



Information Theory and Channel Coding

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VIII. Low-density parity-check (LDPC) codes

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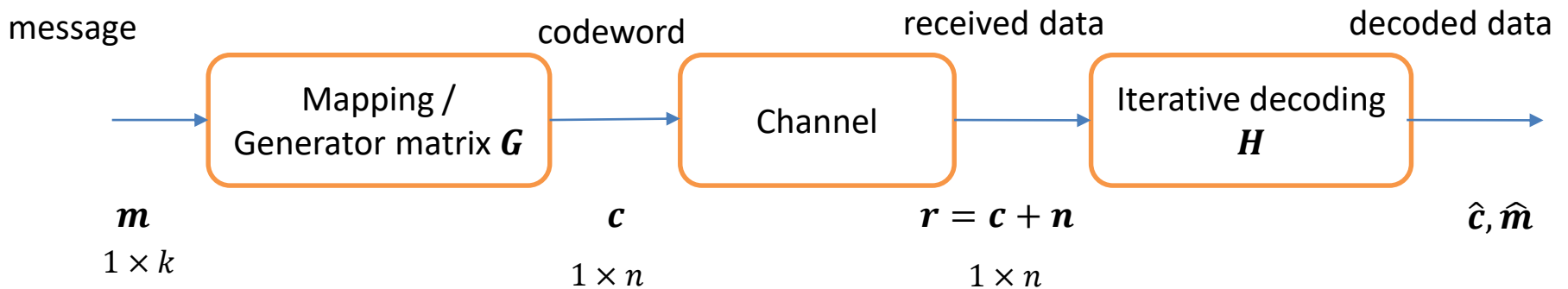
A. Introduction

- LDPC codes are linear block codes that were invented by Robert Gallager in his PhD thesis at MIT in 1960.
- LDPC codes can approach Shannon's theoretical limit by using sparse parity-check matrices and message passing decoding.
- The basic idea consists of designing a parity-check matrix with few ones and many zeros, which would facilitate decoding.
- Decoding for such sparse structures in LDPC codes is carried out by message passing, which is easy to implement.

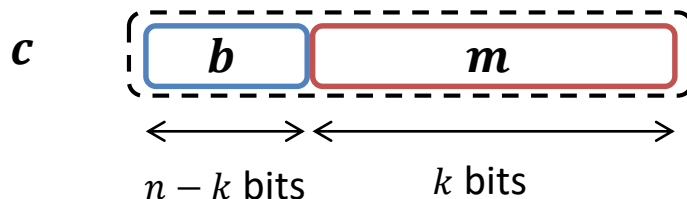




- Let us consider an LDPC system with the following block diagram.



- LDPC codes are denoted by $C(n, k)$ or $C(n, t_c, t_r)$ where k is the length of m in bits, n is the length of c in bits, t_c and t_r are the numbers of ones per column and row of H , respectively.
- The structure of an LDPC code c in systematic form is illustrated by





- LDPC codes are often specified by
 - The parity-check matrix H
 - The block length n
 - The number of ones in each column of H , t_c .
 - The number of ones in each row of H , t_r .
- Unlike other channel codes that focus on the design of the generator matrix G , LDPC codes first design the parity-check matrix H .
- The code rate of LDPC codes is given by

$$R = 1 - \frac{t_c}{t_r} = \frac{k}{n}$$



- The relation of the code rate can be obtained as follows.
- Consider ρ as the density of ones in H , then if we set

$$t_c = \rho \underbrace{(n - k)}_{\text{parity bits}}$$

and

$$t_r = \rho n$$

- We then divide t_c by t_r to obtain

$$\frac{t_c}{t_r} = \frac{\rho(n - k)}{\rho n} = 1 - \frac{k}{n} = 1 - R \rightarrow R = 1 - \frac{t_c}{t_r}$$

- Note that the parity-check matrix H is often not systematic and requires a procedure in its design to ensure a systematic structure for it.



