

## Lecture 33: LDPC Codes

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Low Density Parity Check (LDPC) codes were first introduced by Gallager [1] in his thesis. After the invention, these codes were largely forgotten and there was “no activity” till mid-late 90s when there was a renewed spurt of research activity on LDPC codes. Available resource for the state-of-the-art in LDPC codes can be found in Richardson-Urbanke book [4]. For a quick overview, see the survey by Guruswami [2]. In this lecture, we are going mainly to follow the survey by Guruswami. Unlike other classes of codes that we have seen, LDPC codes are accompanied by very fast encoding and decoding algorithms. For example, we have seen that the capacity of the  $BSC_p$  can be achieved with decoding time complexity  $\text{poly}(n)2^{\text{poly}(\frac{1}{\epsilon})}$ . We want the decoding time ideally to be  $n\text{poly}(\frac{1}{\epsilon})$ . LDPC codes provably achieve this though for  $BEC_\alpha$  (Luby *et al* [3]). This makes LDPC codes not only attractive from the theoretical point of view, but also for practical applications. However, we will not discuss LDPC codes of Luby *et al* in this lecture but go through some of Gallager’s results and some brief idea how his ideas extended. The name of the code suggest that it has to do with parity check codes.

## 1 Low Density Parity Check Codes

Recall that any  $[n, k]_2$  code has an  $(n - k) \times n$  parity check matrix  $H$ . Every parity check matrix  $H$  can be associated with a *factor graph*  $\mathcal{G}$ . In particular  $\mathcal{G}$  is a bipartite graph that has  $n$  left vertices, called *variable* nodes, one for each codeword position. On the right are  $n - k$  vertices, called *check* nodes, one for each parity check (or row in  $H$ ). A check node is adjacent to all variable nodes whose corresponding codeword symbols appear in the corresponding row in  $H$ . In other words,  $H$  is precisely the adjacency matrix of  $\mathcal{G}$ .

Let us fix some notation that we will use later. Recall that a code  $C$  with parity check matrix  $H$  is defined as follows: for every  $\mathbf{c} \in C$ ,  $H\mathbf{c}^T = \mathbf{0}$ . In other words there are  $(n - k)$  constraints on  $(c_1, \dots, c_n)$  variables, where  $c_i$  denotes the  $i$ th bit in a codeword. Let the  $j$ th ( $1 \leq j \leq n - k$ ) parity check (or equivalently  $H_j(c_1, \dots, c_n)^T$ ) be denoted by  $p_j$ . There is an edge  $(c_i, p_j)$  in  $\mathcal{G}$  if  $c_i$  occurs in  $p_j$  or equivalently  $H_{ji} = 1$ , where  $H_{ji}$  is the entry corresponding to row  $j$  and column  $i$  in  $H$ . For example  $[7, 4, 3]_2$  Hamming code has a  $3 \times 7$  parity check matrix  $H$ :

$$\begin{pmatrix} 0 & 0 & 0 & 1 & 1 & 1 & 1 \\ 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{pmatrix}$$

and the corresponding factor graph of  $H$  is:

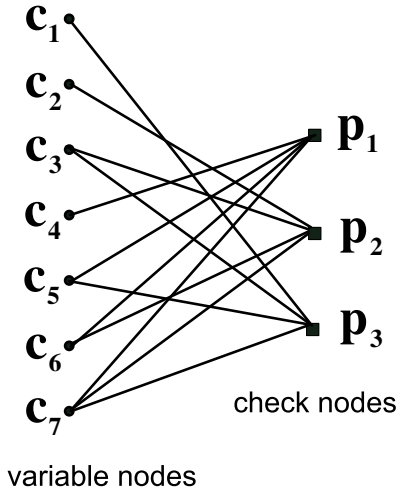


Figure 1: Factor graph of the Hamming code

The number of edges equals the number of ones in the parity check matrix. According to definition of sparse bipartite graph, the number of ones in parity check matrix  $H$  is bounded by  $O(n)$ . Equivalently the number of edges in the factor graph is bounded by  $O(n)$ .

