### LDPC Codes, Construction and Applications

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#### What is an LDPC Code

#### A linear error correcting code

All codewords should satisfy

$$\mathbf{H}\mathbf{x} = \mathbf{0} \tag{1}$$

Any vector that satisfy above equation is a codeword

#### A block code

If **H** is  $M \times N$  and has rank R then the code has dimension

$$K = N - R$$

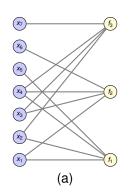
Block length of the code is N.

#### Sparse H

Here we are dealing with binary LDPC codes, so H has few 1s rest all zeros

### Tanner Graph

$$\mathbf{H} = \begin{pmatrix} 1 & 1 & 0 & 1 & 1 & 0 & 0 \\ 1 & 0 & 1 & 1 & 0 & 1 & 0 \\ 0 & 1 & 1 & 1 & 0 & 0 & 1 \end{pmatrix} \tag{2}$$



### **Degree Distribution**

Let  $\lambda$  and  $\rho$  be vectors such that their  $i^{th}$  component  $\lambda_i$  and  $\rho_i$  represent the fraction of edges connecting to a variable node of degree i and check node of degree i respectively. Thus

$$\lambda(x) = \sum_{i} \lambda_i x^{i-1}$$
 and  $\rho(x) = \sum_{i} \rho_i x^{i-1}$ 

are called variable and check degree distribution polynomials respectively.

- Regular LDPC Codes
- Irregular LDPC Codes
- Ensemble LDPC $(n, \lambda, \rho)$

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## Marginalization by Message Passing

For

$$f(x_1, x_2, x_3, x_4, x_5, x_6) = f_1(x_1, x_2, x_3) f_2(x_1, x_4, x_6) f_3(x_4) f_4(x_4, x_5)$$

$$f(x_1) = \sum_{x_2, x_3, x_4, x_5, x_6} f(x_1, x_2, x_3, x_4, x_5, x_6)$$

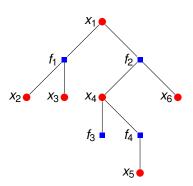
$$= \left[ \sum_{x_2,x_3} f_1(x_1,x_2,x_3) \right] \left[ \sum_{x_4} f_3(x_4) \left( \sum_{x_6} f_2(x_1,x_4,x_6) \right) \left( \sum_{x_5} f_4(x_4,x_5) \right) \right]$$

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### **Factor Graph**

For

$$\left[\sum_{x_2,x_3} f_1(x_1,x_2,x_3)\right] \left[\sum_{x_4} f_3(x_4) \left(\sum_{x_6} f_2(x_1,x_4,x_6)\right) \left(\sum_{x_5} f_4(x_4,x_5)\right)\right]$$



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## Message Passing Rules

$$\mu(x) = f(x) \int_{f}^{x}$$

initialization at leaf nodes

$$\int_{\mathbf{Y}}^{f} \mu(\mathbf{x}) = 1$$

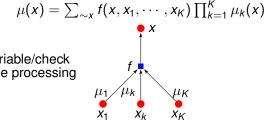
$$\mu(x) = \prod_{k=1}^{K} \mu_k(x)$$

$$f$$

$$\mu_1 \qquad \mu_k \qquad \mu_K$$

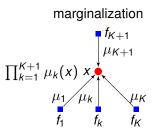
$$f_1 \qquad f_k \qquad f_K$$

variable/check node processing

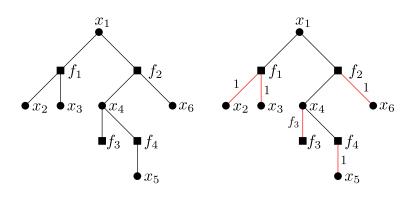


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## Message Passing Rules

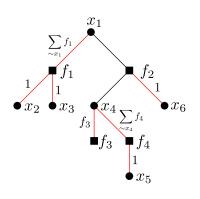


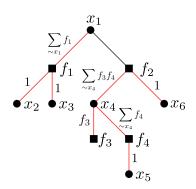
## Message Passing Example



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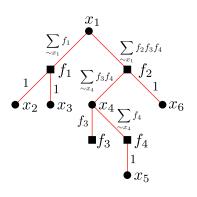
# Message Passing Example