B. Encoding

- Let us now describe the encoding procedure of LDPC codes and focus on the systematic form for the sake of simplicity.
- The parity bits can be described in matrix form by

$$\mathbf{b} = \mathbf{m} \mathbf{P},$$

where $\mathbf{P} = \begin{bmatrix} p_{0,0} & \cdots & p_{0,n-k-1} \\ \vdots & \ddots & \vdots \\ p_{k-1,0} & \cdots & p_{k-1,n-k-1} \end{bmatrix} \in \mathbb{F}^{k \times (n-k)}$ is the parity matrix.

- The codeword is described by

$$\begin{aligned} \mathbf{c} &= [\mathbf{b} \mid \mathbf{m}] \\ &= [\mathbf{m} \mathbf{P} \mid \mathbf{m}] = \mathbf{m} \underbrace{[\mathbf{P} \mid \mathbf{I}_k]}_{\mathbf{G}} = \mathbf{m} \mathbf{G}, \end{aligned}$$

where $\mathbf{G} \in \mathbb{F}^{k \times n}$ is the generator matrix.



- The parity-check matrix $H \in \mathbb{F}^{n-k \times n}$ of LDPC codes is key for both design and decoding, and should be sparse and structured as follows:

$$\begin{aligned} H &= [I_{n-k} \mid P^T] \in \mathbb{F}^{n-k \times n} \\ &= \left[\begin{array}{c|c} \underbrace{H_1^T}_{n-k \times n-k} & \underbrace{H_2^T}_{n-k \times k} \end{array} \right] \end{aligned}$$

- With the above partitioning and by imposing the constraint $cH^T = \mathbf{0}$, we have

$$\begin{aligned} cH^T &= [b \mid m] \begin{bmatrix} H_1 \\ H_2 \end{bmatrix} = \mathbf{0} \\ bH_1 + mH_2 &= \mathbf{0} \\ mPH_1 + mH_2 &= \mathbf{0} \\ m(PH_1 + H_2) &= \mathbf{0} \end{aligned}$$

- The non trivial solution is given by

$$P = H_2 H_1^{-1}$$



- Therefore, the generator matrix G for LDPC codes is given by

$$\begin{aligned} G &= [P \mid I_k] \\ &= [H_2 H_1^{-1} \mid I_k] \in \mathbb{F}^{k \times n} \end{aligned}$$

- In the design of the generator matrix G for LDPC codes we need to take care with H_1 so that it is non singular and its inverse exists.
- In the design of the parity-check matrix H we often take care with the existence of the inverse of H_1 before we proceed to obtain G .



Example 1

Consider an LDPC code with $n = 10$, $t_c = 3$ and $t_r = 5$ and the following parity-check matrix:

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

- Compute the rate of the LDPC code
- Compute the partitioned matrices \mathbf{H}_1 and \mathbf{H}_2
- Determine the generator matrix \mathbf{G}
- Compute the codeword for $\mathbf{m} = [1 \ 0 \ 0 \ 1]$



Solution:

a) The rate of the LDPC code is given by

$$R = 1 - \frac{t_c}{t_r} = 1 - \frac{3}{5} = 0.4$$

b) The partitioned matrices are

$$H_1^T = \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 1 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 1 & 1 \\ 0 & 1 & 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \text{ and } H_2^T = \begin{bmatrix} 0 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 0 & 0 & 1 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 \end{bmatrix}$$



c) The generator matrix can be obtained by

$$\begin{aligned} G &= [P \mid I_k] \\ &= [H_2 H_1^{-1} \mid I_k] \end{aligned}$$

Substituting the values of H_1 and H_2 , we obtain

$$G = \begin{bmatrix} 1 & 0 & 0 & 1 & 1 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 1 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

