

# Design of Coarsely-Quantized Message Passing Decoders

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#### **Binary Message Passing (BMP) for LDPC Codes**

- Gallager, "Low density parity check codes," IRE Trans. IT, 1962
- Kou, Lin, Fossorier, LDPC codes based on finite geometries, IEEE Trans. IT, 2001
- Zhang, Fossorier, Modified bit-flipping decoding, IEEE Comm. Lett., 2004
- Miladinovic, Fossorier, Improved bit-flipping decoding, IEEE Trans. IT, 2005
- Jiang, Zhao, Shi, Chen, Improved bit flipping decoding, IEEE Comm. Lett., 2005
- Ardakani, Kschischang, Properties of binary message-passing, IEEE Trans. IT, 2005
- Sankaranarayanan et al., Failures of the Gallager B decoder, ISIT 2006
- Reliability-based majority-logic decoding for LDPC codes, IEEE Trans. Comm., 2009
- Planjery, Declercq, Danjean, Vasic, Finite alphabet iterative decoders, 2013-
- Many other papers

#### Here Review and Expand on:

 Lechner, Pedersen, Kramer, Analysis and design of binary message passing decoders, IEEE Trans. Comm., 2012 (and ISIT 2007)

#### I. Low-Density Parity-Check (LDPC) Codes





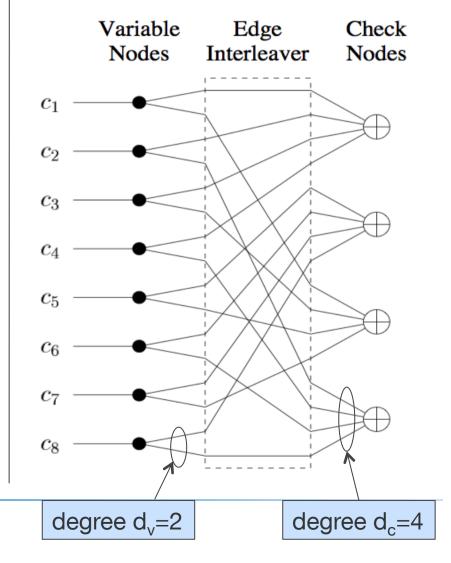
A binary linear block code is the set of binary (row) vectors, or codewords, <u>c</u>, satisfying, e.g.,

$$\underline{c} \begin{bmatrix}
1 & 0 & 1 & 0 \\
1 & 0 & 0 & 1 \\
1 & 0 & 1 & 0 \\
1 & 0 & 0 & 1 \\
0 & 1 & 1 & 0 \\
0 & 1 & 0 & 1 \\
0 & 1 & 0 & 1
\end{bmatrix} = \underline{0}$$

$$\text{matrix } H^{T}$$

where H is a (n-k) x n parity-check-matrix. Rate is R=1-rank(H)/n (example: R=5/8).

Tanner Graph Representation of Parity-Check Constraints





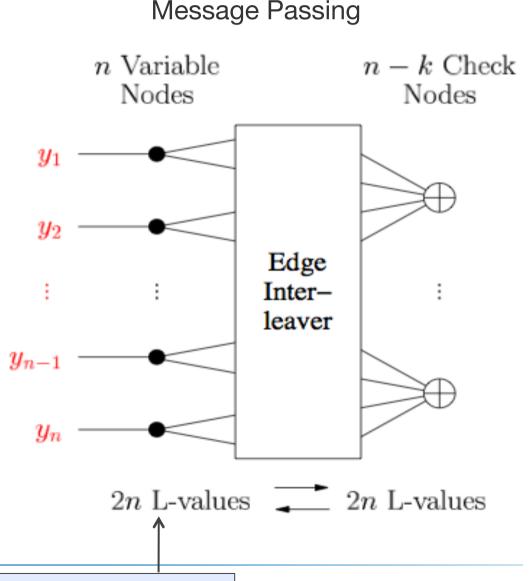
- Code is low-density if each row and column of H<sup>T</sup> has "few" 1's
- Irregular LDPC code: variable number of 1's in every column/row
- Decoding: use message passing on the graph
- Messages may be cond. probabilities

$$\Pr(c_1 = 0 | \underline{y})$$

or log-likelihood ratios (L-values)

$$L_{1} = \log \frac{\Pr(c_{1} = 0 | \underline{y})}{\Pr(c_{1} = 1 | \underline{y})}$$