**Definition 1.1.** A special class of LDPC codes are regular LDPC codes where the factor graph has left vertexes with degree exactly  $d_v$  and every right vertexes with degree exactly  $d_c$  for integers  $d_v, d_c \geq 1$ .

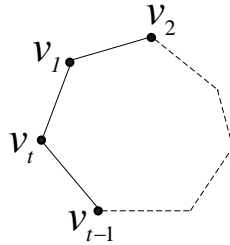
This regular LDPC codes were originally studied by Gallager. We claim that the rate  $R$  of  $(d_v, d_c)$ -regular LDPC code is equal to

$$R = 1 - \frac{d_v}{d_c}.$$

This is because the number of edges of factor graph is equal to  $d_v n = d_c(n - k)$ . Note that this implies  $d_c > d_v$  for positive  $R$ . We will consider families of regular LDPC codes that are obtained by constructing a family of  $(d_v, d_c)$ -regular factor graphs with  $n$  variable nodes with girth greater than  $\Omega(\log n)$ . We will also need the following graph-theoretic notion.

**Definition 1.2.** *Girth of a graph is the size of the smallest cycle.*

Cycle



*In this graph  $v_1, \dots, v_t$  is a cycle.*

Gallager proved in the following results [1]:

- (i) With high probability, for large enough  $d_v$  and  $d_c$ , a random  $(d_v, d_c)$ -regular LDPC code achieves the GV bound (for large enough  $n$ ).
- (ii) Random  $(d_v, d_c)$ -regular LDPC codes (along with MLD) get “close” to  $BSC_p$  capacity.

Since only exponential time solutions to the MLD function are known, Gallager also developed simple, iterative decoding algorithm for LDPC codes. We are not going to look (i) or (ii) and going to study the performance of regular LDPC codes over the  $BEC_\alpha$ .

## 1.1 Decoding on the $BEC_\alpha$

Although Gallager did not explicitly study  $BEC_\alpha$ , his methods do apply to it. For the  $BEC_\alpha$  we will study an iterative message-passing decoding algorithm. General idea of such a decoding algorithm is to pass messages in round. Each round has two phases. In the first phase messages are passed from variable nodes to check nodes (of the factor graph). In the second phase, messages are passed back from the check nodes to the variable nodes. Let the received word be  $\mathbf{y} = (y_1, \dots, y_n) \in \{0, 1, ?\}^n$  and for each round messages are sent as follows:

- (i) variable-to-check message phase: If  $(c_i, p_j)$  is an edge then  $c_i$  sends  $p_j$  a message. In particular, if  $y_i \neq ?$  then  $c_i$  passes along  $y_i$  to all its neighboring check nodes.
- (ii) check-to-variable message phase: If  $(p_j, c_i)$  is an edge, the  $p_j$  sends  $c_i$  a message. In the particular, if the check node  $p_j$  knows the correct value for  $c_i$ , it passes the value to  $c_i$ .

In particular, the messages will have the following semantics:

1. If variable  $c_i$  knows its correct value then the message is the value, else the message is an erasure.
2. If check node  $p_j$  knows the value of  $c_i$  then passes that value, else it passes an erasure.

At end (hopefully) every  $c_i$  knows its value with high probability. The message-passing decoding algorithm has the following *extrinsic* property: the message that is sent from,  $c_i$  to  $p_j$  does not depend on message sent by  $p_j$  in the previous round (a similar property holds for messages sent from the variable nodes to the parity check nodes). Formally the message maps are given as follows, message  $c_i \rightarrow p_j$  in round  $t$  is:

$$\Psi_{c_i}^{t,p_j}(y_i, m_1^{t-1}, \dots, m_{d_v-1}^{t-1}) \rightarrow \{0, 1, ?\},$$