d) The codeword produced by $m = [1 \ 0 \ 01]$ is given by

$$c = mG = [1 \ 1 \ 0 \ 0 \ 0 \ 0 \mid 1 \ 0 \ 01]$$

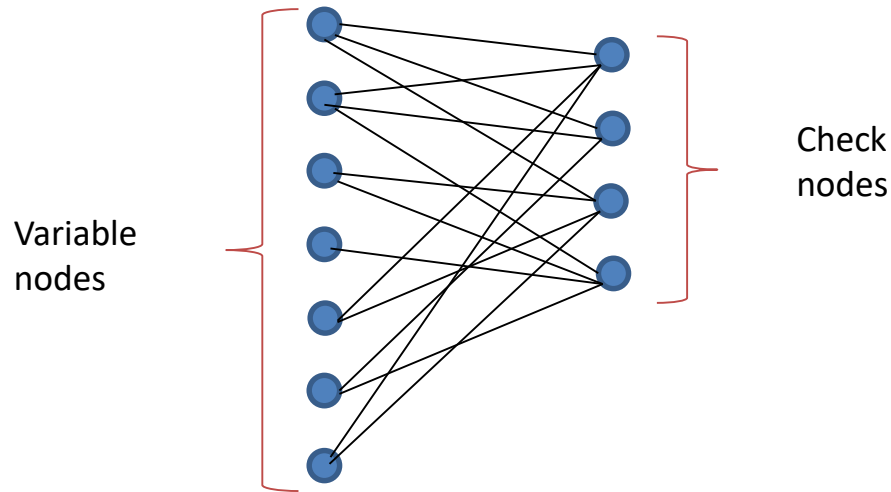


C. Structure and design of LDPC codes

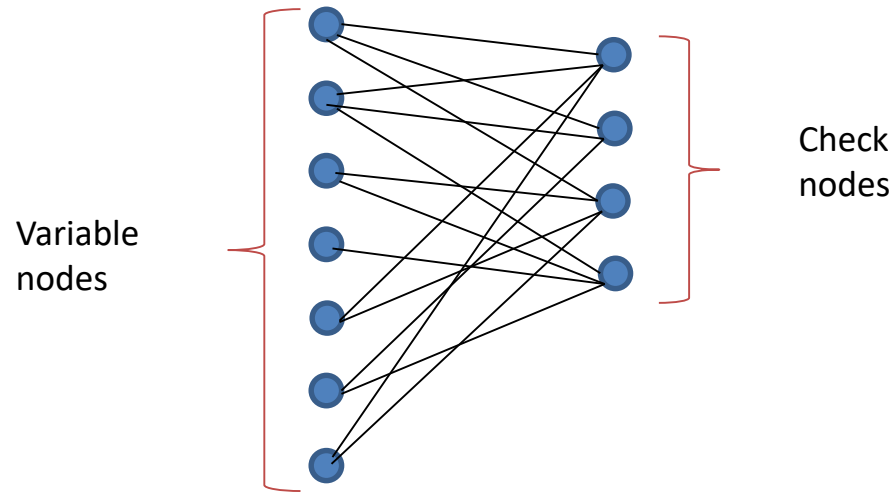
- LDPC codes can be described by bipartite graphs known as Tanner graphs.
- The basic idea is to employ a bipartite graph to describe the parity-check matrix of an LDPC code.
- According to the Tanner graph the variable nodes correspond to the elements of the codeword.
- The check nodes correspond to the parity-check constraints of the LDPC code.

R. M. Tanner, "A recursive approach to low-complexity codes," IEEE Trans. Inform. Theory, vol. 27, pp. 535-547, Sept. 1981

- A Tanner graph can be illustrated by



- A check node j is connected to variable node i when the element of H is a one.
- The $m = n - k$ rows of H specify the m check node connections.
- The n columns of H specify the n variable node connections.



- Cycles: a path comprising ν edges which closes back to itself.
- Girth (γ): it is the minimum cycle length of the Tanner graph.
- Note: the shortest possible girth is 4 $\rightarrow \mathbf{H} = \begin{bmatrix} 0 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 \end{bmatrix}$



Example 2

Consider a (n, t_c, t_r) LDPC code with the following parity-check matrix

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

- Determine the parameters of the code such as rate, t_c and t_r .
- Draw the bipartite graph, also known as as the Tanner graph.
- Compute the girth of the code.