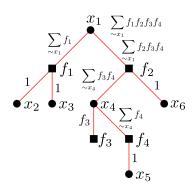




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# Message Passing Example





4 D > 4 A > 4 E > 4 E > 900

## **Belief Propagation Decoding**

Bitwise maximum a posteriori (MAP) decoding for memoryless channel without feedback.

$$\hat{x}_i(y) = \arg \max_{x_i} p_{X_i|Y}(x_i|y)$$
 (3)

$$= \arg \max_{x_i} \sum_{x_i} p_{X|Y}(x|y) \tag{4}$$

$$= \arg \max_{x_i} \sum_{\sim x_i} p_{Y|X}(y|x) p_X(x)$$
 (5)

$$= \arg \max_{x_i} \sum_{\sim x_i} \prod_j p_{Y_j|X_j}(y_j|x_j) p_X(x)$$
 (6)

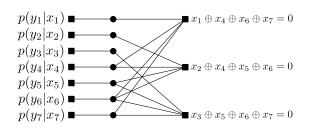
$$= \arg\max_{\mathbf{x}_i} \sum_{\mathbf{x}_i} \prod_j \rho_{Y_j|X_j}(y_j|x_j) \mathbf{1}_{\{x \in C\}}$$
 (7)

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#### An Example

#### Given the parity check matrix

$$\hat{x}_{i}(y) = \arg\max_{x_{j}} \sum_{\sim x_{i}} \prod_{j} p_{Y_{j}|X_{j}}(y_{j}|x_{j}) \mathbf{1}_{\left\{x_{1} + x_{4} + x_{6} + x_{7} = 0\right\}} \mathbf{1}_{\left\{x_{2} + x_{4} + x_{5} + x_{6} = 0\right\}} \mathbf{1}_{\left\{x_{3} + x_{5} + x_{6} + x_{7} = 0\right\}}$$
(9)



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- In the binary case the message is  $(\mu(1), \mu(-1))$ .
- $(\mu(1), \mu(-1)) \leftarrow (p_{y_i|x_i}(y_i|1), p_{y_i|x_i}(y_i|-1))$
- At a variable node of degree K + 1,

$$\mu(1) = \prod_{k=1}^{K} \mu_k(1), \quad \mu(-1) = \prod_{k=1}^{K} \mu_k(-1)$$

•  $r_k \leftarrow \mu_k(1)/\mu_k(-1)$ 

$$r = \frac{\mu(1)}{\mu(-1)} = \frac{\prod_{k=1}^{K} \mu_k(1)}{\prod_{k=1}^{K} \mu_k(-1)} = \prod_k r_k$$

• If  $I_k = \log(r_k)$ , processing rule becomes  $I = \sum_k I_k$ 

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• At a check node of degree J + 1 the computation is

$$\mu(x) = \sum_{x_1, x_2, \dots, x_J} \mathbf{1}_{\{\prod_{j=1}^J x_j = x\}} \prod_{j=1}^J \mu_j(x_j)$$
 (10)

$$r = \frac{\mu(1)}{\mu(-1)} = \frac{\sum_{\{x_1, x_2, \dots, x_J: \prod_{j=1}^J x_j = 1\}} \prod_{j=1}^J \mu_j(x_j)}{\sum_{\{x_1, x_2, \dots, x_J: \prod_{j=1}^J x_j = -1\}} \prod_{j=1}^J \mu_j(x_j)}$$
(11)

$$= \frac{\sum_{\{x_{1},x_{2},\cdots,x_{J}:\prod_{j=1}^{J}x_{j}=1\}} \prod_{j=1}^{J} \frac{\mu_{j}(x_{j})}{\mu_{j}(-1)}}{\sum_{\{x_{1},x_{2},\cdots,x_{J}:\prod_{j=1}^{J}x_{j}=-1\}} \prod_{j=1}^{J} \frac{\mu_{j}(x_{j})}{\mu_{j}(-1)}}$$
(12)

$$r = \frac{\sum_{\{x_1, x_2, \cdots, x_J: \prod_{j=1}^J x_j = 1\}} \prod_{j=1}^J r_j^{(1+x_j)/2}}{\sum_{\{x_1, x_2, \cdots, x_J: \prod_{j=1}^J x_j = -1\}} \prod_{j=1}^J r_j^{(1+x_j)/2}}$$
(13)

$$= \frac{\prod_{j=1}^{J} (r_j + 1) + \prod_{j=1}^{J} (r_j - 1)}{\prod_{j=1}^{J} (r_j + 1) - \prod_{j=1}^{J} (r_j - 1)}$$
(14)

