

# Low Density Parity Check Codes Programming Exercises

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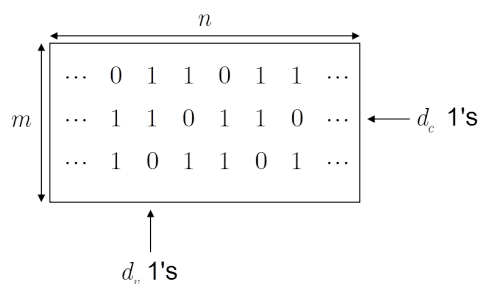
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## LDPC codes

### Definition

A regular- $(d_v, d_c)$  binary low-density parity-check (LDPC) code of length  $n$  is a linear binary block code whose  $m \times n$  parity-check matrix  $\mathbf{H}$  is sparse with exactly  $d_v$  1's on each column and  $d_c$  1's on each row, under the constraints  $d_v \ll n$  and  $d_c \ll n$  and  $md_c = nd_v$ .



The code rate is given by

$$R_c \geq \frac{n-m}{n} = 1 - d_v/d_c$$

## Presentation outline

- 1 LDPC codes and Tanner graphs
- 2 Random generation and storage of LDPC matrices
- 3 Message-passing decoding (MPD)
- 4 Monte Carlo simulation

## LDPC code structure

### struct ldpc

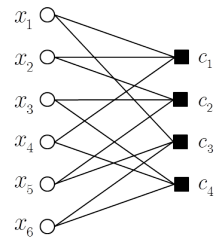
- number  $n$  of columns of LDPC matrix = code length
- number  $m$  of rows of LDPC matrix
- code dimension  $k = \text{rank}(\mathbf{G}) = n - \text{rank}(\mathbf{H})$
- row degree profile  $\mathbf{d}_c$
- column degree profile  $\mathbf{d}_v$
- ldpc matrix, sparse format, non-zero entries
- indicates if the next two fields are empty or not
- submatrix  $\mathbf{P}$  in  $\mathbf{G}$  of dimensions  $k \times m$
- column permutation array (see above)

## Tanner graph

### Definition

- A vector  $\mathbf{x}$  is a valid configuration (codeword) of the behavior defined by the code iff all parity-check equations are simultaneously satisfied.
- The FG represents the factored characteristic function of this behavior.

$$\mathbf{H} = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 1 & 0 & 1 \end{bmatrix}$$



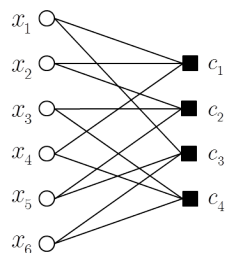
$$\mathcal{N}_v(j) = \{i : \exists \text{ an edge } (x_j, c_i)\} \quad \mathcal{N}_c(i) = \{j : \exists \text{ an edge } (x_j, c_i)\}$$

## Tanner graph structure

### struct bipart

- number  $n_v$  of 1st type vertices (variable nodes)
- number  $n_c$  of 2nd type vertices (check nodes)
- number  $n_e$  of edges
- degree profile  $\mathbf{d}_v$  of 1st type vertices (variable nodes)
- degree profile  $\mathbf{d}_c$  of 2nd type vertices (check nodes)
- neighborhood  $\mathcal{N}_v$  of 1st type vertices (variable nodes)
- neighborhood  $\mathcal{N}_c$  of 2nd type vertices (check nodes)
- edge ensemble (first indexing convention)