Solution:

a) The parameters of the code are

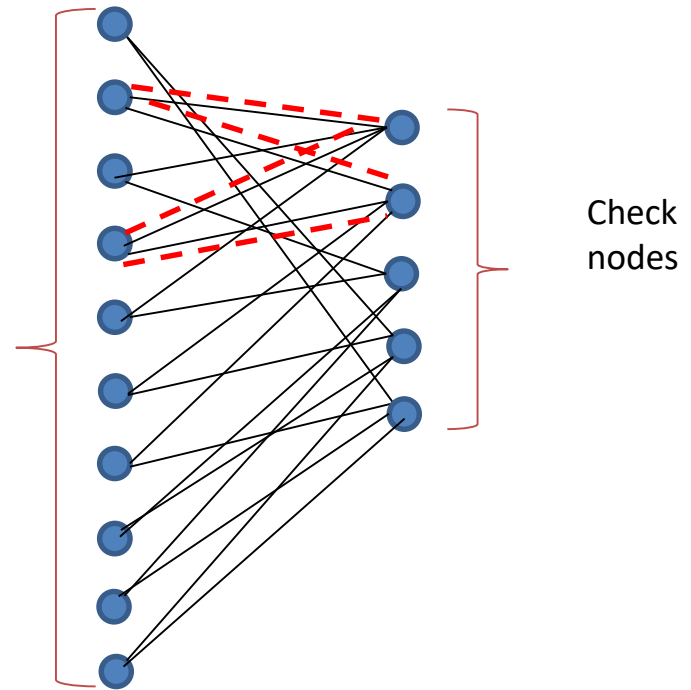
$$t_c = 2, t_r = 4 \text{ and } R = 1 - \frac{t_c}{t_r} = 0.5$$

b) The Tanner graph corresponding to this parity-check matrix is

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

Variable
nodes

c) The girth of the graph is $\gamma = 4$





Design strategies

- The design of LDPC codes is based on the generation of a parity-check matrix H with a given specification.
- In particular, the parameters (n, k, t_c, t_r) used in the specification play a major role in the performance of LDPC codes.
- LDPC codes can also be either regular or irregular depending on the patterns of ones in H .
- Regular codes: the number of ones in each column or row is constant
- Irregular codes: the number of ones in each column or row is variable.



- Specifically, in irregular codes t_c and t_r are functions of the column numbers \rightarrow use of degree distributions.
- The placement of ones is carried out via an optimization procedure known as density evolution.
- The degree distribution polynomials are:
 - The variable node polynomial $\lambda(x)$
 - The check node polynomial $\rho(x)$



- The variable node polynomial is

$$\lambda(x) = \sum_{d=1}^{d_v} \lambda_d x^{d-1},$$

where λ_d is the fraction of all edges connected to degree d variable nodes and d_v denotes the maximum node degree.

- The check node polynomial is

$$\rho(x) = \sum_{d=1}^{d_c} \rho_d x^{d-1},$$

where ρ_d is the fraction of all edges connected to degree d check nodes and d_c denotes the maximum node degree.



Example 3

Consider the design of LDPC codes based on degree polynomials.

- a) Write down the polynomials $\lambda(x)$ and $\rho(x)$ for a regular code with $t_c = 2$ and $t_r = 4$.

- b) Write down and discuss the meaning of an irregular code with $\lambda_1 = 0.00015$, $\lambda_2 = 0.30235$, $\lambda_3 = 0.27132$ and $\lambda_7 = 0.42618$, and $\rho_6 = 0.35559$ and $\rho_7 = 0.64445$.



Solution:

a) By noticing that $t_c = d_v = 2$ and $t_r = d_c = 4$, we have

$$\lambda(x) = \sum_{d=1}^{d_v} \lambda_d x^{d-1} = x$$

and

$$\rho(x) = \sum_{d=1}^{d_c} \rho_d x^{d-1} = x^3$$



b) By substituting $\lambda_1 = 0.00015$, $\lambda_2 = 0.30235$, $\lambda_3 = 0.27132$ and $\lambda_7 = 0.42618$, and $\rho_6 = 0.35559$ and $\rho_7 = 0.64445$ into the expressions of the degree polynomials, we obtain

$$\lambda(x) = \sum_{d=1}^{d_v} \lambda_d x^{d-1} = 0.00015 + 0.30235x + 0.27132x^2 + 0.42618x^6$$

and

$$\rho(x) = \sum_{d=1}^{d_c} \rho_d x^{d-1} = 0.35559x^5 + 0.64445x^6$$

This means that about 42.6% of edges are connected to degree 6 variable degrees, 27,1 % of edges are connected to degree 2 variable degrees and so on.



