

# A Deterministic Construction and Density Evolution Analysis for Generalized Product Codes

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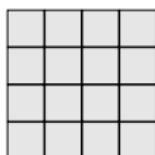
### In This Talk ...

- **Deterministic** code construction that recovers product codes, staircase codes, and block-wise braided codes as special cases
- Rigorous **density evolution analysis** possible over the binary erasure channel
- **Application:** Spatially-coupled product codes and symmetric generalized product codes

# Introduction: Product Codes and Staircase Codes

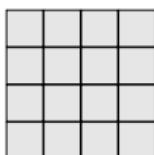
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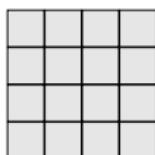
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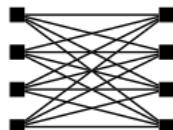
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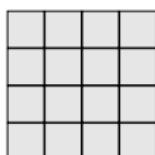
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Tanner graph



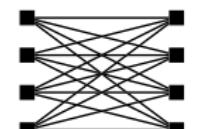
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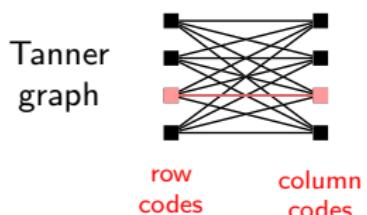
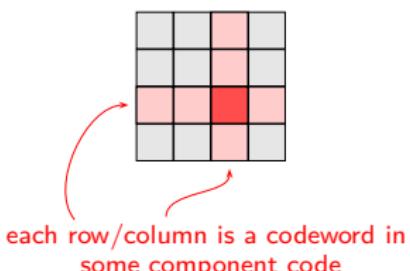


row  
codes      column  
codes

constraint node (CN) degree = component code length

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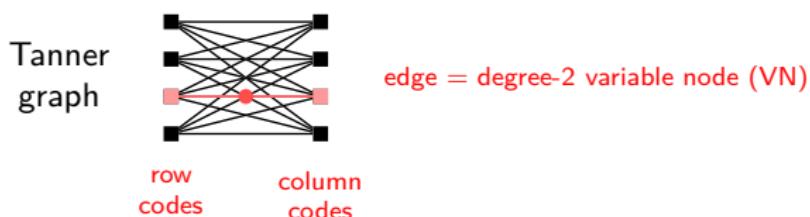
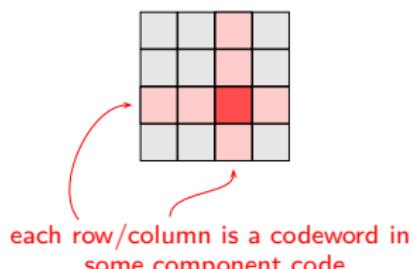
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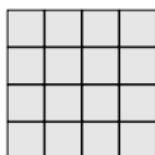
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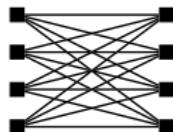
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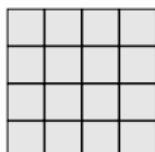


Tanner  
graph

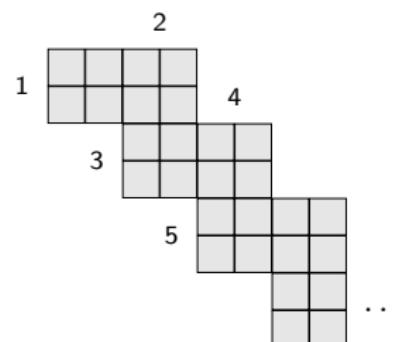


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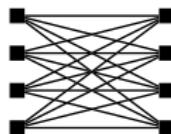
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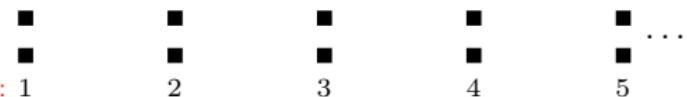
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Tanner graph

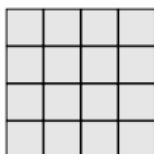


positions: 1

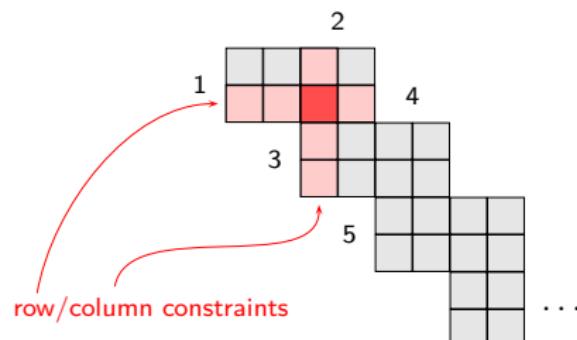


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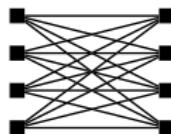
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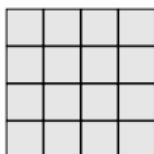


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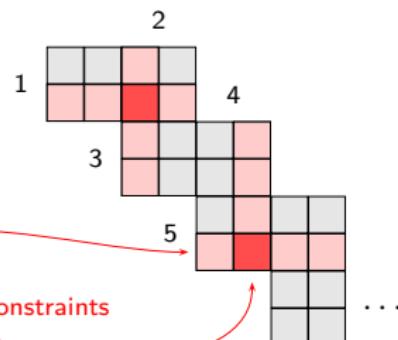


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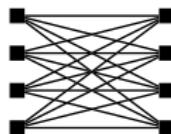
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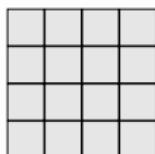
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2

3  
4

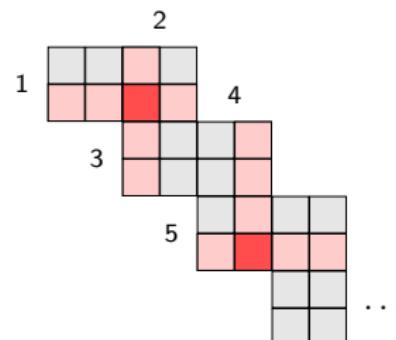
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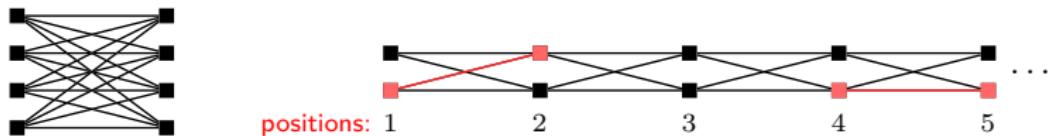
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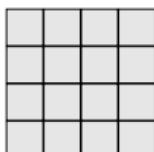


