

On the Typicality of the Linear Code Among the LDPC Coset Code Ensemble

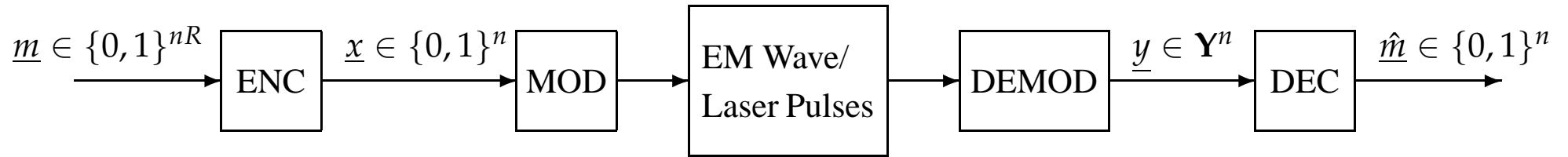
Chih-Chun Wang, Prof. S.R. Kulkarni, and Prof. H.V. Poor

{chihw, kulkarni, poor}@princeton.edu.

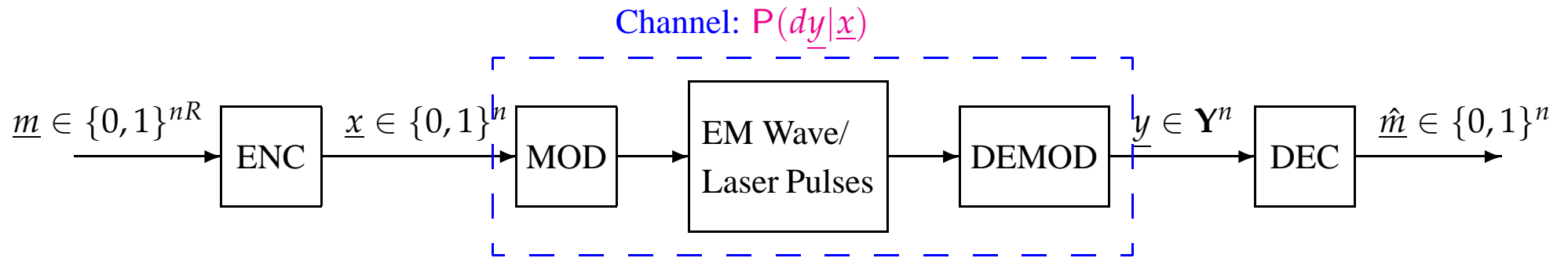
Princeton University



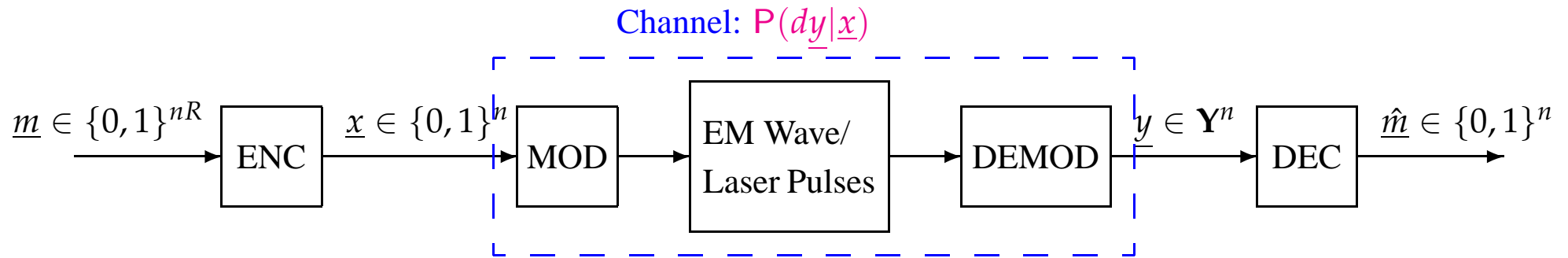
Error Correcting Codes



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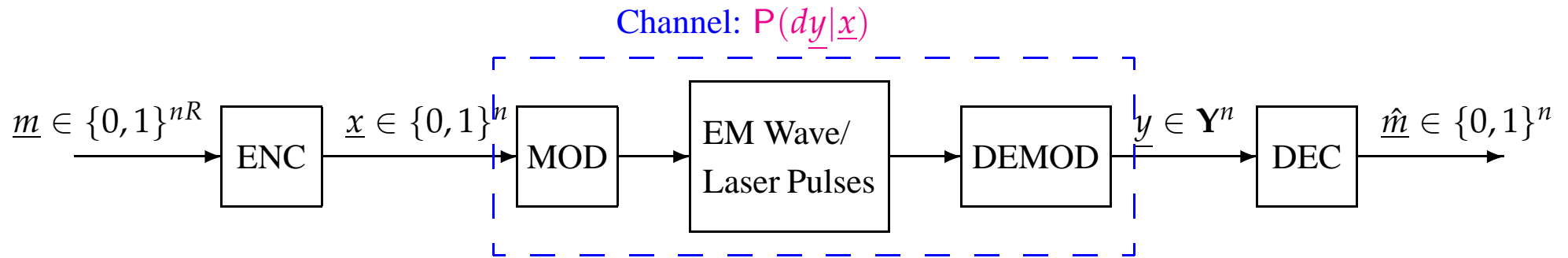
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- Memoryless channels: $P(dy|\underline{x}) = \prod_{i=1}^n P(dy_i|x_i)$



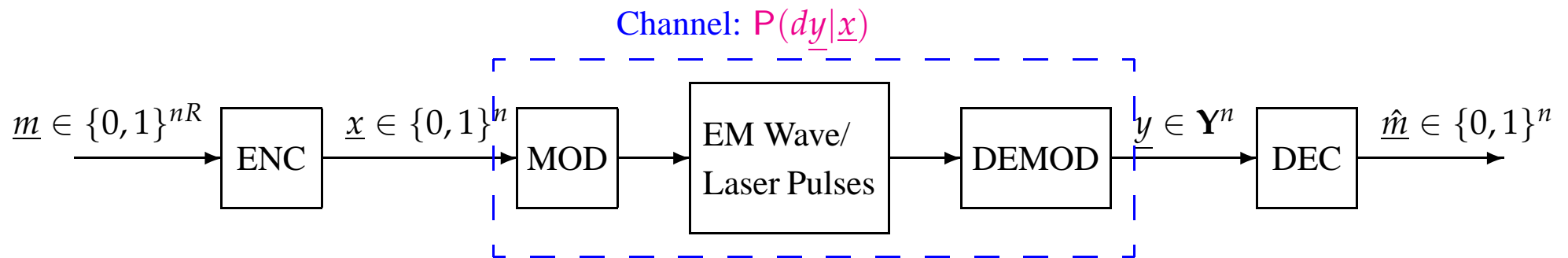
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- Symmetric channels: $\exists \mathcal{T} : \mathbf{Y} \mapsto \mathbf{Y}$ s.t. $\mathcal{T}^2(y) = y, \forall y \in \mathbf{Y}$, and $P(dy|x=0) = P(\mathcal{T}(dy)|x=1)$.



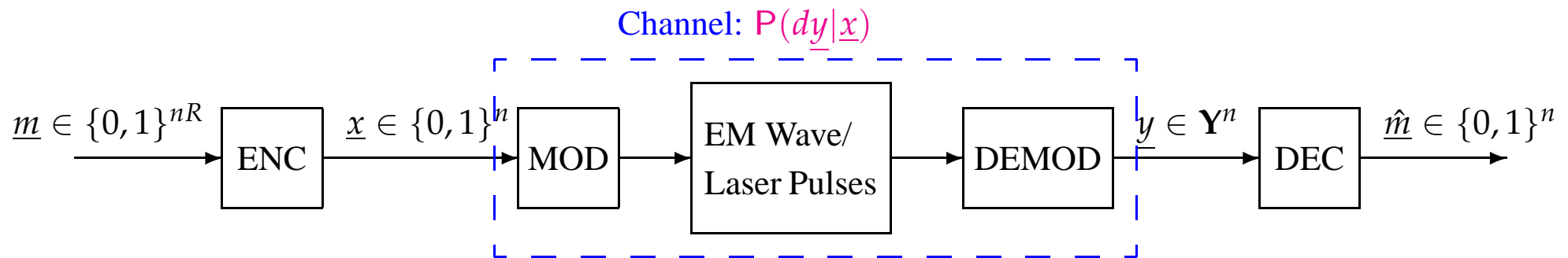
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- Memoryless symmetric channels: Capacity-approaching error correcting codes have been constructed, including turbo codes, low-density parity-check (LDPC) codes, irregular RA codes, LT codes, concatenated tree codes, etc.
- Performance: 0.1~1.5dB away from capacity.



Ultra high performance on almost all symmetric channels.



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Question: how about **non-symmetric** memoryless channels?

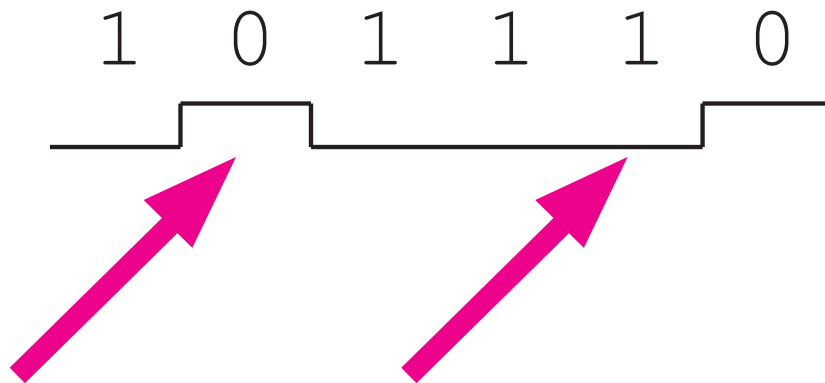


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Examples:

Z-Channels

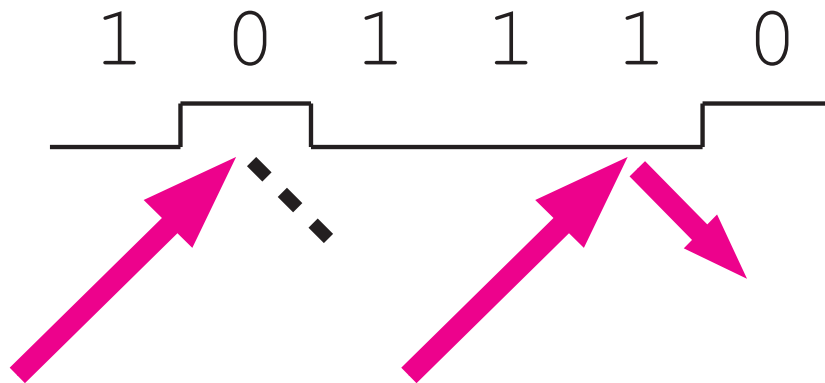


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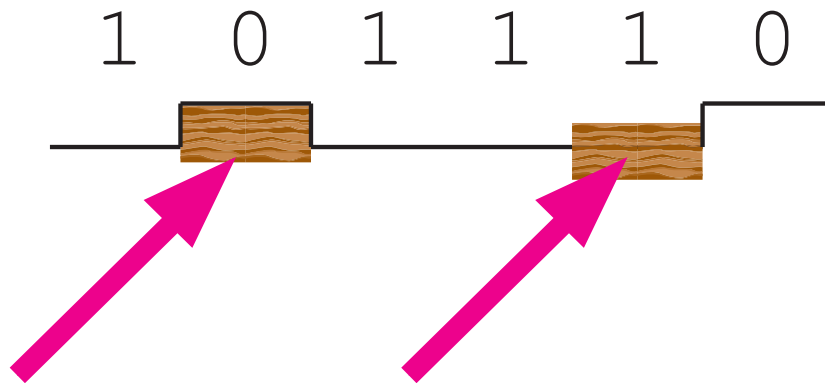


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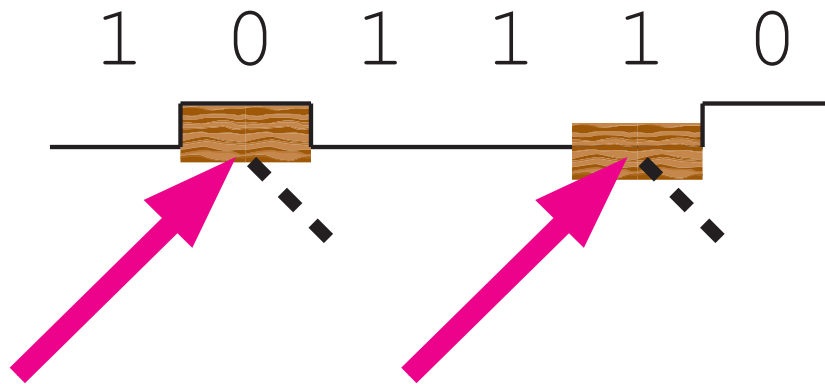
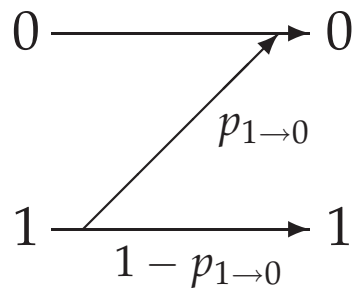


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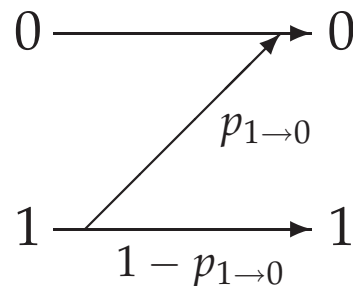


