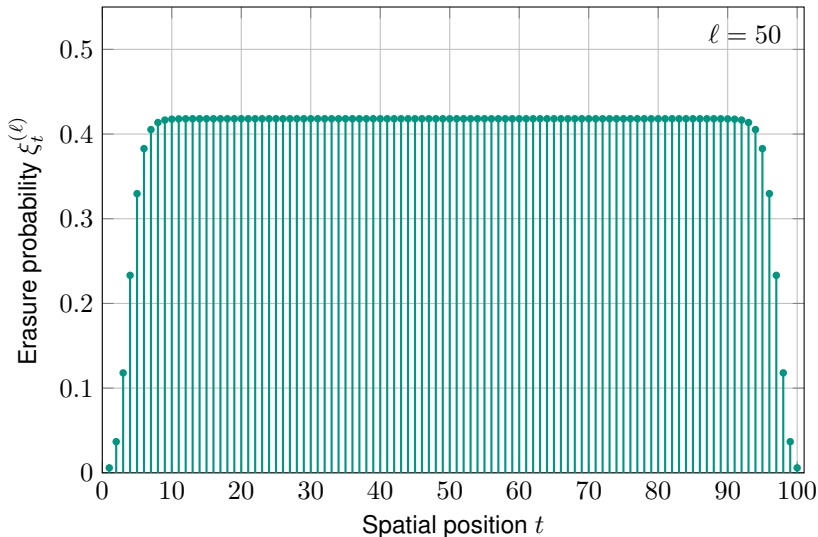
**Example:**  $[d_v = 3, d_c = 6, w = 3, L = 100]$ ,

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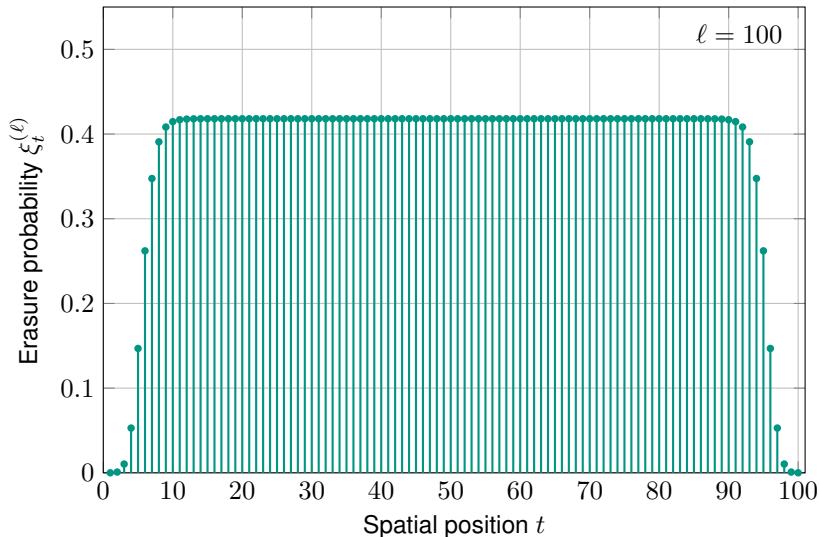
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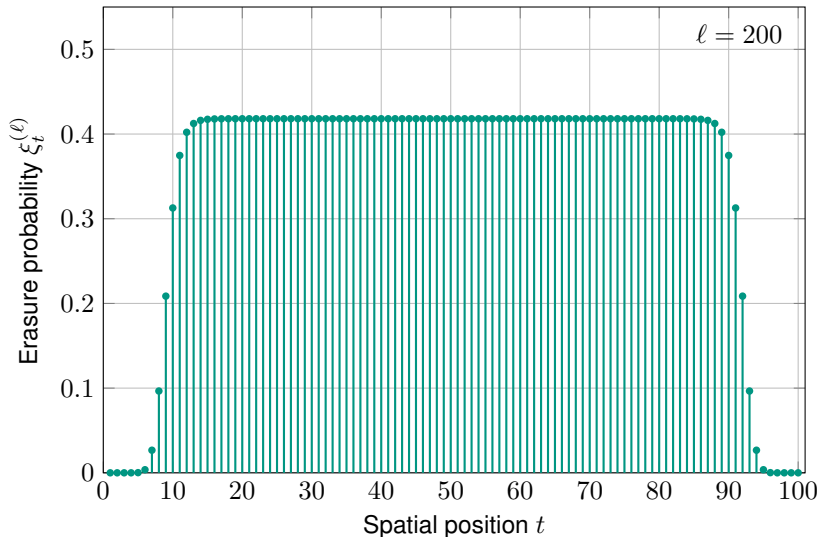
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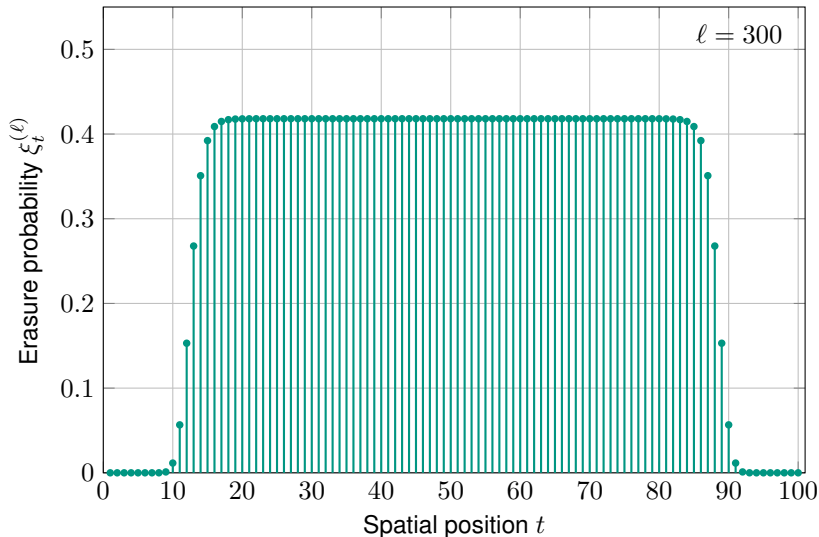
**Example:**  $[d_v = 3, d_c = 6, w = 3, L = 100]$ ,

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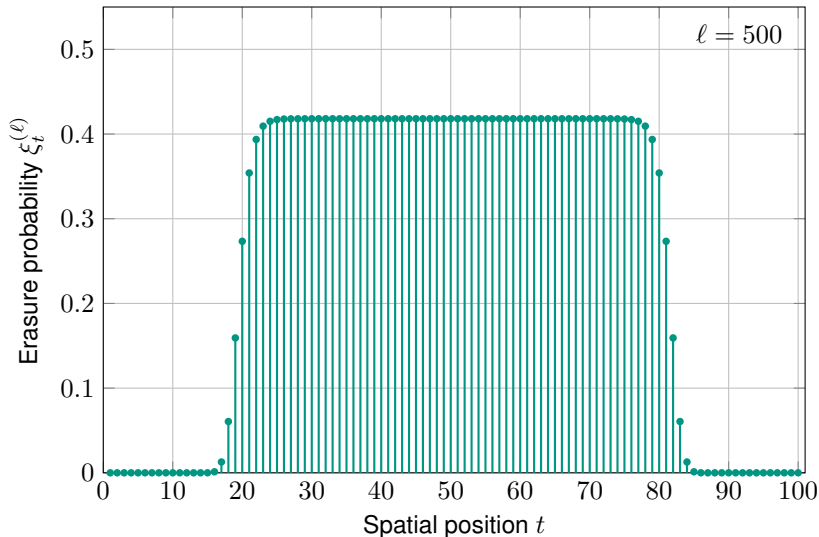
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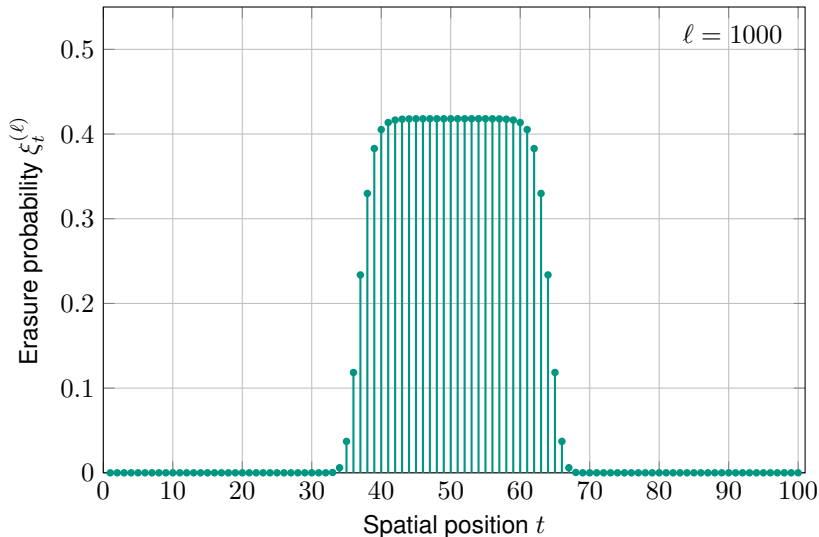
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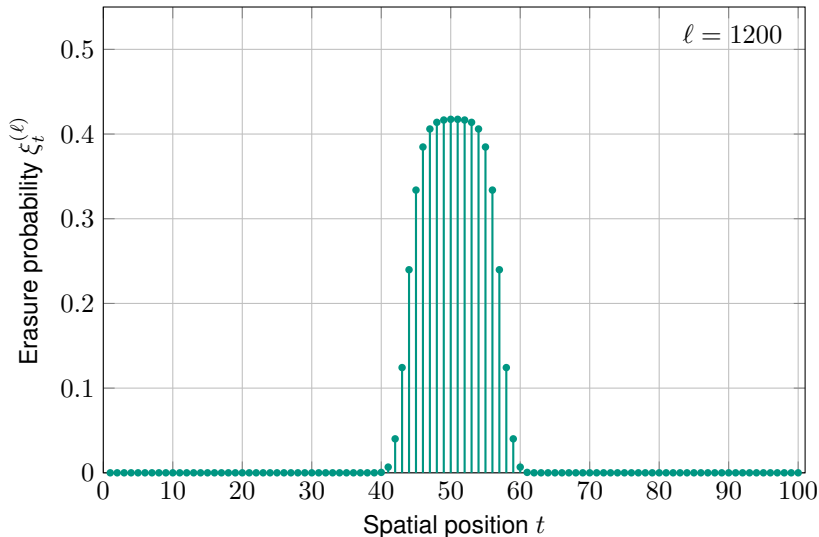
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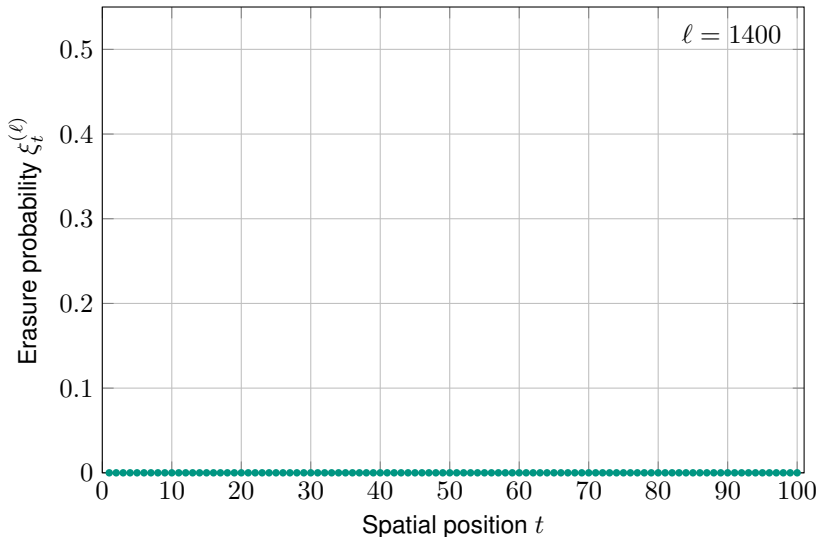
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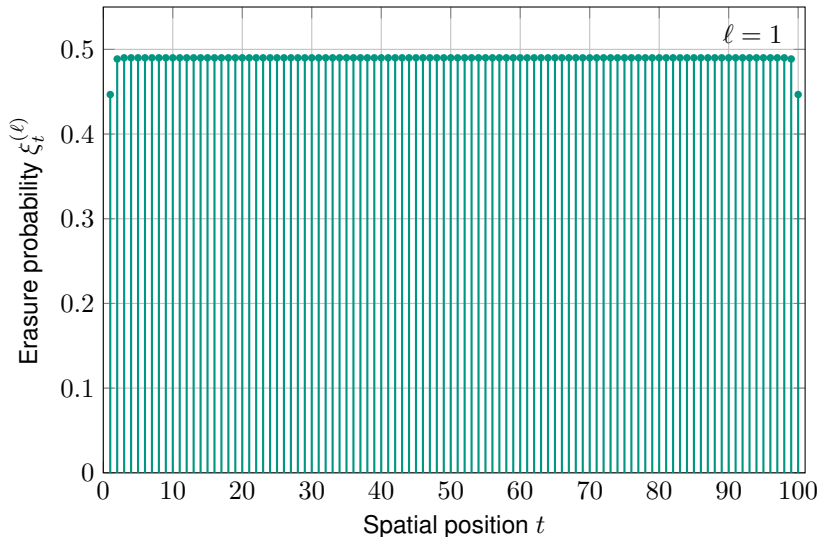
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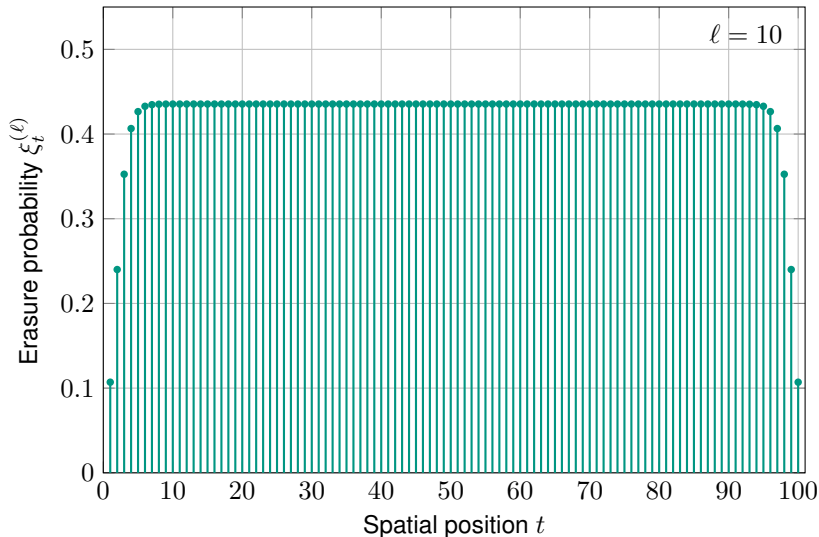