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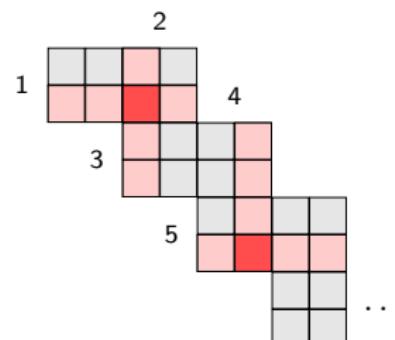


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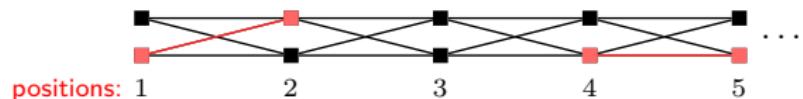
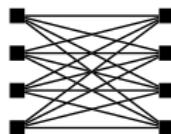
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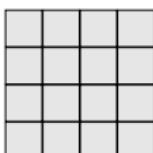
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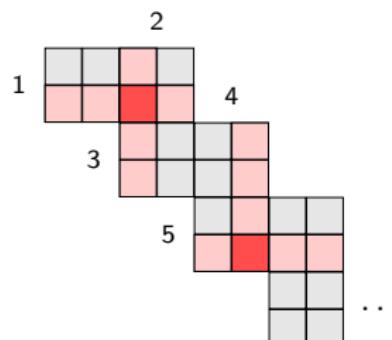
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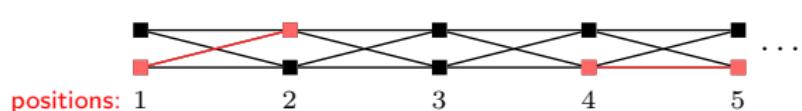
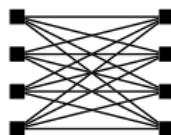
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- **Deterministic** codes with fixed and structured Tanner graph
- Our code construction recovers these (and other) codes as special cases

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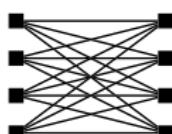
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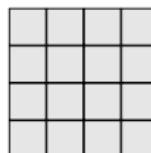
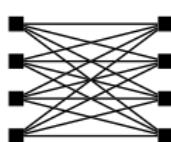
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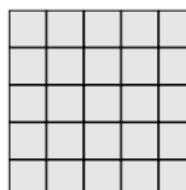
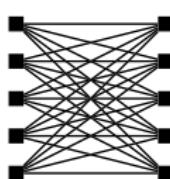
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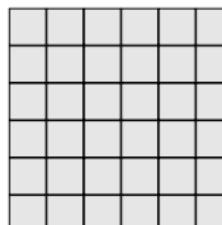
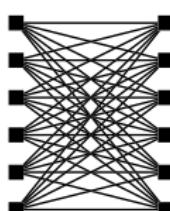
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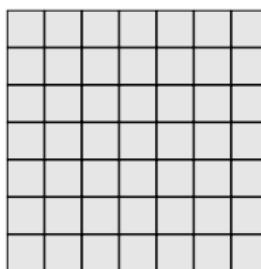
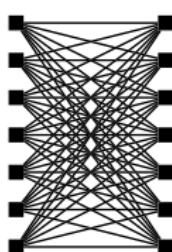
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## Code Construction for $\mathcal{C}_n(\eta)$

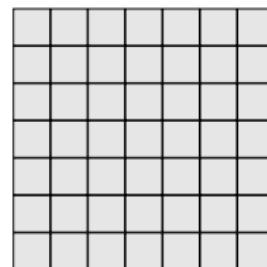
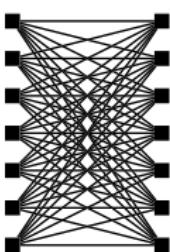
Place  $d$  CNs at each position and connect each CN at position  $i$  to each CN at position  $j$  (through a VN) iff  $\eta_{i,j} = 1$ .

Example:  $\eta = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ ,  $L = 2$ , and  $\gamma = 1$  gives a **product code** with  $n \times n$  array.

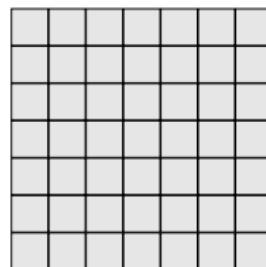
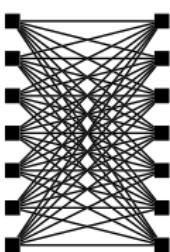
positions:    1                2                 $n = 7 \implies d = 7$



# Channel Model and Decoding

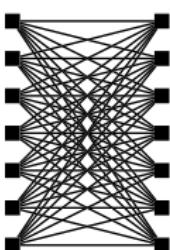


## Channel Model and Decoding



- Each CN corresponds to *t*-erasure correcting component code

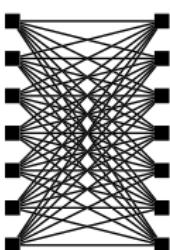
## Channel Model and Decoding



0	1	0	1	0	1	0
0	1	0	1	1	0	1
0	1	0	1	0	1	0
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	0	0	1	1	1
0	1	0	0	0	1	1

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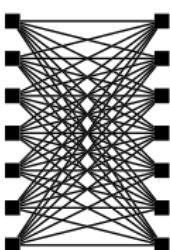
## Channel Model and Decoding



0	1	0	1	0	1	0
0	1	0	1	1	0	1
0	1	0	1	0	1	0
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	0	0	1	1	1
0	1	0	0	0	1	1

- Each CN corresponds to *t*-erasure correcting component code
- Codeword transmission over binary erasure channel with erasure probability  $p$

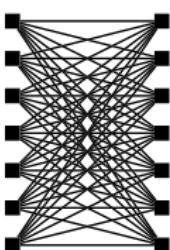
# Channel Model and Decoding



0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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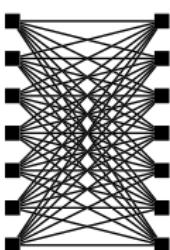
## Channel Model and Decoding



0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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  - If weight of an erasure pattern is  $\leq t$ , correct the pattern
  - If weight is  $> t$ , declare “failure” and do nothing (in that iteration)

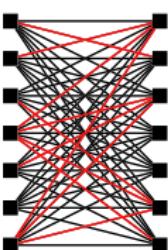
# Channel Model and Decoding



0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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- **Residual graph:** remove known variable nodes (i.e., edges)

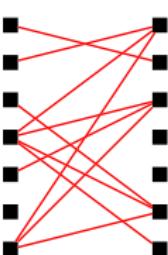
# Channel Model and Decoding



0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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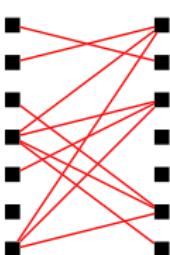
# Channel Model and Decoding