$$= \frac{1 + \frac{\prod_{j=1}^{J} (r_j - 1)}{\prod_{j=1}^{J} (r_j + 1)}}{1 - \frac{\prod_{j=1}^{J} (r_j - 1)}{\prod_{j=1}^{J} (r_j + 1)}}$$
(15)

$$\implies \frac{r-1}{r+1} = \prod_{i=1}^{J} \frac{r_i - 1}{r_i + 1} \tag{16}$$

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$$r = \exp(I) \tag{17}$$

$$r = \exp(I)$$

$$\implies \tanh(I/2) = \frac{r-1}{r+1}$$
(18)

$$= \prod_{j=1}^{J} \frac{r_j - 1}{r_j + 1} \tag{19}$$

$$= \prod_{j=1}^{J} \tanh(I_j/2)$$
 (20)

$$\implies I = 2 \tanh^{-1} \left( \prod_{j=1}^{J} \tanh(I_j/2) \right)$$
 (21)

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### Construction of Parity Check Matrices

- Ideally we need tree.
- In practice we try to achieve a minimum girth in the tanner graph.
- Construction of Parity Check matrix from Reed Solomon codes.
  - Minimum girth of 4 is guaranteed.
  - Constructs a regular LDPC code.
- Progressive Edge Growth construction
  - Starts with a graph with no edges.
  - Progressively introduce edges such that it has minimum effect on the girth
  - Outputs both regular and irregular LDPC codes.

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## **Encoding**

If the parity check matrix is given as [I P], where I is  $N-K \times N-K$  identity matrix, P is some  $N-K \times K$  matrix, a codeword  $\mathbf{x} = \begin{bmatrix} \mathbf{x}_{\mathbf{p}}^\mathsf{T} \ \mathbf{x}_{\mathbf{d}}^\mathsf{T} \end{bmatrix}^\mathsf{T}$  can be formed by assigning

$$\mathbf{x_p} = \mathbf{P} \cdot \mathbf{x_d}$$

We obtain  $\mathbf{G} = [\mathbf{I} \ \mathbf{P}]$  from  $\mathbf{H}$  by

- Column permutation
- Row additions
- Row permutations

Suppose f represents the permutation in obtaining  $\mathbf{G}$  from  $\mathbf{H}$ , and  $\mathbf{G} \cdot \mathbf{x} = \mathbf{0}$ , then  $\mathbf{H} \cdot f(\mathbf{x}) = \mathbf{0}$ 



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## Belief Propagation Decoding for 2 User Gaussian MAC

$$\hat{x}_{i}^{[1]} \stackrel{\Delta}{=} \arg \max_{x_{i}} \rho_{X_{i}^{[1]}|Y}(x_{i}^{[1]}|y)$$
 (22)

$$= \arg \max_{x_i} \sum_{\sim x_i^{[1]}} \rho_{X^{[1]}, X^{[2]}|Y}(x^{[1]}, x^{[2]}|Y)$$
 (23)

$$= \arg \max_{x_i} \sum_{\sim x_i^{[1]}} \rho_{Y|X^{[1]},X^{[2]}}(y|x^{[1]},x^{[2]}) \rho_{X^{[1]},X^{[2]}}(x^{[1]},x^{[2]})$$
(24)

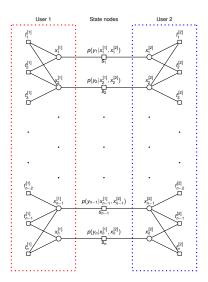
$$= \arg\max_{x_i} \sum_{x_i^{[1]}} p_{Y|X^{[1]},X^{[2]}}(y|x^{[1]},x^{[2]}) p_{X^{[1]}}(x^{[1]}) p_{X^{[2]}}(x^{[2]})$$
 (25)

$$= \arg\max_{x_i} \sum_{\sim x_i^{[1]}} \prod_j \rho_{Y_j|X_j^{[1]},X_j^{[2]}}(y_j|x_j^{[1]},x_j^{[2]}) \mathbf{1}_{\{x^{[1]}\in\mathcal{C}^{[1]}\}} \mathbf{1}_{\{x^{[2]}\in\mathcal{C}^{[2]}\}} (26)$$

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# Factor Graph for Joint Decoding