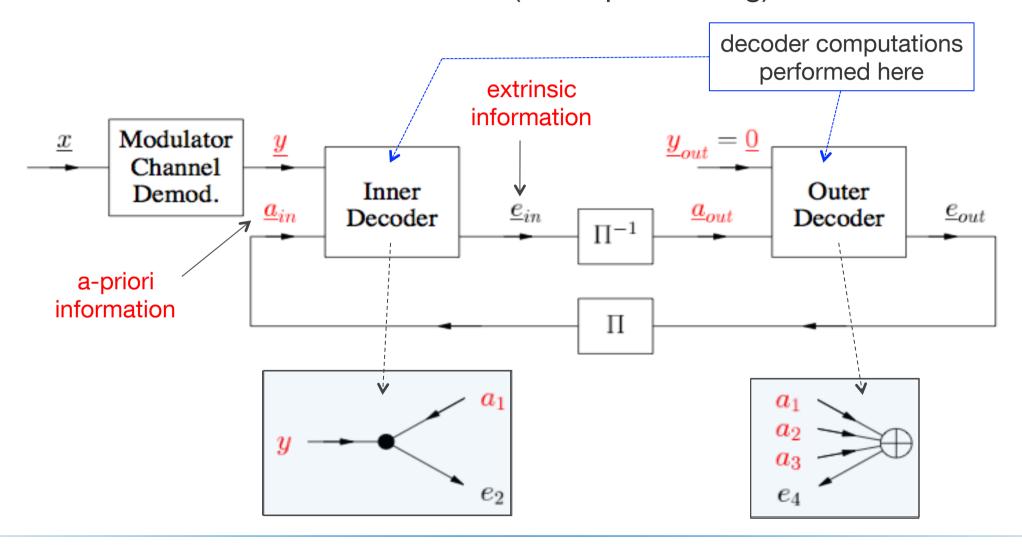
or, in practice, quantized L-values



## **II. Iterative Decoding**



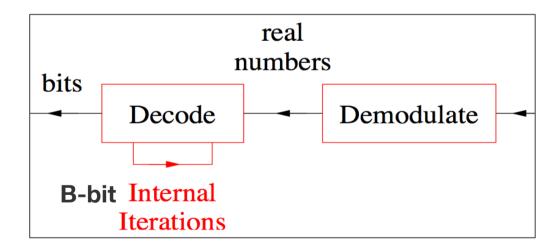
## LDPC code decoder iterations (turbo processing):



## III. Demodulation and Decoding

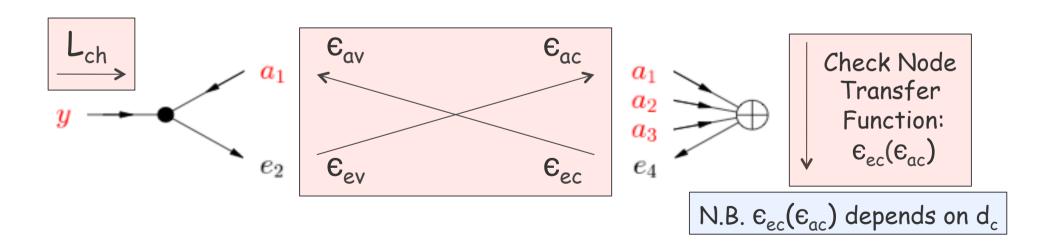


- L-values are real but must be quantized, see figure below
  - Demodulator: can put out soft decisions (>log<sub>2</sub>(M) bits/symbol) or hard decisions (=log<sub>2</sub>(M) bits/symbol)
  - 2) Decoder iterations: B-bit message passing: binary message passing (BMP, B=1) ternary message passing (TMP, B≈2)
- Motivation: high-speed devices (>100 Gb/s) need simplifications