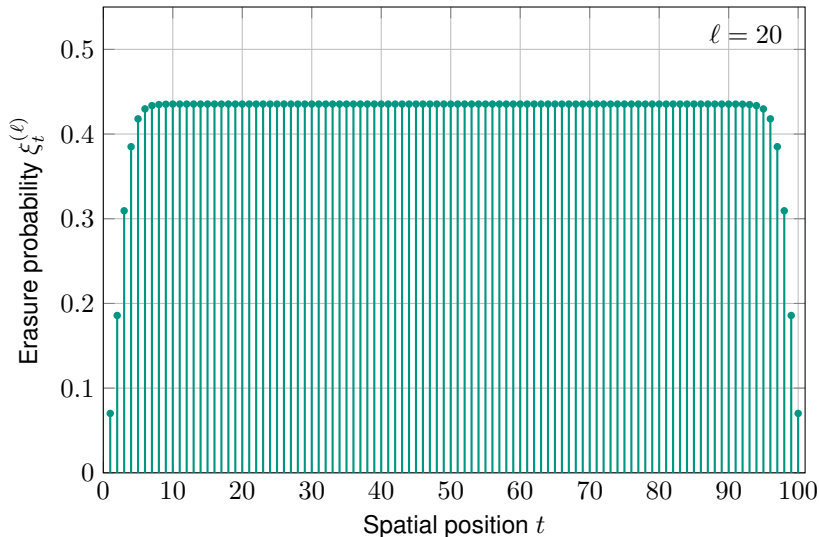
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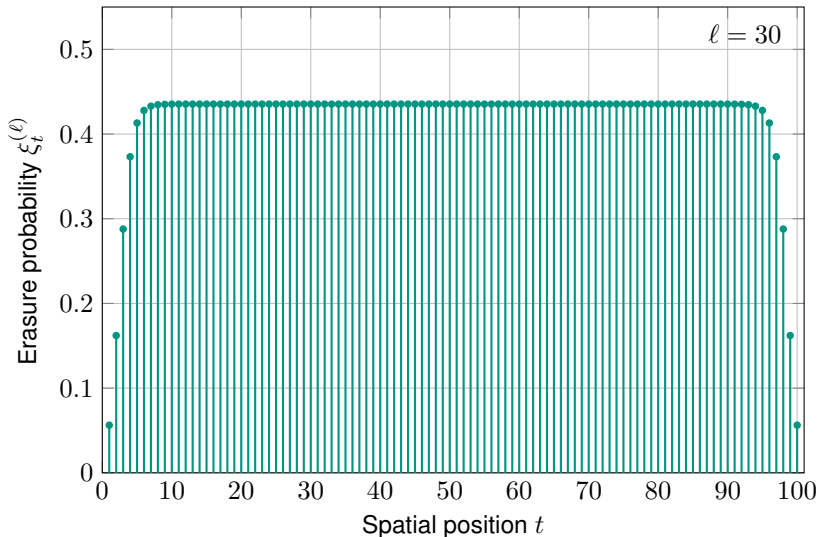
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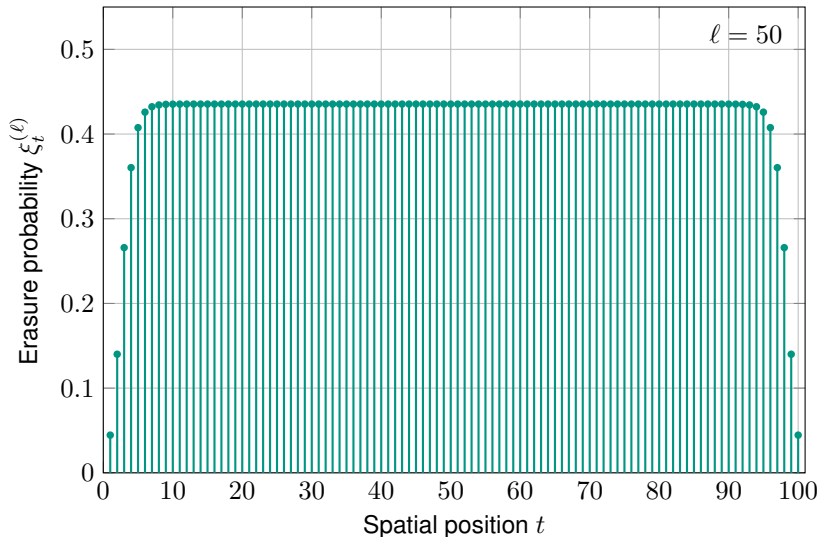
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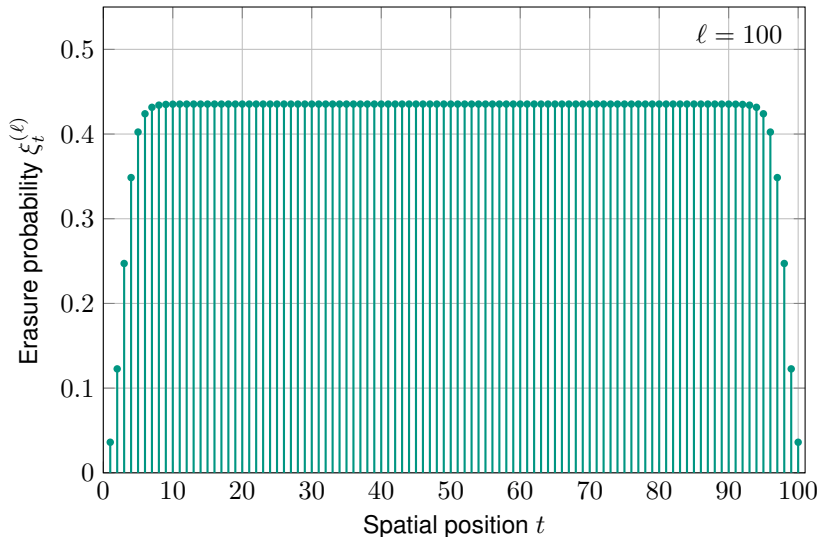
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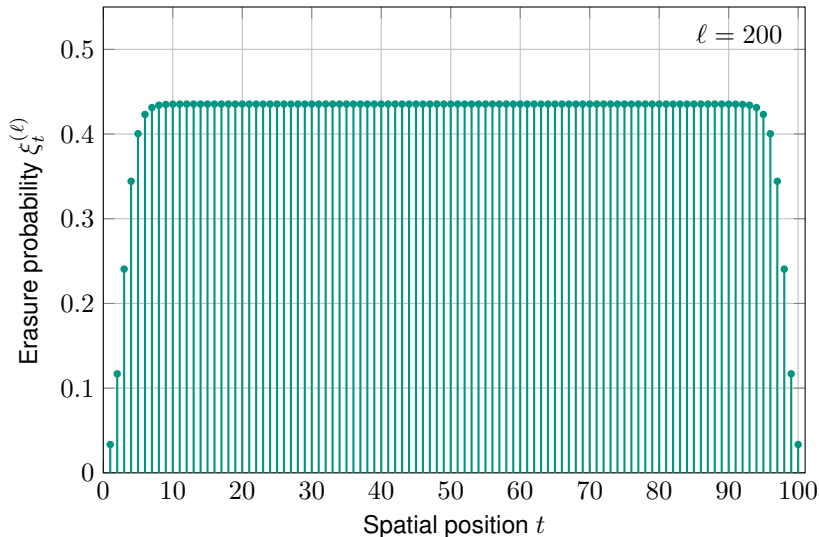
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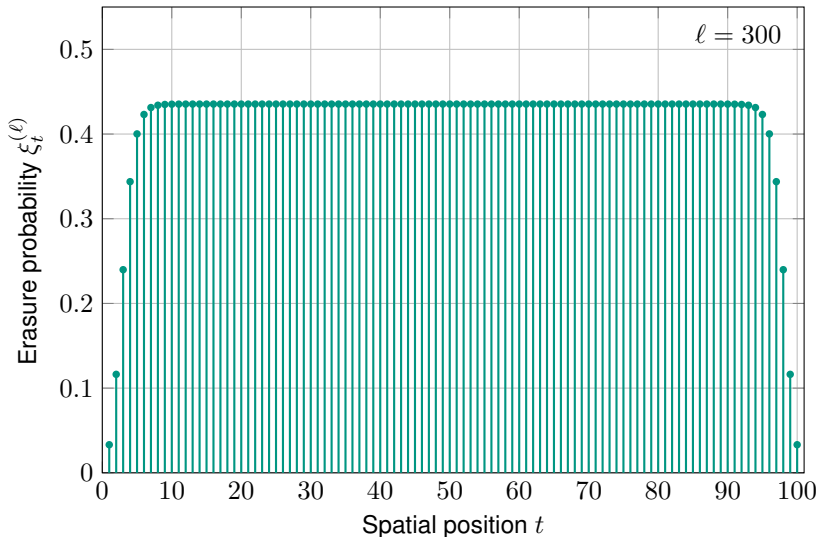
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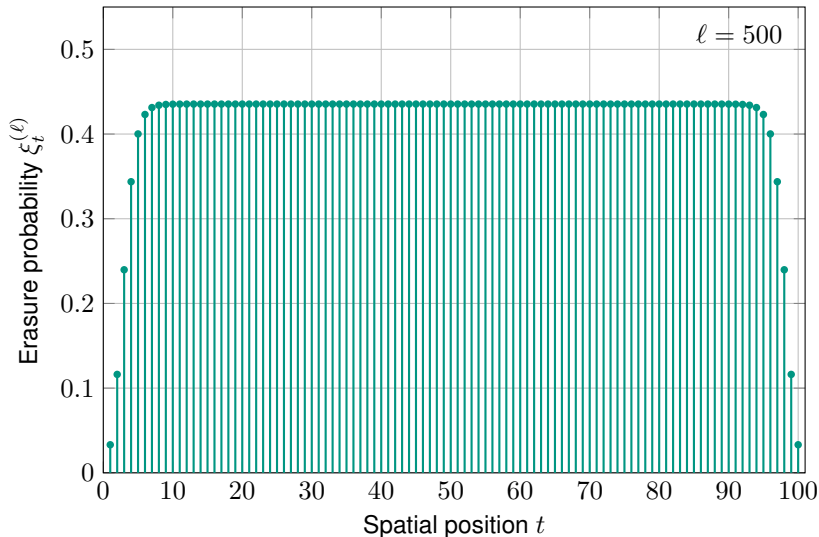
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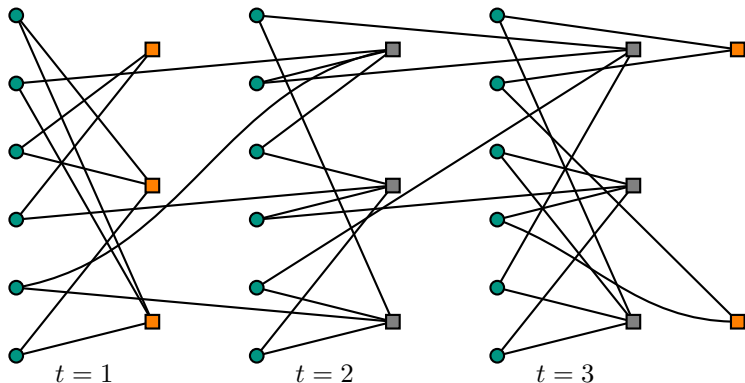


- Similarly to LDPC codes, we define the decoding threshold  $\epsilon_{\text{SC}}^*$  as

$$\epsilon_{\text{SC}}^* = \sup \left\{ \epsilon \in [0, 1] : \xi_t^{(\ell)} \xrightarrow{\ell \rightarrow \infty} 0, \forall t \in \{1, \dots, L\}, \right. \\ \left. \xi_t^{(0)} = 1 \forall t \in \{1, \dots, L\}, \text{ and } \xi_t^{(0)} = 0 \forall t \notin \{1, \dots, L\} \right\}$$

- Recall that the threshold of the  $[3, 6]$  regular LDPC code ensemble was 0.4292
- Spatial coupling dramatically improves the threshold
- The key are slightly different check node degrees at the boundaries
- Nucleation effect, also observed in physics (e.g., supercooled water)
- Video ...

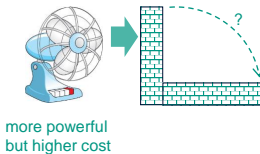
# Boundary Effects of Spatial Coupling: Example














- Check nodes at boundaries have lower degrees than other check nodes
- Lower degree check nodes decrease the possibility of an erasure
- Better protection of bits at boundaries

# Triggering the Decoding Wave

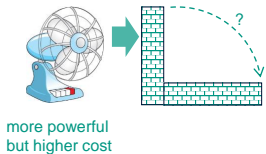
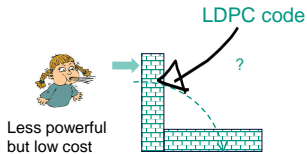
## ■ A Toy Example














Wall width			
			
			

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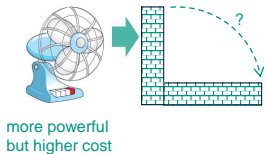
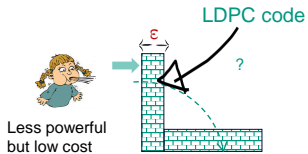
## ■ A Toy Example














Wall width			
			
			

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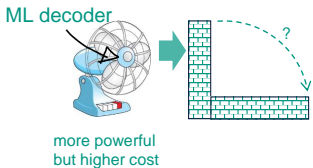
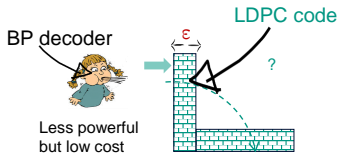
## ■ A Toy Example














Wall width			
			
			

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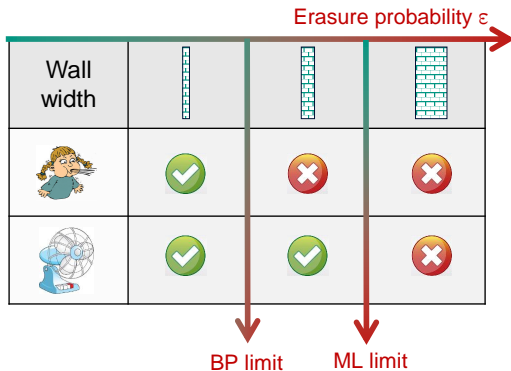
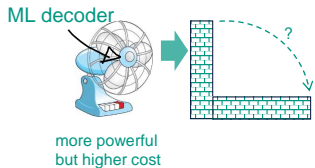
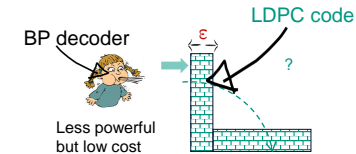
## ■ A Toy Example



Wall width			
			
			

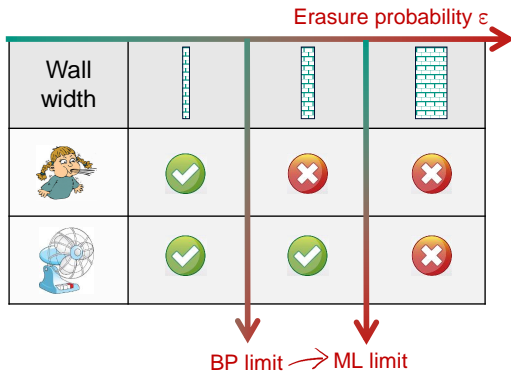
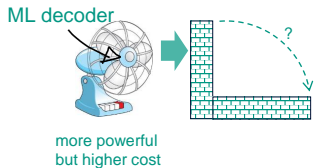
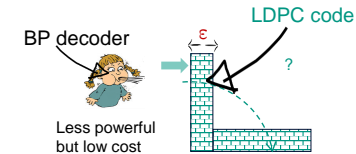
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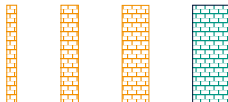
## ■ A Toy Example





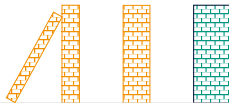
# Triggering the Decoding Wave

- Triggering similar to dominoes



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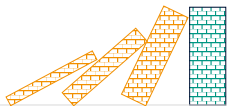
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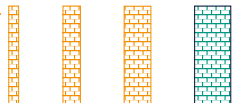
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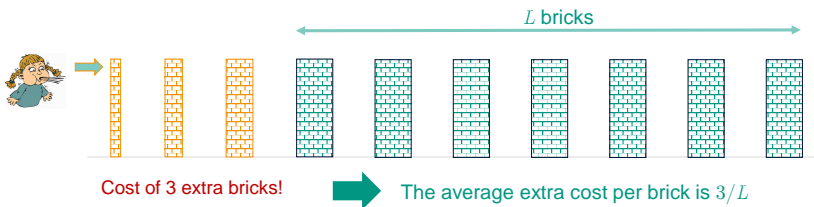
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Cost of 3 extra bricks!