## Tanner graph structure



$$\mathcal{N}_v(1) = \{1, 3\}$$

$$\mathcal{N}_v(2) = \{1, 2\}$$

$$\mathcal{N}_v(3) = \{2, 4\}$$

$$\mathcal{N}_v(4) = \{1, 4\}$$

$$\mathcal{N}_v(5) = \{2, 3\}$$

$$\mathcal{N}_v(6) = \{3, 4\}$$

$$\mathcal{N}_c(1) = \{1, 2, 4\}$$

$$\mathcal{N}_c(2) = \{2, 3, 5\}$$

$$\mathcal{N}_c(3) = \{1, 5, 6\}$$

$$\mathcal{N}_c(4) = \{3, 4, 6\}$$

$$e(1) = (v_1, c_1)$$

$$e(2) = (v_2, c_1)$$

$$e(3) = (v_4, c_1)$$

$$e(4) = (v_2, c_2)$$

$$e(5) = (v_3, c_2)$$

$$e(6) = (v_5, c_2)$$

$$e(7) = (v_1, c_3)$$

$$e(8) = (v_5, c_3)$$

$$e(9) = (v_6, c_3)$$

$$e(10) = (v_3, c_4)$$

$$e(11) = (v_4, c_4)$$

$$e(12) = (v_6, c_4)$$

## Gallager's generation method I

### Algorithm 1 Generation of a $(d_v, d_c)$ -regular binary LDPC matrix

**Require:**  $m, n, d_v, d_c$  under condition  $m \cdot d_c = n \cdot d_v$

**Ensure:** binary  $(d_v, d_c)$ -regular LDPC matrix of dimensions  $m \times n$

- 1: generate the first submatrix  $\mathbf{H}_1$  of dimensions  $m/d_v \times n$  such that the  $i$ -th row,  $i \leq m/d_v$ , has non-zero entries in the  $((i-1)d_c + 1)$ -th to  $id_c$ -th columns
- 2: **for**  $j = 2$  to  $d_v$  **do**
- 3:   construct  $\mathbf{H}_j$  by randomly permutated the columns of  $\mathbf{H}_1$
- 4: **end for**

## MacKay and Neal's generation method

### Definition

- Consider the partial matrix  $\mathbf{H}^{(j)}$  of dimensions  $m \times j$  made of  $j \leq n$  already generated columns. The row degree profile  $\mathbf{d}_c^{(j)}$  of  $\mathbf{H}^{(j)}$  is the vector defined as

$$d_{c,i}^{(j)} = w_H(\mathbf{h}_i^{(j)}), \forall i \in \{1, \dots, m\} \quad (1)$$

where  $\mathbf{h}_i^{(j)}$  is the  $i^{\text{th}}$  row of  $\mathbf{H}^{(j)}$

- The row degree profile  $\mathbf{d}_c^{(j)}$  satisfies the degree constraints if

$$d_{c,i}^{(j)} \leq d_c, \forall i \in \{1, \dots, m\} \quad (2)$$

- The row  $i$  is degree-saturated if  $d_{c,i}^{(j)} = d_c$

## MacKay and Neal's generation method II

```

13:   else
14:     randomly generate weight- $d_v$  column  $\tilde{\mathbf{h}}^j$ 
15:     evaluate  $\tilde{\mathbf{d}}_c^{(j)}$  for  $\tilde{\mathbf{H}}^{(j)} = \begin{bmatrix} \mathbf{H}^{(j-1)} & \tilde{\mathbf{h}}^j \end{bmatrix}$ 
16:   end if
17:   until  $\tilde{\mathbf{d}}_c^{(j)}$  satisfies the degree constraints
18:      $\mathbf{H}^{(j)} \leftarrow \tilde{\mathbf{H}}^{(j)}$ 
19:      $\mathbf{d}_c^{(j)} \leftarrow \tilde{\mathbf{d}}_c^{(j)}$ 
20: end for