- BMP/TMP: natural approaches are as follows:
  - 1) Every edge bit represents a hard decision on an extrinsic L-value
  - Variable nodes convert apriori bits to L-values, add L-values, and make binary (hard) or ternary decisions on output L-values
  - 3) Check nodes perform (extrinsic) XORs for binary message passing, and (extrinsic) XORs and erasures for ternary message passing
- Analysis: use distribution evolution (DE) to track extrinsic probabilities.
   BMP: track error probabilities; TMP: also track erasures

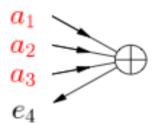


#### **BMP** Distribution Evolution



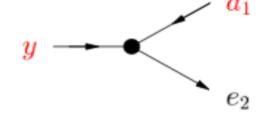
Check node (degree d<sub>c</sub>) and binary messages:

$$\epsilon_{ec} = f_c(\epsilon_{ac}; d_c) = \frac{1 - (1 - 2\epsilon_{ac})^{d_c - 1}}{2}$$



Variable node (degree d<sub>v</sub>): suppose x<sub>i</sub>=±1 (BPSK)

$$\varepsilon_{ev} = \sum_{j=1}^{d_v} \frac{1}{d_v} \Pr\left[ \operatorname{sgn}(L_{ev,j}) \neq x_j \right]$$



$$L_{ev,j} = L_{ch} + \sum_{i=1:i\neq j}^{d_{v}} L_{av,i} \quad j = 1, 2, \dots, d_{v}$$

$$L_{av,i} = a_i \log \frac{1 - \varepsilon_{av}}{\varepsilon_{av}}, \ a_i = \pm 1, \text{ but what is } \varepsilon_{av}?$$

 $L_{ch}$  depends on channel quantization  $\leftarrow$  Two design issues



#### **Issue 1: Variable Node Processing**

- Processing depends on  $\epsilon_{av}$  which
  - Varies from iteration to iteration
  - Is unknown, unless the codes have infinite length in which case  $\epsilon_{av}$  can be computed from EXIT chart (see below)
- Two other approaches:
  - Optimize "choice" of eav offline by numerical simulation
  - Estimate ε<sub>av</sub> online based on the number of unsatisfied checks
- 1<sup>st</sup> approach is complex, but likely very good. This variant was used to design certain deployed LDPC codes
- 2<sup>nd</sup> approach is used here



## **Issue 2: Channel Outputs**

- Consider an AWGN channel,  $x_j=\pm 1$ , noise variance  $\sigma_n^2$
- Let  $D_{ch} = |L_{ch}|$  ... called the reliability of the L-value
- For soft decisions:

$$D_{ch} = \frac{2}{\sigma_n^2}$$

■ For <u>hard decisions</u> get a <u>binary symmetric channel (BSC)</u> with crossover probability  $\epsilon_{ch}$  (0 ≤  $\epsilon_{ch}$  ≤ 0.5)

