

Lecture : Peeling decoder (contd.) and an introduction to message-passing decoding

Recall the definition of the peeling decoder, from last time. We now present a bound on its performance, via the idea of using **stopping sets**.

Def 1 (Stopping set): A stopping set (ss) $\mathcal{A} \subseteq \mathcal{V}$ is a collection of variable nodes whose induced subgraph has no check nodes with degree 1.

By definition, we let the empty set be a ss.

Remark: Let $\text{supp}(\underline{c})$ denote the support of a codeword \underline{c} in the code of interest. In order to satisfy the parity-check equations, all check nodes in $N(\text{supp}(\underline{c}))$ must be connected at least twice to variable nodes in $\text{supp}(\underline{c})$. Hence, $\text{supp}(\underline{c})$, for any codeword \underline{c} , is a stopping set.

Stopping sets have the following properties:

Proposition 1: (1) If \mathcal{A}_1 and \mathcal{A}_2 are ss, then so is $\mathcal{A}_1 \cup \mathcal{A}_2$.

(2) Each subset $W \subseteq \mathcal{V}$ has a unique maximal ss.

(3) If $W \subseteq \mathcal{V}$ is the collection of erased variable nodes, then the peeling decoder recovers the value of all variable nodes except those in the unique maximal ss. in W .

Proof:

- (1) Note that if $i \in N(d_1, U d_2)$, then $i \in N(d_1)$ or $i \in N(d_2)$, implying that i is connected at least twice to nodes in $d_1, U d_2$.
- (2) Define Z to be the union of all ss contained in W ; by (1), it follows that Z is the unique maximal ss in W as any ss in W is contained in Z .
- (3) Note that since $Z = Z(W)$ is a ss, the variable nodes in W will not be recovered by the peeling decoder. Further, any $T \subseteq W \setminus Z$ is such that $T \cup Z$ is not a s.s. and will be recovered in part by one step of the peeling decoder. \square

Def 2: A ss is minimal if the only ss it contains besides itself is the empty set.

For a given code \mathcal{L} with a fixed parity-check matrix H (that defines a Tanner graph), let $A_{ss}(h)$ denote the number of minimal stopping sets of cardinality h . Now, over a BEC, note that the peeling decoder is unsuccessful at decoding from the erasures introduced (at some set $\eta \subseteq V$), only if η contains some minimal ss [Why?].

Hence, it follows that

$$P_u[\hat{\underline{c}} \neq \underline{c}] = P_{\text{err, peeling}}(\epsilon) \leq \sum_{h=1}^n A_{ss}(h) \epsilon^h.$$

FACT [Luby et al. (2001), "Efficient erasure correcting codes"]: Cascaded LDPC codes (followed by an MDS code) achieve the capacity of an erasure channel, when a peeling decoder is used for decoding the LDPC codes.

An aside: MAP decoding over the BEC(ϵ)

Recall that the MAP decoder (or equivalently, the ML decoder for equiprobable codewords) computes

$$\begin{aligned}\hat{c}^{\text{MAP}} &= \underset{x \in \mathcal{L}}{\text{argmax}} P(x|y) \\ &= \underset{x \in \mathcal{L}}{\text{argmax}} P(y|x).\end{aligned}$$

For an erasure channel, given y , let $\mathcal{X}(y)$ denote the set of codewords "compatible" with y , i.e.,

$$\mathcal{X}(y) = \{c \in \mathcal{L} : c_j = y_j, \text{ if } y_j \neq ?\}.$$

If this set $\mathcal{X}(y)$ has cardinality 1, it follows that MAP decoding will be successful. [why?]

Now, in order to find the set $\mathcal{X}(y)$, note that we only need to solve a system of linear equations (over \mathbb{F}_2):

Let η be the set of erased locations & for any set \mathcal{S} , let $H_{\mathcal{S}}$ denote the submatrix of H obtained by retaining the columns in \mathcal{S} .

Then, note that

$$\mathcal{X}(y) = \{x \in \mathcal{C} : x_{[\omega] \setminus \eta} = y_{[\omega] \setminus \eta} \text{ and } H_{\eta} x_{\eta}^T = H_{[\omega] \setminus \eta} y_{[\omega] \setminus \eta}^T\}.$$

Solving this system of linear equations takes time $O(n^3)$, but decoding via a peeling decoder is expected to be faster.

Remark: Note that the MAP decoder fails if and only if η contains the support of a codeword. However, the peeling decoder fails if and only if η contains a stopping set.