where  $m_1^{t-1}, \dots, m_{d_v-1}^{t-1}$  are the  $d_v - 1$  messages received by  $c_i$  from all its neighboring check nodes other than  $p_j$ , in the previous round. Generally the parameter  $p_j$  is dropped because the mapping is the same for every  $p_j$ . The mapping of message  $p_j \rightarrow c_i$  in round  $t$  is:

$$\Psi_{p_j}^{t,c_i}(m_1^t, \dots, m_{d_c-1}^t) \rightarrow \{0, 1, ?\},$$

where again  $m_1^t, \dots, m_{d_c-1}^t$  are the  $d_c - 1$  messages received from all the variable nodes other than  $c_i$  in the current round (recall that the check node to variable nodes phase takes place after variable-to-check node phase in the same round). Similarly for check nodes the parameter  $c_i$  is dropped because generally it is the same for every  $c_i$ .

We now define the message functions. In round  $t = 1$ , message  $c_i \rightarrow p_j$  is computed as follows:

$$\Psi_{c_i}^{0,p_j}(y_i, m_1^0, \dots, m_{d_v-1}^0) = y_i$$

and for  $t \geq 2$  we have:

$$\Psi_{c_i}^{t,p_j}(y_i, m_1^{t-1}, \dots, m_{d_v-1}^{t-1}) = \begin{cases} b & , \text{if at least one of } \{y_i, m_1^{t-1}, \dots, m_{d_v-1}^{t-1}\} \text{ is } b \in \{0, 1\} \\ ? & , \text{if } y_i = m_1^{t-1} = \dots = m_{d_v-1}^{t-1} = ? \end{cases}$$

and message  $p_j \rightarrow c_i$  when  $t \geq 1$  we have:

$$\Psi_{p_j}^{t,c_i}(m_1^t, \dots, m_{d_c-1}^t) = \begin{cases} ? & , \text{if any one of } m_i^t = ? \\ m_1^t \oplus \dots \oplus m_{d_c-1}^t & , \text{otherwise} \end{cases}$$

Recall that the codewords are those vectors  $(c_1, \dots, c_n)$  such that for all check nodes the sum of the neighboring positions among the message nodes is zero, i.e.

$$c_i \oplus m_1^{t-1} \oplus \dots \oplus m_{d_c-1}^{t-1} = 0.$$

In other words if  $\Psi_{c_i}^{t,p_j}(\cdot)$  only transmits correct values in  $\{0, 1\}$ , then  $\Psi_{p_j}^{t,c_i}(\cdot)$  computes the correct bit for  $c_i$  if all its inputs are in  $\{0, 1\}$ .

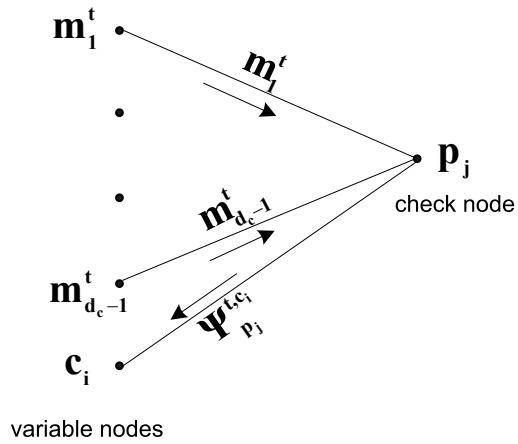


Figure 2: Message computation for  $c_i$

## References

- [1] R.G. Gallager. *Low-Density Parity-Check Codes*. M.I.T. Press, Cambridge, MA, 1963.
- [2] V. Guruswami. Iterative decoding of low-density parity check codes (a survey). *Bulletin of the EATCS*, September 2006.
- [3] M. Luby, M. Mitzenmacher, A. Shokrollahi, and D. Spielman. Efficient erasure correcting codes. *IEEE Transactions on Information Theory*, 47(2):569–584, 2001.
- [4] T. Richardson and R. Urbanke. *Modern Coding Theory*. Cambridge University Press, 2007.