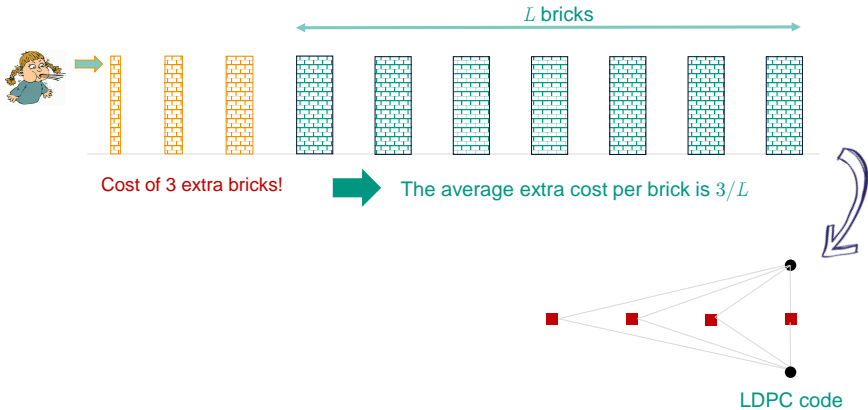
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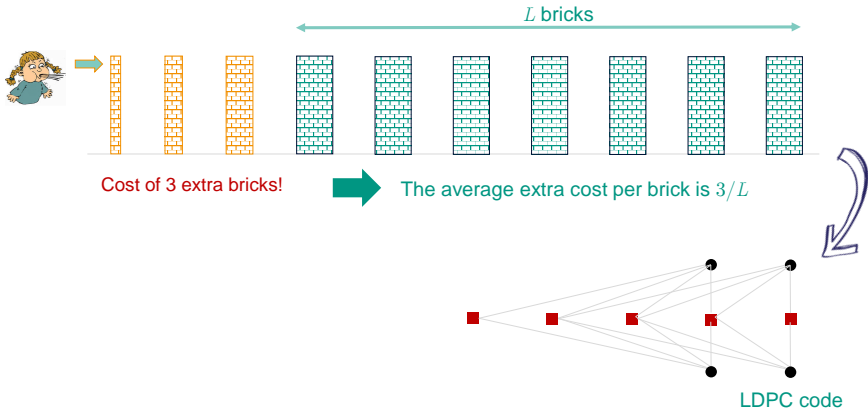
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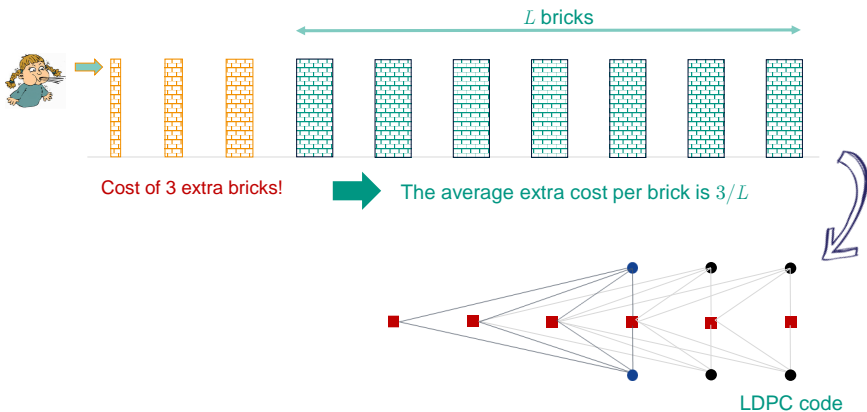
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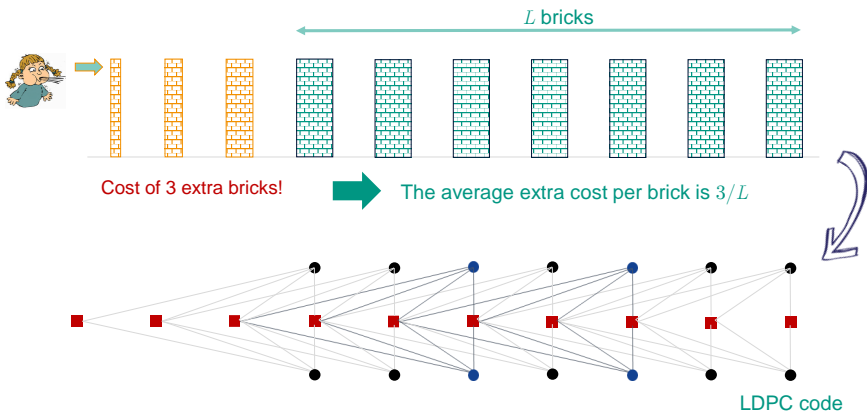
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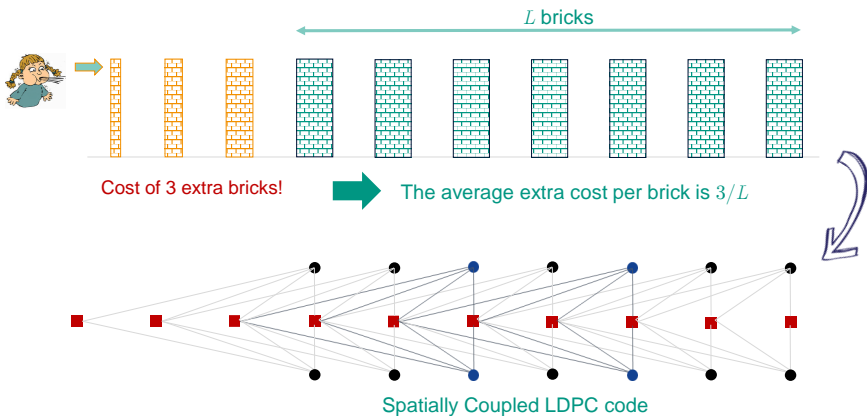
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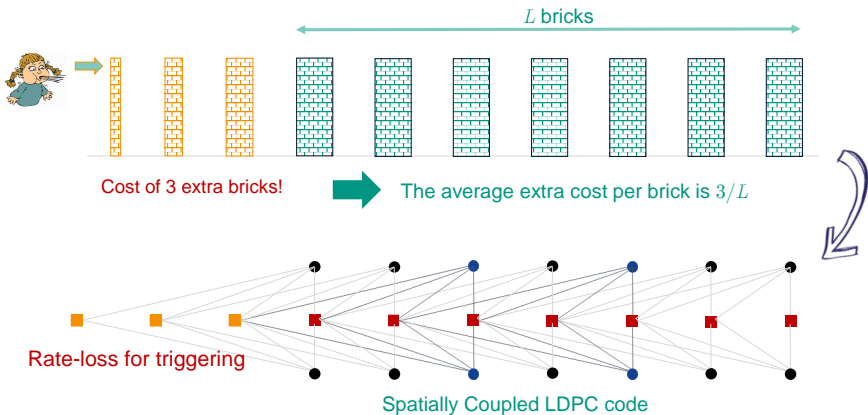
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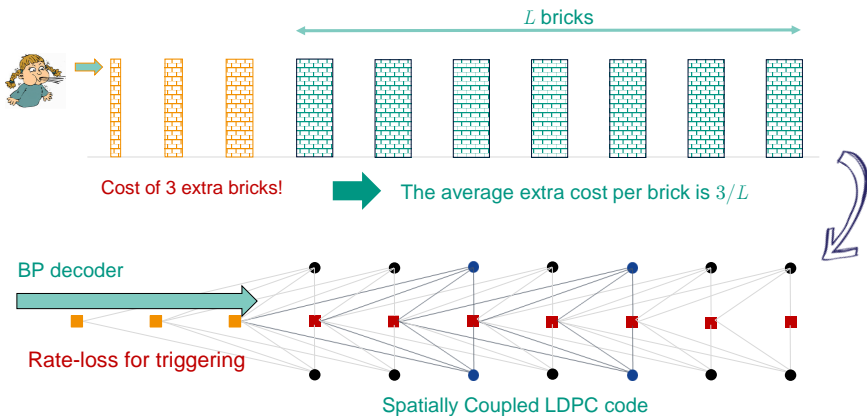
# Triggering the Decoding Wave

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# Triggering the Decoding Wave

- Triggering similar to dominoes



- By running density evolution, we can compute the thresholds

$d_v$	$d_c$	$w$	$L$	$r_{d,SC}$	$r_d$	$\epsilon_{SC}^*$	$\epsilon^*$
3	6	2	100	0.49516	0.5	0.48837	0.42944
3	6	3	100	0.49089	0.5	0.48833	0.42944
4	8	2	100	0.49504	0.5	0.49446	0.38344
4	8	3	100	0.49039	0.5	0.49788	0.38344
5	10	2	100	0.49501	0.5	0.48268	0.34154
5	10	3	100	0.49017	0.5	0.49895	0.34154
5	10	4	100	0.48557	0.5	0.49971	0.34154

- Thresholds of SC-LDPC codes are very close to capacity limit  $1 - r_{d,SC}$
- Note that  $r_{d,SC}$  can be made arbitrarily close to  $r_d$  by increasing  $L$
- Note that  $\epsilon^*$  decreases with increasing  $d_v$  while  $\epsilon_{SC}^*$  increases!

# Threshold Saturation of Spatially Coupled LDPC Codes

- It has been recently shown [KRU11]

$$\lim_{w \rightarrow \infty} \lim_{L \rightarrow \infty} r_{d, \text{SC}} = 1 - \frac{d_v}{d_c}$$
$$\lim_{w \rightarrow \infty} \lim_{L \rightarrow \infty} \epsilon_{\text{SC}}^* = \lim_{w \rightarrow \infty} \lim_{L \rightarrow \infty} \epsilon_{\text{SC}}^{\text{ML}} = \epsilon^{\text{ML}}$$

- This means that in the limit of  $L$  and  $w$  (in that order!) the threshold **saturates** to the ML performance of the spatially coupled LDPC code
- The ML performance of the spatially coupled code equals (in the limit of  $L$  and  $w$ ) the ML performance of the uncoupled LDPC code!



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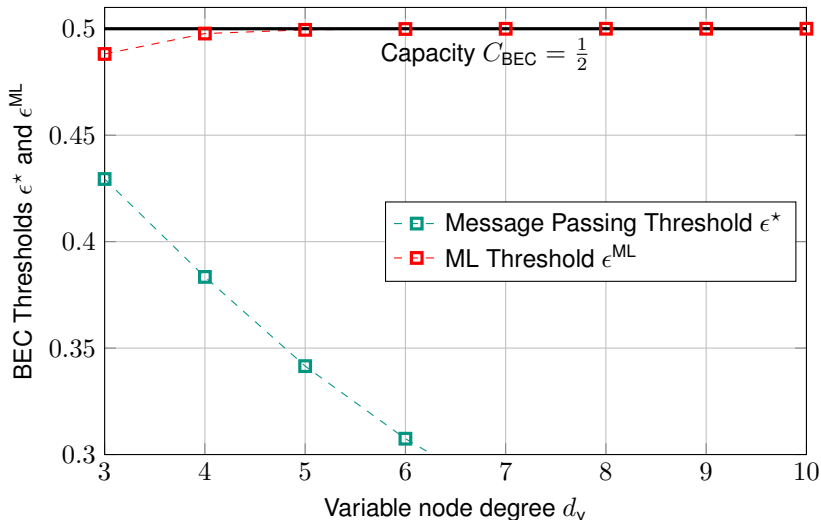
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- This means that in the limit of  $L$  and  $w$  (in that order!) the threshold **saturates** to the ML performance of the spatially coupled LDPC code
- The ML performance of the spatially coupled code equals (in the limit of  $L$  and  $w$ ) the ML performance of the uncoupled LDPC code!
- This **threshold saturation** allows us to
  - Take a block LDPC code which we know has good ML performance
  - Apply spatial coupling with large enough  $w$  and  $L$
  - With simple message passing decoding, ML performance can be approached

[KRU11] S. Kudekar, T. Richardson and R. Urbanke, "Threshold Saturation via Spatial Coupling: Why Convolutional LDPC Ensembles Perform So Well over the BEC," IEEE Trans. Inform. Theory, vol. 57, no. 2, Feb. 2011

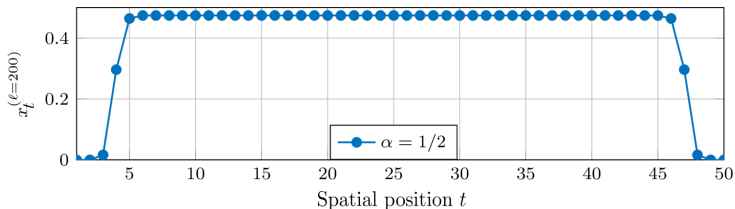
# ML Performance of Regular $[d_v, d_c]$ LDPC Codes



- 1 Introduction: Channel Coding and LDPC Codes
- 2 **Spatially Coupled LDPC Codes**
  - Motivation and Definition
  - Performance of SC-LDPC Codes
  - **Practical Implementation of SC-LDPC Codes**
  - Improvement of SC-LDPC Codes by Non-Uniform Coupling
- 3 Burst Correction Capabilities of Spatially Coupled LDPC Codes
  - Error Probability after Burst Erasures
  - Application Example
- 4 Conclusions

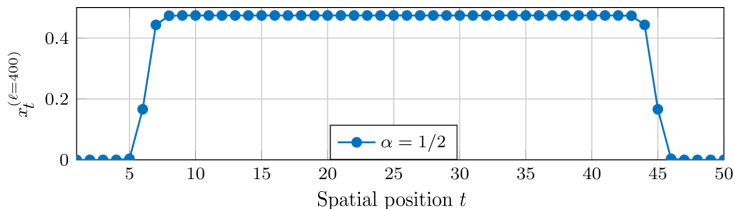
# Density Evolution ( $\varepsilon = 0.48, L = 50$ )

- Spatially coupled LDPC Code [ $d_v = 5, d_c = 10, w = 2, L = 50$ ]
- Performance after decoding iterations



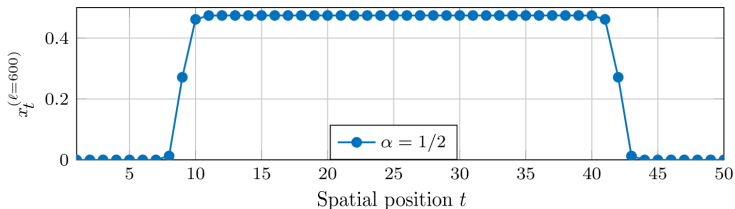
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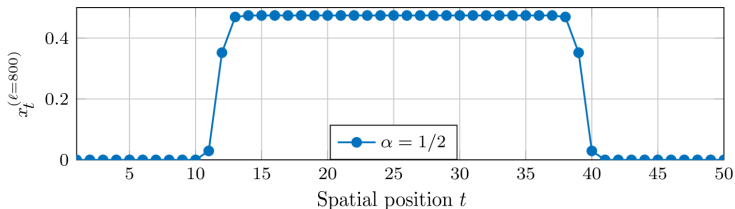
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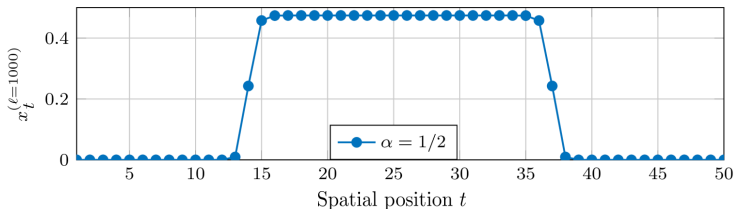
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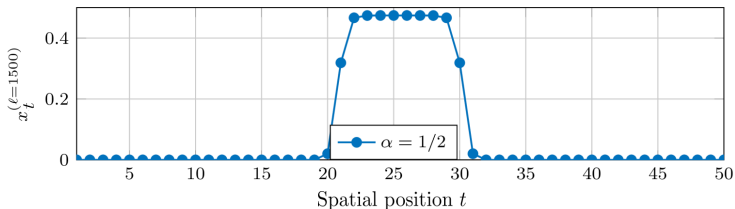
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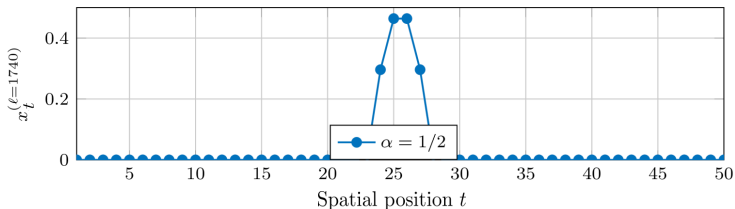
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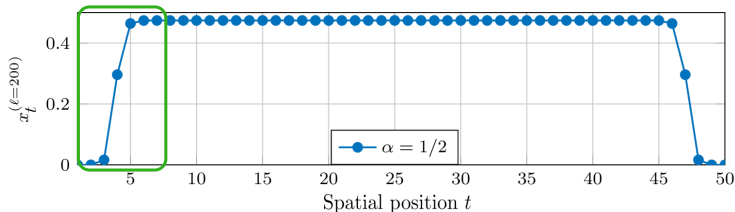


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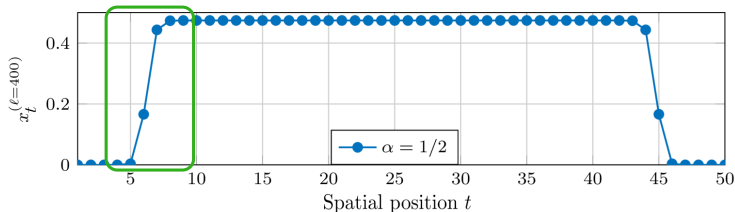
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- Windowed decoding sufficient to achieve capacity [ISU+13]
- To save latency, we are only interested in left-most portion of wave and use windowed decoder of size  $W_D$  for this part (decode while receive)
- Window latency of order  $W_D + w$  ( $W_D + w - 1$  SPs in window)
- Decoding complexity of order  $(W_D + w) \cdot I$  ( $I$ : number iterations per window)