0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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# Channel Model and Decoding

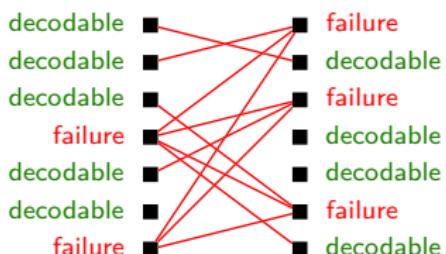


0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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- **Peeling** of vertices with degree  $\leq t$  (in parallel)

# Channel Model and Decoding

1st iteration ( $t = 2$ )

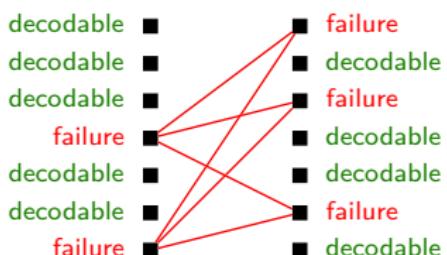


0	?	0	?	0	1	?
?	1	0	1	1	0	1
0	1	0	?	0	?	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	?	?	1	1	?
0	1	0	?	0	1	1

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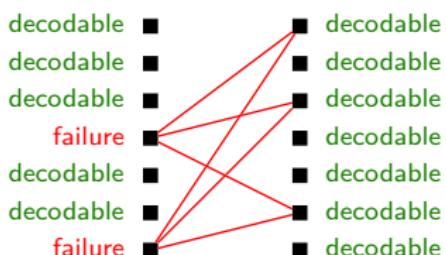


0	1	0	?	0	1	?
0	1	0	1	1	0	1
0	1	0	?	0	1	?
1	1	1	0	1	1	0
0	0	1	0	0	0	1
1	0	0	?	1	1	?
0	1	0	0	0	1	1

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# Channel Model and Decoding

2nd iteration ( $t = 2$ )



0	1	0	?	0	1	?	
0	1	0	1	1	0	1	
0	1	0	?	0	1	?	
1	1	1	0	1	1	0	
0	0	1	0	0	0	1	
1	0	0	?	1	1	?	
0	1	0	0	0	1	1	

- Each CN corresponds to  $t$ -erasure correcting component code
- Codeword transmission over **binary erasure channel** with erasure probability  $p$
- $\ell$  iterations of **bounded-distance decoding** for all CNs:
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# Channel Model and Decoding

2nd iteration ( $t = 2$ )

decodable	■	■ decodable	0 1 0 1 0 1 0
decodable	■	■ decodable	0 1 0 1 1 0 1
decodable	■	■ decodable	0 1 0 1 0 1 0
failure	■	■ decodable	1 1 1 0 1 1 0
decodable	■	■ decodable	0 0 1 0 0 0 1
decodable	■	■ decodable	1 0 0 0 1 1 1
failure	■	■ decodable	0 1 0 0 0 1 1

- Each CN corresponds to  $t$ -erasure correcting component code
- Codeword transmission over **binary erasure channel** with erasure probability  $p$
- $\ell$  iterations of **bounded-distance decoding** for all CNs:
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# Density Evolution

## Density Evolution

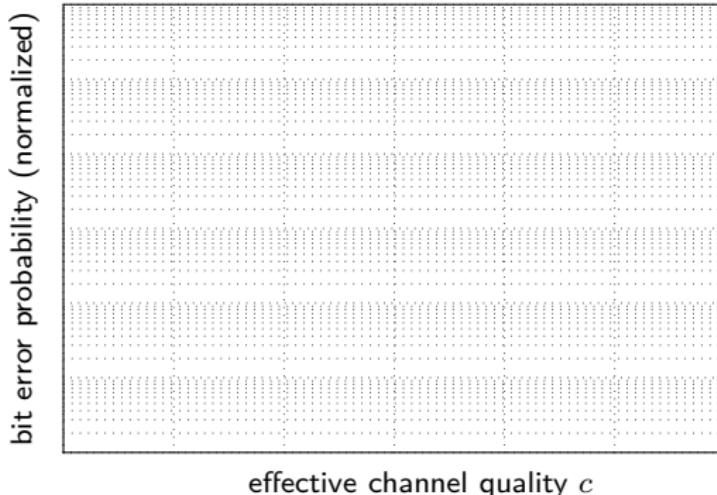
- What happens **asymptotically** for  $n \rightarrow \infty$ ?

## Density Evolution

- What happens **asymptotically** for  $n \rightarrow \infty$ ?
- Let  $p = c/n$  for  $c > 0$ , where  $c$  is the **effective channel quality**