```

## MacKay and Neal's generation method I

### Algorithm 2 Generation of a $(d_v, d_c)$ -regular binary LDPC matrix

**Require:**  $m, n, d_v, d_c$  under condition  $m \cdot d_c = n \cdot d_v$

**Ensure:** binary  $(d_v, d_c)$ -regular LDPC matrix of dimensions  $m \times n$

```

1: set  $t \leq n$ , set  $\theta_1$  to a large value
2:  $j \leftarrow 1$ 
3: backtrack:
4:  $s \leftarrow \max\{1, j - t\}$ 
5: optional: increase  $t$  ( $t \leq n$ )
6: delete the last  $t + 1$  columns of  $\mathbf{H}^{(j)}$  and store it as  $\mathbf{H}^{(s-1)}$ 
7: for  $j = s$  to  $n$  do
8:    $c1 \leftarrow 0$ 
9:   repeat
10:     $c1 \leftarrow c1 + 1$ 
11:    if  $c1 > \theta_1$  then
12:      goto backtrack

```

## Storage of sparse LDPC matrices

Example: a length-12  $(3, 4)$ -regular binary matrix

$$\mathbf{H} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 \end{bmatrix} \quad (3)$$

## Storage of sparse LDPC matrices

### Principle

- For each row (or each column), store only the non-zero positions

Example: sparse storage scheme by row.

$$\mathbf{H} = \begin{array}{c|cccc} r1 & 1 & 6 & 8 & 10 \\ r2 & 1 & 4 & 5 & 11 \\ r3 & 2 & 5 & 7 & 9 \\ r4 & 3 & 6 & 11 & 12 \\ r5 & 3 & 7 & 8 & 12 \\ r6 & 2 & 5 & 9 & 11 \\ r7 & 1 & 4 & 7 & 10 \\ r8 & 2 & 6 & 8 & 10 \\ r9 & 3 & 4 & 9 & 12 \end{array} \quad (4)$$

## MacKay and Neal's generation method I

**Algorithm 3** Generation of a  $(\mathbf{d}_v, \mathbf{d}_c)$ -irregular binary LDPC matrix

**Require:**  $m, n, \mathbf{d}_v, \mathbf{d}_c$  under condition  $\sum_{i=1}^m d_{c,i} = \sum_{j=1}^n d_{v,j}$

**Ensure:** binary irregular LDPC matrix of dimensions  $m \times n$

```

1: set  $t \leq n$ , set  $\theta_1$  to a large value
2:  $j \leftarrow 1$ 
3: backtrack:
4:  $s \leftarrow \max\{1, j - t\}$ 
5: optional: increase  $t$  ( $t \leq n$ )
6: delete the last  $t + 1$  columns of  $\mathbf{H}^{(j)}$  and store it as  $\mathbf{H}^{(s-1)}$ 
7: for  $j = s$  to  $n$  do
8:    $c1 \leftarrow 0$ 
9:   repeat
10:     $c1 \leftarrow c1 + 1$ 
11:    if  $c1 > \theta_1$  then
12:      goto backtrack

```

## MacKay and Neal's generation method

### Definition

- The row degree profile  $\mathbf{d}_c^{(j)}$  satisfies the degree constraints if

$$d_{c,i}^{(j)} \leq d_{c,i}, \forall i \in \{1, \dots, m\} \quad (5)$$

- The row  $i$  is degree-saturated if  $d_{c,i}^{(j)} = d_{c,i}$

## MacKay and Neal's generation method II

```

13: else
14:   randomly generate weight- $d_v$  column  $\tilde{\mathbf{h}}^j$ 
15:   evaluate  $\tilde{\mathbf{d}}_c^{(j)}$  for  $\tilde{\mathbf{H}}^{(j)} = \begin{bmatrix} \mathbf{H}^{(j-1)} & \tilde{\mathbf{h}}^j \end{bmatrix}$ 
16: end if
17: until  $\tilde{\mathbf{d}}_c^{(j)}$  satisfies the degree constraints
18:  $\mathbf{H}^{(j)} \leftarrow \tilde{\mathbf{H}}^{(j)}$ 
19:  $\mathbf{d}_c^{(j)} \leftarrow \tilde{\mathbf{d}}_c^{(j)}$ 
20: end for