## Message Passing Rules for Joint Decoding

- At variable nodes and at Check nodes the rules are the same.
- Rule for the function node

$$\mu_{s_{i} \to x_{i}^{[2]}} = \log \left( \frac{\exp(\mu_{x_{i}^{[1]} \to s_{i}}) p(y|x_{i}^{[1]} = -1, x_{i}^{[2]} = -1) + p(y|x_{i}^{[1]} = 1, x_{i}^{[2]} = -1)}{\exp(\mu_{x_{i}^{[1]} \to s_{i}}) p(y|x_{i}^{[1]} = -1, x_{i}^{[2]} = 1) + p(y|x_{i}^{[1]} = 1, x_{i}^{[2]} = 1)} \right)$$

$$\mu_{s_{i} \to x_{i}^{[1]}} = \log \left( \frac{\exp(\mu_{x_{i}^{[2]} \to s_{i}}) p(y|x_{i}^{[1]} = -1, x_{i}^{[2]} = -1) + p(y|x_{i}^{[1]} = -1, x_{i}^{[2]} = 1)}{\exp(\mu_{x_{i}^{[2]} \to s_{i}}) p(y|x_{i}^{[1]} = 1, x_{i}^{[2]} = -1) + p(y|x_{i}^{[1]} = 1, x_{i}^{[2]} = 1)} \right)$$

$$(28)$$

$$\mu_{s_{i} \to x_{i}^{[1]}} = \log \left( \frac{\exp(\mu_{x_{i}^{[2]} \to s_{i}}) \rho(y|x_{i}^{[1]} = -1, x_{i}^{[2]} = -1) + \rho(y|x_{i}^{[1]} = -1, x_{i}^{[2]} = 1)}{\exp(\mu_{x_{i}^{[2]} \to s_{i}}) \rho(y|x_{i}^{[1]} = 1, x_{i}^{[2]} = -1) + \rho(y|x_{i}^{[1]} = 1, x_{i}^{[2]} = 1)} \right)$$
(28)

#### Results - Bit Error Rate Performance

(3, 6) Regular Code of block length N = 96, dimension K = 50, Number of parity checks M = 48,  $10^7$  Bytes Transferred

	No Code		With LDPC	
sigma	Errors	BER	Errors	BER
0.3	4338	0.0004338	0	0
0.4	61814	0.0061814	0	0
0.5	227228	0.0227228	31	3.1e-06
0.6	477744	0.0477744	65855	0.0065855
0.7	765326	0.0765326	572393	0.0572393
0.8	1055848	0.1055848	968856	0.0968856
0.9	1332039	0.1332039	1291208	0.1291208
1.0	1585653	0.1585653	1565803	0.1565803

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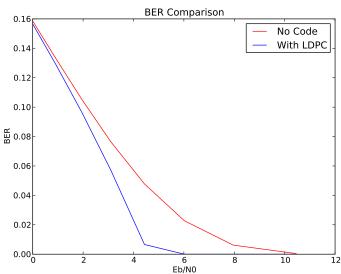
#### Results - Block Error Rate Performance

LDPC generated using PEG Algorithm, Block length N = 1008, Dimension K = 504Number of Parity Checks M = 504

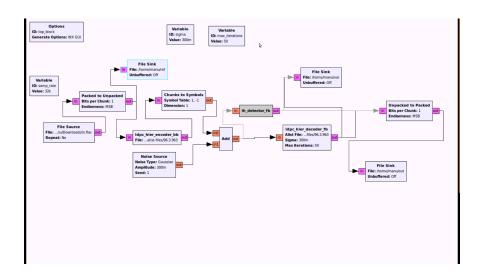
	No Code		With LDPC	
sigma	Errors	BER	Errors	BER
0.3	3898	0.196461	0	0
0.4	18938	0.954488	0	0
0.5	19840	0.999994	0	0
0.6	19841	1.0	17	0.0008568
0.7	19841	1.0	19841	1.0

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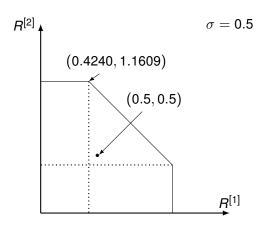
#### Results - Plots



## gr-Idpc module in GNU Radio



#### Rate - Capacity comparison



For single user at  $\sigma=$  0.8 (capacity = 0.678) all blocks were decoded correctly.

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#### **Future Work**

- 2 users with sample offset of 0.5 symbol duration.
- Encoder using sparse LU decomposition
- Using VOLK