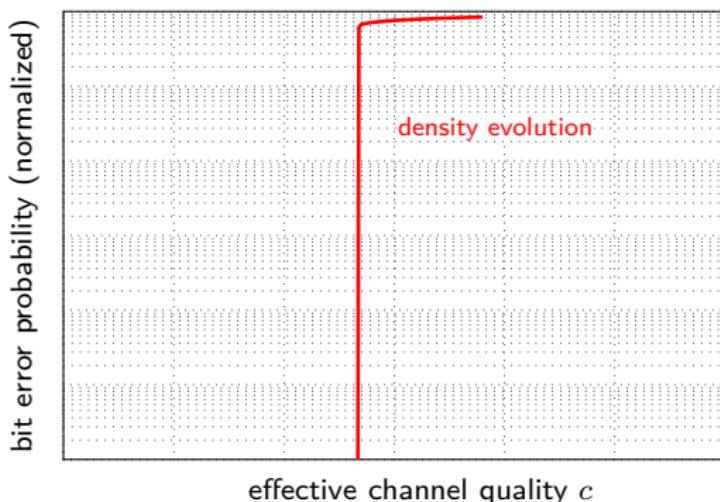
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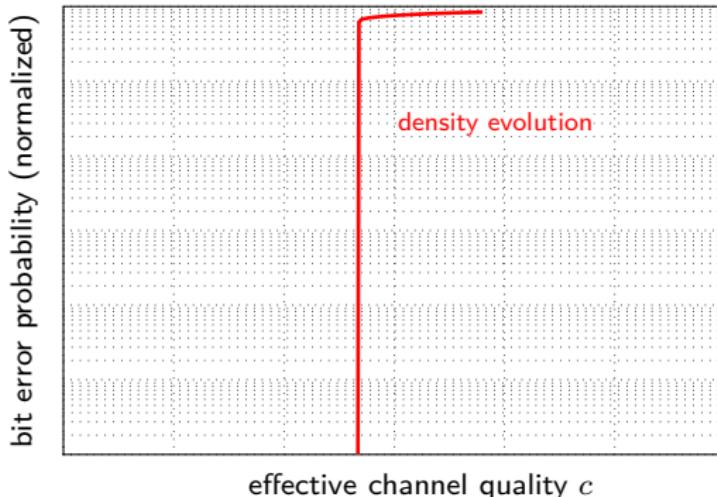
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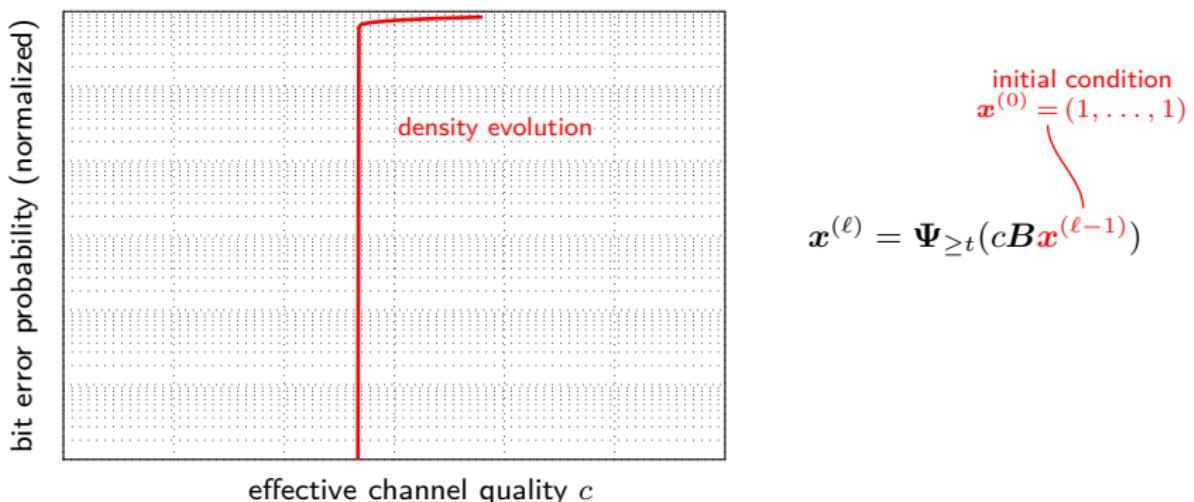
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$$\boldsymbol{x}^{(\ell)} = \Psi_{\geq t}(cB\boldsymbol{x}^{(\ell-1)})$$