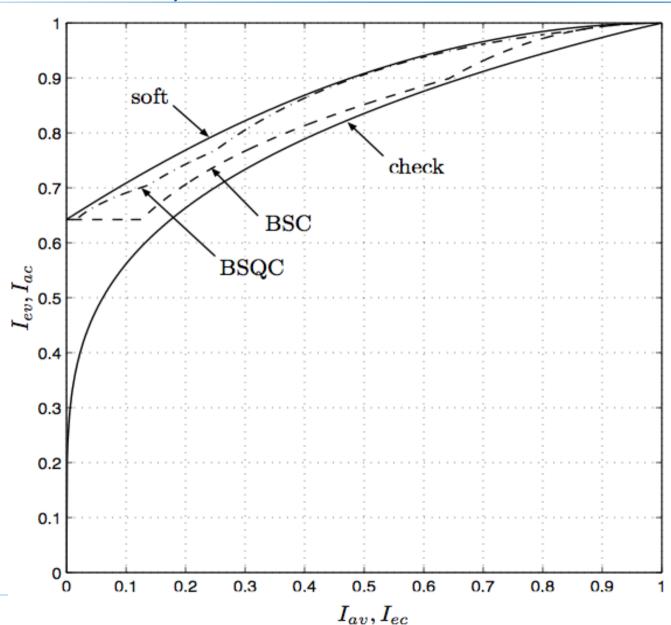
$$D_{ch} = \log \frac{1 - \varepsilon_{ch}}{\varepsilon_{ch}}$$
, where  $\varepsilon_{ch} = Q(1/\sigma_n)$ 

 For <u>b-bit quantization</u>: use mixture of b hard-decision channel reliabilities, e.g., <u>2-bit</u> quantization with a <u>binary symmetric quaternary</u> output (BSQC) channel

#### **Transfer Functions for (4,6)-Regular Code**



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Channel:  $\sigma_n$ =0.67 x-axis to y-axis: variable nodes y-axis to x-axis: check nodes

 $I_{ac}$ =1-h( $\varepsilon_{ac}$ ) where h(x) is the binary entropy function: h(x) = -xlog<sub>2</sub>x-(1-x)log<sub>2</sub>(1-x) Example: h(0.11)=0.5 Similar for  $I_{ec}$ ,  $I_{av}$ ,  $I_{ev}$ 

#### **Comments:**

BSC quantization same as
 Gallager B algorithm
 BSQC quantization
 thresholds at 0 and ±1.9

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## IV. Optimization: Irregular LDPC Codes



■ Each node's  $\epsilon_{ev}$  depends on  $d_v$ : write as  $\epsilon_{ev}(\epsilon_{av}, d_v)$ . Now use different degrees to shape avg. variable node curve:

$$\varepsilon_{ev}(\varepsilon_{av}) = \sum_{i} \lambda_{i} \varepsilon_{ev}(\varepsilon_{av}, i)$$

with  $\lambda_i$ =fraction of edges connected to var. nodes of degree i

- Can similarly shape the check node function  $\epsilon_{\rm ec}(\epsilon_{\rm ac})$
- Degree distribution {λ<sub>i</sub>} design: use EXIT chart
  - $\epsilon_{ev}(\epsilon_{av})$  curve should lie above  $\epsilon_{ec}(\epsilon_{ac})$  curve for convergence (and  $n=\infty$ )
  - L-value messages: Matching EXIT curves maximizes rate.
- BMP: new issues vs. L-value messages
  - Stability (decoder convergence when  $\epsilon_{av}$  or  $\epsilon_{ac}$  are small)
  - Cycles related to "absorbing sets" cause decoder to get stuck
- Approach: build optimization & remedies into a linear program

## Rate, Stability, Cycles



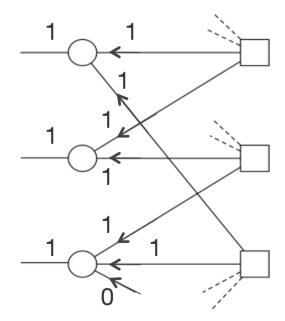
Design Rate:

$$R = 1 - \frac{1/d_c}{\sum_{i} \lambda_i / i}$$

• Stability: satisfied for binary message passing and hard or soft channel messages if and only if (try  $\lambda_2 = 1$ )

$$(\lambda_2 + 2\varepsilon_{ch}\lambda_3)(d_c - 1) < 1$$

- Cycles:
  - Structure on right causes decoding failure if all channel messages in error, and if all other incoming messages correct
  - Obvious idea: avoid cycles of degree 2 or 3 variable nodes