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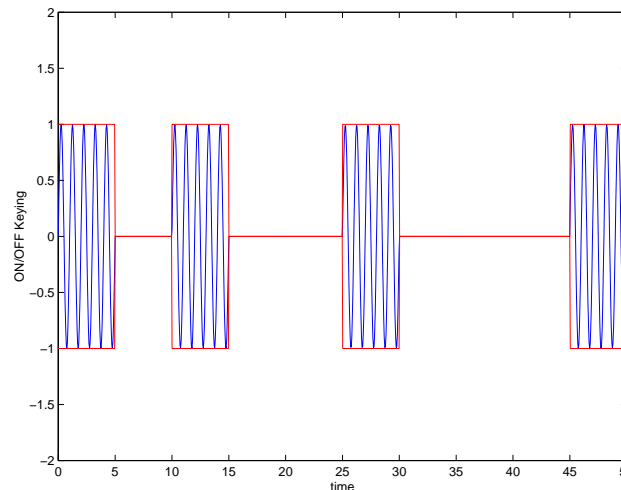
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Examples:

Z-Channels



On/Off Keying w. Rayleigh Fading

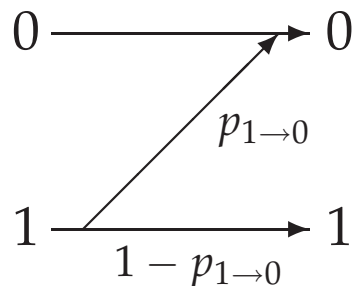


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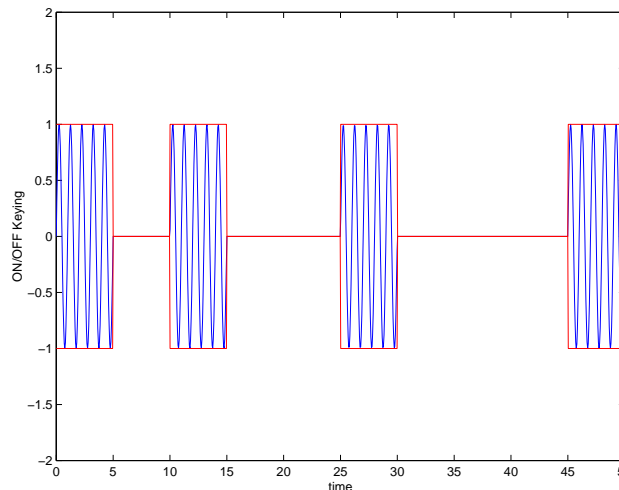
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BICM



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Exam Z-Ch

0

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time

[Majani & Rumsey 91] showed that the ratio between the symmetric mutual information rate and the capacity is lower bounded by $\frac{e \ln 2}{2} \approx 0.942$.

[Shulman & Feder 04] further proved that the absolute difference is upper bounded by 0.011 bit/sym.

M

1

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Content

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 - Codeword-dependent error resiliency



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LDPC Codes & the Graph-based Code Ensemble

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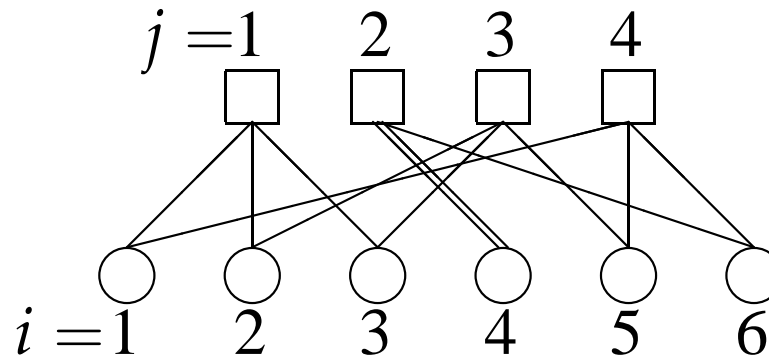
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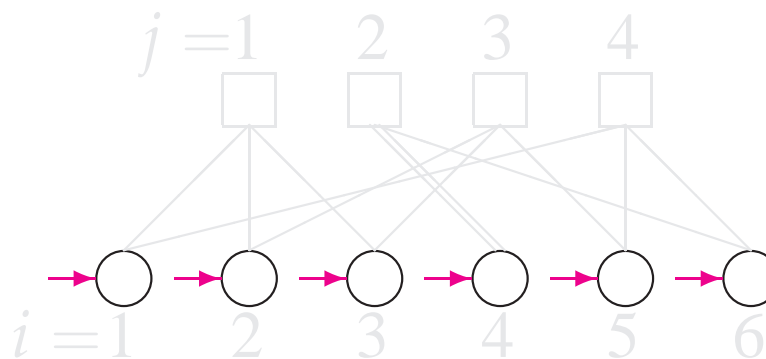
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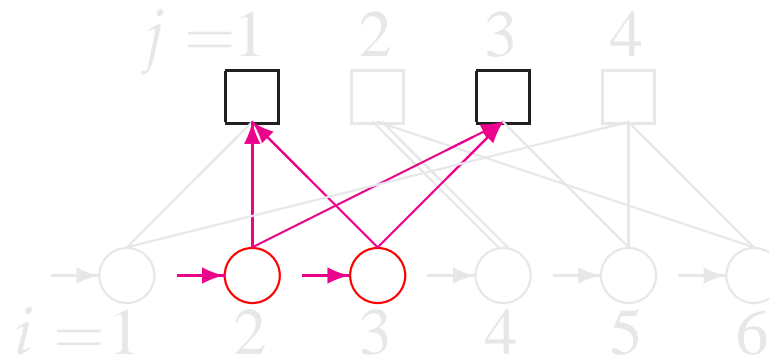
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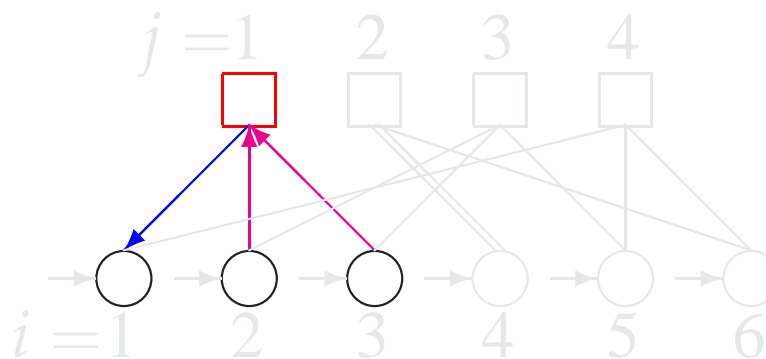
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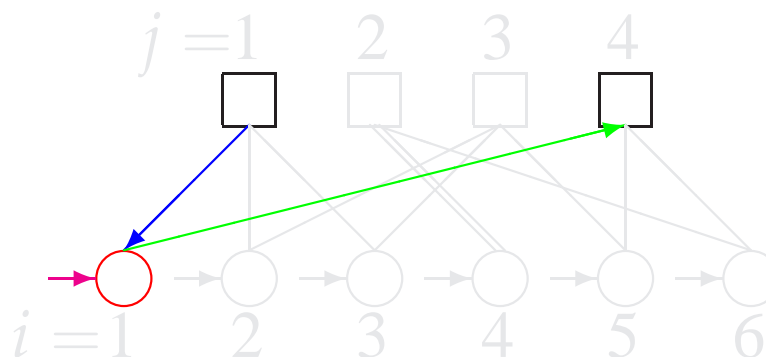
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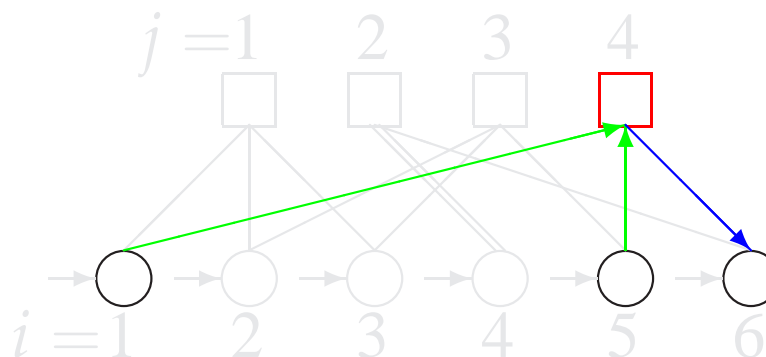
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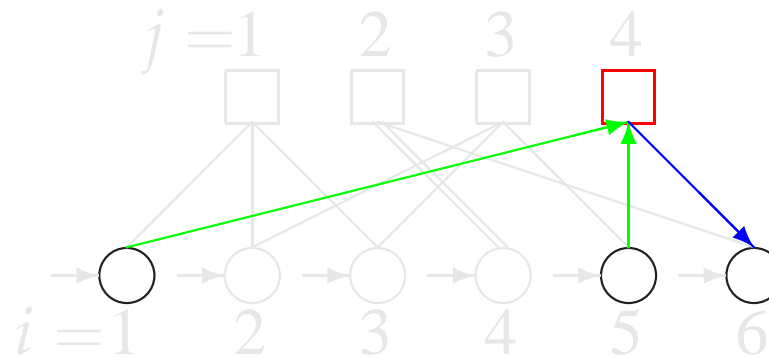
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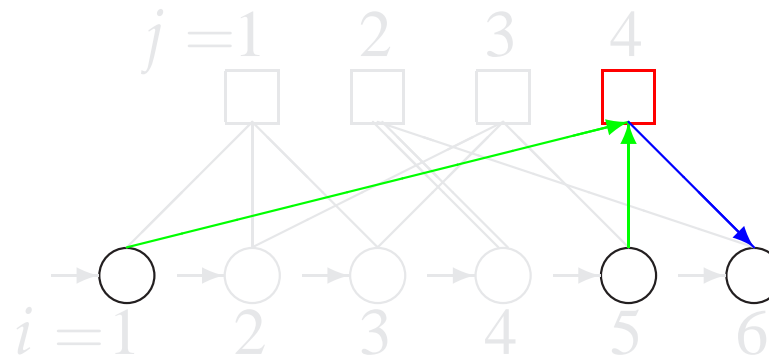
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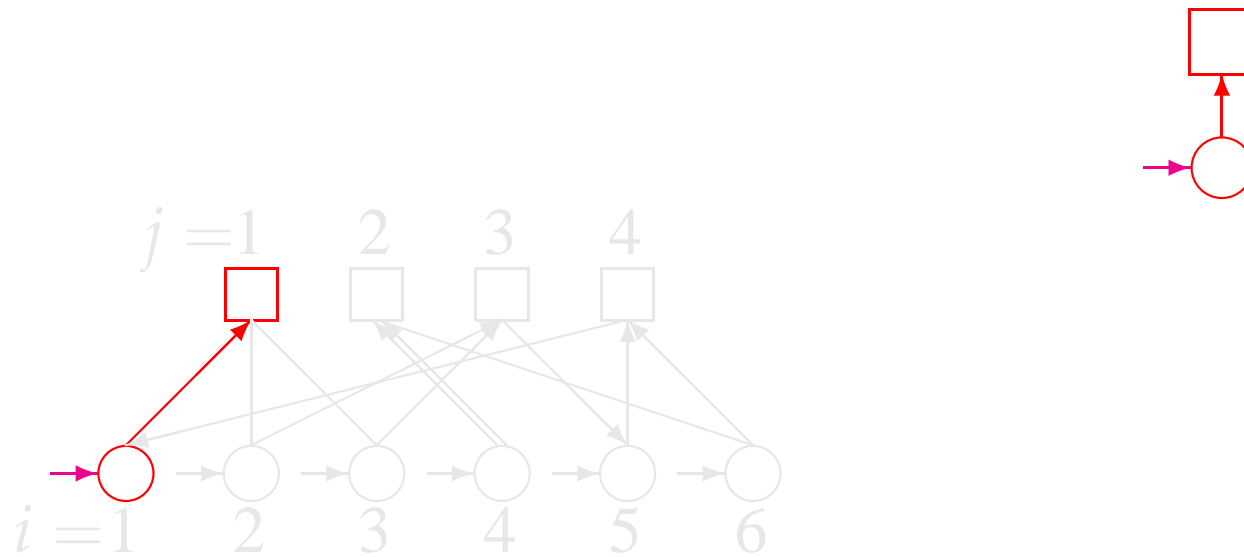
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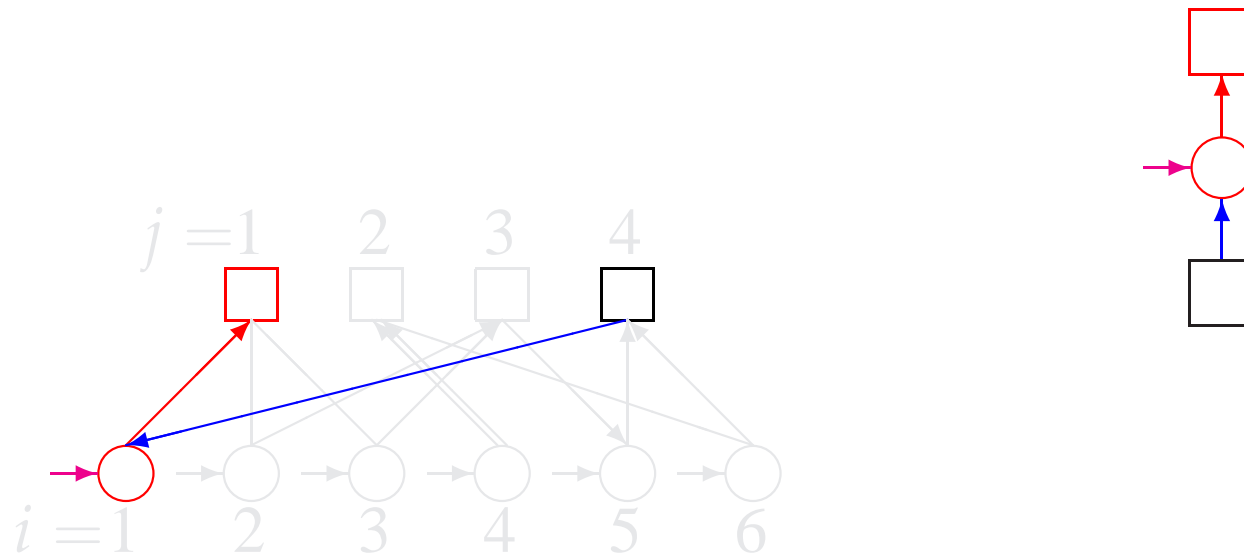
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- By simulation, belief propagation + LDPC codes also have outstanding performance for non-symmetric channels.