Gallager codes

- a) The original design approach to LDPC codes is known as Gallager codes, which rely on

$$H = \begin{bmatrix} H_1 \\ H_2 \\ \vdots \\ H_{t_c} \end{bmatrix}$$

The submatrices H_d are computed as follows.

For any μ and t_r greater than 1, each H_d has dimensions $\mu \times \mu t_r$ with row weight w_r and column weight 1.

The submatrix H_1 has the form

For $i = 1, 2, \dots, \mu$

The i th row contains all t_r ones in columns $(i - 1)t_r + 1$ to it_r

end



The remaining submatrices are permutations of H_1 as illustrated by

$$H = \begin{bmatrix} H_1 \\ H_2 \\ \vdots \\ H_{t_c} \end{bmatrix} = \begin{bmatrix} H_1 \\ \phi_1(H_1) \\ \vdots \\ \phi_{t_c-1}(H_1) \end{bmatrix}$$

where $\phi (.)$ is a permutation operation.



McKay codes

b) The design of McKay codes relies on the partitioning of the parity-check matrix

$$\mathbf{H} = [\mathbf{H}_1^T \mid \mathbf{H}_2^T]$$

The design algorithm for regular codes has the following steps:

1. \mathbf{H} is generated randomly using columns with column weights t_c and uniform row weights.
2. \mathbf{H} is constrained such that no columns have overlap greater than one.
3. \mathbf{H} is constrained to avoid short cycles.
4. \mathbf{H} is constructed such that \mathbf{H}_1 is invertible.



McKay codes lack structure for low-complexity encoding.

The encoding is performed as follows:

We obtain

$$\begin{aligned} G &= [P \mid I_k] \\ &= [H_2 H_1^{-1} \mid I_k] \in \mathbb{F}^{k \times n} \end{aligned}$$

using Gauss-Jordan elimination.



Other designs

c) Irregular LDPC designs

- Optimization of $\lambda(x)$ and $\rho(x)$ using density evolution.
- Adjustment by other methods.

d) Progressive edge growth designs

- Optimization of connections or edges between variable and check nodes in a progressive fashion.

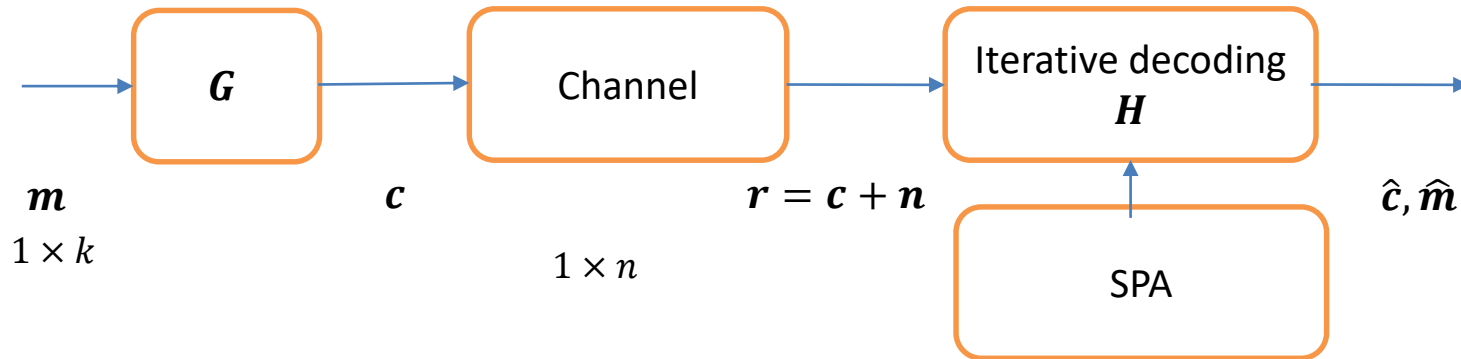
e) Repeat accumulate (RA) and irregular RA (IRA) designs