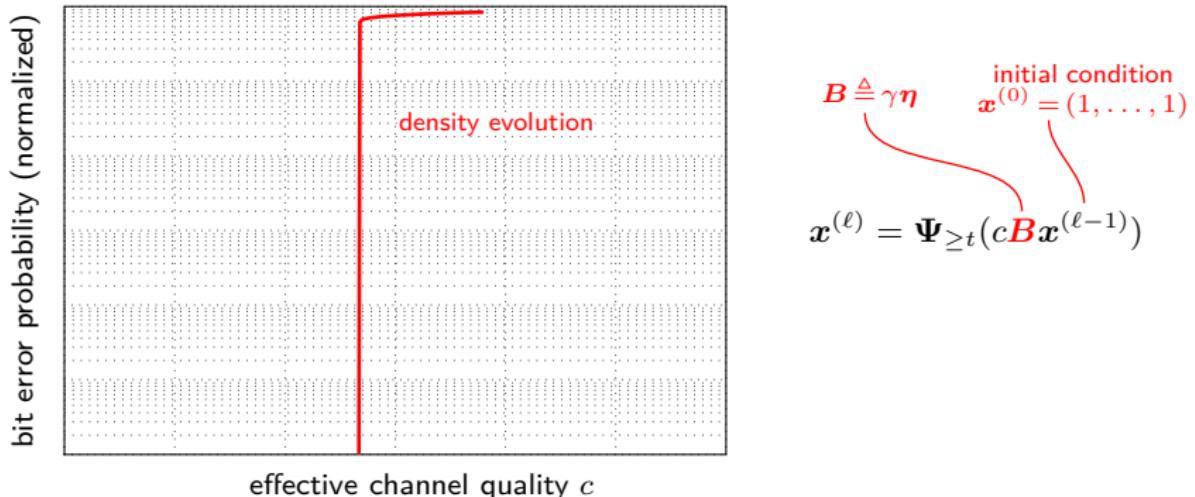
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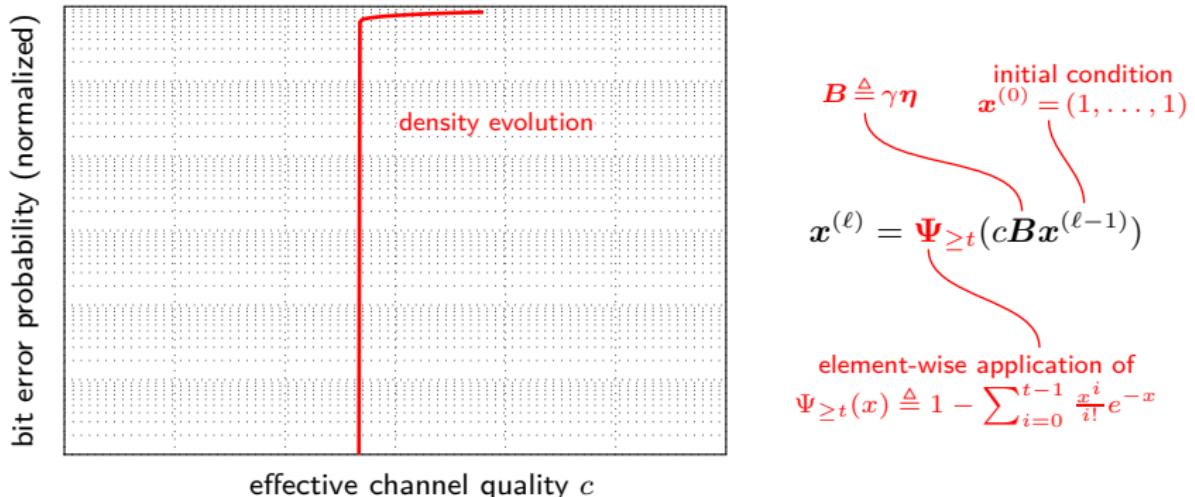
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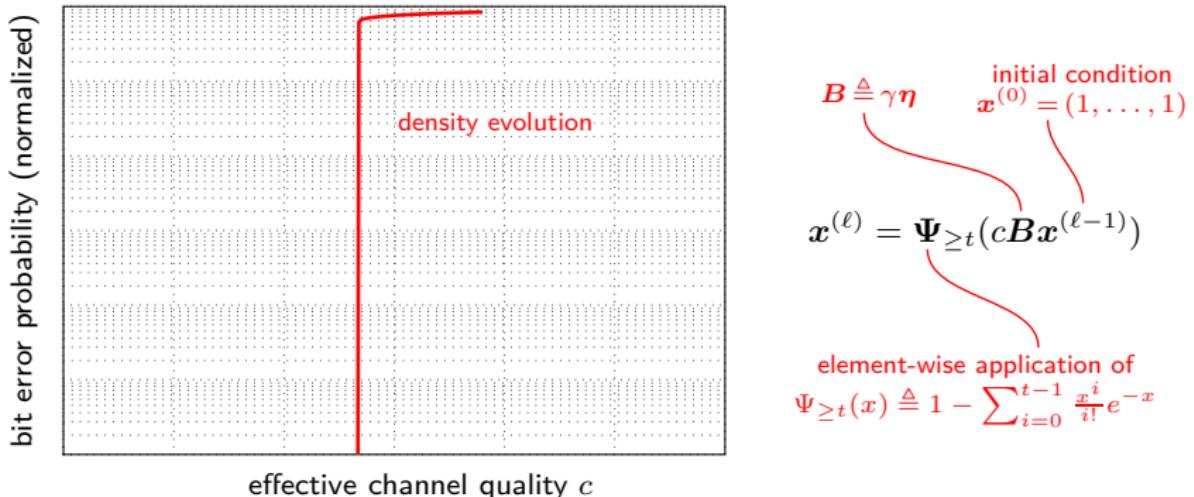
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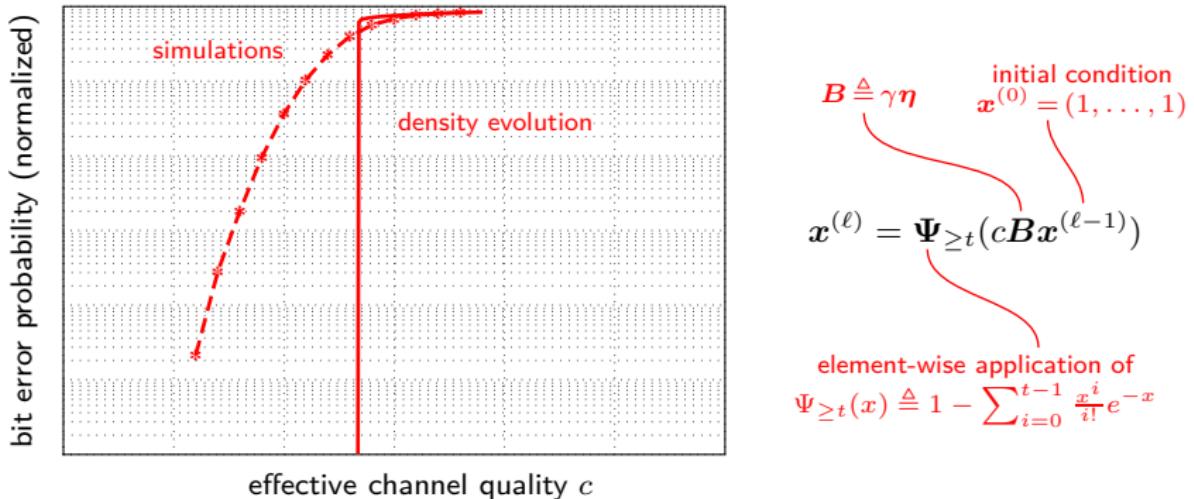
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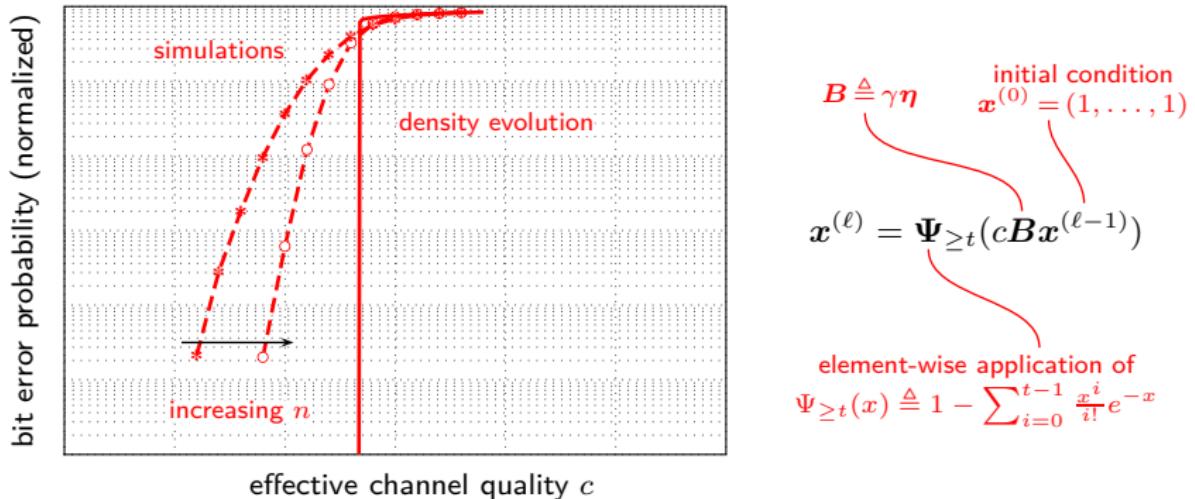
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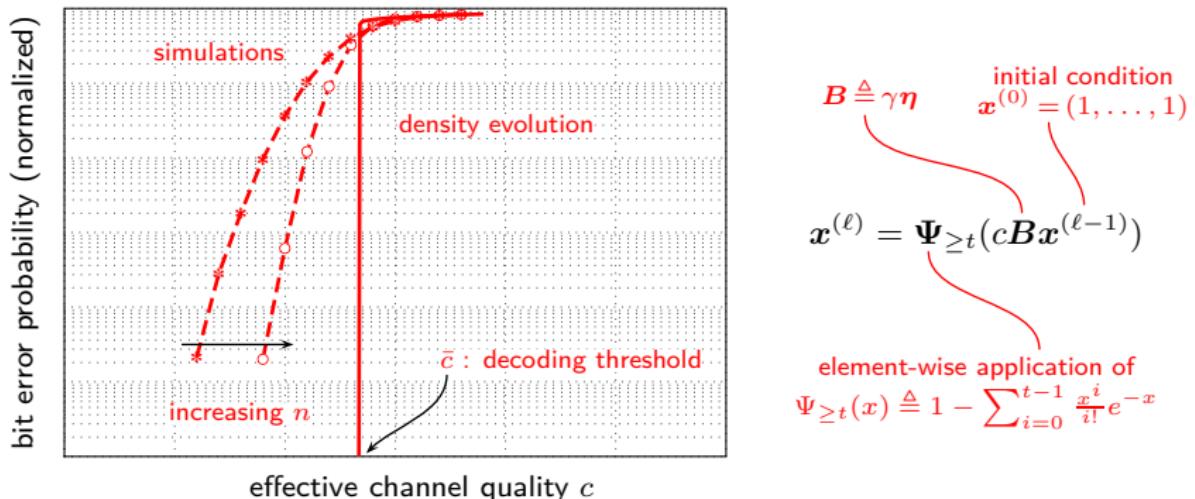
## Density Evolution