```

## Storage of sparse LDPC matrices

Example: A length-10 irregular binary matrix

$$\mathbf{H} = \begin{bmatrix} 1 & 1 & 0 & 1 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 1 & 1 \\ 1 & 1 & 0 & 0 & 0 & 1 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 0 & 1 & 0 & 1 \end{bmatrix} \quad (6)$$

## Storage of sparse LDPC matrices

### Principle

For each row (or each column), store only the non-zero positions

Example: sparse storage scheme by row.

$$\mathbf{H} = \begin{bmatrix} r1 & 1 & 2 & 4 & 5 & 8 \\ r2 & 2 & 3 & 5 & 6 & 7 \\ r3 & 4 & 8 & 9 & 10 & \\ r4 & 1 & 2 & 6 & 7 & 9 \\ r5 & 3 & 6 & 8 & 10 & \end{bmatrix} \quad (7)$$

## MacKay and Neal's generation method I

**Algorithm 4** Generation of a full-rank  $(\mathbf{d}_v, \mathbf{d}_c)$ -irregular binary LDPC matrix

**Require:**  $m, n, \mathbf{d}_v, \mathbf{d}_c$  under condition  $\sum_{i=1}^m d_{c,i} = \sum_{j=1}^n d_{v,j}$

**Ensure:** binary irregular LDPC matrix of dimensions  $m \times n$

- 1: set  $t \leq n$ , set  $\theta_1, \theta_2$  to large values
- 2:  $j \leftarrow 1$
- 3:  $c1 \leftarrow 0$
- 4: **repeat**
- 5:    $c1 \leftarrow c1 + 1$
- 6:   **if**  $c1 > \theta_1$  **then**
- 7:     **exit with failure**
- 8:   **end if**
- 9:   **backtrack:**
- 10:    $s \leftarrow \max\{1, j - t\}$
- 11:   optional: increase  $t$  ( $t \leq n$ )

## MacKay and Neal's generation method II

- 12: delete the last  $t + 1$  columns of  $\mathbf{H}^{(j)}$  and store it as  $\mathbf{H}^{(s-1)}$
- 13: **for**  $j = s$  to  $n$  **do**
- 14:    $c2 \leftarrow 0$
- 15:   **repeat**
- 16:      $c2 \leftarrow c2 + 1$
- 17:     **if**  $c2 > \theta_2$  **then**
- 18:       **goto** backtrack
- 19:     **else**
- 20:       randomly generate weight- $d_v$  column  $\tilde{\mathbf{h}}^j$
- 21:       evaluate  $\tilde{\mathbf{d}}_c^{(j)}$  for  $\tilde{\mathbf{H}}^{(j)} = \begin{bmatrix} \mathbf{H}^{(j-1)} & \tilde{\mathbf{h}}^j \end{bmatrix}$
- 22:       **end if**
- 23:       **until**  $\tilde{\mathbf{d}}_c^{(j)}$  satisfies the degree constraints
- 24:        $\mathbf{H}^{(j)} \leftarrow \tilde{\mathbf{H}}^{(j)}$
- 25:        $\mathbf{d}_c^{(j)} \leftarrow \tilde{\mathbf{d}}_c^{(j)}$
- 26:   **end for**

## MacKay and Neal's generation method III

```

27:   compute rank(H)
28: until rank(H) =  $m$ 

```

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## MacKay and Neal's generation method II

```

12:   delete the last  $t + 1$  columns of  $\mathbf{H}^{(j)}$  and store it as  $\mathbf{H}^{(s-1)}$ 
13:   for  $j = s$  to  $n$  do
14:      $c2 \leftarrow 0$ 
15:     repeat
16:        $c2 \leftarrow c2 + 1$ 
17:       if  $c2 > \theta_2$  then
18:         goto backtrack
19:       end if
20:        $c3 \leftarrow 0$ 
21:       repeat
22:          $c3 \leftarrow c3 + 1$ 
23:         if  $c3 > \theta_3$  then
24:           goto backtrack
25:         else
26:           randomly generate weight- $d_v$  column  $\tilde{\mathbf{h}}^j$ 