- Low-complexity designs that are competitive for low code rates.

f) Quasi-cyclic and finite geometry designs

- Low-complexity encoding \rightarrow reduction from $O(n^2)$ to $O(n \log_{10} n)$

D. Decoding



- Given the received data vector r , the decoder must find the most probable \hat{c} that satisfies

$$\hat{c}H^T = \mathbf{0}$$

- According to the MAP decoding rule, we have the log likelihood ratios (LLRs)

$$l(c_j) = l(c_j | \mathbf{r}) = \log \left(\frac{P(c_j = +1 | \mathbf{r})}{P(c_j = -1 | \mathbf{r})} \right) \underset{<}{\overset{\geq}{\geq}} 0$$

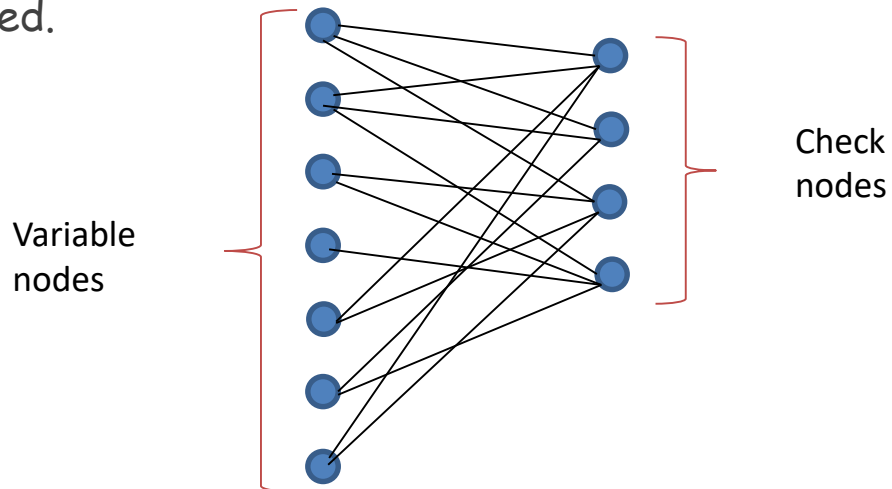
- A message passing strategy known as sum-product algorithm (SPA) is employed to compute the LLRs.



- SPA has two alternating steps:
 - The horizontal step runs along the rows of H .
 - The vertical step runs along the columns of H .
- The variable or bit nodes are elements of r
- The check nodes are the rows of H .
- $J(i)$ is the set of bits that participate in check node i .
- $J(j)$ is the set of nodes in which variable node j participates.
- $J(i)/j$ is the set $J(i)$ that excludes variable node j .
- $J(j)/i$ is the set $J(j)$ that excludes check node i .

The SPA decoding strategy is as follows:

- To exchange LLRs associated with the ones of H in an alternating way.
- We compute the LLR of check node i $l(q_{ij})$ (horizontal step) and send it to variable node j $l(r_{ij})$ → This checks the probability that bit j is +1 or -1.
- We compute the LLR of variable node j $l(r_{ij})$ (vertical step) and send it to check node i $l(q_{ij})$ → This checks the probability that check node i is satisfied.





SPA

- The SPA first considers the LLR

$$l(c_j) = l(c_j|\mathbf{r}) = \log \left(\frac{P(c_j = +1|\mathbf{r})}{P(c_j = -1|\mathbf{r})} \right),$$

where c_j is the j th element of the codeword c .

- We compute the LLR of check node i and send it to variable node j as described by

$$l(r_{ij}) = 2 \tanh^{-1} \left\{ \prod_{i \in \mathcal{J}(j)/i} \tanh \left[\frac{1}{2} l(q_{ij}) \right] \right\},$$

where $i = 1, 2, \dots, n - k$ and $j = 1, \dots, n$.