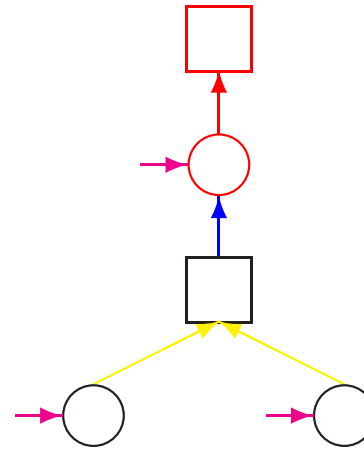
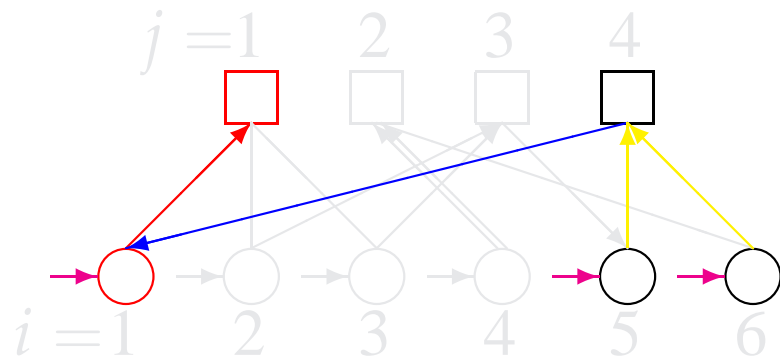
The Density Evolution



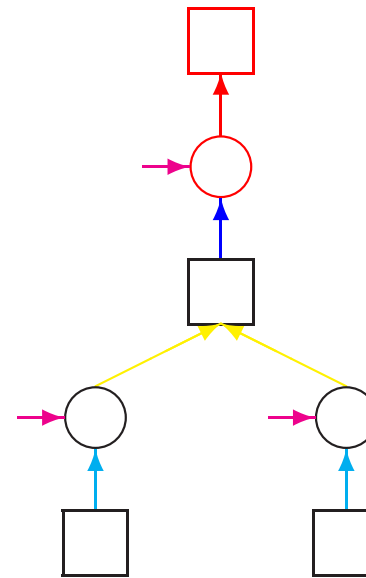
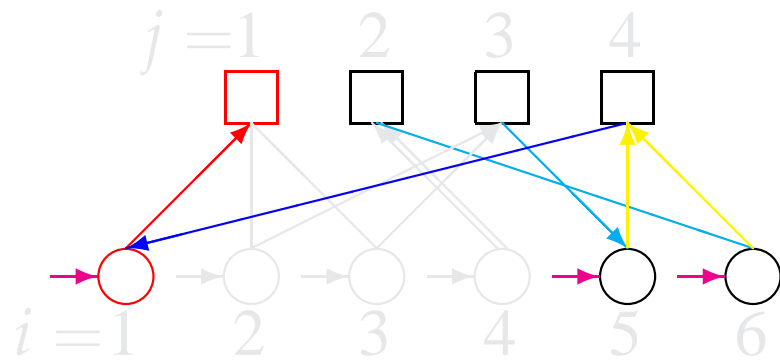
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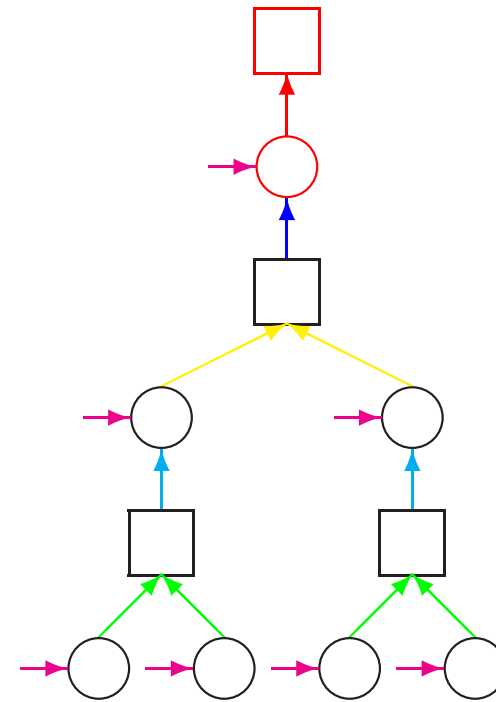
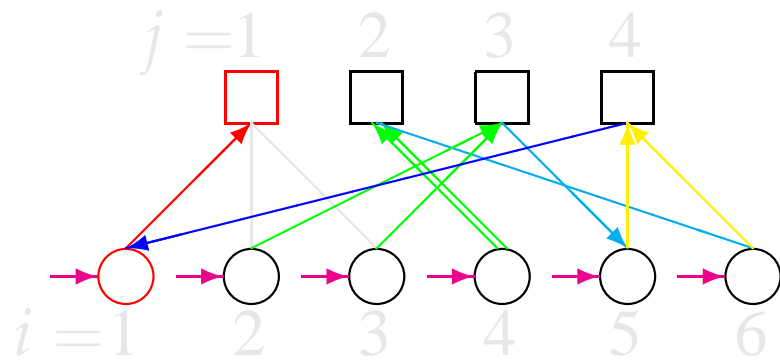
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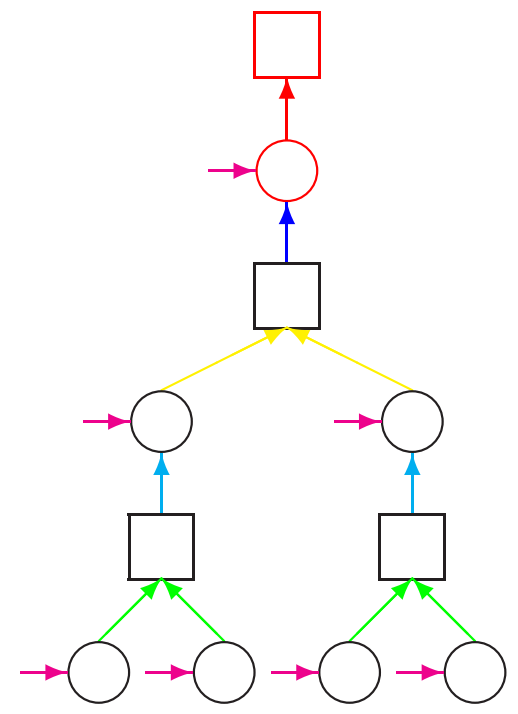
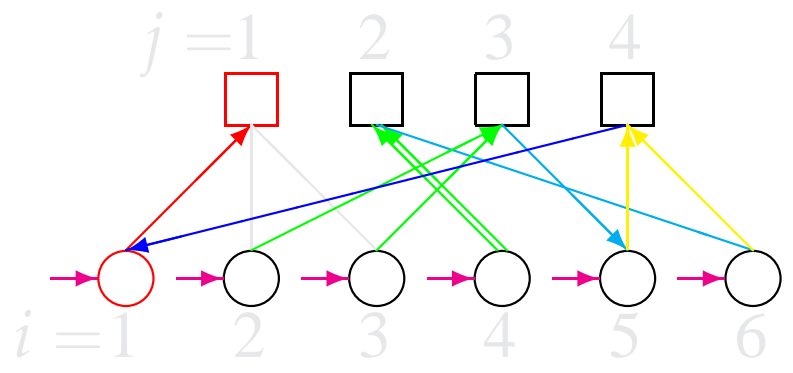


The Density Evolution



The Density Evolution

Sym. Chs: Assuming $\mathbf{x} = \mathbf{0}$.



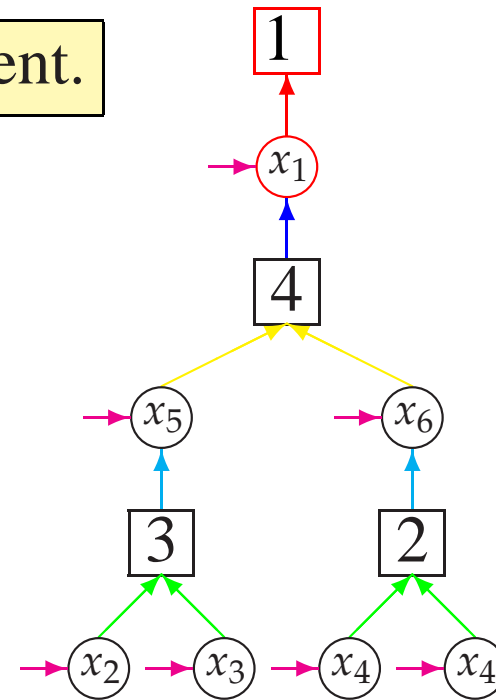
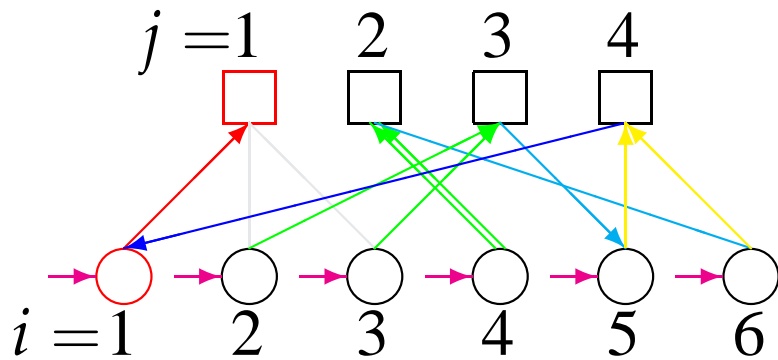
$$P^{(l)} = P^{(0)} \otimes \left(Q^{(l-1)} \right)^{\otimes (d_v - 1)}$$

$$Q^{(l-1)} = \Gamma^{-1} \left(\left(\Gamma \left(P^{(l-1)} \right) \right)^{\otimes (d_c - 1)} \right),$$



The Density Evolution

Non-sym. Chs: Codeword-dependent.



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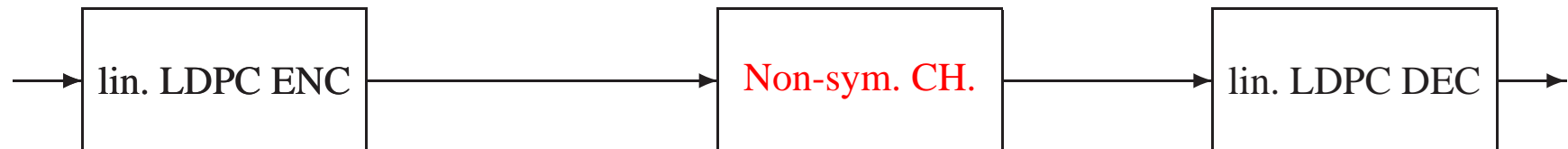
First Approach: the LDPC Coset Ensemble

- **Coset codes:** Any valid codeword \mathbf{x} satisfies $\mathbf{Ax} = \mathbf{s}$, where \mathbf{A} is from the same equiprobable bipartite graph ensemble, and the coset-defining syndrome \mathbf{s} is uniformly drawn from $\{0, 1\}^{n(1-R)}$.