[ISU+13] A. Iyengar, P. Siegel, R. Urbanke, J. Wolf, "Windowed decoding of spatially coupled codes," *IEEE Trans. Inf. Theory*, 2013

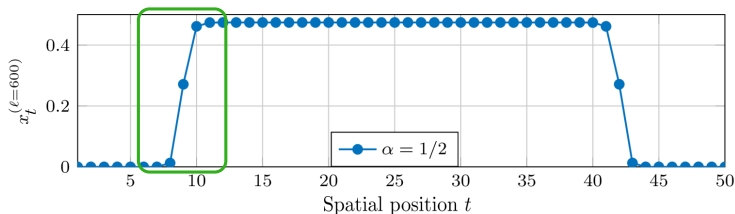
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- Decoding complexity of order  $(W_D + w) \cdot I$  ( $I$ : number iterations per window)

[ISU+13] A. Iyengar, P. Siegel, R. Urbanke, J. Wolf, "Windowed decoding of spatially coupled codes," *IEEE Trans. Inf. Theory*, 2013

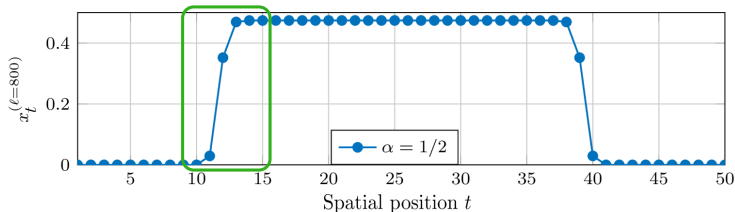
- Spatially coupled LDPC Code [ $d_v = 5$ ,  $d_c = 10$ ,  $w = 2$ ,  $L = 50$ ]
- Performance after decoding iterations



- Windowed decoding sufficient to achieve capacity [ISU+13]
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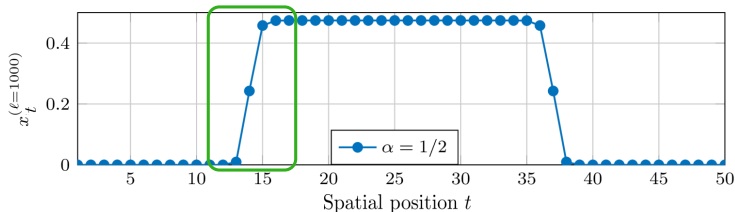
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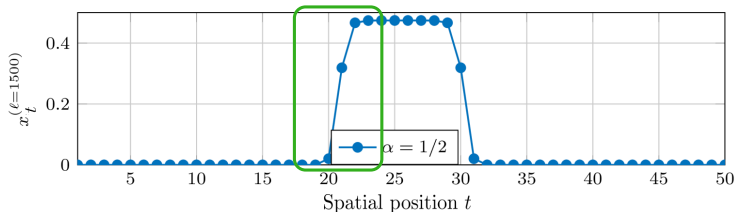
- Spatially coupled LDPC Code [ $d_v = 5$ ,  $d_c = 10$ ,  $w = 2$ ,  $L = 50$ ]
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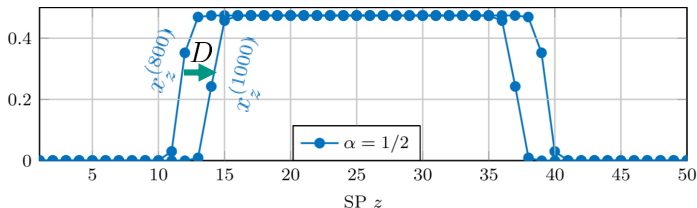
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- Performance after decoding iterations



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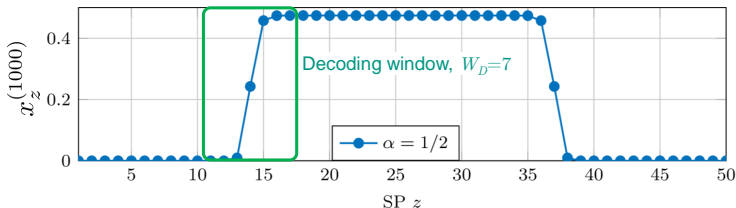
- Decoding velocity as displacement of erasure profile per decoding iteration [ASTB13], [EM16]
- Decoding velocity  $v$  defined as  $D/I$ , where  $I$  is the number of iterations required to advance the profile by  $D$ , i.e., here  $v = D/200$

[ASTB13] V. Aref, L. Schmalen, S. ten Brink, "On the convergence speed of spatially coupled LDPC ensembles," *Proc. Allerton Conf.*, 2013

[EM16] R. E-Khatib, N. Macris, "The velocity of the decoding wave for spatially coupled codes on BMS channels," *Proc. ISIT*, 2016

[ISU+13] A. Iyengar, P. Siegel, R. Urbanke, J. Wolf, "Windowed decoding of spatially coupled codes," *IEEE Trans. Inf. Theory*, 2013

# Decoding Velocity and Windowed Decoding



- Windowed decoding only carries out decoding operations on  $W_D$  spatial positions that benefit from decoding [ISU+13]
- Complexity of windowed decoding directly linked to the velocity of the profile

[AS1B13] V. Aref, L. Schmalen, S. ten Brink, "On the convergence speed of spatially coupled LDPC ensembles," *Proc. Allerton Conf.*, 2013

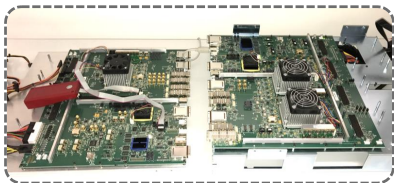
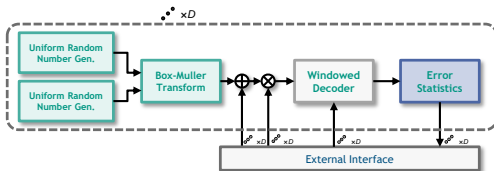
[EM16] R. E-Khatib, N. Macris, "The velocity of the decoding wave for spatially coupled codes on BMS channels," *Proc. ISIT*, 2016

[ISU+13] A. Iyengar, P. Siegel, R. Urbanke, J. Wolf, "Windowed decoding of spatially coupled codes," *IEEE Trans. Inf. Theory*, 2013

- Use your favorite LDPC decoding algorithm
- Only operate on **parts of the Tanner graph** corresponding to  $W_D + w - 1$  spatial positions (the **window**)
- After carrying out  $I$  iterations on the window, shift the window by 1 spatial position
  - **Important:** Keep all the edge messages, do not re-initialize messages
- Virtual positions assumed to be received without error (shortened)

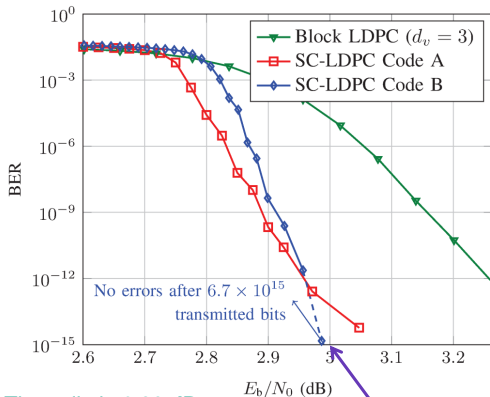
# FPGA-Based Code Evaluation

- High throughputs & large coding gains necessary in optical communication networks & submarine cables
- Required BER: less than 0.00000000000001% ( $10^{-15}$ )
- Less than 10 bit errors per day at line rate of 100 Gbit/s
- Requirements might become more strict in the future (now 1 Tbit/s)
  
- Simulation of codes using FPGA-platform with windowed decoder



# Result of FPGA Decoding Platform

## Codes of rate $r = 0.8$

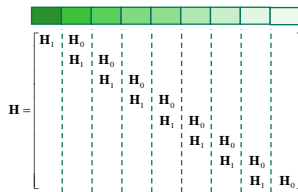


Theor. limit: **2.06 dB**  
(asymptotic)

Net coding gain **12.01 dB**

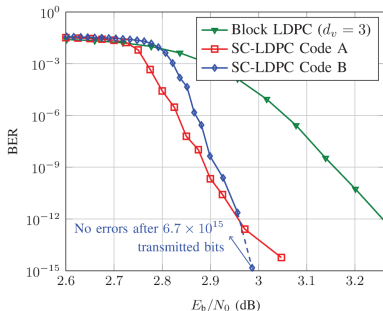
## Comparison of two different codes

- Code A: optimized velocity
  - Code B: optimized error floor
- Dual engine decoder, extra iteration without extra latency [SSR+15]



[SSR+15] L. Schmalen, D. Suikat, D. Rösener, V. Aref, A. Leven, S. ten Brink "Spatially coupled codes and optical fiber communications: An ideal match?," Proc. Workshop on Signal Processing Advances in Wireless Communications (SPAWC), 2015

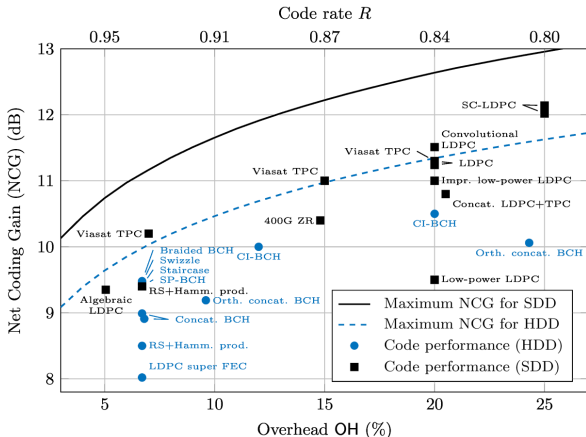
# Result of FPGA Decoding Platform



- **0.4dB** gain correspond to **900km reach increase** in trans-pacific cables
- Optical fiber communication systems **age** (material, lasers, photodiodes) and the **SNR will decay over time**
- In this case, additional gains increase lifetime/reduce margins of a system
- **More gains** are possible with **higher decoding complexity**
- However, **we want even more gains!**

[SSR+15] L. Schmalen, D. Suikat, D. Rösener, V. Aref, A. Leven, S. ten Brink "Spatially coupled codes and optical fiber communications: An ideal match?," *Proc. Workshop on Signal Processing Advances in Wireless Communications (SPAWC)*, 2015

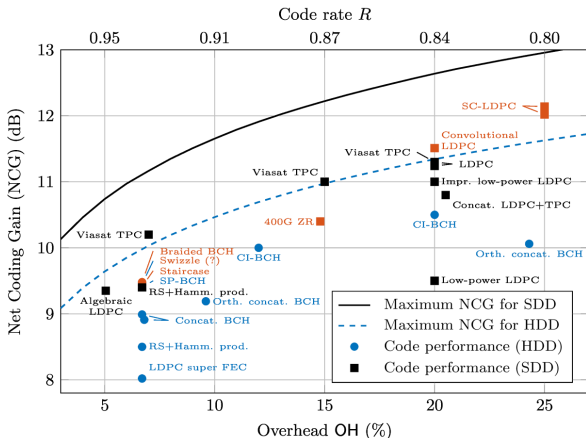
# Comparison of Coding Schemes used in Optical Communications



- State-of-the-art coding schemes proposed for practical implementation
- Performance verified or reasonably estimated at  $10^{-15}$  BER

[GS20] A. Graell i Amat and L. Schmalen, "Forward Error Correction for Optical Transponders", *Springer Handbook of Optical Networks*, B. Mukherjee, I. Tomkos, M. Tomatore, P. Winzer, and Y. Zhao (editors), Springer, October 2020

# Comparison of Coding Schemes used in Optical Communications



- State-of-the-art coding schemes proposed for practical implementation
- Performance verified or reasonably estimated at  $10^{-15}$  BER
- The best performing schemes are **spatially coupled codes**



- 1 Introduction: Channel Coding and LDPC Codes
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  - Performance of SC-LDPC Codes
  - Practical Implementation of SC-LDPC Codes
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- 4 Conclusions

- By running density evolution, we can compute the thresholds

$d_v$	$d_c$	$w$	$L$	$r_{d,SC}$	$r_d$	$\epsilon_{SC}^*$	$\epsilon^*$
3	6	2	100	0.49516	0.5	0.48837	0.42944
3	6	3	100	0.49089	0.5	0.48833	0.42944
4	8	2	100	0.49504	0.5	0.49446	0.38344
4	8	3	100	0.49039	0.5	0.49788	0.38344
5	10	2	100	0.49501	0.5	0.48268	0.34154
5	10	3	100	0.49017	0.5	0.49895	0.34154
5	10	4	100	0.48557	0.5	0.49971	0.34154

- We see that there is **still a gap to capacity**

If you do not get what you expect...

Find ways to generalize the construction!

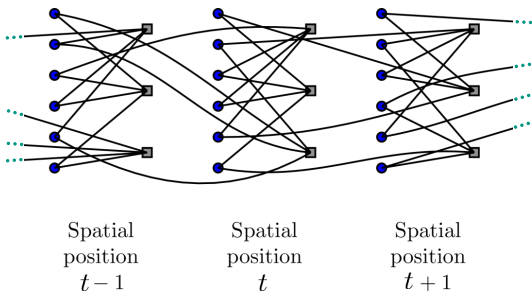
# New: Non-Uniformly Coupled LDPC Codes

- Start with SC-LDPC Code ensemble with  $w = 2$
- We fixed: probability of coupling an edge with neighboring position ( $1/w$ )
- **New: generalization of coupling**

[SAJ17] L. Schmalen, V. Aref, F. Jardel, "Non-Uniformly Coupled LDPC Codes: Better Thresholds, Smaller Rate-loss, and Less Complexity," *Proc. ISIT*, 2017  
[SA19] L. Schmalen and V. Aref, "Spatially Coupled LDPC Codes with Non-uniform Coupling for Improved Decoding Speed," *Proc. ITW*, 2019, available online at <https://arxiv.org/abs/1904.07026>

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## Definition

Connect each edge from variable node at SP  $z$  to

- check node at position  $z$  with **probability  $a$**  and to
- Check node at position  $z + 1$  with **probability  $1 - a$**

[SAJ17] L. Schmalen, V. Aref, F. Jardel, "Non-Uniformly Coupled LDPC Codes: Better Thresholds, Smaller Rate-loss, and Less Complexity," *Proc. ISIT*, 2017  
[SA19] L. Schmalen and V. Aref, "Spatially Coupled LDPC Codes with Non-uniform Coupling for Improved Decoding Speed," *Proc. ITW*, 2019, available online at <https://arxiv.org/abs/1904.07026>

- Let  $\boldsymbol{\nu} = (\nu_0, \dots, \nu_{w-1})$  denote the coupling vector with  $\sum_{i=0}^{w-1} \nu_i = 1$
- In uniform coupling, we have  $\boldsymbol{\nu} = (\frac{1}{w}, \dots, \frac{1}{w})$
- For  $w = 2$ , we have  $\boldsymbol{\nu} = (\alpha, 1 - \alpha)$

## Density Evolution of Non-Uniformly Coupled Codes

The DE update equation for a  $[d_v, d_c, \boldsymbol{\nu}, L]$  SC-LDPC code ensemble becomes

$$\xi_t^{(\ell)} = \epsilon \left( 1 - \sum_{i=0}^{w-1} \nu_i \left( 1 - \sum_{j=0}^{w-1} \nu_j \xi_{t+i-j}^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1}$$

- Free parameters  $\nu_i$  can be optimized to yield good thresholds

- Non-uniform coupling has an impact on the rate

## Rate of Non-Uniformly Coupled Codes

The rate of a non-uniformly coupled  $[d_v, d_c, \nu, L]$  SC-LDPC code ensemble amounts  $r_{d,SC} = \left(1 - \frac{d_v}{d_c}\right) - \frac{1}{L}\Delta$ , where

$$\Delta = \frac{d_v}{d_c} \left( w - 1 - \sum_{k=0}^{w-2} \left[ \binom{k}{\sum_{i=0}^k \nu_i}^{d_c} + \binom{w-1}{\sum_{i=k+1}^{w-1} \nu_i}^{d_c} \right] \right)$$

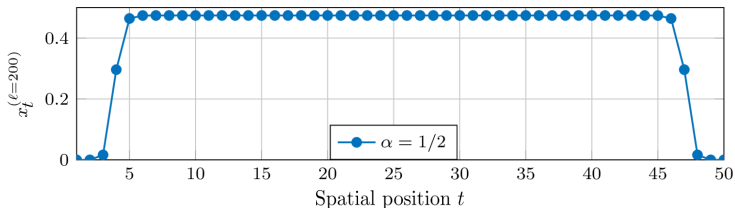
- Interestingly, for  $w = 2$ ,  $\Delta$  becomes maximal for  $\alpha = \nu_0 = \nu_1 = \frac{1}{2}$ , i.e., for  $w = 2$ , non-uniform coupling reduces the rate loss!
- For general  $w > 3$ , non-uniform coupling may also lead to larger rate losses

# Density Evolution ( $\varepsilon = 0.48, L = 50$ )

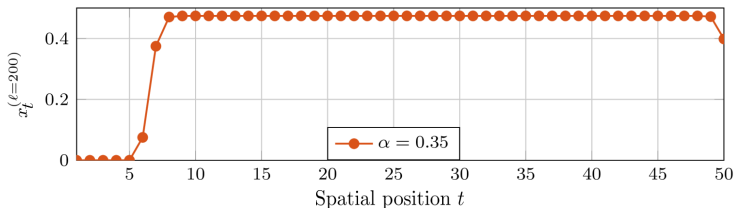
Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

$I = 200$  iter.