## **Cycles and Linear Program**



• Result: a Tanner graph with no cycles having degree 2 and 3 variable nodes exists if and only if (try  $\lambda_3 = 1$ )

$$3\lambda_2 + 4\lambda_3 \le \frac{6}{d_c} \left( 1 - \frac{1}{(1-R)N} \right) < \frac{6}{d_c}$$

• Linear Program:  $\lambda = \{\lambda_i\}$  is variable node degree distribution

$$\lambda^* = \underset{\lambda}{\operatorname{arg\,max}} R = \underset{\lambda}{\operatorname{arg\,max}} \left( 1 - \frac{1/d_c}{\sum_i \lambda_i / i} \right) = \underset{\lambda}{\operatorname{arg\,max}} \sum_i \lambda_i / i$$

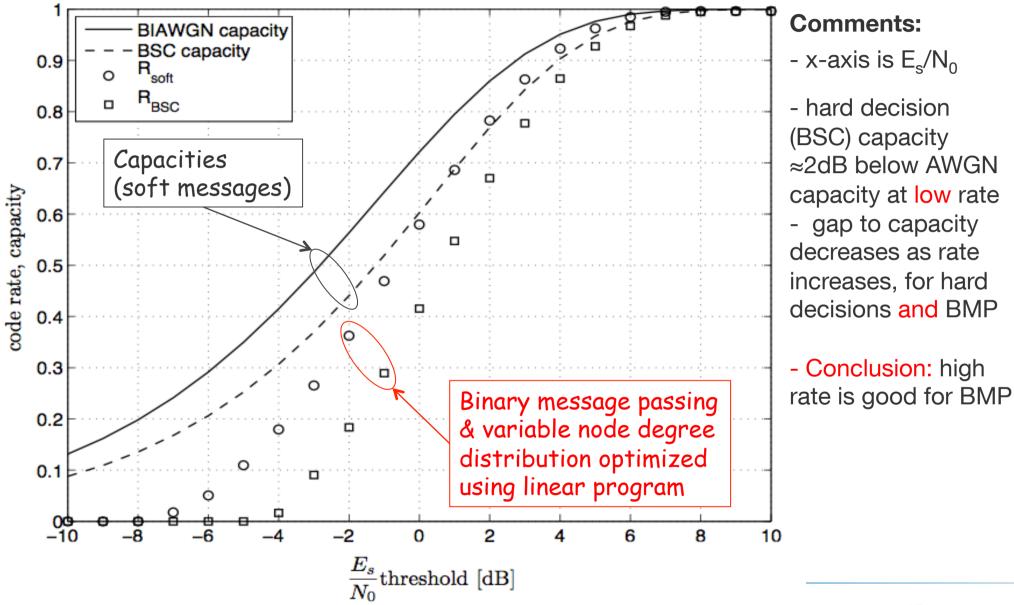
subject to [variable node EXIT curve above check node EXIT curve]

$$\sum_{i} \lambda_{i} = 1, \quad 0 \le \lambda_{i} \le 1$$

$$(\lambda_2 + 2\varepsilon_{ch}\lambda_3)(d_c - 1) < 1, \quad 3\lambda_2 + 4\lambda_3 < \frac{6}{d_c}$$