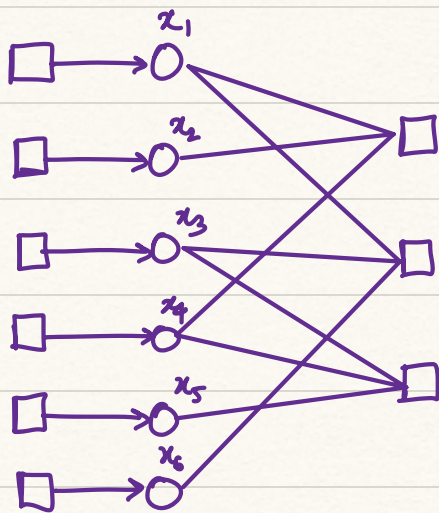
The peeling decoder is hence suboptimal compared to the MAP decoder.

An introduction to message-passing decoding for general channels

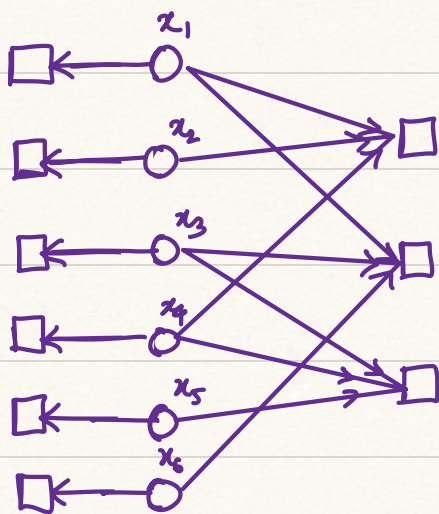
The peeling decoder structure informs a more general algorithm for decoding codes on graphs (such as LDPC codes), over channels other than the BEC.

This general algorithm, called the **message-passing decoder**, is based on the well-studied **sum-product algorithm** (or "generalized distributive law") that is applied to the problem of computing marginals in probabilistic graphical models. → also called a belief-propagation (BP) decoder

An illustration of the BP decoder : Fix a received sequence y over some channel $P_{y|x}$ and let the Tanner graph of the code be as shown below.



Initial messages from the channel



Variable to factor node messages

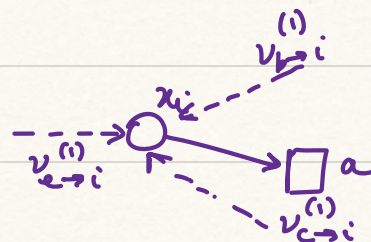
Each message is a vector $\underline{u} = [u(0), u(1)]$.



The channel forwards the message

$$\underline{m}_i = [P(y_i|0), P(y_i|1)]$$

1. (a)

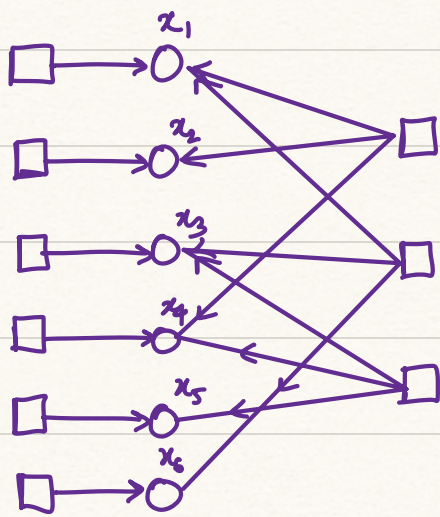


The variable nodes forwards the message

$$\mu_{i \rightarrow a}^{(i)}(x_i) = \prod_{b \in N(i) \setminus \{a\}} v_{b \rightarrow i}^{(i)}(x_i)$$

to the factor node $a \in N(i)$, for each $x_i \in \{0, 1\}$.

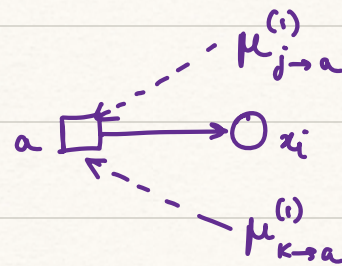
[The vector $\mu_{i \rightarrow a}^{(i)}$ can be thought of as the msg. vector]



Factor to variable node messages

1. (b)

Here, $v_{b \rightarrow i}^{(1)}(x)$ are initialized to 1, for all b .



The factor nodes forward the message

$$v_{a \rightarrow i}^{(2)}(x_i)$$

$$= \sum_{x_{N(a) \setminus \{i\}}} 1 \left\{ \bigoplus_{j \in N(a)} x_j = 0 \right\} \cdot \prod_{j \in N(a) \setminus \{i\}} \mu_{j \rightarrow a}^{(1)}(x_j)$$

to variable node x_i , for each $x_i \in [0, 1]$.

[Here again, $v_{a \rightarrow i}^{(2)}$ can be thought of as the message vector]

... and the steps 1.(a) and 1.(b) are recursed over until convergence of the messages or until a fixed no. of iterations.