- What happens **asymptotically** for  $n \rightarrow \infty$ ?
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## Density Evolution

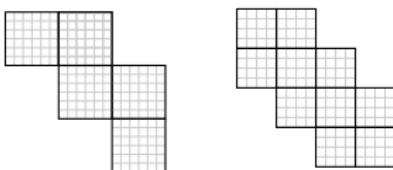
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# Spatially-Coupled Product Codes

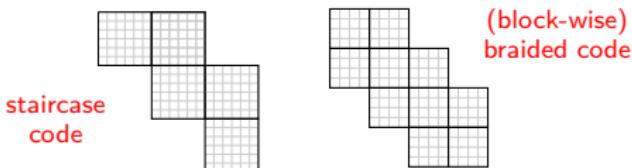
# Spatially-Coupled Product Codes

Deterministic



# Spatially-Coupled Product Codes

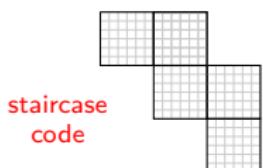
Deterministic



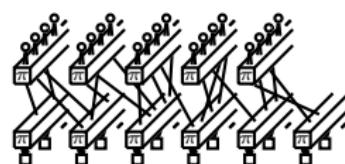
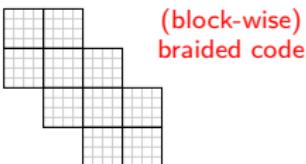
"convolutional-like"  
structure

# Spatially-Coupled Product Codes

Deterministic



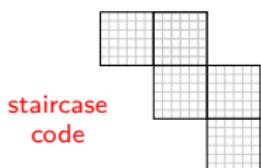
Ensemble-Based [Jian et al., 2012]



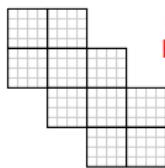
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# Spatially-Coupled Product Codes

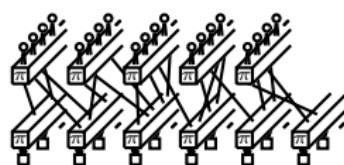
## Deterministic



"convolutional-like" structure



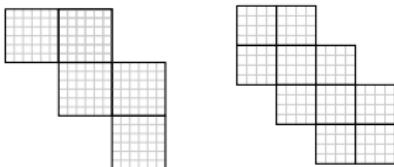
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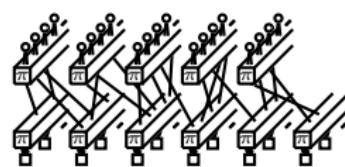
capacity-achieving at high rates over the binary symmetric channel

# Spatially-Coupled Product Codes

Deterministic

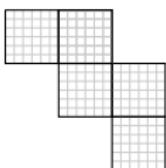


Ensemble-Based [Jian et al., 2012]



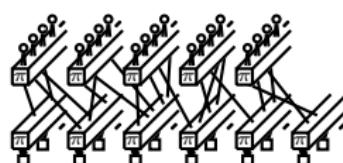
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Deterministic



$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(cB\mathbf{x}^{(\ell-1)})$$

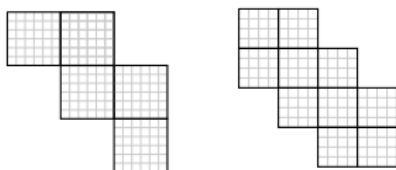
Ensemble-Based [Jian et al., 2012]



$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(c\tilde{B}\mathbf{x}^{(\ell-1)})$$

# Spatially-Coupled Product Codes

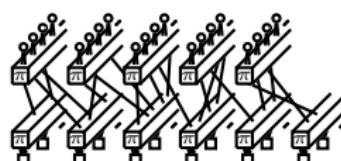
Deterministic



$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(cB\mathbf{x}^{(\ell-1)})$$

(B = \gamma\eta)

Ensemble-Based [Jian et al., 2012]



$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(c\tilde{B}\mathbf{x}^{(\ell-1)})$$

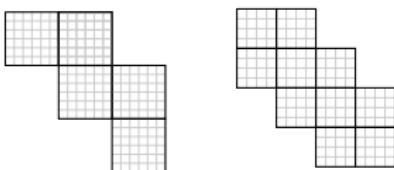
(\tilde{B} = \mathbf{A}^\top \mathbf{A})

$$\mathbf{A} = \frac{1}{w} \begin{pmatrix} 1 & 1 & \cdots & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & \cdots & 1 & 0 & 0 \\ 0 & 0 & \ddots & \ddots & \ddots & \ddots & 0 \\ 0 & 0 & 0 & 1 & 1 & \cdots & 1 \end{pmatrix}$$

$w$  (coupling width)

# Spatially-Coupled Product Codes

Deterministic



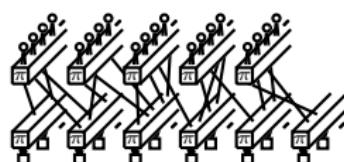
$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(cB\mathbf{x}^{(\ell-1)})$$

$$(B = \gamma\eta)$$

$$\frac{1}{2} \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}, \frac{1}{3} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix},$$

staircase                          braided (simplified)

Ensemble-Based [Jian et al., 2012]



$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(c\tilde{B}\mathbf{x}^{(\ell-1)})$$

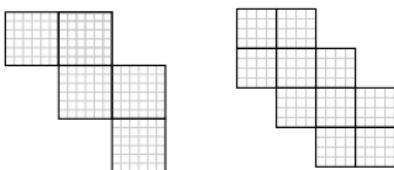
$$(\tilde{B} = A^\top A)$$

$$\frac{1}{4} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 2 & 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 1 & 2 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix}, \frac{1}{9} \begin{pmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 2 & 2 & 1 & 0 & 0 \\ 1 & 2 & 3 & 2 & 1 & 0 \\ 0 & 1 & 2 & 3 & 2 & 1 \\ 0 & 0 & 1 & 2 & 2 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 \end{pmatrix}$$

$w = 2$                                    $w = 3$

# Spatially-Coupled Product Codes

Deterministic



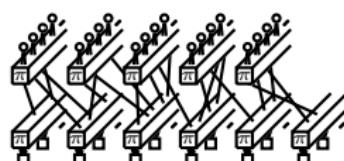
$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(cB\mathbf{x}^{(\ell-1)})$$

$$(B = \gamma\eta)$$

$$\frac{1}{2} \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}, \frac{1}{3} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix},$$

staircase                          braided (simplified)