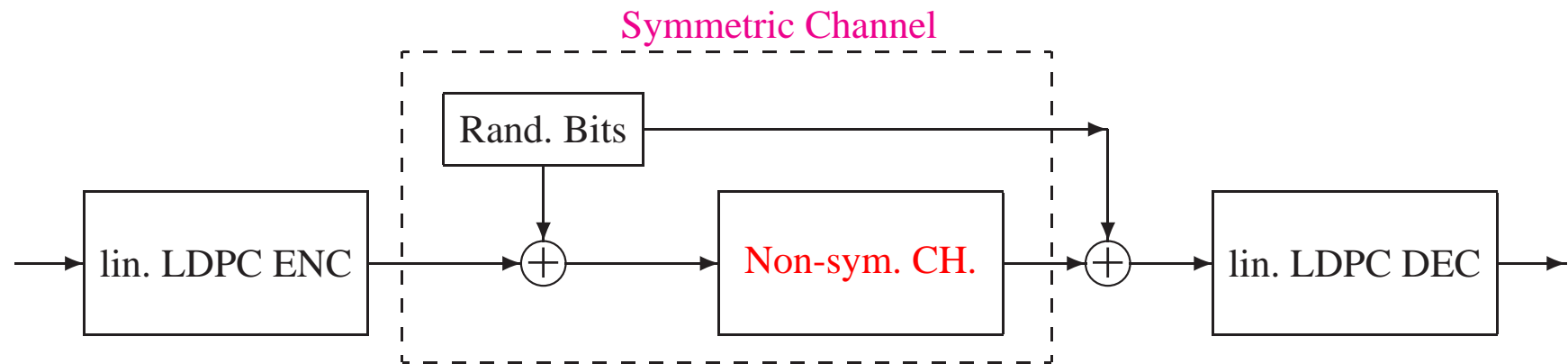
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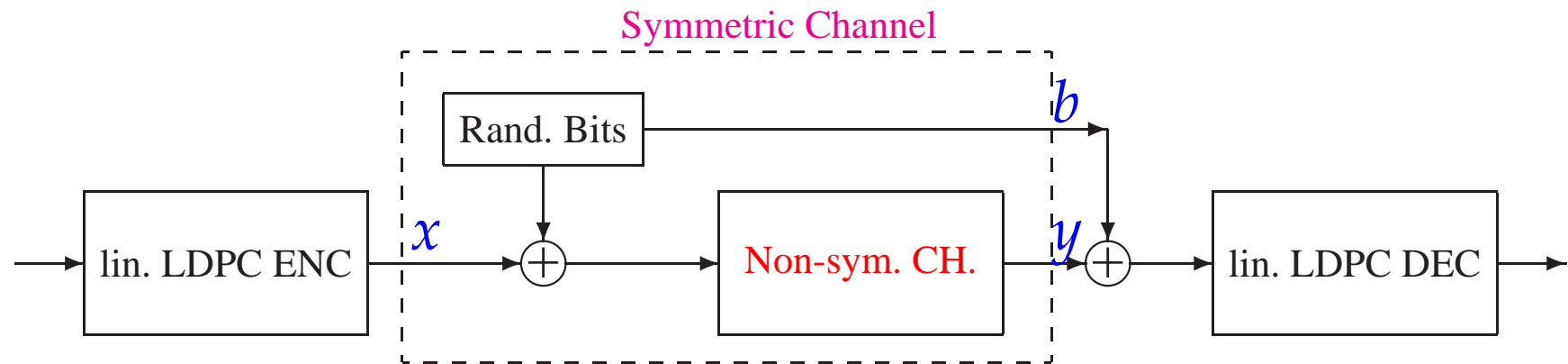
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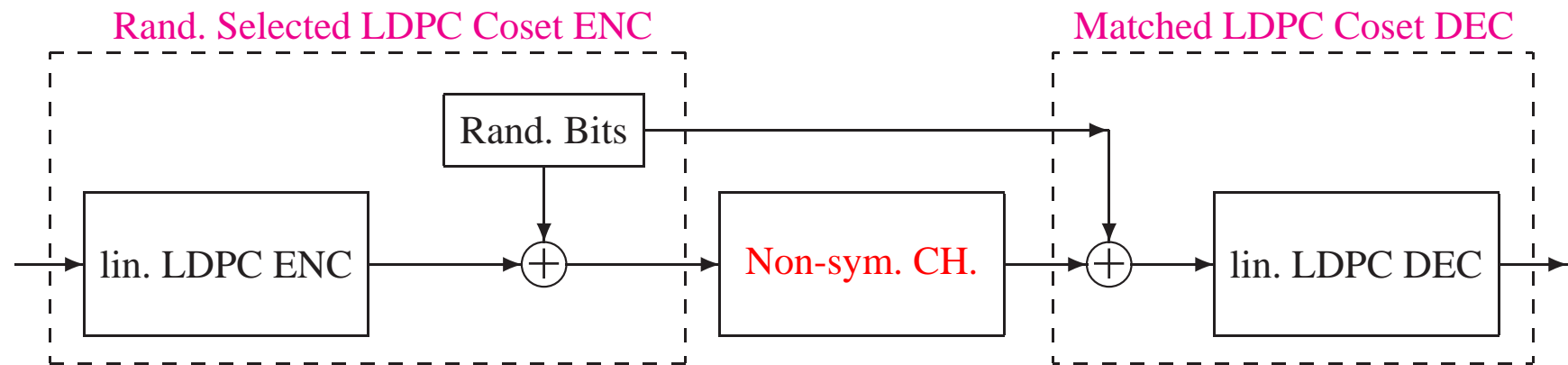


$$P(b, y|x = 0) = P(b - 1, y|x = 1) = P(\mathcal{T}(b, y)|x = 1).$$



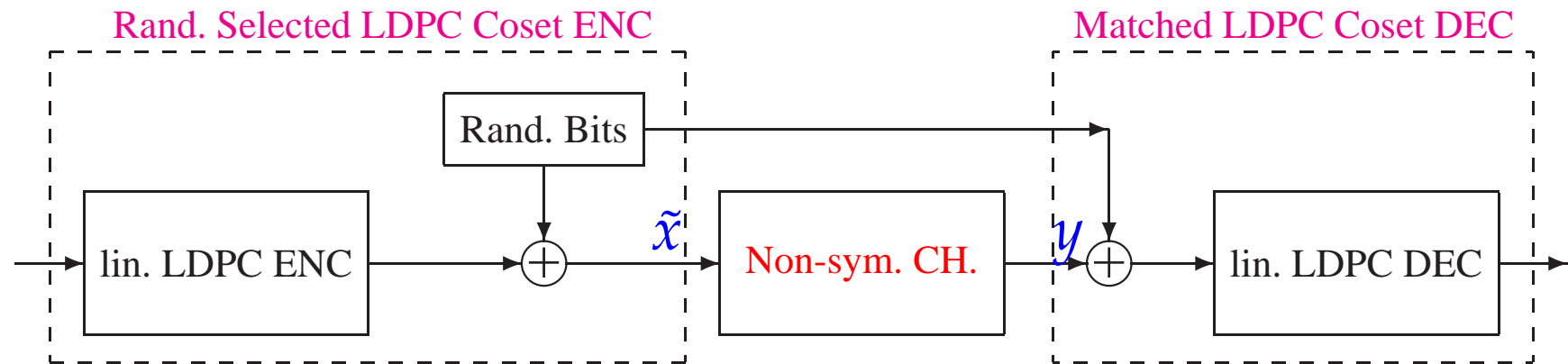
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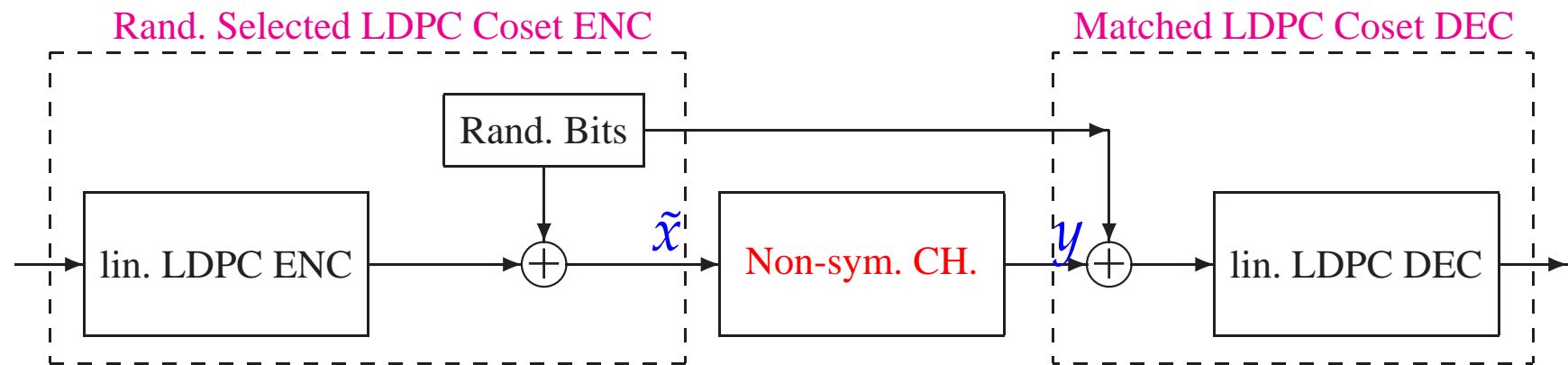


$$\mathbf{A}\tilde{\mathbf{x}} = \mathbf{A}(\mathbf{x} + \mathbf{b}) = \mathbf{Ax} + \mathbf{Ab} = \mathbf{Ab} = \mathbf{s}$$



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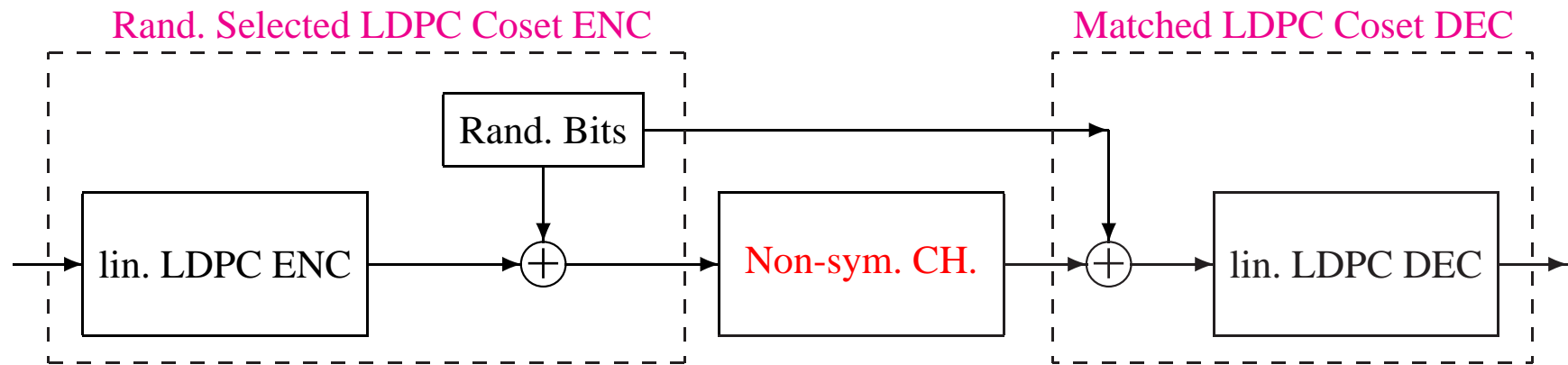


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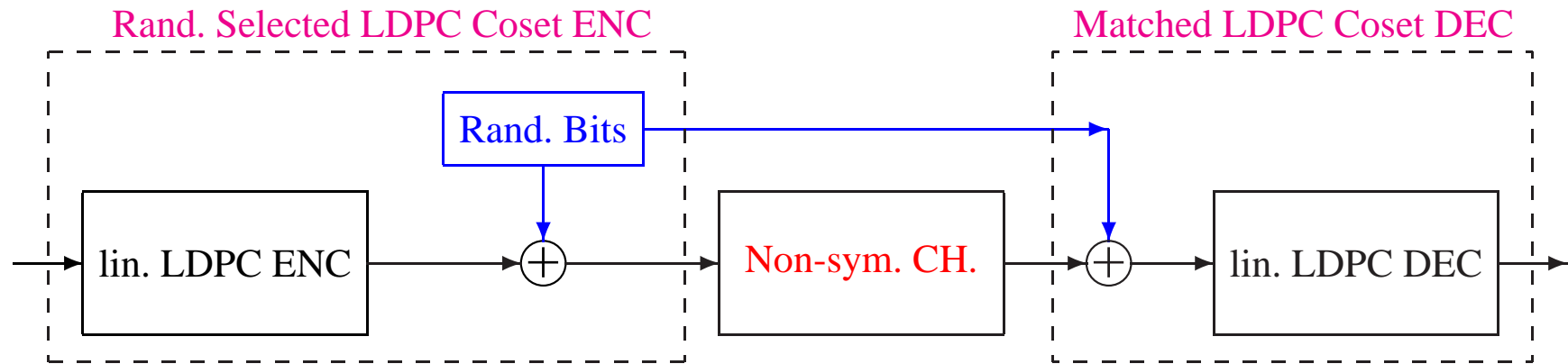
- Averaged performance of the LDPC coset code ensemble



The Existing Typicality Result



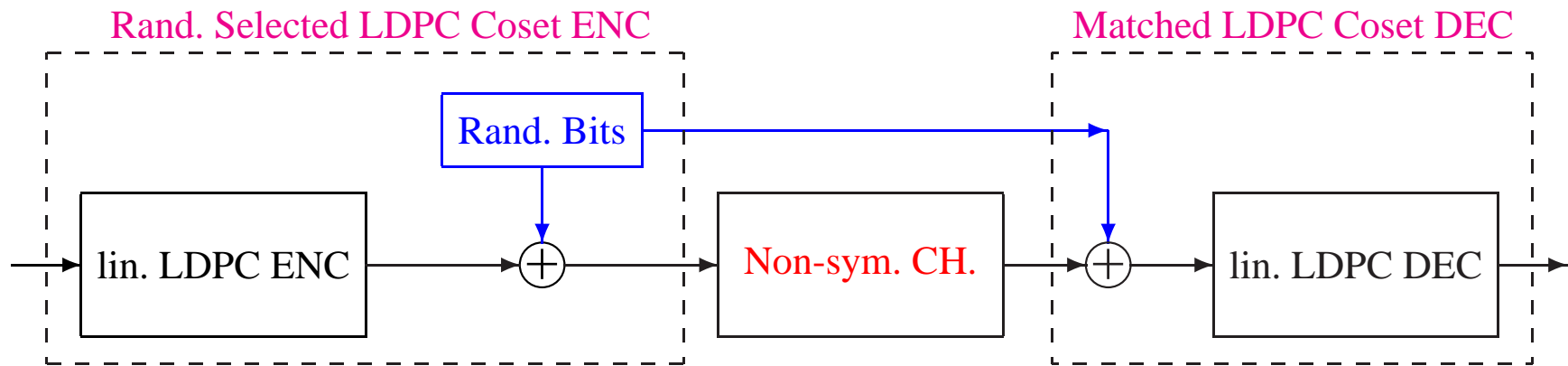
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- The difficulty: Maintaining the synchronization of two random strings.