**New:** Non-uniformly  
coupled code  
with  
 $d_v = 5, d_c = 10$   
**Single-side  
convergence**

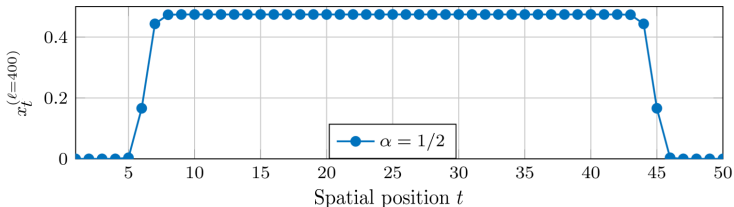


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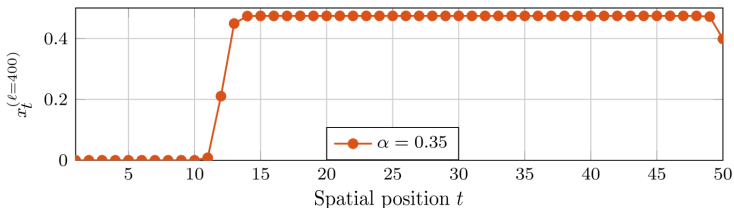
Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

$I = 400$  iter.



**New:** Non-  
uniformly  
coupled code  
with  
 $d_v = 5, d_c = 10$   
**Single-side  
convergence**



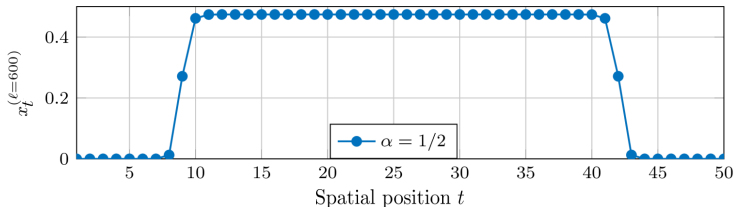


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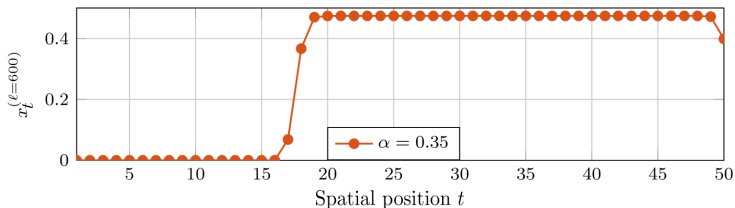
Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

$I = 600$  iter.



**New:** Non-  
uniformly  
coupled code  
with  
 $d_v = 5, d_c = 10$   
**Single-side  
convergence**

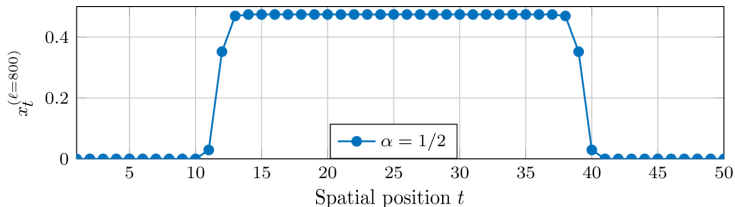


# Density Evolution ( $\varepsilon = 0.48, L = 50$ )

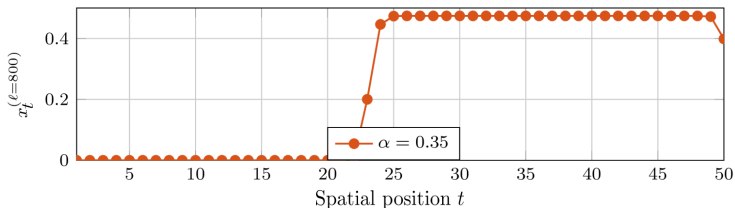
Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

$I = 800$  iter.



**New:** Non-uniformly  
coupled code  
with  
 $d_v = 5, d_c = 10$   
**Single-side  
convergence**

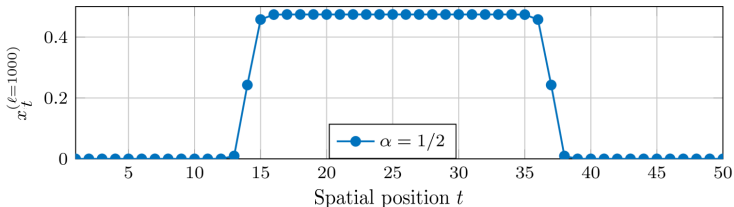


# Density Evolution ( $\varepsilon = 0.48, L = 50$ )

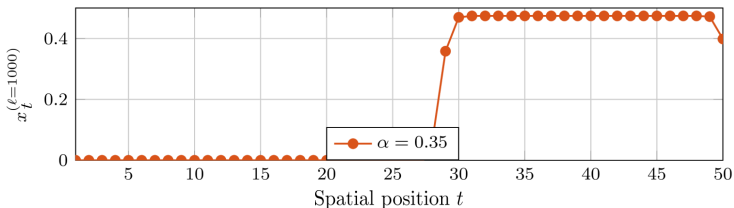
Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

$I = 1000$  iter.



**New:** Non-  
uniformly  
coupled code  
with  
 $d_v = 5, d_c = 10$   
**Single-side  
convergence**

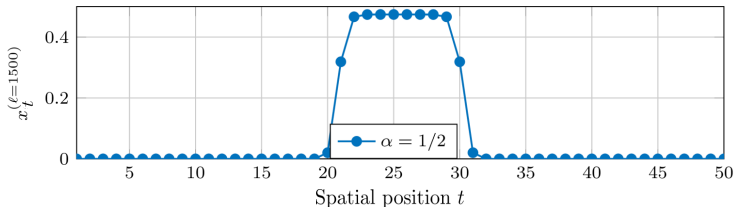


# Density Evolution ( $\varepsilon = 0.48, L = 50$ )

Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

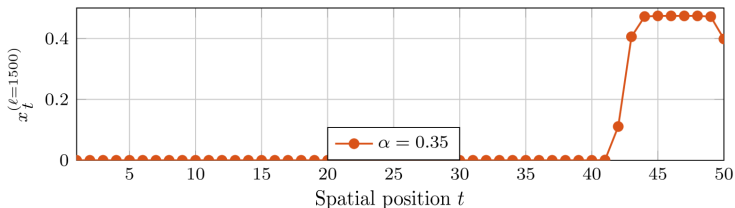
$I = 1500$  iter.



**New:** Non-  
uniformly  
coupled code  
with

$d_v = 5, d_c = 10$

**Single-side  
convergence**

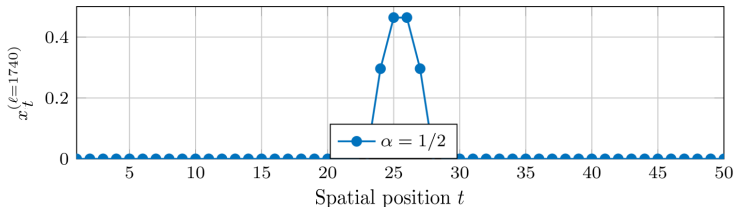


# Density Evolution ( $\varepsilon = 0.48, L = 50$ )

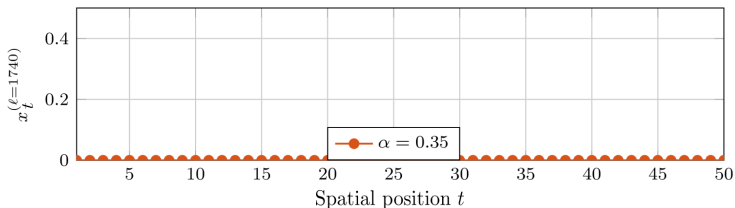
Conventional  
Spatially  
Coupled LDPC  
Code

$d_v = 5, d_c = 10$

$I = 1740$  iter.



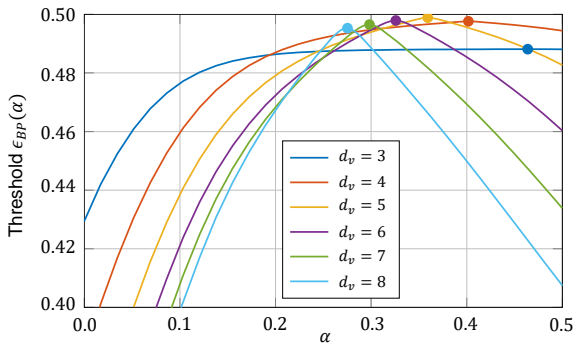
**New:** Non-uniformly coupled code with  $d_v = 5, d_c = 10$   
**Single-side convergence**



**Less iterations required for full convergence!**

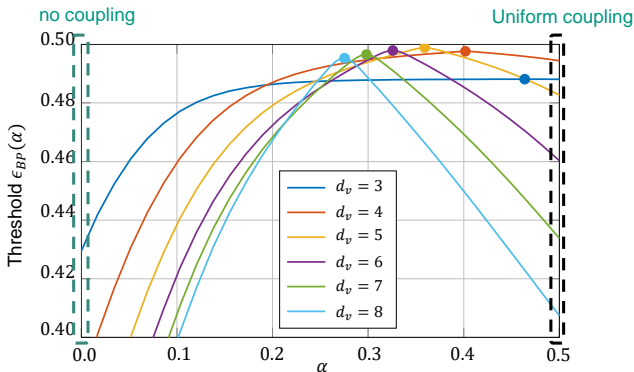
# Non-Uniform Coupling vs. Conventional Uniform Coupling (1)

- Decoding thresholds: SC-LDPC [ $d_v, 2d_v, w = 2, \alpha, L = 100$ ] over the BEC



# Non-Uniform Coupling vs. Conventional Uniform Coupling (1)

- Decoding thresholds: SC-LDPC [ $d_v, 2d_v, w = 2, \alpha, L = 100$ ] over the BEC



# Non-Uniform Coupling with $w > 2$

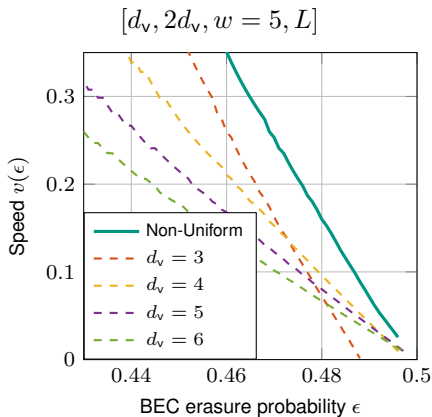
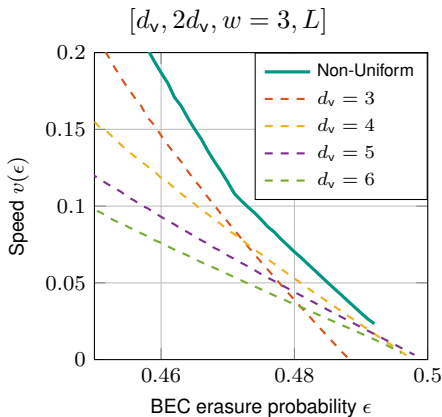
- Thresholds already sufficiently close to capacity with  $w > 2$
- Improvement of decoding speed (speed of wave)
  
- Optimizing of coupling vectors  $\nu$  using numerical methods
- Using tool of **differential evolution**
- Cost function is speed of decoding wave (estimated by curve fitting from profile)
- Details omitted here but can be found in [SA19]

[SA19] L. Schmalen and V. Aref, "Spatially Coupled LDPC Codes with Non-uniform Coupling for Improved Decoding Speed," *Proc. ITW*, 2019, available online at <https://arxiv.org/abs/1904.07026>



# Non-Uniform Coupling with $w > 2$

- Thresholds already sufficiently close to capacity with  $w > 2$
- Improvement of decoding speed (speed of wave)



- Demonstration of SC-LDPC Code Decoding

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- SC-LDPC codes have some other unique properties, we will highlight just one of them
- Consider the following scenario:
  - All spatial positions have been correctly received with the exception of position  $t_b$ , which is *completely erased*
  - This can model different scenarios, for instance packet loss in data networks, or server failure in distributed storage, or impulse noise in xDSL systems
- When can we recover from this burst?

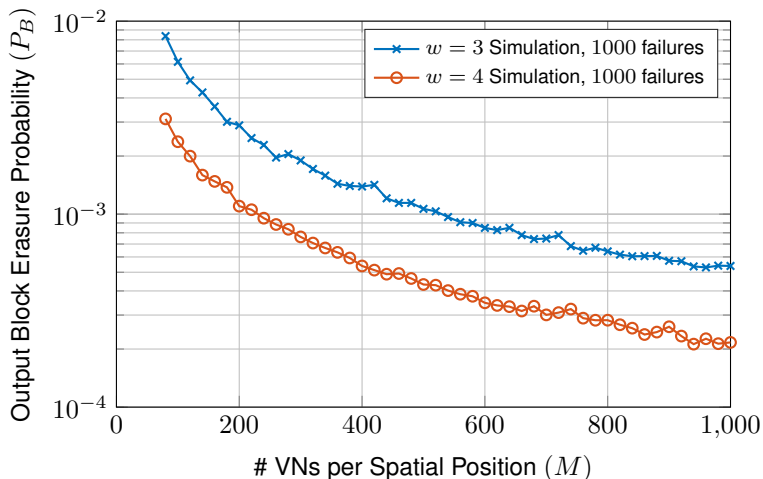
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- When can we recover from this burst?

## Theorem

*Consider an  $[d_v, d_c, w, L]$  SC-LDPC code where  $L - 1$  spatial positions have been correctly received and one spatial position is completely erased. A necessary condition for recovering this spatial position is that  $w > (\epsilon^*)^{-1}$ , where  $\epsilon^*$  is the BEC decoding threshold of the regular  $[d_v, d_c]$  LDPC code.*

## Example: $[d_v = 3, d_c = 6, w, L]$ SC-LDPC

- Construct codes with  $M$  variable nodes per spatial position
- One position erased, other positions without error



# What is going on?

- We should be able to recover the burst
- What happened?



# What is going on?

- We should be able to recover the burst
- What happened?
  