Ensemble-Based [Jian et al., 2012]



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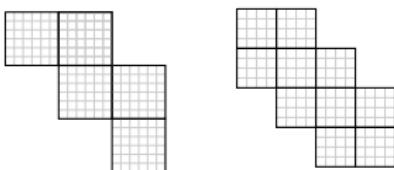
$$\frac{1}{4} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 2 & 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 1 & 2 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix}, \frac{1}{9} \begin{pmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 2 & 2 & 1 & 0 & 0 \\ 1 & 2 & 3 & 2 & 1 & 0 \\ 0 & 1 & 2 & 3 & 2 & 1 \\ 0 & 0 & 1 & 2 & 2 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 \end{pmatrix}$$

w = 2                                  w = 3

- Equations have the **same form**, but **different averaging matrices  $B$  and  $\tilde{B}$**

# Spatially-Coupled Product Codes

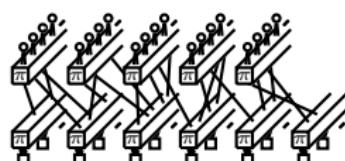
Deterministic



$$\mathbf{x}^{(\ell)} = \Psi_{\geq t}(cB\mathbf{x}^{(\ell-1)})$$

$$(B = \gamma\eta)$$

Ensemble-Based [Jian et al., 2012]



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$$\frac{1}{2} \begin{pmatrix} 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \end{pmatrix}, \frac{1}{3} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix},$$

staircase

braided (simplified)

$$\frac{1}{4} \begin{pmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 2 & 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 1 & 0 & 0 \\ 0 & 0 & 1 & 2 & 1 & 0 \\ 0 & 0 & 0 & 1 & 2 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 \end{pmatrix}, \frac{1}{9} \begin{pmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 2 & 2 & 1 & 0 & 0 \\ 1 & 2 & 3 & 2 & 1 & 0 \\ 0 & 1 & 2 & 3 & 2 & 1 \\ 0 & 0 & 1 & 2 & 2 & 1 \\ 0 & 0 & 0 & 1 & 1 & 1 \end{pmatrix}$$

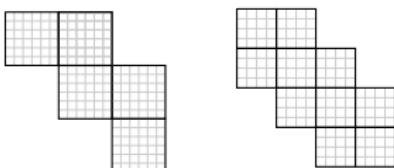
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- Equations have the **same form**, but **different averaging matrices  $B$  and  $\tilde{B}$**
- One can show that ensemble performance can be “emulated”

# Spatially-Coupled Product Codes

Deterministic



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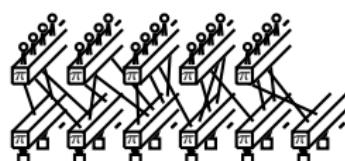
$$(B = \gamma\eta)$$

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staircase

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$w = 2$

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- Equations have the **same form**, but **different averaging matrices  $B$  and  $\tilde{B}$**
- One can show that ensemble performance can be “emulated”
- $\Rightarrow$  ensemble **threshold bounds** in [Jian et al., 2012] **apply to deterministic codes!**

# Symmetric Generalized Product Codes

## Symmetric Generalized Product Codes

- So far,  $\eta_{i,i} = 0$  for all  $i \in \{1, 2, \dots, L\}$ . What about  $\eta_{i,i} = 1$ ?

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

product code

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

staircase code

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

(block-wise) braided code

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$$n = 5 \implies d = 5$$

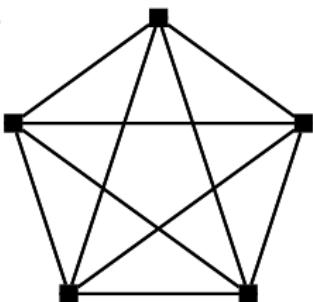


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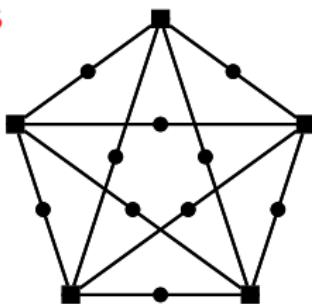


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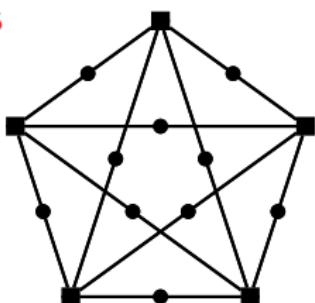


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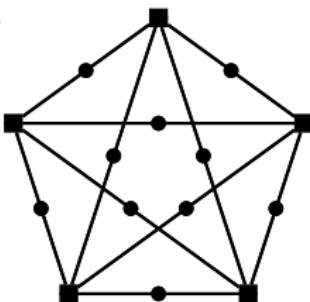
array representation?

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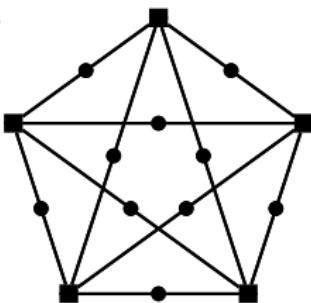
0	*	*	*	*
$c_1$	0	*	*	*
$c_2$	$c_3$	0	*	*
$c_4$	$c_5$	$c_6$	0	*
$c_7$	$c_8$	$c_9$	$c_{10}$	0

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0	*	*	*	*
$c_1$	0	*	*	*
$c_2$	$c_3$	0	*	*
$c_4$	$c_5$	$c_6$	0	*
$c_7$	$c_8$	$c_9$	$c_{10}$	0

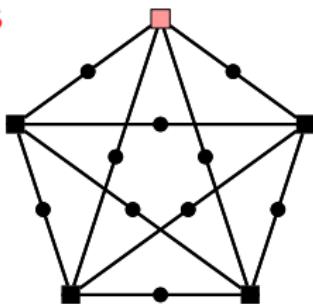
\* = not used

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$c_7$	$c_8$	$c_9$	$c_{10}$	0