```

## MacKay and Neal's generation method I

**Algorithm 5** Generation of a full-rank  $(\mathbf{d}_v, \mathbf{d}_c)$ -irregular binary LDPC matrix without length-4 cycles

**Require:**  $m, n, \mathbf{d}_v, \mathbf{d}_c$  under condition  $\sum_{i=1}^m d_{c,i} = \sum_{j=1}^n d_{v,j}$

**Ensure:** binary irregular LDPC matrix of dimensions  $m \times n$

```

1: set  $t \leq n$ , set  $\theta_1, \theta_2, \theta_3$  to large values
2:  $j \leftarrow 1$ 
3:  $c1 \leftarrow 0$ 
4: repeat
5:    $c1 \leftarrow c1 + 1$ 
6:   if  $c1 > \theta_1$  then
7:     exit with failure
8:   end if
9:   backtrack:
10:   $s \leftarrow \max\{1, j - t\}$ 
11:  optional: increase  $t$  ( $t \leq n$ )

```

## MacKay and Neal's generation method III

```

27:           evaluate  $\tilde{\mathbf{d}}_c^{(j)}$  for  $\tilde{\mathbf{H}}^{(j)} = \begin{bmatrix} \mathbf{H}^{(j-1)} & \tilde{\mathbf{h}}^j \end{bmatrix}$ 
28:           end if
29:           until  $\tilde{\mathbf{d}}_c^{(j)}$  satisfies the degree constraints
30:           check the presence of length-4 cycles in  $\tilde{\mathbf{H}}^{(j)}$ 
31:           until no length-4 cycle detected
32:            $\mathbf{H}^{(j)} \leftarrow \tilde{\mathbf{H}}^{(j)}$ 
33:            $\mathbf{d}_c^{(j)} \leftarrow \tilde{\mathbf{d}}_c^{(j)}$ 
34:         end for
35:       compute rank(H)
36: until rank(H) =  $m$ 

```

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## Pseudo-code I

**Algorithm 6** LBP for the BEC**Require:** Tanner graph of the code, received word  $\mathbf{y}$ 

```

1: preliminary test:
2: if  $\forall j \in \{1, \dots, n\} \ y_j \neq e$  then
3:   return  $\mathbf{y}$ 
4: end if
5: define two arrays  $\Phi$  and  $\Psi$  of dimensions  $n \times m$  and  $m \times n$ 
6:  $l \leftarrow 0$ 
7: init: compute beliefs
8: for  $j = 1$  to  $n$  do
9:   if  $y_j \neq e$  then
10:     $\beta_j \leftarrow y_j$ 
11:   end if
12: end for
13: init: compute bit-to-check messages

```

## Pseudo-code II

```

14: for  $j = 1$  to  $n$  do
15:   for all  $i \in \mathcal{N}_v(j)$  do
16:      $\Phi_{j,i} \leftarrow y_j$ 
17:   end for
18: end for
19: repeat
20:    $l \leftarrow l + 1$ 
21:   step: compute check-to-bit messages
22:   for  $i = 1$  to  $m$  do
23:     for all  $j \in \mathcal{N}_c(i)$  do
24:       if  $\forall k \in \mathcal{N}_c(i) \setminus \{j\}, \Phi_{k,i} \neq e$  then
25:          $\Psi_{i,j} \leftarrow \sum_{k \in \mathcal{N}_c(i) \setminus \{j\}} \Phi_{k,i} \pmod{2}$ 
26:       else
27:          $\Psi_{i,j} \leftarrow e$ 
28:       end if