- **Finite-length effect due to the construction**
- A burst on some structures is not recoverable
- We will just give a few hints on what is going on
- Full analysis and many extensions in [ARS18]

[ARS18] V. Aref, N. Rengaswamy and L. Schmalen, "Finite-length analysis of spatially-coupled regular LDPC ensembles on burst erasure channels," *IEEE Transactions on Information Theory*, vol. 64, no. 5, pp. 3431–3449, 2018, available online at <https://arxiv.org/abs/1611.08267>

## Lemma (Probability that two variable nodes form a stopping set [ARS18])

Consider a code sampled randomly from the uniformly coupled  $[d_v, d_c, w, L]$  ensemble with  $M$  variable nodes per spatial position (SP). The probability that two variable nodes from SP  $t$  form a stopping set is

$$P_{\mathcal{R}} = \frac{\left(1 - \frac{1}{d_c}\right)^{d_v}}{\sum_{\ell=0}^{d_v} \binom{d_v}{\ell} \binom{wM \frac{d_v}{d_c} - d_v}{d_v - \ell} \left(1 - \frac{1}{d_c}\right)^{\ell}} \quad (1)$$

- The proof is based on a counting argument and purely combinatoric [ARS18]
- If  $M$  grows, the probability  $P_{\mathcal{R}}$  decreases

[ARS18] V. Aref, N. Rengaswamy and L. Schmalen, "Finite-length analysis of spatially-coupled regular LDPC ensembles on burst erasure channels," *IEEE Transactions on Information Theory*, vol. 64, no. 5, pp. 3431–3449, 2018, available online at <https://arxiv.org/abs/1611.08267>

## Lemma (Second moment method)

Let  $X \geq 0$  be a positive random variable with finite variance. Then

$$P(X > 0) \geq \frac{(\mathbb{E}\{X\})^2}{\mathbb{E}\{X^2\}}$$

## Lemma (Joint expectation of two stopping sets [ARS18])

Consider one SP of a code sampled from uniformly coupled  $[d_v, d_c, w, L]$  ensemble with  $M$  variable nodes per spatial position (SP). Recall the indicator function  $U_{ij} = 1$  if VNs  $v_i$  and  $v_j$  in that SP form a size-2 stopping set. Assuming  $d_c > 2$  and  $wM \geq 2(d_v + 1)d_c$ , we have the following bound on the joint expectation

$$\mathbb{E}\{U_{ij}U_{kl}\} \leq \frac{2\mathbb{E}\{U_{ij}\}}{\left( wM \frac{d_v}{d_c} - 2d_v \right)}$$

## Theorem

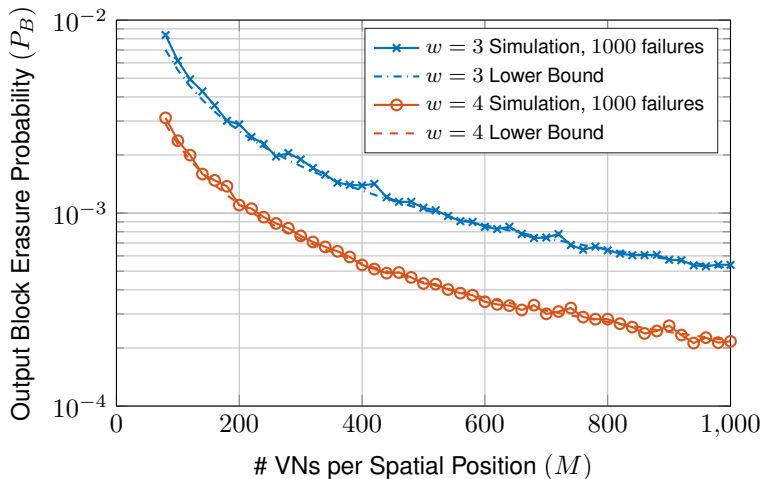
Consider a code sampled uniformly from the  $[d_v, d_c, w, L]$  ensemble with  $M$  variable nodes per spatial position (SP) with  $w d_c > 2$  and  $wM \geq 2(d_v + 1)d_c$ . If all variable nodes of a randomly chosen SP are erased, the (average) probability of BP decoding failure is lower-bounded by

$$P_B^{\text{SPBC}} \geq \binom{M}{2} \left( 1 - \frac{M^2}{\left(\frac{w}{d_c}M - 3\right)^{d_v}} \right) P_{\mathcal{R}}$$

where  $P_{\mathcal{R}}$  is the probability that two variable nodes from an SP of the code form a stopping set, given by (1).

# Example: $[d_v = 3, d_c = 6, w, L]$ SC-LDPC

- Construct codes with  $M$  variable nodes per spatial position



# Can We Do Better?

- The error rates are pretty depressing, even with large  $w$
- How can we do better?

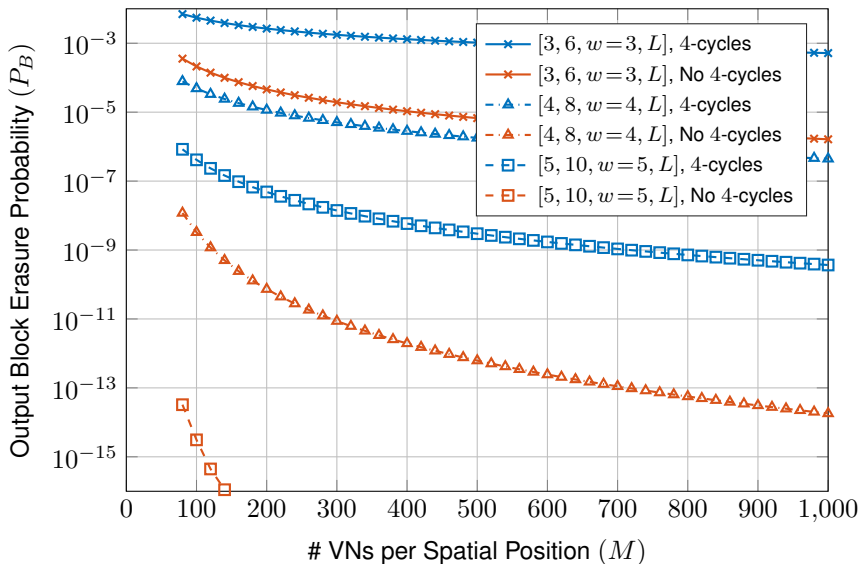
- The error rates are pretty depressing, even with large  $w$
- How can we do better?
- If size-2 stopping sets cause most of the error floor, **expurgate them!**
- Increasing the girth of the graph to 6 (no length-4 cycles) yields to a minimum stopping set size of

$$s_{\min} = d_v + 1$$

- We can approximate the probability that  $d_v + 1$  VNs form a stopping set as

$$P_{\mathcal{R},6} \approx \frac{\left(1 - \frac{1}{d_c}\right)^{\frac{1}{2}d_v(d_v+1)} \frac{(d_v!)^{d_v+1}}{\left(\frac{1}{2}d_v(d_v+1)\right)!}}{\binom{wM \frac{d_v}{d_c} - \frac{1}{2}d_v(d_v+1)}{\frac{1}{2}d_v(d_v+1)}}$$

# Effect of Expurgation





# Additional Erasures in Remaining SPs

- We now consider the case where

$$\epsilon_{t_b} = 1$$

$$\epsilon_{\sim t_b} = \epsilon$$

i.e., one SP is completely erased, all other SPs transmitted over BEC( $\epsilon$ )

# Additional Erasures in Remaining SPs

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## Error Probability for Both Errors and Erasures

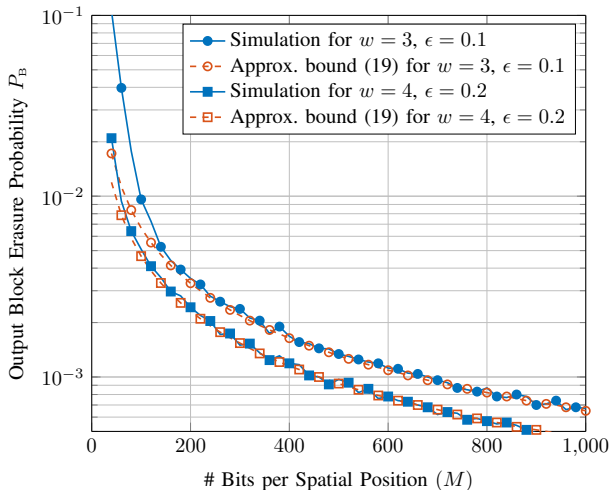
By counting all possibilities for stopping sets (which can now form between all possible positions), we get the approximation

$$P_B \gtrsim \lambda_0 + \epsilon(2 - \epsilon) \sum_{k=1}^w \lambda_k + \epsilon^2 \sum_{k=0}^{w-1} (L - k - 1) \lambda_k$$

with

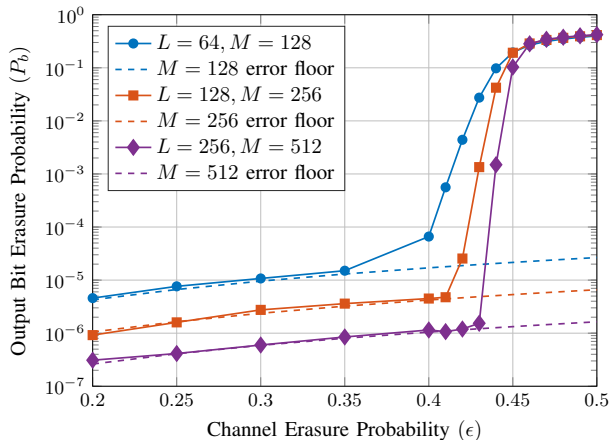
$$\lambda_0 = \binom{M}{2} P_{\mathcal{R}} \quad \text{and} \quad \lambda_k = M^2 \left(1 - \frac{k}{w}\right)^{d_v} P_{\mathcal{R}} \quad (k \in \{1, \dots, w-1\})$$

# Simulation Results (1)



- $[3, 6, w, L = 10]$  ensemble (design rate  $\frac{1}{2}$ )
- $\epsilon_{t_b} = 1, \epsilon_{\sim t_b} = \epsilon$
- Again, we can accurately predict the error floor

# Simulation Results (2)



- $[3, 6, w, L]$  ensemble (design rate  $\frac{1}{2}$ )
- $\epsilon_{t_b} = 1, \epsilon_{\sim t_b} = \epsilon$
- Again, we can accurately predict the error floor

# Random Burst Channel

- Now: Burst of length  $b$  starts at variable node  $S$  at SP  $t_b$
- Erased bit positions in codeword

$$E = \{(t_b - 1)M + S, \dots, (t_b - 1)M + S + b - 1\}$$

- We introduce the normalized quantities

$$\beta = \frac{b}{M} \quad \text{and} \quad s = \frac{S}{M}$$

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## Theorem

Consider the  $(d_v, d_c, w, L)$  SC-LDPC ensemble and transmission over a channel where a random selection of  $b$  consecutive VNs are erased and all other VNs are received without erasure. Let  $\beta = \frac{b}{M}$  denote the normalized burst length. Furthermore, let  $w \geq 1 + \lceil \beta \rceil$ . Asymptotically (in the limit of  $M$ ), the erased VNs can be recovered when  $w \geq \left\lceil (1 + \lceil \beta \rceil) / \epsilon_{LDPC[d_v, d_c]}^* \right\rceil$ , where  $\epsilon_{LDPC[d_v, d_c]}^*$  is the BP threshold of the underlying  $(d_v, d_c)$  uncoupled LDPC ensemble.

- We consider a random starting position  $\beta$
- A burst of length  $\beta M$  and random starting position is recoverable in the limit of  $M$  if

$$\beta < \beta_{\max} = \min_{0 \leq s < 1} \beta(s)$$

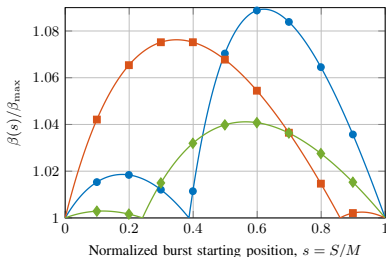
- We can determine  $\beta_{\max}$  using density evolution [ARS16]

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- We can determine  $\beta_{\max}$  using density evolution [ARS16]

—●—  $w=3, \beta_{\max}=1.61$  —■—  $w=4, \beta_{\max}=2.14$  —◆—  $w=5, \beta_{\max}=2.76$



- $[d_v = 3, d_c = 6, w, L]$  ensemble
- Normalized maximum correctable burst length for a given starting position  $S = sM$

[ARS16] V. Aref, N. Rengaswamy, and L. S., "Spatially coupled LDPC codes affected by a single random burst of errors," in *Proc. ISTC*, Sep. 2016



## Theorem

Consider the  $[d_v, d_c, w, L]$  SC-LDPC ensemble affected by a burst of length  $b = \beta M$  with a random starting bit  $M(t_b - 1) + S$ , where  $1 \leq S \leq M$  and  $\beta \ll \beta_{\max}$ . The number of erased VNs in SP  $z$  is given by

$$m_z = \begin{cases} 0 & z < t_b \\ \min\{b, M - S + 1\} & z = t_b \\ \max\{0, \min\{M, b + S - 1 - (z - t_b)M\}\} & z > t_b. \end{cases}$$

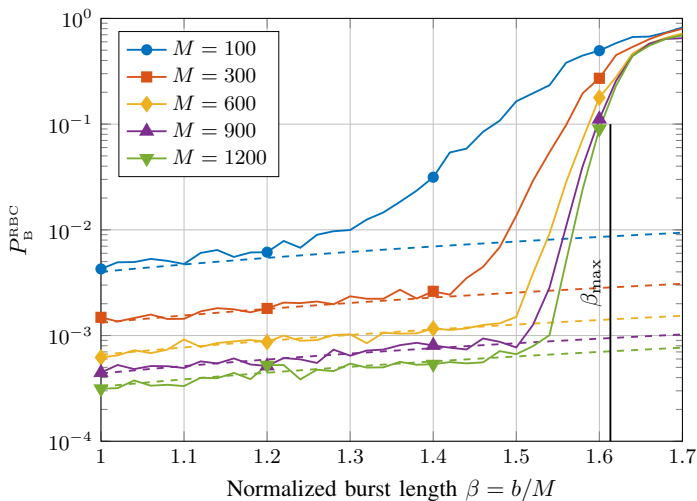
Then the expected number of size-2 stopping sets formed by VNs erased by the burst is given by

$$\mathbb{E}[N_2(S, t_b, b)] \approx \frac{L - \lceil \beta \rceil}{(L - \beta)M + 1} \sum_{S=1}^M \sum_{z=1}^{\lceil \beta \rceil + 1} \left( \binom{m_z}{2} q_0 + \sum_{k=1}^{w-1} m_z m_{z+k} q_k \right)$$

with  $q_k = P_{\mathcal{R}} \left(1 - \frac{k}{w}\right)^{d_v} \mathbb{1}_{\{k \in \{0, \dots, w-1\}\}}$ .

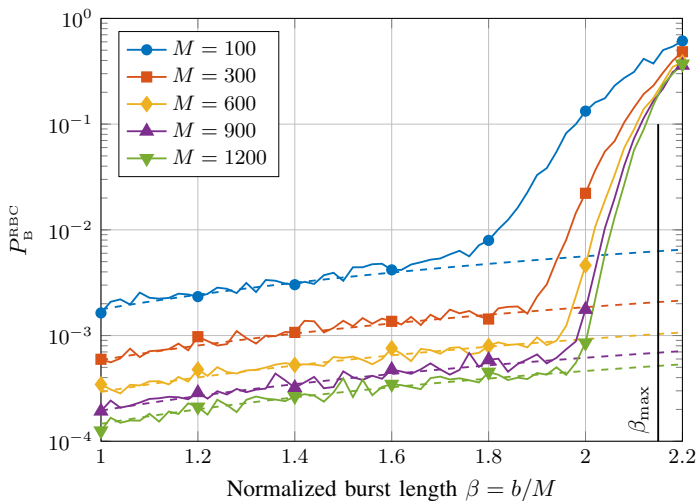
# Monte-Carlo Simulations

- The  $[d_v = 3, d_c = 6, w = 3, L]$  ensemble



# Monte-Carlo Simulations

- The  $[d_v = 3, d_c = 6, w = 4, L]$  ensemble

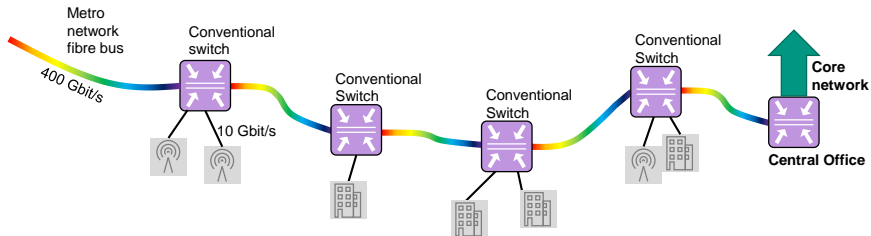


- 1 Introduction: Channel Coding and LDPC Codes
- 2 Spatially Coupled LDPC Codes
  - Motivation and Definition
  - Performance of SC-LDPC Codes
  - Practical Implementation of SC-LDPC Codes
  - Improvement of SC-LDPC Codes by Non-Uniform Coupling
- 3 Burst Correction Capabilities of Spatially Coupled LDPC Codes
  - Error Probability after Burst Erasures
  - Application Example
- 4 Conclusions

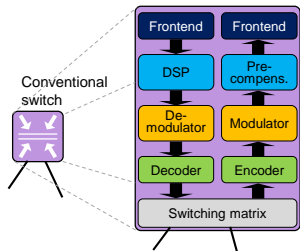
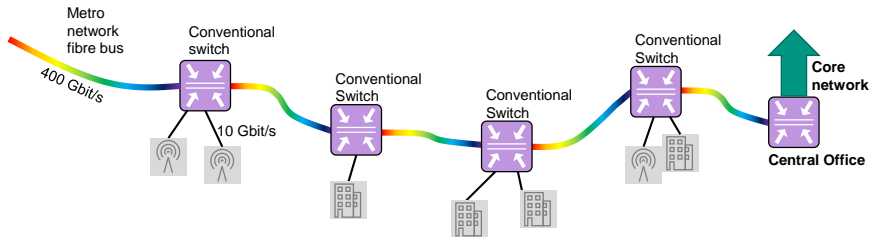
- Burst erasures come in many varieties
- Collision in coded random access schemes (e.g., slotted ALOHA) [LPL+12]
- Strong fading in block fading channels [ULA+14]
- Error due to misestimation of channel
- Impulse noise in wireline communications (e.g., DSL)
- Server failure in distributed storage (cloud storage) [JBS+15]
  