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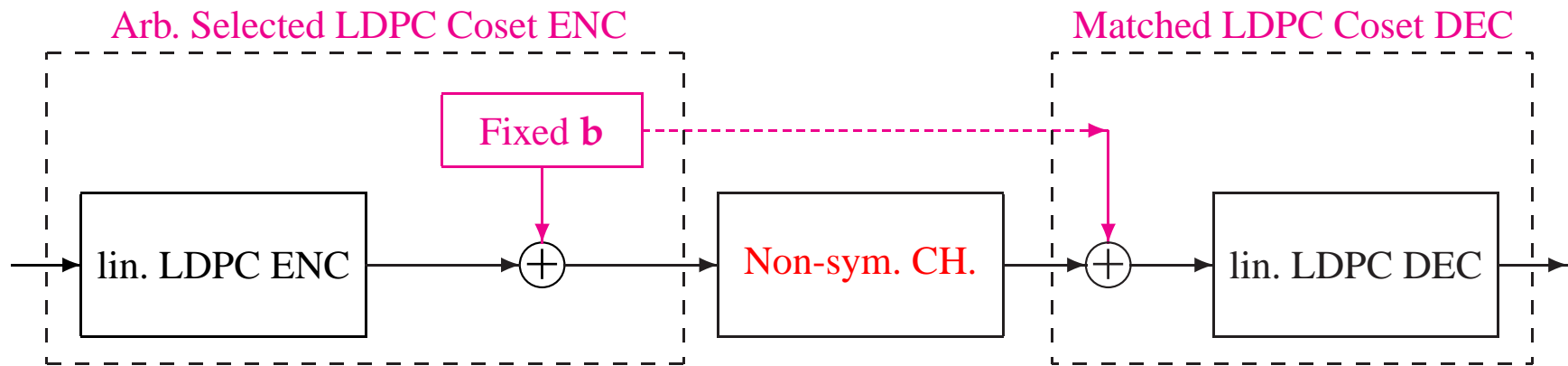


- The difficulty: Maintaining the synchronization of two random strings.
- [Kavčić 03]: Asymptotically, almost all $\mathbf{s} \in \{0, 1\}^{n(1-R)}$ are typical. Namely,

$$\lim_{n \rightarrow \infty} P(\mathbf{s} : |p_e(\mathbf{s}) - E_{\mathbf{s}}\{p_e(\mathbf{s})\}| < \epsilon) = 1, \quad \forall \epsilon > 0.$$



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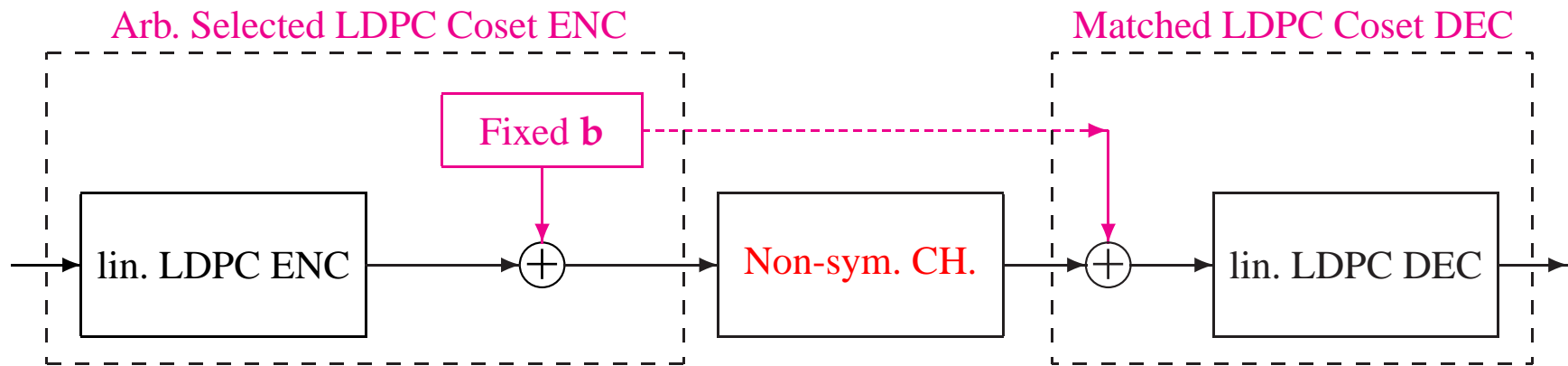


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The Existing Typicality Result



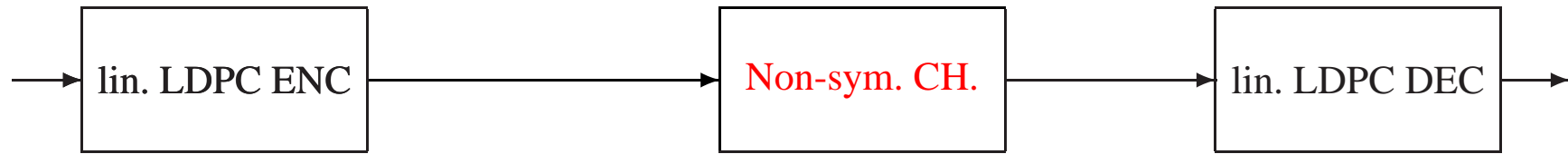
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- Due to their **hardware uniformity**, linear codes are always the superior choice.



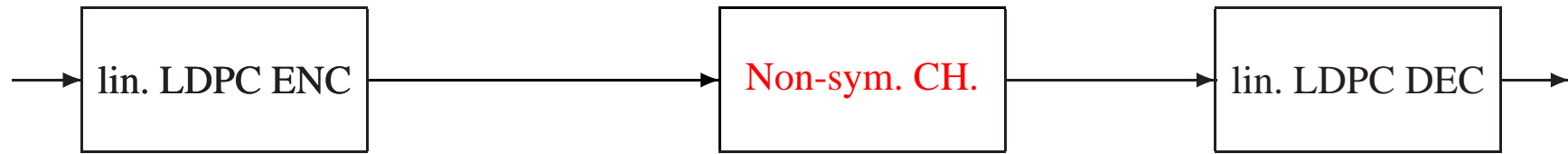
Second Approach: Linear Codes w. Codeword Averaging



- Codeword-dependent performance



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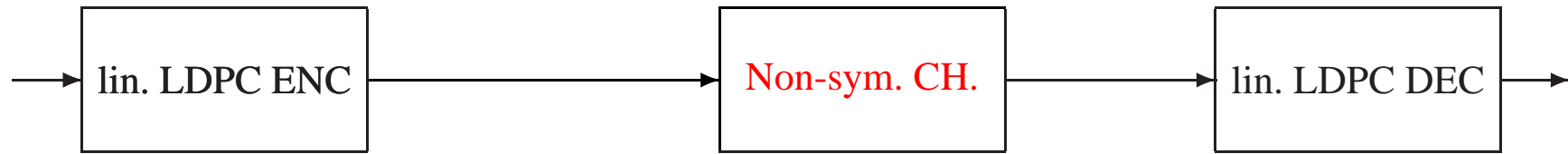


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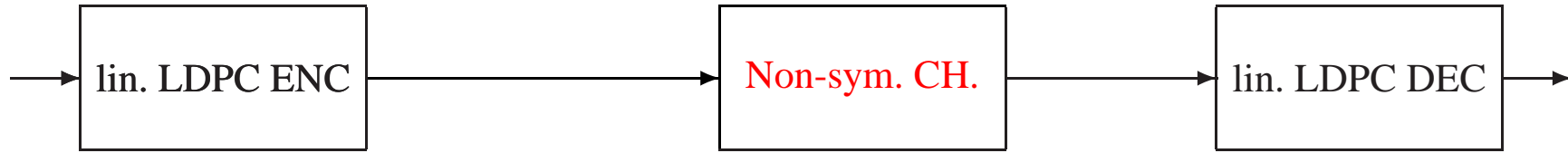


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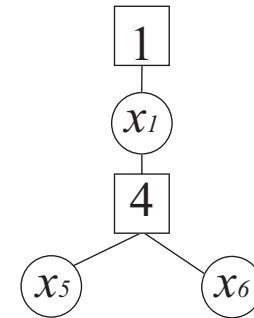
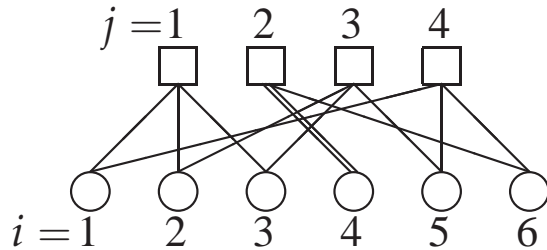


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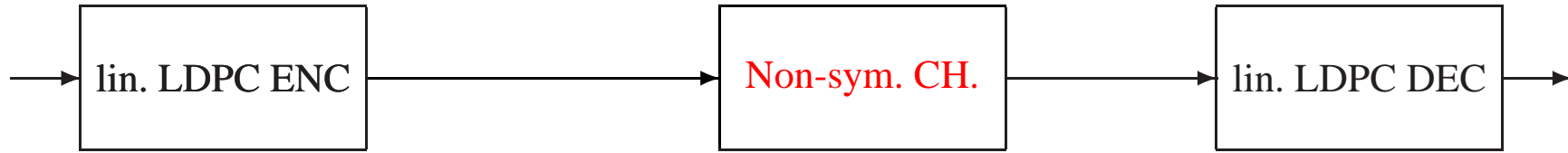


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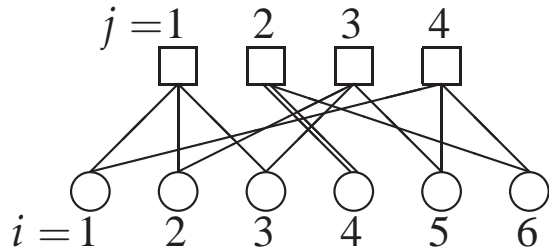


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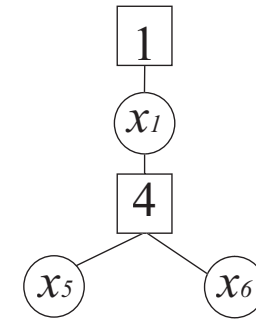
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$$\mathbf{A} = \begin{pmatrix} 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 & 1 \end{pmatrix}$$

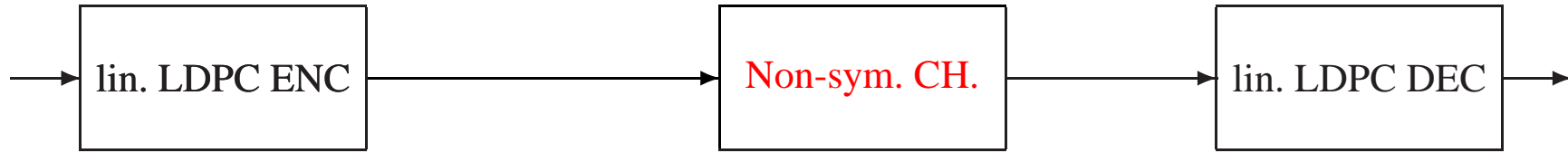
$$\mathbf{Ax} = \mathbf{0} \iff x_6 = 0$$



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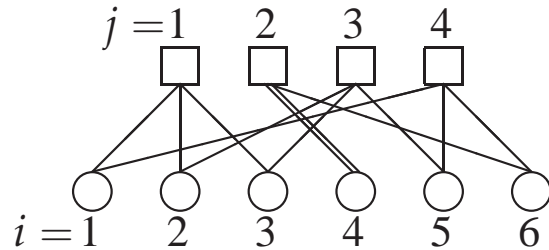


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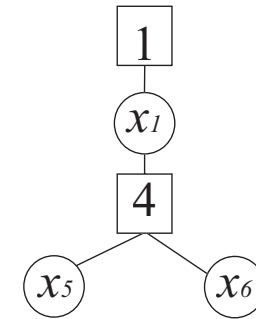
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Averaging the trimmed tree code is not equivalent to averaging the original code.