```

## Pseudo-code III

```

29:   end for
30: end for
31: step: compute beliefs
32: for  $j = 1$  to  $n$  do
33:   if  $y_j \neq e$  then
34:      $\beta_j \leftarrow y_j$ 
35:   else
36:     for all  $i \in \mathcal{N}_v(j)$  do
37:       if  $\forall k \in \mathcal{N}_v(j), \Psi_{k,j} = e$  then
38:          $\beta_j \leftarrow e$ 
39:       else if  $\exists k \in \mathcal{N}_v(j)$  s.t.  $\Psi_{k,j} \neq e$  then
40:          $\beta_j \leftarrow \Psi_{k,j}$ 
41:       end if
42:     end for
43:   end if

```

## Pseudo-code IV

```

44: end for
45: step: compute bit-to-check messages
46: for  $j = 1$  to  $n$  do
47:   if  $y_j \neq e$  then
48:     for all  $i \in \mathcal{N}_v(j)$  do
49:        $\Phi_{j,i} \leftarrow y_j$ 
50:     end for
51:   else
52:     for all  $i \in \mathcal{N}_v(j)$  do
53:       if  $\forall k \in \mathcal{N}_v(j) \setminus \{i\}, \Psi_{k,j} = e$  then
54:          $\Phi_{j,i} \leftarrow e$ 
55:       else if  $\exists k \in \mathcal{N}_v(j) \setminus \{i\}$  s.t.  $\Psi_{k,j} \neq e$  then
56:          $\Phi_{j,i} \leftarrow \Psi_{k,j}$ 
57:       end if
58:     end for

```

## Pseudo-code V

```

59:   end if
60: end for
61: test: stopping criterion
62: if all bit-to-check messages are either 0 or 1  $\vee l = l_{\max}$  then
63:    $STOP \leftarrow 1$ 
64: end if
65: until  $STOP = 1$ 
66: return  $\beta$ 

```

## Pseudo-code II

```

11: end if
12: define two arrays  $\Phi$  and  $\Psi$  of dimensions  $n \times m$  and  $m \times n$ 
13: init: get LLRs from channel (in parallel)
14: for  $j = 1$  to  $n$  do
15:   for all  $k \in \mathcal{N}_v(j)$  do
16:      $\Phi_{j,k} \leftarrow \Gamma_j$ 
17:   end for
18: end for
19: for  $it = 1$  to  $N_{it}$  do
20:   step 1: compute check-to-bit messages
21:   for  $i = 1$  to  $m$  do
22:     for all  $k \in \mathcal{N}_c(i)$  do
23:        $\Psi_{i,k} \leftarrow 2 \tanh^{-1} \left( \prod_{k' \in \mathcal{N}_c(i) \setminus \{k\}} \tanh \left( \frac{\Phi_{k',i}}{2} \right) \right)$ 
24:     end for
25:   end for

```

## Pseudo-code I

## Algorithm 7 LBP for BMC

**Require:** Tanner graph, vector  $\Gamma$  of log likelihood ratios (LLRs) of coded bits, maximum number  $N_{it}$  of iterations

**Ensure:** vector  $\Lambda$  of log (approximated) APPs of coded bits and/or estimated codeword  $\hat{\mathbf{x}}$

```

1: preliminary test: check syndrome
2: for  $j = 1$  to  $n$  do
3:   if  $\Gamma_j \geq 0$  then
4:      $\hat{x}_j \leftarrow 0$ 
5:   else
6:      $\hat{x}_j \leftarrow 1$ 
7:   end if
8: end for
9: if  $\mathbf{z} = \hat{\mathbf{x}}\mathbf{H}^\top = \mathbf{0}$  then
10:  return  $\hat{\mathbf{x}}$ 

```

## Pseudo-code III

```

26: stop criterion: check syndrome
27: step: compute beliefs
28: for  $j = 1$  to  $n$  do
29:    $\Lambda_j \leftarrow \Gamma_j + \sum_{k \in \mathcal{N}_v(j)} \Psi_{k,j}$ 
30:   if  $\Lambda_j \geq 0$  then
31:      $\hat{x}_j \leftarrow 0$ 
32:   else
33:      $\hat{x}_j \leftarrow 1$ 
34:   end if
35: end for
36: if  $\mathbf{z} = \hat{\mathbf{x}}\mathbf{H}^\top = \mathbf{0}$  then
37:   return  $\Lambda$  and/or  $\hat{\mathbf{x}}$ 
38: end if
39: step 2: compute bit-to-check messages
40: for  $j = 1$  to  $n$  do

