Optimized Bit Mappings for Spatially Coupled LDPC Codes over Parallel Binary Erasure Channels

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System Model 000	Threshold and Optimization	Optimization Results 00000	Conclusions O	CHALMERS
		Motivation		
		parallel channels ← channel 1 ÷ ← channel m		

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		channel 1		

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Main question

How much gain is possible by optimizing the bit mapper compared to a uniformly random mapper in the asymptotic setting (infinite block length)?

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		Outline		

1. System Model

- 2. Decoding Threshold and Optimization
- 3. Results
- 4. Conclusions

System Model		
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AWGN channel using a labeled signal constellation:

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 $X \in \mathcal{X} \xrightarrow{N} Y$

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 $\varepsilon_1 \triangleq 1 - I(B_1; Y), \ \varepsilon_2 \triangleq 1 - I(B_2; Y)$ with average $\overline{\varepsilon} = (\varepsilon_1 + \varepsilon_2)/2$.

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System Model	Threshold and Optimization	Optimization Results	Conclusions	CHALMERS
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• Two-sided and circular spatially coupled (d_v, d_c, L, w) code ensembles

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- Design rate $R = 1 d_v/d_c R_{loss}(L)$, where $R_{loss}(L) \to 0$ as $L \to \infty$
- For circular ensemble, $R = 1 d_v/d_c$ (no rate loss)

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Bit Mapper







• VNs (i.e., code bits) at different positions belong to different equivalence classes



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- Resulting VN erasure probablities $(\varepsilon^1, \dots, \varepsilon^L) = (\varepsilon_1, \dots, \varepsilon_m) \cdot \mathbf{A}$





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		Bit Mapper		
			$\begin{array}{c} & \\ & \\ & \\ & \\ & \\ & \\ & \\ & \\ & \\ & $	
• Exa	mple: $\mathbf{A} = \begin{pmatrix} 1.0 & 0.0 \\ 0.0 & 1.0 \end{pmatrix}$	$\begin{array}{cccc} 0.5 & 0.75 & 0.25 \\ 0.5 & 0.25 & 0.75 \end{array}$)	





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• Set of valid assignment matrices $\mathcal{A}^{m \times L}$: columns sum to 1 (all VNs are assigned), rows sum to L/m (all channels are used equally often)



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- Baseline bit mapper \mathbf{A}_{uni} with $a_{i,j} = 1/m$, $\forall i, j$

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Density Evolution for BECs

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Density Evolution for BECs



















decoding failure

For BECs, density evolution is simple: Track the evolution of the VN erasure probabilities at all positions j.

0.2 0.4 0.6 0.8 1.0

ω 0.5 0.4 0.3



 $0.2 \ 0.4 \ 0.6 \ 0.8 \ 1.0$ $\bar{\varepsilon}$ decoding failure

$$p_j^{(l)} = \varepsilon^j \left(\frac{1}{w} \sum_{a=0}^w \left(1 - \left(1 - \frac{1}{w} \sum_{b=0}^w p_{j+a-b}^{(l-1)} \right)^{d_{\mathsf{c}}-1} \right) \right)^{d_{\mathsf{v}}-1}$$

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initial VN erasure probability -

$$p_{j}^{(l)} = \varepsilon^{j} \left(\frac{1}{w} \sum_{a=0}^{w} \left(1 - \left(1 - \frac{1}{w} \sum_{b=0}^{w} p_{j+a-b}^{(l-1)} \right)^{d_{c}-1} \right) \right)^{d_{c}-1}$$



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• $\bar{\varepsilon}^*(\mathbf{A}) \triangleq \text{largest } \bar{\varepsilon} \in [0,1] \text{ such that } \lim_{l \to \infty} p_j^{(l)} \to 0, \, \forall j$

 $0.4 \\ 0.3$

 $\begin{array}{c}
 0.2 \\
 0.1 \\
 0
\end{array}$

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Example for Baseline Bit Mapper $\mathbf{A}_{\mathsf{uni}}$

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Threshold and Optimization		
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Threshold and Optimization	Optimization Results 00000	CHALMERS

• Ideally, we would like to solve the problem

 $\label{eq:argmax} \mathbf{A}_{\mathsf{opt}} = \underset{\mathbf{A} \in \mathcal{A}^{m \times L}}{\operatorname{argmax}} \quad \bar{\varepsilon}^*(\mathbf{A}).$

Threshold and Optimization ○○●	Optimization Results 00000	CHALMERS

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Threshold and Optimization ○○●	Optimization Results 00000	CHALMERS

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Threshold and Optimization	Optimization Results	CHAI MERS
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- Significantly reduced computational complexity, however, $\mathbf{A}_{\text{opt}} \neq \mathbf{A}^{\!*}$ in general

System Model 000	Threshold and Optimization	Optimization Results ●0000	Conclusions O	CHALMERS

	Optimization Results •0000	CHALMERS













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	Optimization Results •0000	CHALMERS

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System Model	Threshold and Optimization	Optimization Results	Conclusions	CHALMERS
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	Optimization Resul	ts Conclusion O	

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	Optimization Results	CHALMERS

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• Different channel qualities can be exploited to create a boundary-like termination effect



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		Optimization Results	
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BICM Verification

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BICM Verification

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BIC	M Verification	

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Thank you!