Perfect Projection Condition

Definition 1 (Perfect Projection) *The supporting tree \mathcal{N}^{2l} is perfectly projected, if for any codeword \mathbf{x}_t of the tree code \mathbf{X}_t ,*

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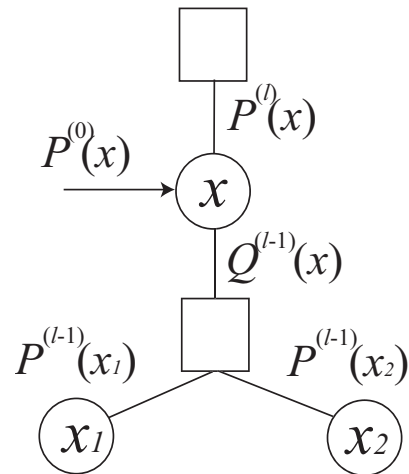
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We then have

$$P^{(l)}(x) := \left\langle P^{(l)}(\mathbf{x}) \right\rangle_{\{\mathbf{x} \in \mathbf{X} : \mathbf{x}|_0 = x\}} = \left\langle P^{(l)}(\mathbf{x}_t) \right\rangle_{\{\mathbf{x}_t \in \mathbf{X}_t : \mathbf{x}_t|_0 = x\}},$$
$$p_e^{(l)} = \frac{1}{2} \left(\int_{m=-\infty}^0 P^{(l)}(\mathbf{0})(dm) + \int_{m=-\infty}^0 P^{(l)}(\mathbf{1})(dm) \right).$$



New Iterative Formula for DE



$$x = 0, \quad x_1 x_2 = 00, 11$$

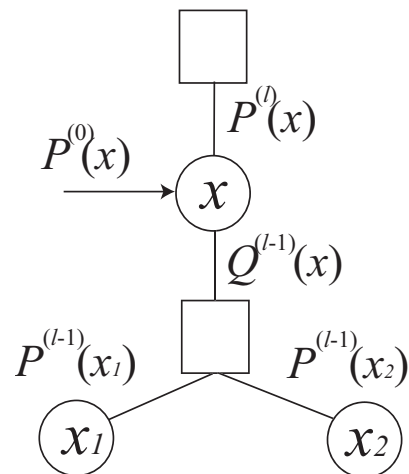
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Similarities

- **Symmetry:**
- **Monotonicity:**
- **Stability Conditions:**



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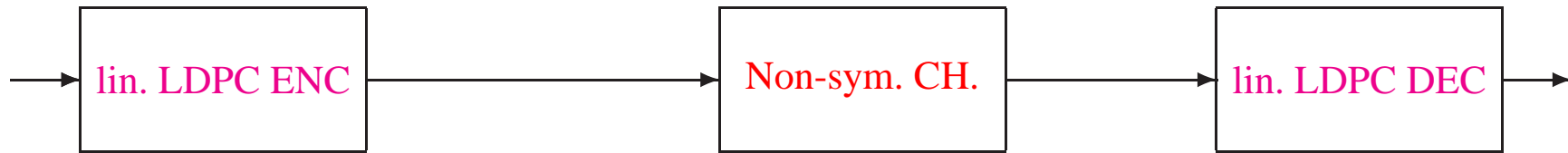
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- Having the same stability conditions: $\langle CB^{(0)} \rangle < \frac{1}{\lambda_2 \rho'(1)}$.

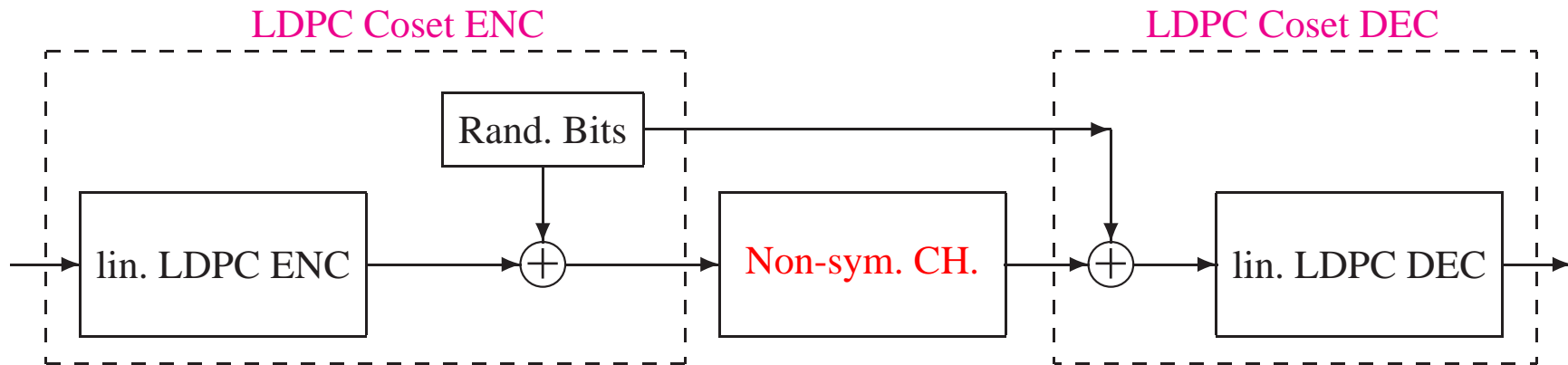


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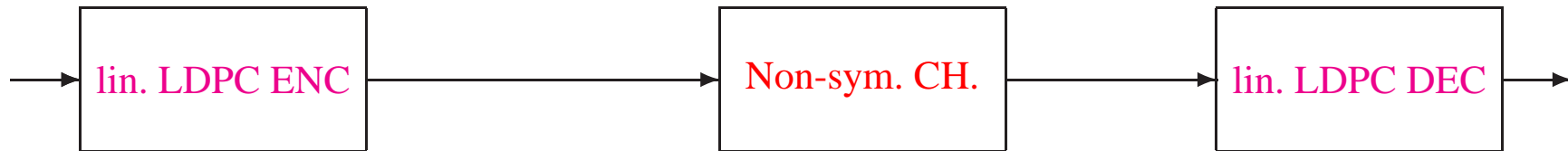


LDPC coset code ensemble: $\mathbf{Ax} = \mathbf{s}$ and $\mathbf{s} \in_{\text{rand.}} \{0, 1\}^{n(1-R)}$.

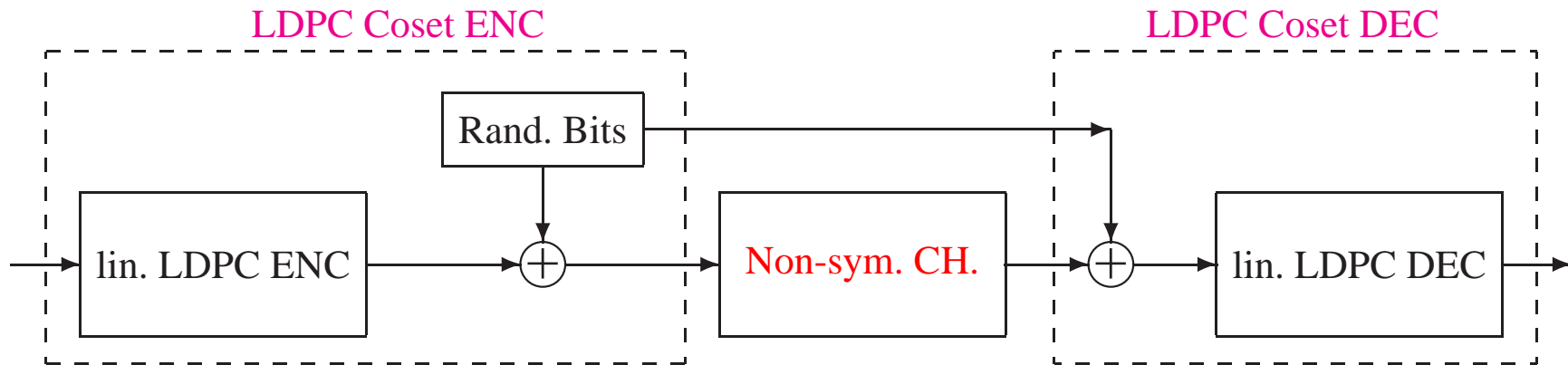


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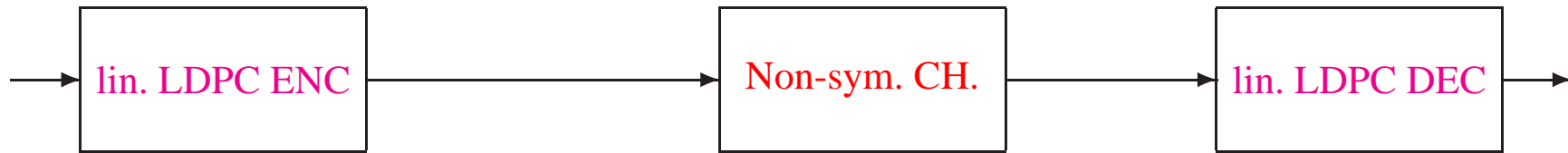
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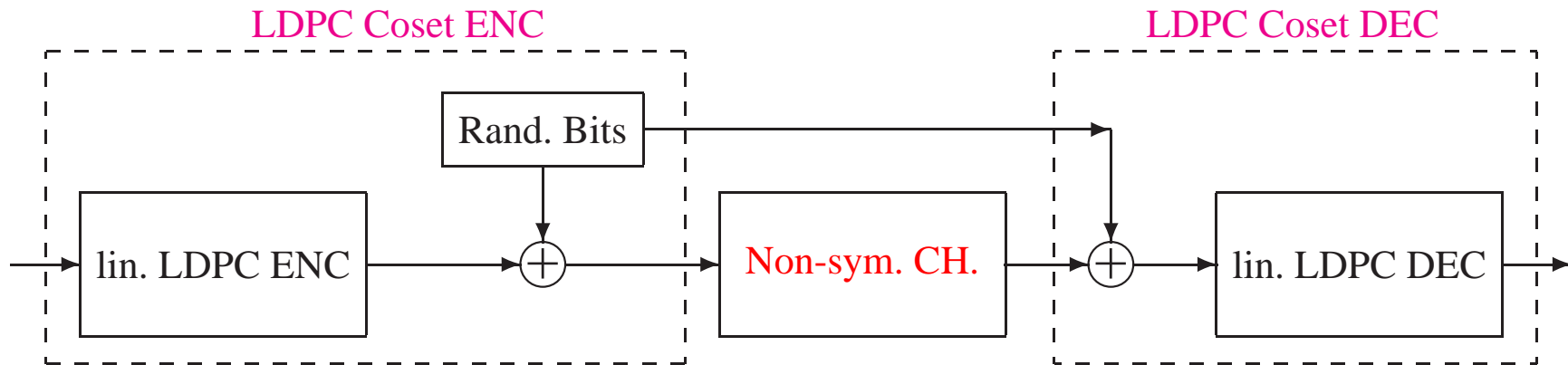


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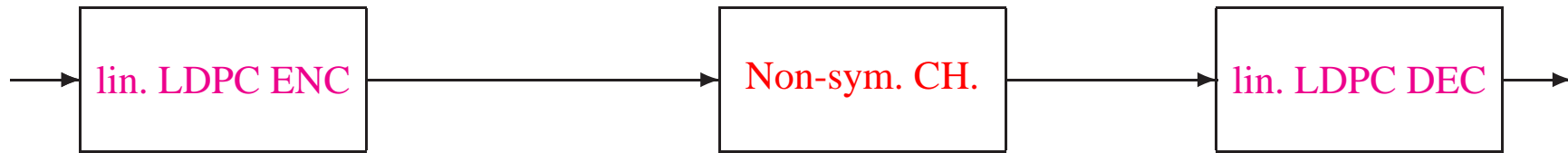
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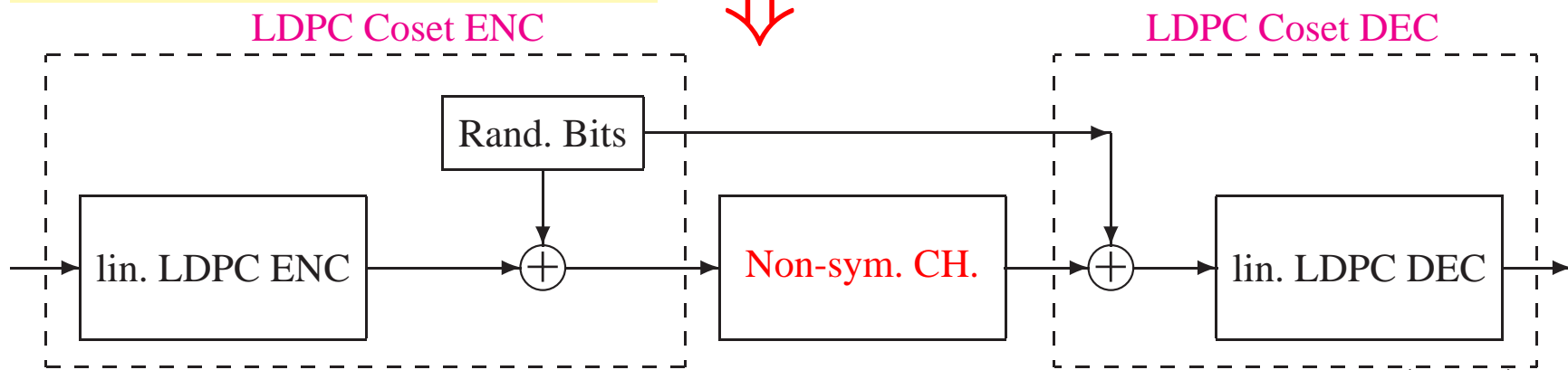