- Application: **Distributed coding**

- [LPL+12] G. Liva, E. Paolini, M. Lentmaier, and M. Chiani, "Spatially-coupled random access on graphs," in Proc. IEEE Int. Symp. Inform. Theory (ISIT), pp. 478–482, 2012.
- [ULA+14] N. Ul Hassan, M. Lentmaier, I. Andriyanova, and G. Fettweis, "Improving code diversity on block-fading channels by spatial coupling," in Proc. IEEE Int. Symp. Inform. Theory (ISIT), pp. 2311–2315, June 2014
- [JBS+15] F. Jardel, J. J. Boutros, M. Sarkiss, and G. Rekaya-Ben Othman, "Spatial coupling for distributed storage and diversity applications," in Proc. IEEE International Conference in Communications and Networking (ComNet), (Hammamet, Tunisia), Nov. 2015

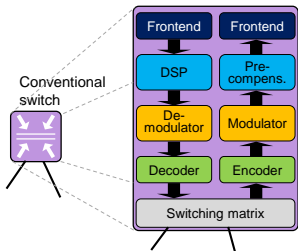
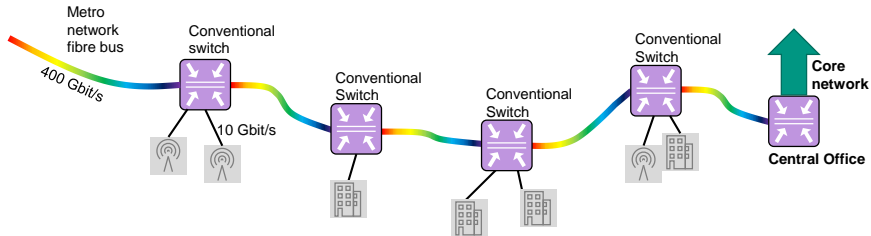
# Multi-Point-to-Point Aggregation Bus



# Multi-Point-to-Point Aggregation Bus



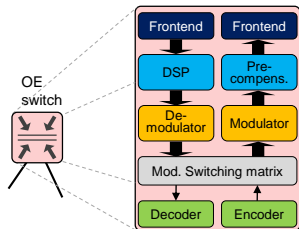
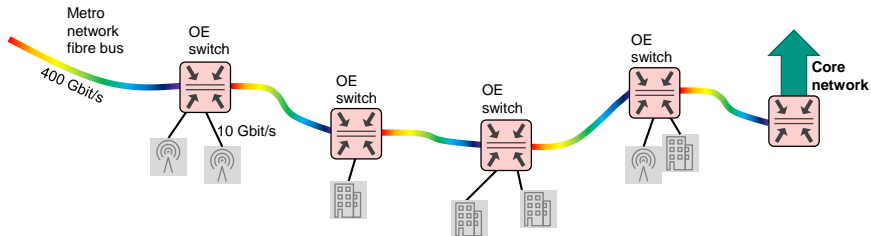
# Multi-Point-to-Point Aggregation Bus



- Switch has an optical coherent front-end with full encoding and decoding in each hop
- Significant amount of data will not be dropped, but re-encoded for next hop
- Full decoding often not necessary in each hop, but only at destination

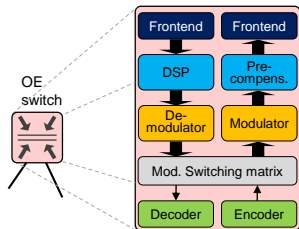
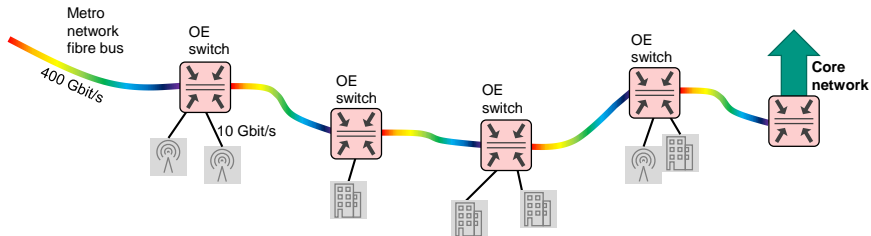


# Multi-Point-to-Point Aggregation Bus



[LBB+17] W. Lautenschläger, N. Benzaoui, F. Buchali, L. Dembeck, R. Dischler, B. Franz, U. Gebhard, J. Milbrandt, Y. Pointurier, D. Rösener, L. S. and A. Leven "Optical Ethernet – Flexible Optical Metro Networks," *IEEE/OSA J. Lightw. Technol.*, 2017

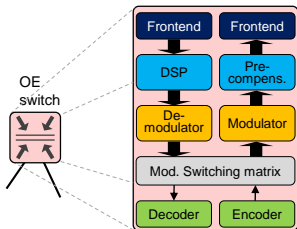
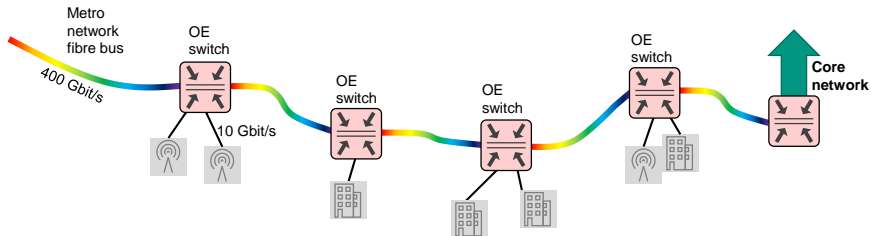
# Multi-Point-to-Point Aggregation Bus



- Based on header, perform switching on noisy frames after hard-decision

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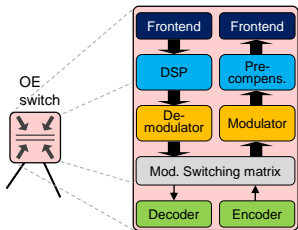
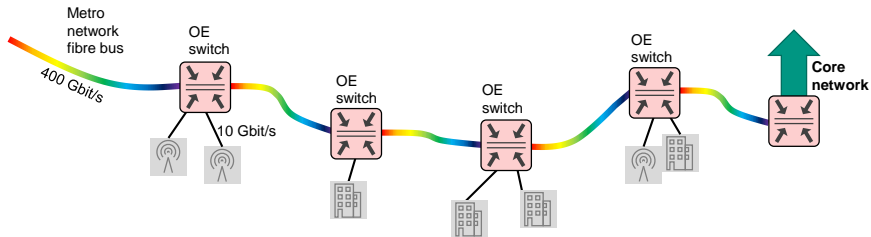
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- Based on header, perform switching on noisy frames after hard-decision
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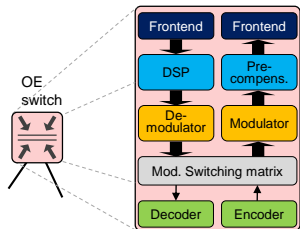
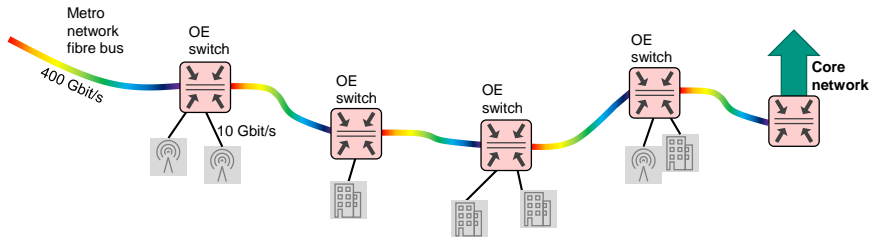
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- Based on header, perform switching on noisy frames after hard-decision
- Carry out decoding only for data that are being switched
- Decision-and-forward relay

[LBB+17] W. Lautenschläger, N. Benzaoui, F. Buchali, L. Dembeck, R. Dischler, B. Franz, U. Gebhard, J. Milbrandt, Y. Pointurier, D. Rösener, L. S. and A. Leven "Optical Ethernet – Flexible Optical Metro Networks," *IEEE/OSA J. Lightw. Technol.*, 2017

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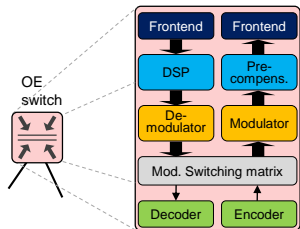
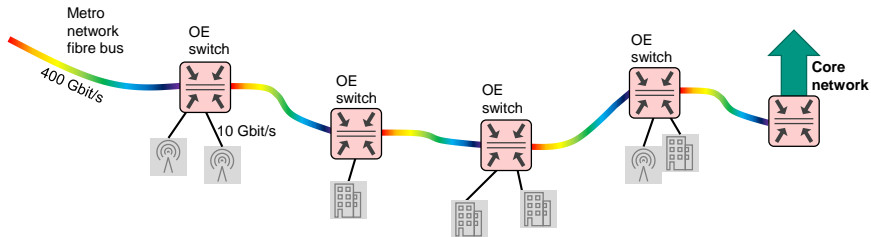


- Container-based transmission of codewords
- Adding data to empty parts of codeword

"0"

[LBB+17] W. Lautenschläger, N. Benzaoui, F. Buchali, L. Dembeck, R. Dischler, B. Franz, U. Gebhard, J. Milbrandt, Y. Pointurier, D. Rösener, L. S. and A. Leven "Optical Ethernet – Flexible Optical Metro Networks," *IEEE/OSA J. Lightw. Technol.*, 2017

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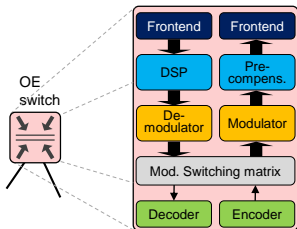
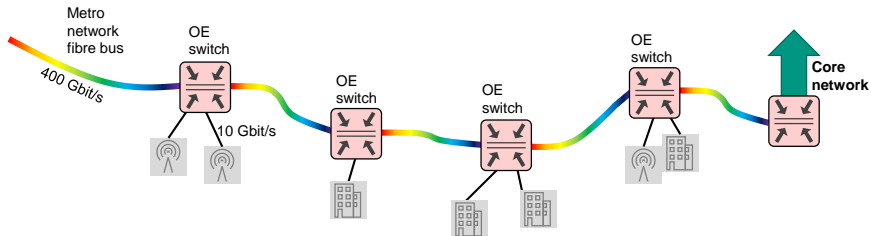


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- Container-based transmission of codewords
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Container 

Data 1	"0"	Parity 1
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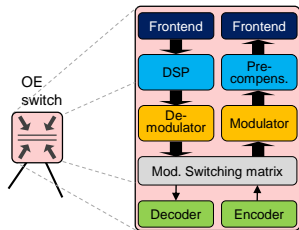
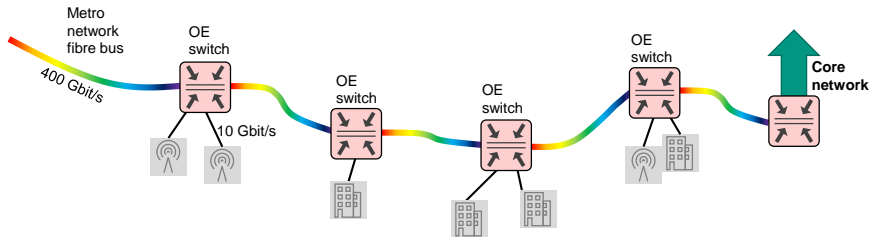
+

New data 

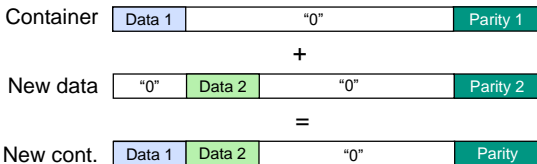
"0"	Data 2	"0"	Parity 2
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[LBB+17] W. Lautenschläger, N. Benzaoui, F. Buchali, L. Dembeck, R. Dischler, B. Franz, U. Gebhard, J. Milbrandt, Y. Pointurier, D. Rösener, L. S. and A. Leven "Optical Ethernet – Flexible Optical Metro Networks," *IEEE/OSA J. Lightw. Technol.*, 2017

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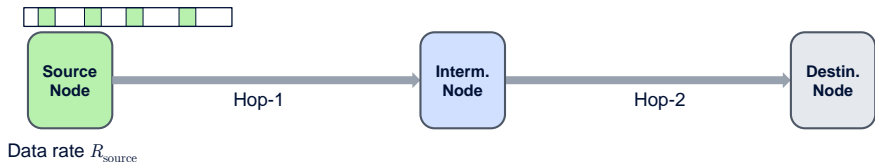
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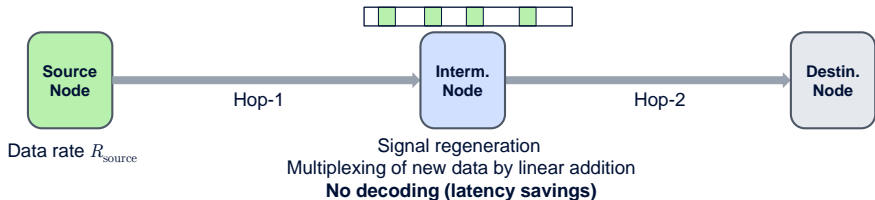
- Many applications require **ultra-low networking latency**, e.g., industrial control (Industrie 4.0), high-speed trading, sensor networks, etc...
- New networking architecture with end-to-end channel coding for such applications [LBB+17], [SEB+17]
- Toy Example:



[LBB+17] W. Lautenschläger, N. Benzaoui, F. Buchali, L. Dembeck, R. Dischler, B. Franz, U. Gebhard, J. Milbrandt, Y. Pointurier, D. Rösener, L. S. and A. Leven "Optical Ethernet – Flexible Optical Metro Networks," *IEEE/OSA J. Lightw. Technol.*, 2017

[SEB+17] L. S. T. A. Eriksson, F. Buchali, R. Dischler, U. Gebhard, "Distributed Transmission and Spatially Coupled Forward Error Correction in Regenerative Multipoint-to-Point Networks," *Proc. ECOC*, 2017

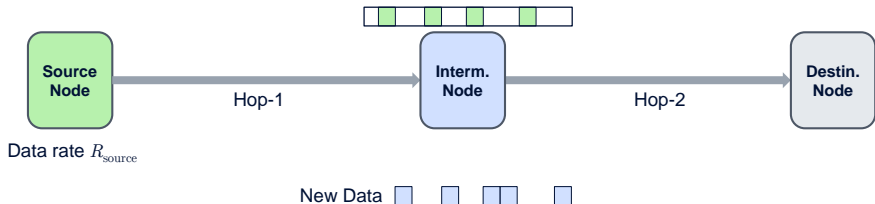
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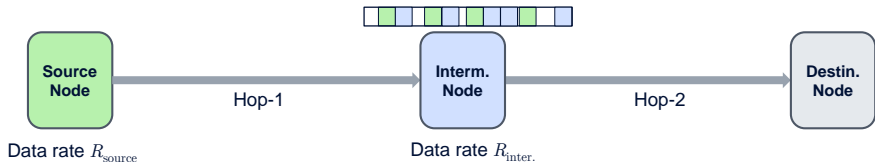
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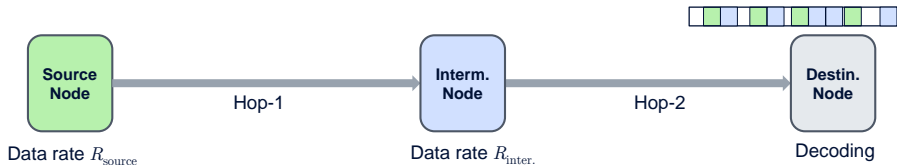
- Many applications require **ultra-low networking latency**, e.g., industrial control (Industrie 4.0), high-speed trading, sensor networks, etc...
- New networking architecture with end-to-end channel coding for such applications [LBB+17], [SEB+17]
- Toy Example:



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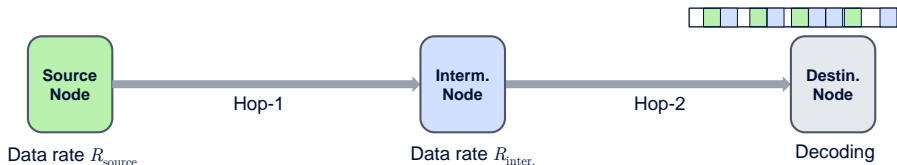
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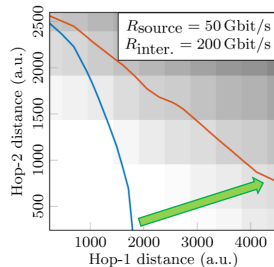
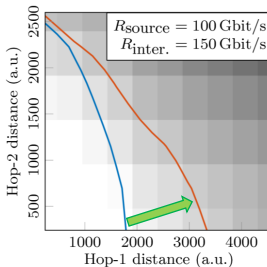
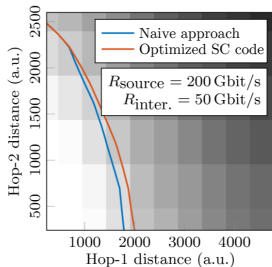
- **Packets traveling from source to destination see worse channel than other packets** ( $\rightarrow$  burst!)