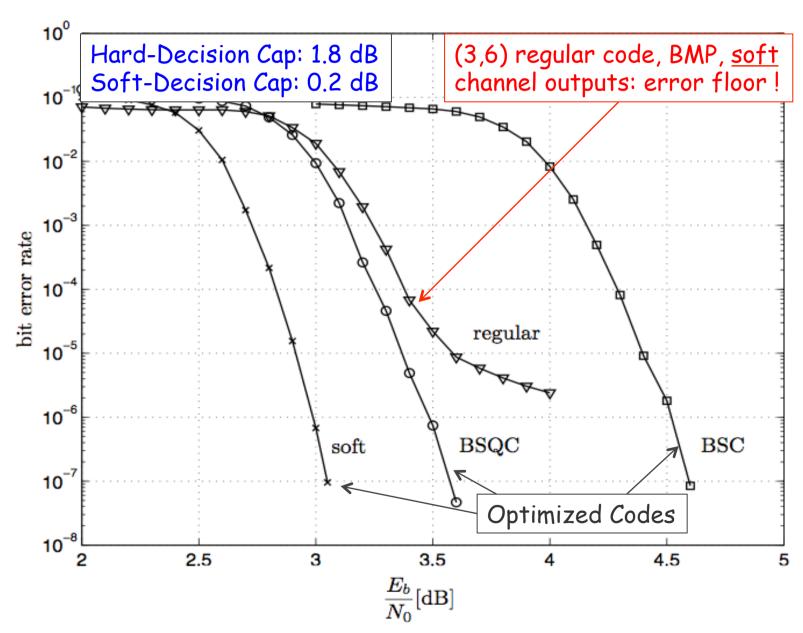
#### **BMP Thresholds**





## Performance: Rate 1/2, BMP





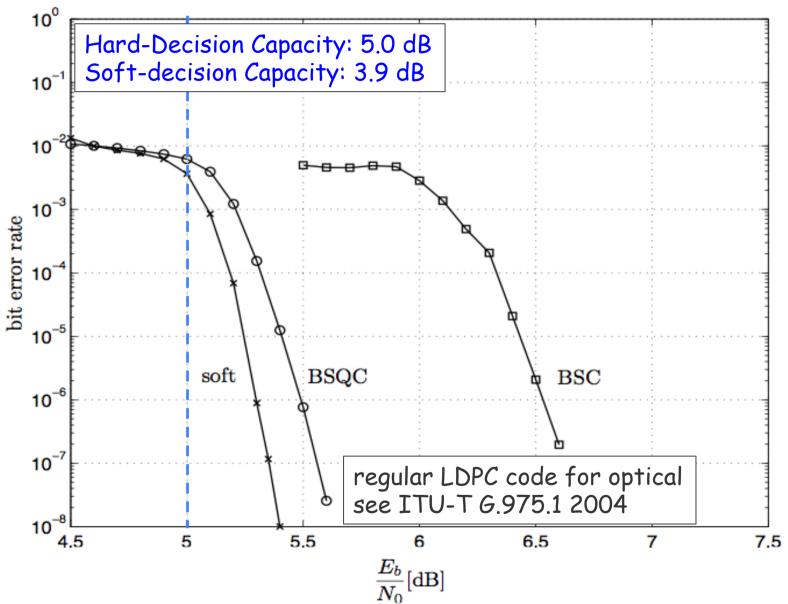
#### **Comments:**

- x-axis is  $E_b/N_0$
- PEG interleavers automatically avoid undesirable cycles
- n = 10,000
- 2-bit quant. gains
  ≈1dB over Gallager
  B and loses ≈0.5dB
  as compared to soft outputs