```



## Pseudo-code IV

---

```

41:   for all  $k \in \mathcal{N}_v(j)$  do
42:      $\Phi_{j,k} \leftarrow \Gamma_j + \sum_{k' \in \mathcal{N}_v(j) \setminus \{k\}} \Psi_{k',j}$ 
43:     //  $\Phi_{j,k} \leftarrow \Lambda_j - \Psi_{k,j}$  (to optimize the execution speed) //
44:   end for
45: end for
46: end for

```

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## Monte Carlo simulation I

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**Algorithm 8** Performance of an LDPC codes over a BIAWGNC under message-passing decoding (SPA) (length  $2^8 \leq n \leq 2^{20}$ )

---

**Require:** generator matrix  $\mathbf{G}_{rr}$ , Tanner graph for LDPC matrix  $\mathbf{H}$

**Ensure:** bit error probability  $P_b$ , block error probability  $P_e$

```

1: init: open and read files, allocate workspace, store all structures, etc.
2: for all  $SNR$  in  $SNR$  range (dB) do
3:   compute the noise variance  $\sigma$ 
4:   initialize bit and block error counters
5:   repeat
6:     randomly generate1 an information vector  $\mathbf{u} \in \mathbb{F}_2^k$ 
7:     encode  $\mathbf{u}$  into codeword  $\mathbf{x} = \mathbf{u}\mathbf{G}_{rr} \in \mathbb{F}_2^n$  (systematic encoding)
8:     BPSK modulate2 codeword  $\mathbf{x}$  into vector  $\mathbf{s} \in \{\pm 1\}^n$ 
9:     transmit  $\mathbf{s}$  on the BIAWGNC( $\sigma$ ) and generate the output3  $\mathbf{y} \in \mathbb{R}^n$ 
10:    compute the vector of LLRs4  $\mathbf{\Gamma} \in \mathbb{R}^n$ 
11:    permute the components of  $\mathbf{\Gamma}$  if necessary5

```

## Monte Carlo simulation II

---

```

12:   activate LBP6 and find the vector of (approx) LAPPs  $\mathbf{\Lambda} \in \mathbb{R}^n$ 
13:   deduce from  $\mathbf{\Lambda}$  an estimate  $\hat{\mathbf{u}} \in \mathbb{F}_2^k$  of  $\mathbf{u}$ 
14:   update bit and block error counters
15:   until  $10^6$  codewords simulated or 500 codeword errors
16:   compute bit error probability and block error probability
17: end for
18: plot  $P_b = f(SNR)$  and  $P_e = g(SNR)$ 

```

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<sup>1</sup>Due to symmetries (channel and message-passing decoder), the average bit (resp. block) error probability is equal to the bit (resp. block) error probability conditional on all-zero codeword transmitted.

<sup>2</sup> $\forall l \in \{1, \dots, n\}$ ,  $s_l = (-1)^{x_l}$ .

<sup>3</sup> $\forall l \in \{1, \dots, n\}$ ,  $y_l = s_l + w_l$  where  $w_l \sim \mathcal{N}(0, \sigma^2)$  with  $\sigma^2 = \frac{N_0}{2R_c E_b}$ .

<sup>4</sup> $\forall l \in \{1, \dots, n\}$ ,  $\Gamma_l = \log \frac{p(y_l | x_l=0)}{p(y_l | x_l=1)} = \frac{2}{\sigma^2} y_l$  (prove it).

<sup>5</sup>The codes used to encode and decode are equivalent but not necessarily identical.

<sup>6</sup> $N_{it}$  is fixed and typically large, e.g.,  $N_{it} = 100$ , to ensure LBP convergence.