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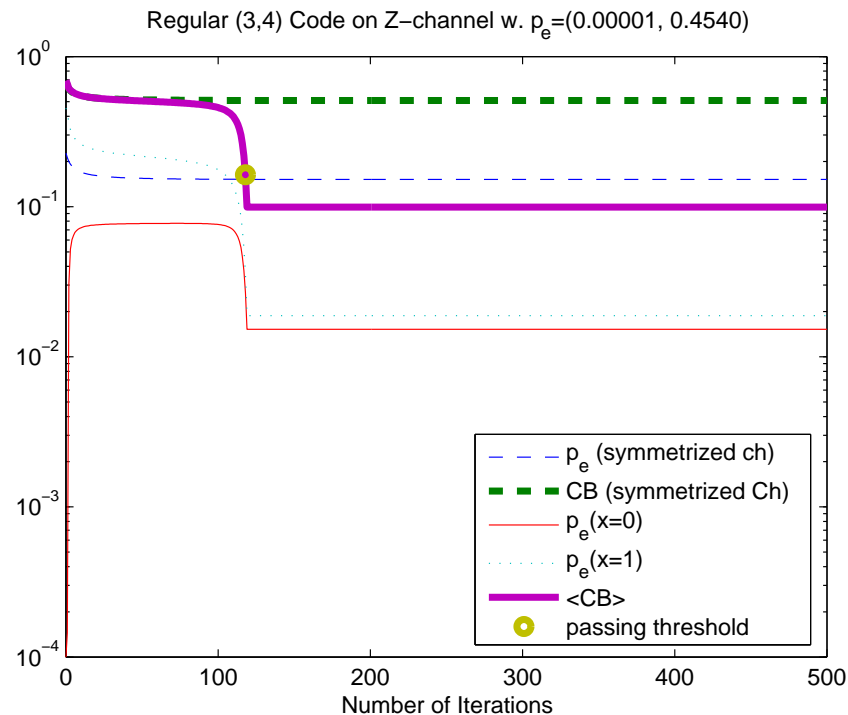


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Typicality of Linear LDPC Codes

(λ, ρ)	(x^2, x^3)	(x^2, x^5)	$(x^2, 0.5x^2 + 0.5x^3)$	$(x^2, 0.5x^4 + 0.5x^5)$
Linear	0.4540	0.2305	0.5888	0.2689
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Theorem 1 (Main Theorem) Consider *non-symmetric channels* and a fixed pair of degree polynomials λ and ρ . The shifted check node polynomial is denoted by $\rho_\Delta = x^\Delta \cdot \rho$. Let $P_{linear}^{(l)}$ and $P_{coset}^{(l)}$ denote the evolved densities of the linear and coset code ensemble with degrees (λ, ρ_Δ) . Then, $\forall l_0 \in \mathbb{N}, \lim_{\Delta \rightarrow \infty} \langle P_{linear}^{(l_0)} \rangle \stackrel{\mathcal{D}}{=} P_{coset}^{(l_0)}$ in distribution, with the convergence rate being $\mathcal{O}(\text{const}^\Delta)$ for some $\text{const} < 1$.



Intuition

- Consider the 1st check node iteration. Suppose for all possible $\{P_{linear}^{(0)}(x)\}$, $Q_{linear}^{(0)}(0) = Q_{linear}^{(0)}(1) = Q_{coset}^{(0)}$. Then

$$P_{linear}^{(1)}(x) = P_{linear}^{(0)}(x) \otimes \left(Q_{coset}^{(0)}\right)^{\otimes (d_v - 1)}$$

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- Since the iterative equations of DE are continuous, we need only to prove that for all possible $\{P_{linear}^{(0)}(x)\}$,

$$\lim_{\Delta \rightarrow \infty} Q_{linear}^{(0)}(0) \stackrel{\mathcal{D}}{=} \lim_{\Delta \rightarrow \infty} Q_{linear}^{(0)}(1) \stackrel{\mathcal{D}}{=} Q_{coset}^{(0)}$$



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Proof of the Main Theorem

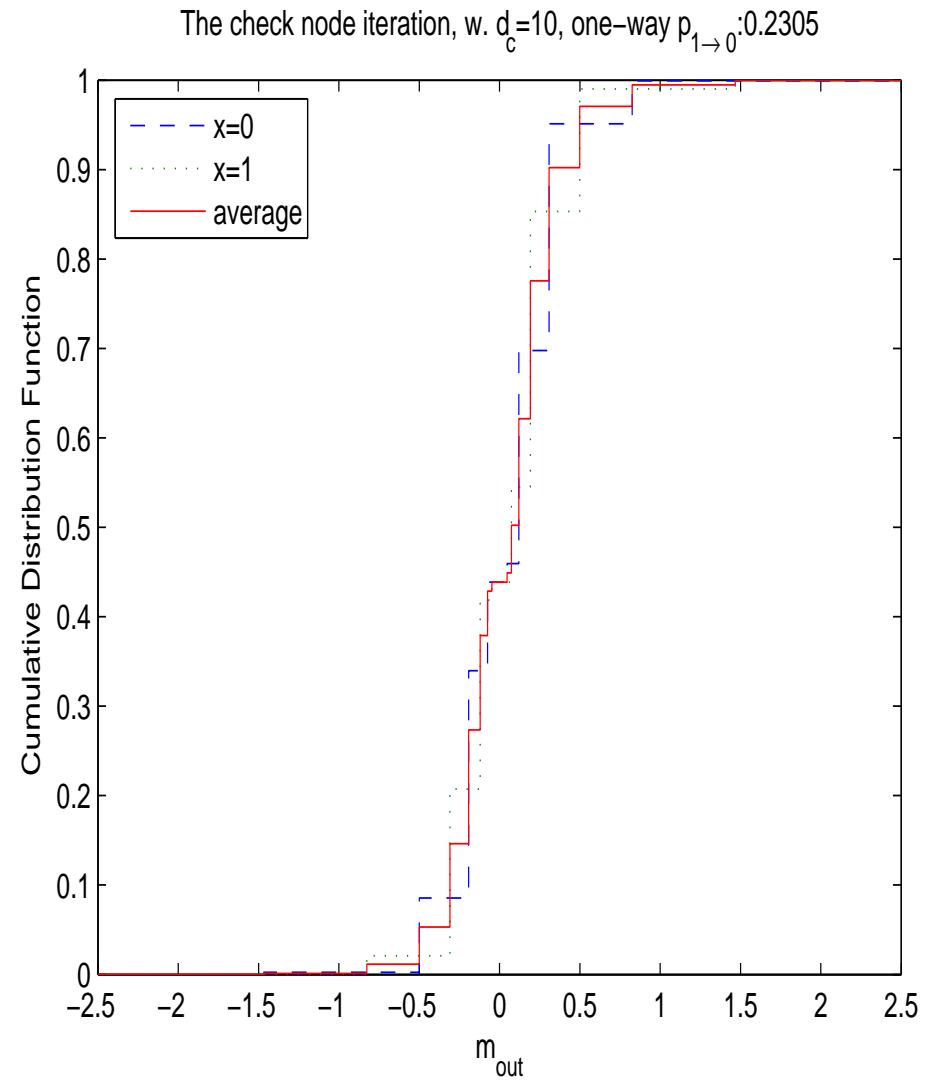
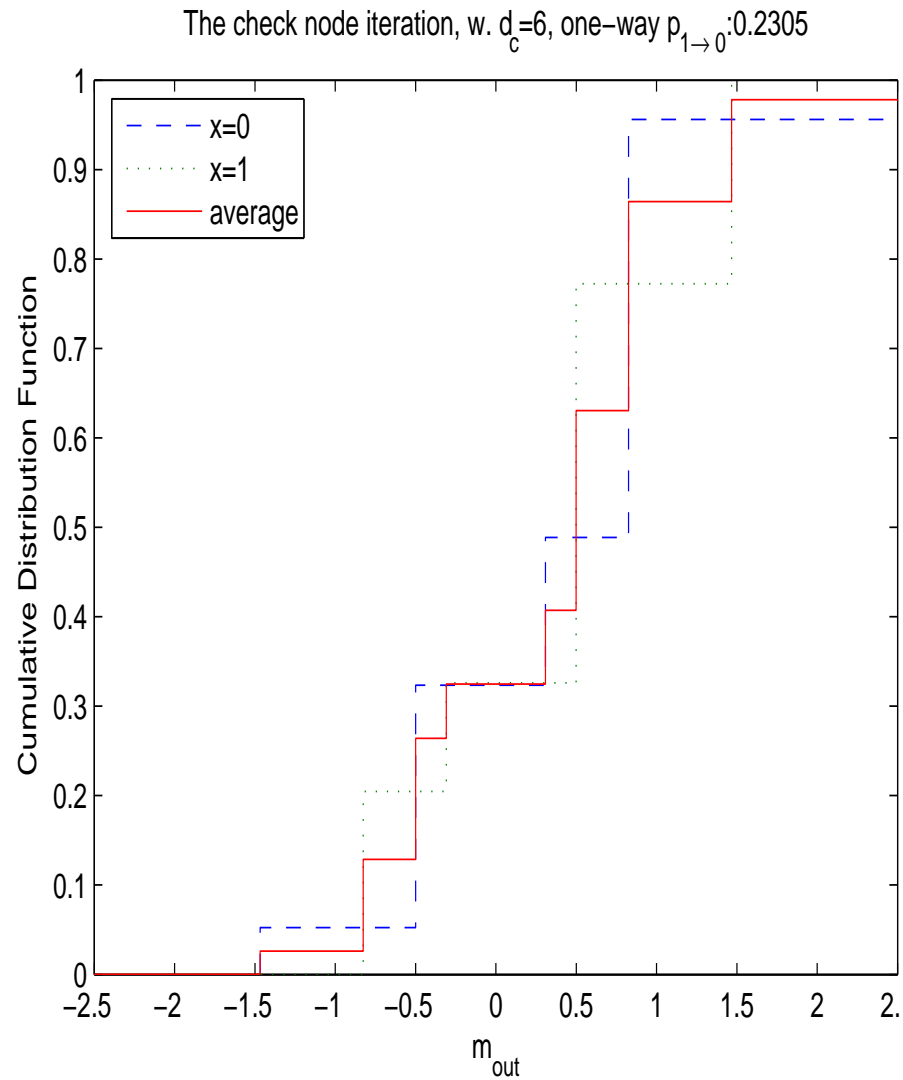
- $m = \log \frac{P(X=0|Y)}{P(X=1|Y)}.$
- $\gamma(m) = (1_{\{m < 0\}}, \log \cot \left| \frac{m}{2} \right|)$
- $P_{linear}^{(0)}(x)(dm) \mapsto P'_x(d\gamma)$ and $Q_{linear}^{(0)}(x)(dm) \mapsto Q'_x(d\gamma).$
- The characteristic function: $\Phi_P(\lambda_1, \lambda_2) = E(-1)^{\lambda_1 \gamma_1} e^{i \lambda_2 \gamma_2}.$
- By induction,

$$\Phi_{Q'_0}(\lambda_1, \frac{\lambda_2}{\Delta}) - \Phi_{Q'_1}(\lambda_1, \frac{\lambda_2}{\Delta}) = 2 \left(\frac{\Phi_{P'_0}(\lambda_1, \frac{\lambda_2}{\Delta}) - \Phi_{P'_1}(\lambda_1, \frac{\lambda_2}{\Delta})}{2} \right)^\Delta.$$

- By Taylor's expansion, the RHS converges to zero for all λ_1, λ_2 and the convergence rate is exponential.

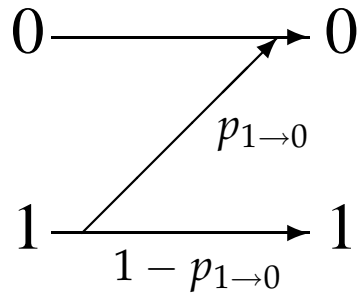


The Weak Convergence



Typicality in Terms of the Decodable Threshold

Z-Channels:



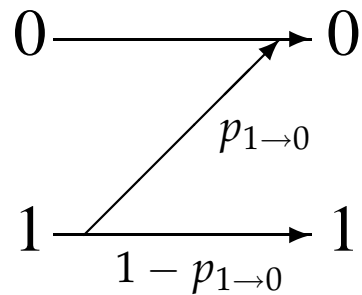
$$p_{1 \rightarrow 0, linear}^* := \sup \left\{ p_{1 \rightarrow 0} > 0 : \lim_{l \rightarrow \infty} p_{e, linear}^{(l)} = 0 \right\}$$

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Corollary 1 (Typicality Results for Z-Channels) *For any $\epsilon > 0$, there exists a $\Delta \in \mathbb{N}$ such that*

$$\left| p_{1 \rightarrow 0, linear}^* - p_{1 \rightarrow 0, coset}^* \right| < \epsilon.$$



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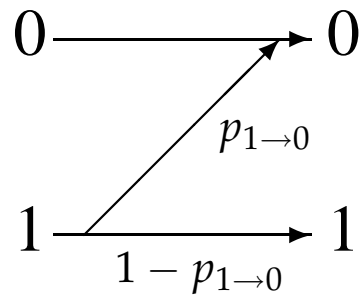
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Typicality in Terms of the Decodable Threshold

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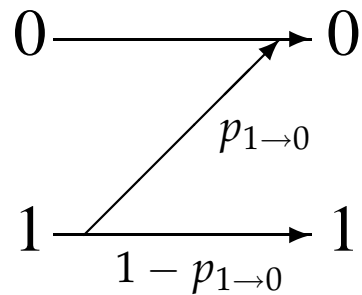
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Similar corollaries can be easily derived for other types of channel models.