## Performance: Rate 15/16, BMP



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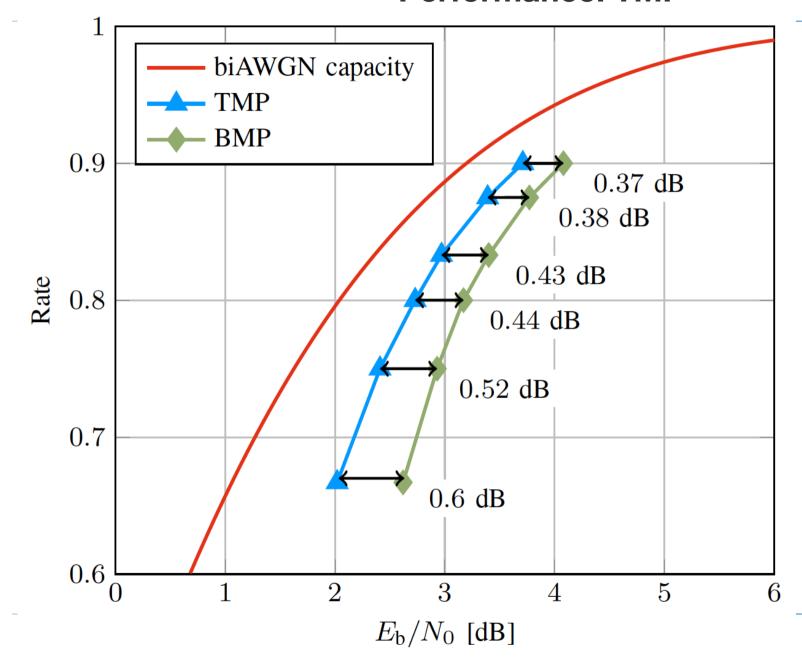


#### **Comments:**

- x-axis is  $E_b/N_0$
- interleaver taken from standard
   2-bit quant. gains
  ≈1dB over Gallager
  B and loses ≈0.2dB
  vs. soft outputs
   BMP is ≈1.5dB
  from L-value
  message capacity
   longer & irregular
  codes get closer

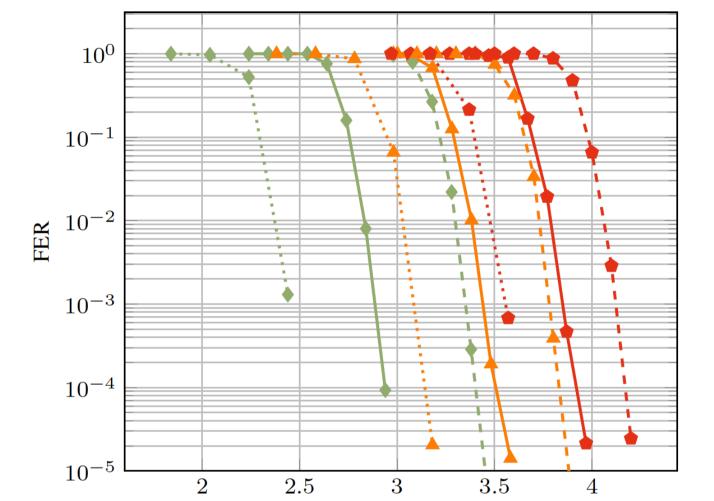
#### **Performance: TMP**





#### **Comments:**

- Figure taken from Emna Ben Yacoub's Master Thesis, Oct. 2018
- Curves show decoding thresholds with BMP and TMP for optimized protograph LDPC code ensembles





#### **Comments:**

- Figure from E. Ben Yacoub et al.'s arxiv paper, Sep. 2018
- Curves show frame error rate (FER) of AR4JA and optimized codes

 $E_{\rm b}/N_0~{\rm [dB]}$ 



#### For More Details:

G. Lechner, T. Pedersen, and G. Kramer, "Analysis and design of binary message passing decoders," IEEE Trans. Commun., 60(3), 601-607, 2012. See also: <a href="http://arxiv.org/pdf/1004.4020v1">http://arxiv.org/pdf/1004.4020v1</a>

E. Ben Yacoub, "LDPC Decoding Algorithms Based on Ternary Message Passing," Master's Thesis, Technical University of Munich, Oct. 2018

E. Ben Yacoub, F. Steiner, B. Matuz, G. Liva, "Protograph-Based LDPC Code Design for Ternary Message Passing Decoding," Sep. 2018 <a href="https://arxiv.org/abs/1809.10910v2">https://arxiv.org/abs/1809.10910v2</a>

## See the Posters! And the First Talk Tomorrow!