- We then compute the LLR of the variable node j updated and send it to check node i as follows:

$$l(q_{ij}) = l(c_j) + \sum_{j \in \mathcal{J}(i)/j} l(r_{ij})$$

- At the end of each iteration, variable node j computes the total LLR given by

$$l(Q_j) = l(c_j) + \sum_{j \in \mathcal{J}(i)} l(r_{ij})$$

- The decisions about the bits are obtained by

$$\hat{c}_j = \begin{cases} 1, & \text{if } l(Q_j) \geq 0 \\ -1, & \text{otherwise} \end{cases}$$



Summary of SPA

Initialization: $l(r_{ij}) = l(c_j)$ (equal probabilities)

Goal: to compute $l(c_j) = l(c_j|\mathbf{r}) = \log\left(\frac{P(c_j = +1|\mathbf{r})}{P(c_j = -1|\mathbf{r})}\right)$

For each iteration, update

$$1. \quad l(r_{ij}) = 2 \tanh^{-1} \left\{ \prod_{i \in \mathcal{J}(j)/i} \tanh \left[\frac{1}{2} l(q_{ij}) \right] \right\},$$

$$2. \quad l(q_{ij}) = l(c_j) + \sum_{j \in \mathcal{J}(i)/j} l(r_{ij})$$

$$3. \quad l(Q_j) = l(c_j) + \sum_{j \in \mathcal{J}(i)} l(r_{ij})$$

Stop if $\hat{\mathbf{c}}\mathbf{H}^T = \mathbf{0}$ or if the maximum number of iterations is reached

Make decision

$$\hat{c}_j = \begin{cases} 1, & \text{if } l(Q_j) \geq 0 \\ -1, & \text{otherwise} \end{cases}, j = 1, 2, \dots, n$$

$$l(c_j) = l(c_j|\mathbf{r}) = \log\left(\frac{P(c_j = +1|\mathbf{r})}{P(c_j = -1|\mathbf{r})}\right),$$



Example 4

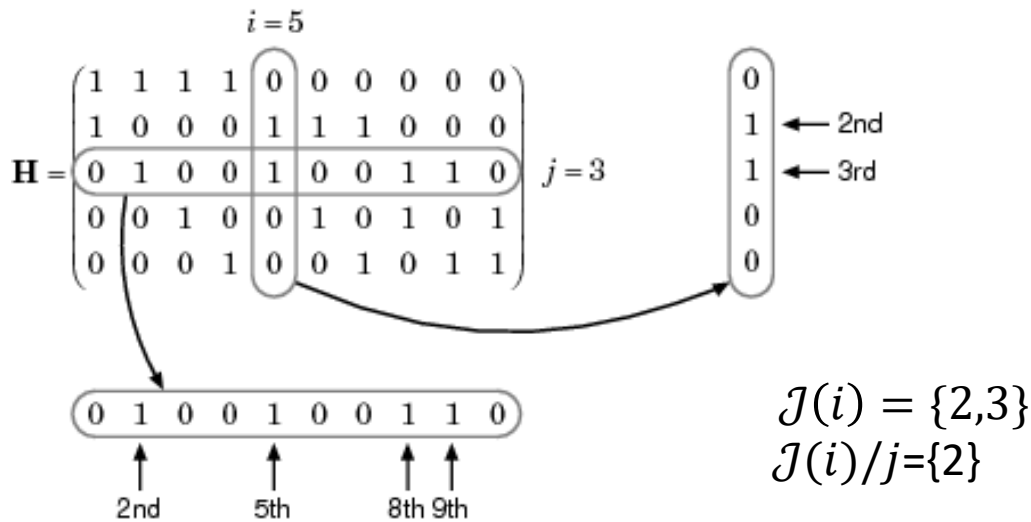
Illustrate the SPA decoding of the following parity-check matrix

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



Solution:

For $i = 5$ and $j = 3$, the index sets would be



$$J(j) = \{2,5,8,9\} \\
 J(j)/i = \{2,8,9\}$$

At the end of each iteration, $l(Q_j)$ provides an updated estimate of the a posteriori log-likelihood ratio for the transmitted bit .