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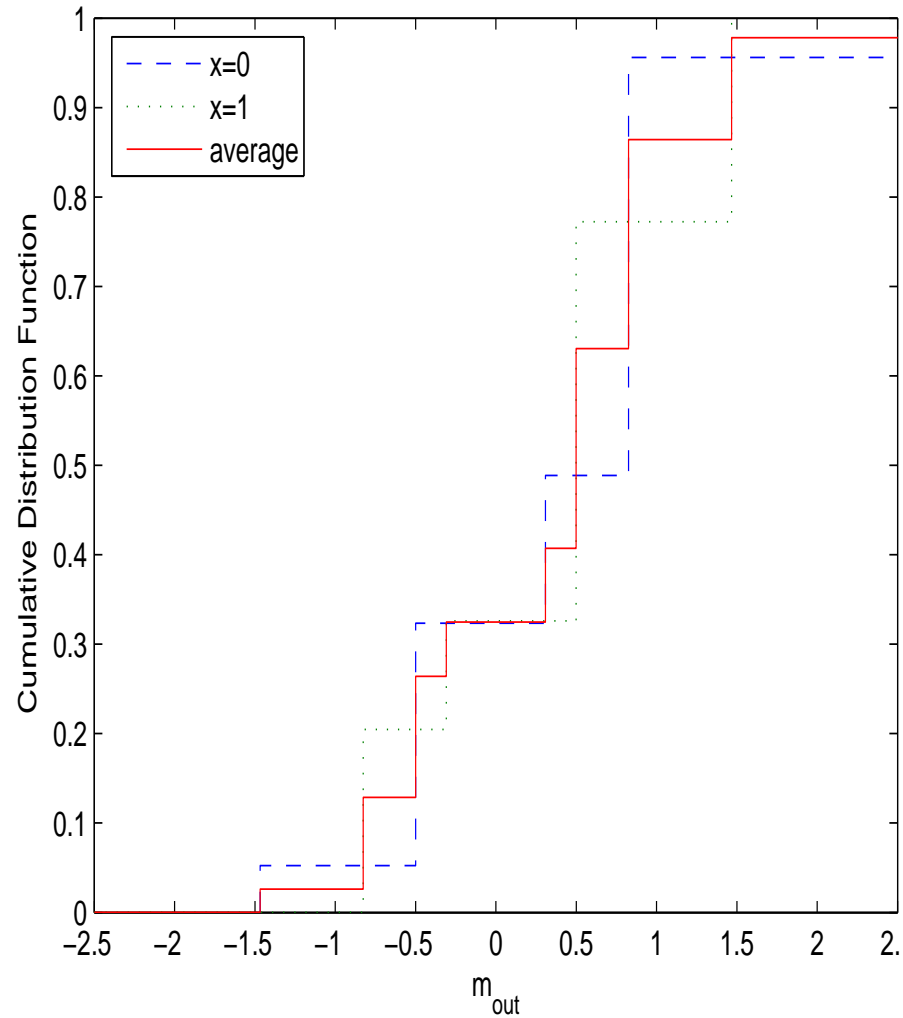
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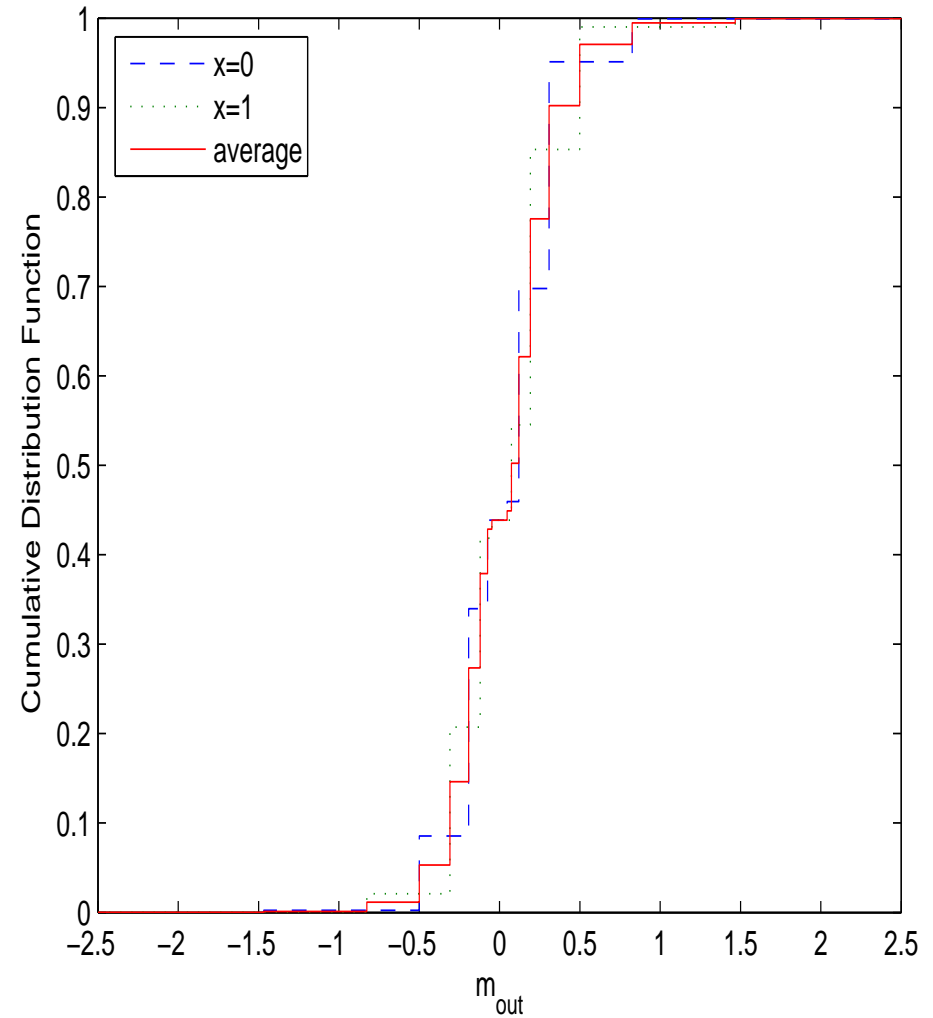


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The check node iteration, w. $d_c=6$, one-way $p_{1 \rightarrow 0}:0.2305$

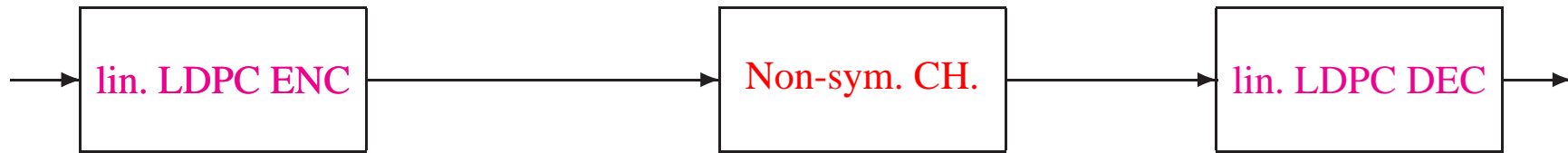


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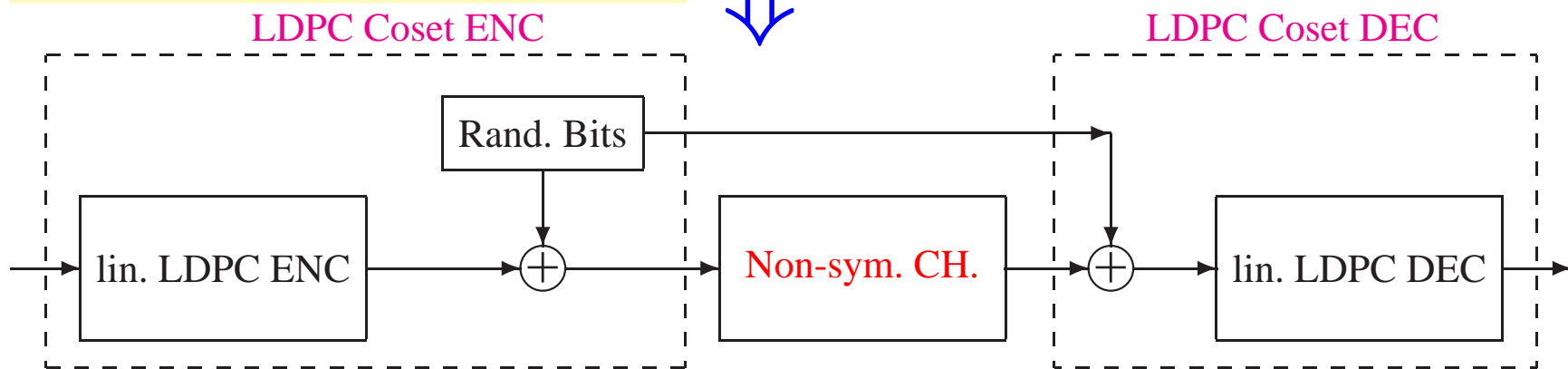


Comparisons

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The LDPC code ensemble:

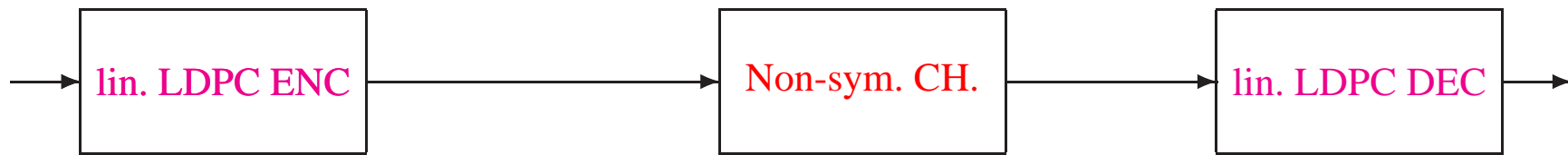


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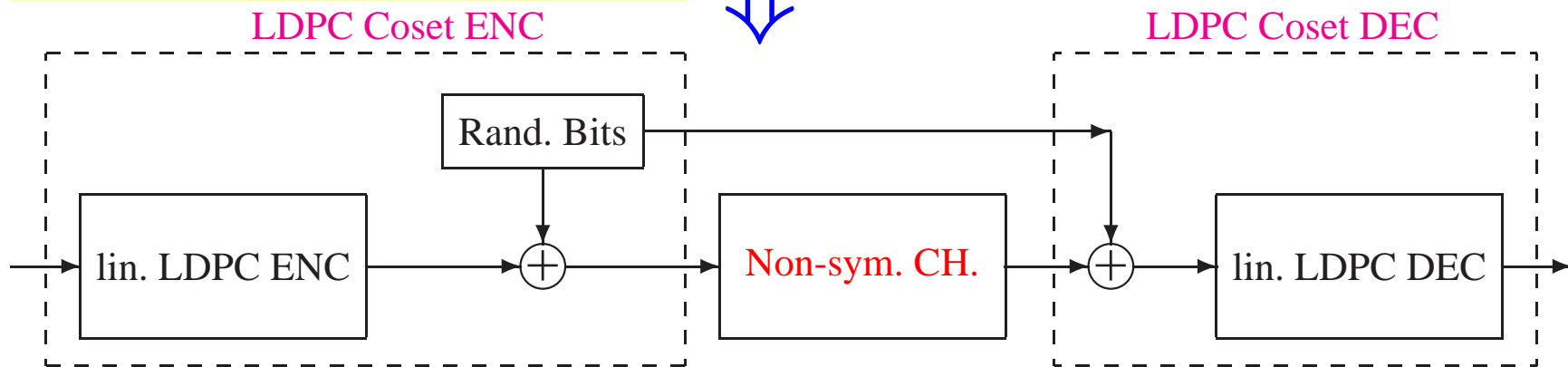
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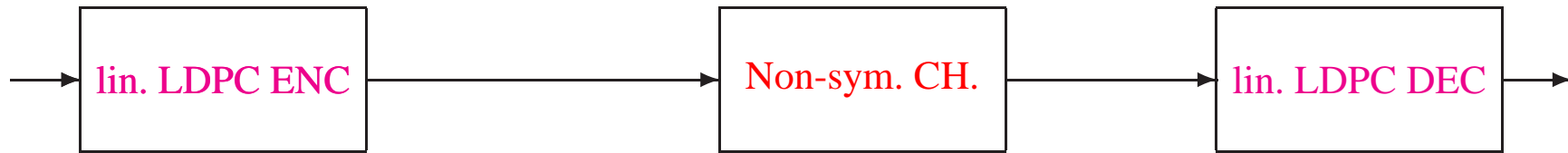


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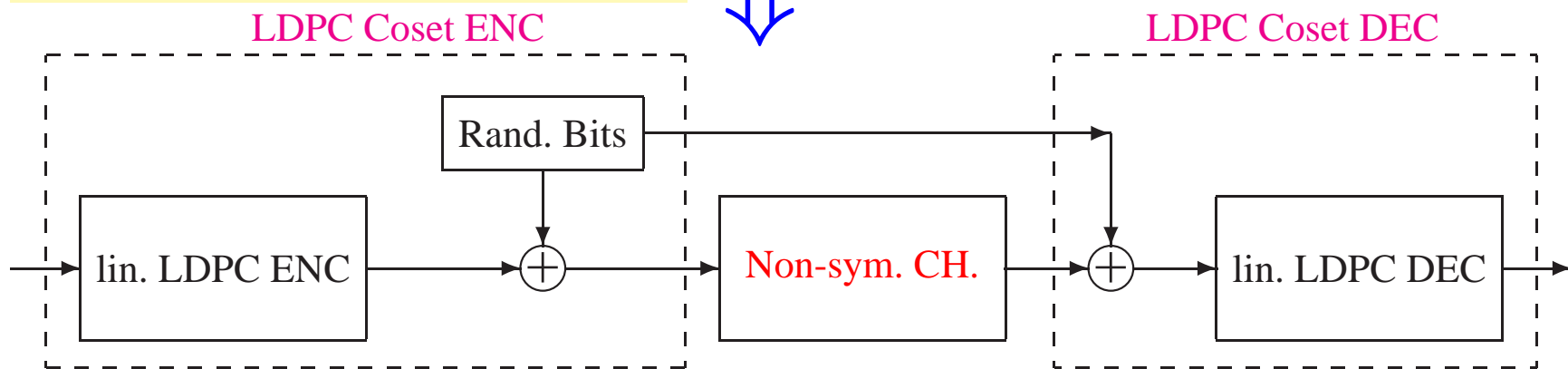


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