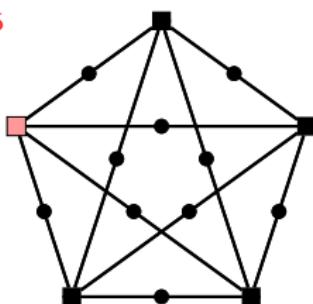
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$c_4$	$c_5$	$c_6$	0	*
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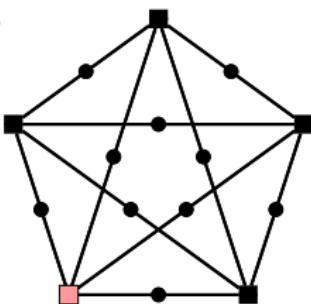
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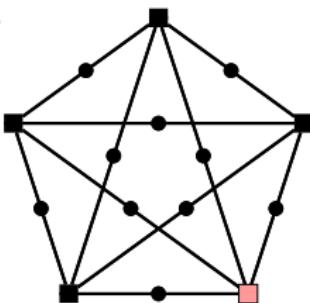
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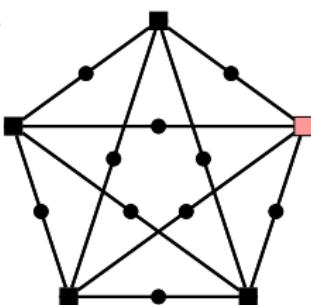
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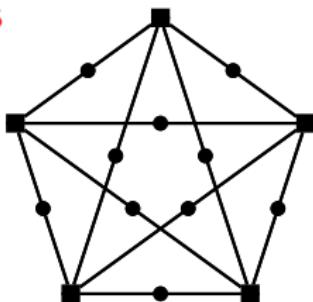
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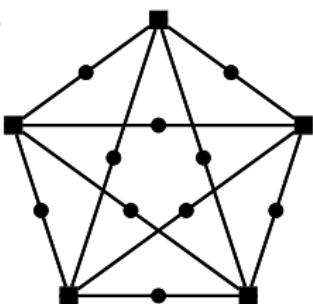
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0	$c_1$	$c_2$	$c_4$	$c_7$
$c_1$	0	$c_3$	$c_5$	$c_8$
$c_2$	$c_3$	0	$c_6$	$c_9$
$c_4$	$c_5$	$c_6$	0	$c_{10}$
$c_7$	$c_8$	$c_9$	$c_{10}$	0

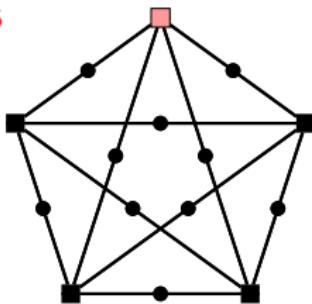
symmetric array

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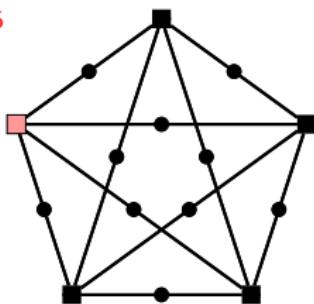
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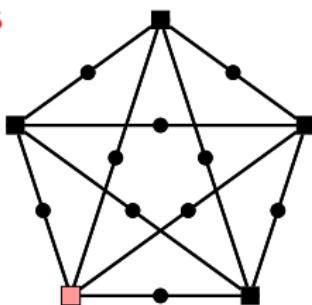
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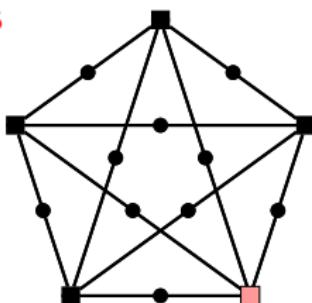
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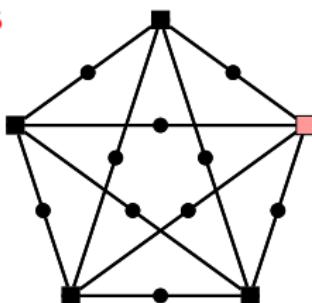
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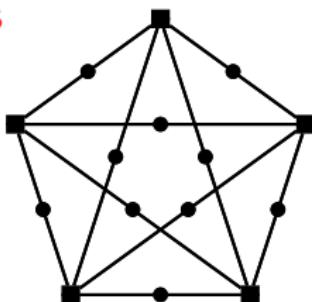
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$c_4$	$c_5$	$c_6$	0	$c_{10}$
$c_7$	$c_8$	$c_9$	$c_{10}$	0

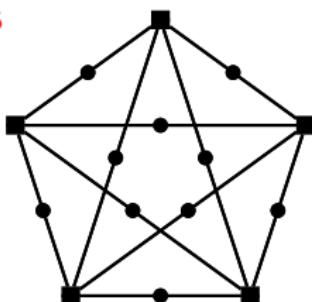
symmetric array

## Symmetric Generalized Product Codes

- So far,  $\eta_{i,i} = 0$  for all  $i \in \{1, 2, \dots, L\}$ . What about  $\eta_{i,i} = 1$ ?

Example:  $L = 1$ ,  $\eta = 1$ , and  $\gamma = 1$  gives a **half-product code** [Justesen, 2011]

$$n = 5 \implies d = 5$$



0	$c_1$	$c_2$	$c_4$	$c_7$
$c_1$	0	$c_3$	$c_5$	$c_8$
$c_2$	$c_3$	0	$c_6$	$c_9$
$c_4$	$c_5$	$c_6$	0	$c_{10}$
$c_7$	$c_8$	$c_9$	$c_{10}$	0

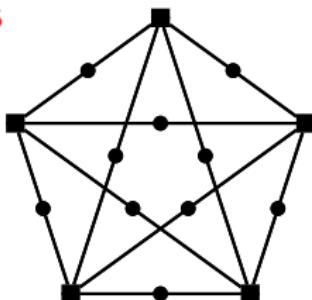
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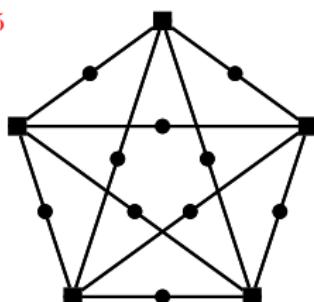
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- A **half-product code** has the **same threshold** as a product code, but less than **half the block length**

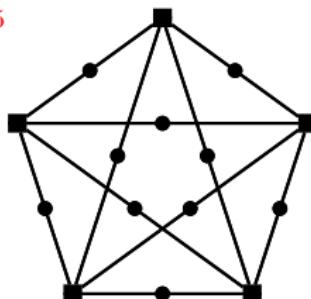
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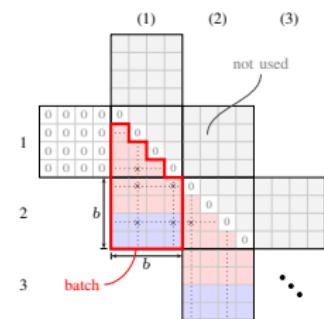
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- A **half-product code** has the **same threshold** as a product code, but less than **half the block length**
- Half-braided codes** can **outperform staircase and braided codes** in the waterfall region, at a lower error floor and decoding delay [Häger et al., 2016]



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Thank you!



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