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# End-to-end Coding in Low-Latency Networks

- Optimization of code parameters using burst correction theory
- Optimization of media access (data placement) by burst theory
- Flexibility and adaptivity of effective rates to channel conditions
- Hop-1 and Hop-2 achievable distance pairs significantly improved



- 1 Introduction: Channel Coding and LDPC Codes
- 2 Spatially Coupled LDPC Codes
  - Motivation and Definition
  - Performance of SC-LDPC Codes
  - Practical Implementation of SC-LDPC Codes
  - Improvement of SC-LDPC Codes by Non-Uniform Coupling
- 3 Burst Correction Capabilities of Spatially Coupled LDPC Codes
  - Error Probability after Burst Erasures
  - Application Example
- 4 Conclusions



- Spatially Coupled LDPC Codes are a class of powerful codes that **extend** LDPC codes
- Record coding gains in optical communications
- Best approach to design parameters still unknown (various open research problems)
- General concept that applies to other areas of science as well (compressed sensing, statistical physics, ...)
- Applications in distributed storage, random access, fading channels, subscriber lines, etc...
  
- **Active area of research**



# Overview

- Parity-Check Matrix of a Spatially Coupled LDPC Code
- ML Performance of LDPC Codes
- Analysis of SC-LDPC Codes over General Channels
- Simulation Example of Non-Uniform Coupling
- Burst erasure of complete spatial position
- Error Probability after Decoding with Burst Erasure
- The Block Erasure Channel



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# ML Dec. of Regular $[d_v, d_c]$ LDPC Codes

- The ML decoding threshold  $\epsilon^{\text{ML}}$  (i.e., the  $\epsilon$  up to which an ideal ML decoder successfully works) of a regular  $[d_v, d_c]$  LDPC code on the BEC can be described as follows
- Let  $x_{\text{ML}}$  be the unique positive solution of the equation

$$x + \frac{1}{d_c}(1-x)^{d_c-1}(d_v + d_v(d_c-1)x - d_c x) - \frac{d_v}{d_c} = 0$$

- Then,

$$\epsilon^{\text{ML}} = \frac{x_{\text{ML}}}{(1 - (1 - x_{\text{ML}})^{d_c-1})^{d_v-1}}$$

- The proof of this fact is involved and beyond the scope of this lecture, however, we have

$$\lim_{d_v \rightarrow \infty} \epsilon^{\text{ML}} = \frac{d_v}{d_c} = 1 - r_d = C_{\text{BEC}}$$

with exponentially fast convergence

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- We can use spatial coupling to construct capacity achieving codes (with  $w \rightarrow \infty, L \rightarrow \infty, d_v \rightarrow \infty, n \rightarrow \infty, \text{ and } \ell \rightarrow \infty$ )

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# Analysis of SC-LDPC Codes for General Channels

- For more general channels, e.g., the BI-AWGN channel, we can use the Gaussian approximation to analyze the convergence behavior
- For every spatial position  $t$  we define mean values  $\mu_{\xi,t}^{(\ell)}$  and  $\mu_{\chi,t}^{(\ell)}$  which we initialize as

$$\mu_{\xi,t}^{(1)} = \begin{cases} \mu_c & \text{if } t \in \{1, 2, \dots, L\} \\ +\infty & \text{otherwise} \end{cases}$$

where in practice, we replace  $+\infty$  by an appropriately large number.

- The variable node outgoing message mean can be computed as

$$\mu_{\xi,t}^{(\ell)} = \mu_{c,t} + \frac{d_v - 1}{w} \sum_{i=0}^{w-1} \mu_{\chi,t+i}^{(\ell)}$$

where

$$\mu_{c,t} = \begin{cases} \mu_c & \text{if } t \in \{1, 2, \dots, L\} \\ +\infty & \text{otherwise} \end{cases}$$

# Analysis of SC-LDPC Codes for General Channels (2)

- Note that the *incoming* message to the check node has a distribution that is a sum of Gaussian pdfs with different means (and variances) and with weight  $1/w$ .
- The mean of the check node outgoing message at position  $t$  can be computed as

$$1 - \varphi(\mu_{\chi,t}) = \left[ 1 - \frac{1}{w} \sum_{j=0}^{w-1} \varphi(\mu_{\xi,t-j}^{(\ell-1)}) \right]^{d_c-1}$$

- This leads to the overall update equation, combining variable and check node update

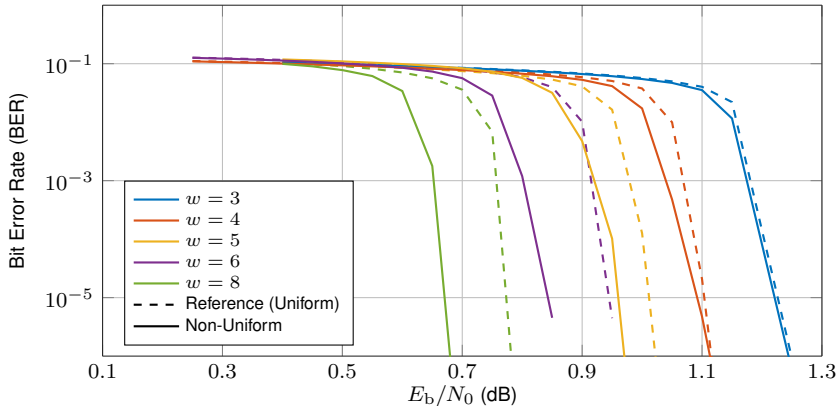
$$\mu_{\chi,t} = \varphi^{-1} \left( 1 - \left[ 1 - \frac{1}{w} \sum_{j=0}^{w-1} \varphi \left( \mu_{c,t-j} + \frac{d_v-1}{w} \sum_{i=0}^{w-1} \mu_{\chi,t+i-j}^{(\ell-1)} \right) \right]^{d_c-1} \right)$$

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# Simulation Example

- Simulation with windowed decoder ( $W_D = 5w$  and  $I = 1$  iteration)
- Code details and optimal values  $\nu$  can be found in [SA19]



[SA19] L. Schmalen and V. Aref, "Spatially Coupled LDPC Codes with Non-uniform Coupling for Improved Decoding Speed," *Proc. ITW*, 2019, available online at <https://arxiv.org/abs/1904.07026>

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(2)

- We have the update equation

$$\xi_t^{(\ell)} = \epsilon_t \left( 1 - \frac{1}{w} \sum_{i=0}^{w-1} \left( 1 - \frac{1}{w} \sum_{j=0}^{w-1} \xi_{t+i-j}^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1} \quad \forall t \in \{1, \dots, L\}$$

with

$$\epsilon_t = \begin{cases} 1 & \text{if } t = t_b \\ 0 & \text{otherwise} \end{cases}$$

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with

$$\epsilon_t = \begin{cases} 1 & \text{if } t = t_b \\ 0 & \text{otherwise} \end{cases}$$

- This implies that  $\xi_t^{(\ell)} = 0$ , if  $t \neq t_b$ , and we only need to consider the case  $t = t_b$  with

$$\xi_{t_b}^{(\ell)} = \left( 1 - \frac{1}{w} \sum_{i=0}^{w-1} \left( 1 - \frac{1}{w} \sum_{j=0}^{w-1} \xi_{t_b+i-j}^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1}$$

(3)

- In the inner sum, we only need to consider the case  $j = i$ , as the other contributions are zero, hence

$$\begin{aligned}\xi_{tb}^{(\ell)} &= \left( 1 - \frac{1}{w} \sum_{i=0}^{w-1} \left( 1 - \frac{1}{w} \xi_{tb}^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1} \\ &= \left( 1 - \left( 1 - \frac{1}{w} \xi_{tb}^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1}\end{aligned}$$

- We now make a change of variable  $\kappa^{(\ell)} = \frac{1}{w} \xi_{tb}^{(\ell)}$  leading to

$$\begin{aligned}w\kappa^{(\ell)} &= \left( 1 - \left( 1 - \kappa^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1} \\ \Rightarrow \kappa^{(\ell)} &= \frac{1}{w} \left( 1 - \left( 1 - \kappa^{(\ell-1)} \right)^{d_c-1} \right)^{d_v-1}\end{aligned}$$



- This is exactly the update equation of  $[d_v, d_c]$  regular LDPC codes on the BEC (with  $\epsilon$  replaced by  $\frac{1}{w}$ ).
- We know that  $\kappa^{(\ell)}$  converges to 0 iff  $\frac{1}{w} < \epsilon^*$  where  $\epsilon^*$  is the threshold of the  $[d_v, d_c]$  LDPC code
- This leads to the following result

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## Theorem

Consider a code sampled uniformly from the  $[d_v, d_c, w, L]$  ensemble with  $M$  variable nodes per spatial position (SP) with  $w d_c > 2$  and  $wM \geq 2(d_v + 1)d_c$ . If all variable nodes of a randomly chosen SP are erased, the (average) probability of BP decoding failure is lower-bounded by

$$P_B^{\text{SPBC}} \geq \binom{M}{2} \left( 1 - \frac{M^2}{\left(\frac{w}{d_c}M - 3\right)^{d_v}} \right) P_{\mathcal{R}}$$

where  $P_{\mathcal{R}}$  is the probability that two variable nodes from an SP of the code form a stopping set, given by (1).

Proof:

- Let  $N_2^{\text{SP}}$  denote the number of size-2 stopping sets formed by two VNs in an SP

- Note that that

$$N_2^{\text{SP}} = \sum_{0 \leq i < j \leq M} U_{ij} \quad \text{and} \quad \mathbb{E}\{N_2^{\text{SP}}\} = \binom{M}{2} P(U_{ij} = 1) = \binom{M}{2} \mathbb{E}\{U_{ij}\}$$

and

$$\mathbb{E}\{U_{ij}\} = P_{\mathcal{R}}$$

- We have

$$\begin{aligned} P_{\text{B}}^{\text{SPBC}} &= P(\text{At least one stopping set in a SP}) \\ &\geq P(N_2^{\text{SP}} \geq 1) \\ &\geq \frac{(\mathbb{E}\{N_2^{\text{SP}}\})^2}{\mathbb{E}\{(N_2^{\text{SP}})^2\}} \end{aligned}$$

■ Furthermore

$$\begin{aligned}\mathbb{E} \left\{ (N_2^{\text{SP}})^2 \right\} &= \mathbb{E} \left\{ \left( \sum_{1 \leq i < j \leq M} U_{ij} \right)^2 \right\} \\ &= \sum_{1 \leq i < j \leq M} \mathbb{E} \{ U_{ij}^2 \} + \sum_{\substack{(i,j) \neq (k,l) \\ i < j, k < l}} \mathbb{E} \{ U_{ij} U_{kl} \} \\ &\leq \binom{M}{2} \mathbb{E} \{ U_{ij} \} + \frac{2 \binom{M}{2} \left( \binom{M}{2} - 1 \right)}{\left( wM \frac{d_v}{d_c} - 2d_v \right)} \mathbb{E} \{ U_{ij} \}\end{aligned}$$

■ Finally

$$\begin{aligned} P_B^{\text{SPBC}} &\geq \frac{\binom{M}{2}^2 \mathbb{E}\{U_{ij}\}^2}{\binom{M}{2} \mathbb{E}\{U_{ij}\} + \frac{2\binom{M}{2}(\binom{M}{2}-1)}{(wM \frac{d_v}{d_c} - 2d_v)} \mathbb{E}\{U_{ij}\}} \\ &\geq \frac{\binom{M}{2} \mathbb{E}\{U_{ij}\}}{1 + \frac{2\binom{M}{2}}{(wM \frac{d_v}{d_c} - 2d_v)}} \\ &\stackrel{(a)}{\geq} \mathbb{E}\{N_2^{\text{SP}}\} \left( 1 - \frac{2\binom{M}{2}}{(wM \frac{d_v}{d_c} - 2d_v)} \right) \\ &\geq \mathbb{E}\{N_2^{\text{SP}}\} \left( 1 - \frac{M^2}{\left(\frac{wM}{d_c} - 3\right)^{d_v}} \right) \end{aligned}$$

where (a) is due to the fact that  $\frac{1}{1+\tau} \geq 1 - \tau$  for  $\tau > -1$ .



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# The Block Erasure Channel (BLEC)

- Every SP is erased independently with probability  $p$

$$\epsilon_t = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1 - p \end{cases}$$

- Erasures far apart (in spatial dimension) do not interfere



# The Block Erasure Channel (BEC)

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- Erasures far apart (in spatial dimension) do not interfere
- Analysis consists of three parts
  - Partition SPs into segments of non-interfering erasures (at least  $w - 1$  non-erased positions separating segments). The expected number of SPs in such a segment is given by

$$\mathbb{E}[\tau_i] = \frac{1}{p} \left( \frac{1}{(1-p)^{w-1}} - 1 \right) \approx (w-1) \left( 1 + \frac{1}{2}(w-1)p + \frac{5}{6}(w-1)^2 p^2 \right)$$

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- There are  $N$  segments with  $\sum_{i=1}^N \tau_i = L + w - 1$ . The average decoding failure is obtained as

$$P_B^{\text{BLEC}} = 1 - \mathbb{E} \left[ \prod_{i=1}^N (1 - P_e^{(i)}) \right] \lesssim \frac{L + w - 1}{\mathbb{E}[\tau_i]} \mathbb{E} [P_e^{(1)}]$$

where  $P_e^{(1)}$  is the error probability of a segment

- Analysis consists of three parts

- (iii) Finally, we compute  $P_e^{(1)} = \sum_{i=1}^{\infty} Q_i$  where  $Q_i$  is the probability that a segment cannot be recovered if  $i$  SPs are erased. We consider decoding errors due to small stopping sets and decoding sets that cause the decoder to fail for any  $M$  (burst  $> \beta_{\max}$ ). Putting all together

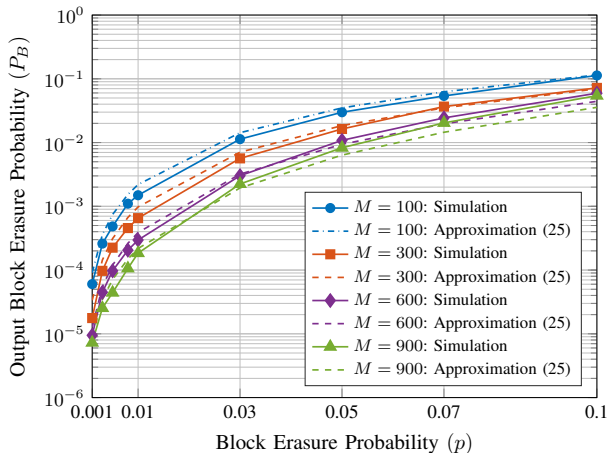
$$P_B^{\text{BLEC}} \approx (L + w - 1)p(1-p)^{w-1} \left( (1-p)^{w-1} \left[ P_B^{\text{SPBC}} + pP_B^{2\text{PBC}} \right] + \frac{p^2}{(1-p)^{w-1} + p} \right)$$

with

$$P_B^{2\text{PBC}} \gtrsim p(1-p)^{w-1} [1 - (1-p)^{w-1}] \left( 2 \binom{M}{2} q_0 + M^2 q_1 \right)$$

# The Block Erasure Channel

- $[d_v = 3, d_c = 6, w = 4, L = 30]$  code ensemble



- Approximations track true behavior closely, even for  $p$  as large as 0.1 and  $M$  as small as 100

- With toolbox of approximations, we can approximate codes for the multi-point-to-point scheme