



§ 4.5* Maximum Likelihood (ML) Decoding

Maximum Likelihood (ML) Decoding: Given a received word / symbol sequence \bar{y} , the codeword \bar{c} (or message \bar{u}) that maximizes the channel transition probability $P(\bar{y}|\bar{c})$ is the decoding output which is denoted as $\hat{\bar{c}}$ (or $\hat{\bar{u}}$). That says

$$\hat{\bar{c}} = \operatorname{argmax}_{\bar{c} \in \mathcal{C}} P(\bar{y}|\bar{c}).$$

- Based on Bayes' theorem, the a posteriori probability can be determined as

$$P(\bar{c}|\bar{y}) = \frac{P(\bar{y}|\bar{c})P(\bar{c})}{P(\bar{y})}.$$

Maximum A Posteriori (MAP) Decoding: Given \bar{y} , the codeword \bar{c} (or message \bar{u}) that maximizes the MAP $P(\bar{c}|\bar{y})$ is the decoding output. That says

$$\hat{\bar{c}} = \operatorname{argmax}_{\bar{c} \in \mathcal{C}} P(\bar{c}|\bar{y}).$$

- By assuming equiprobable codeword as $P(\bar{c}) = |\mathcal{C}|^{-1}$, the ML decoding output coincides with the MAP decoding.



§ 4.5* Maximum Likelihood (ML) Decoding

Union Bound

- Union bound can be used to characterize the ML decoding performance of codes, which requires knowledge of the code's weight spectrum (distribution of codewords of different weights).
- The codeword $c_1^N = \bar{c} = \{c_1, c_2, \dots, c_N\} \in \mathcal{C}$ of a linear block code has discrete weight values, denoted as $\{d_0, d_1, d_2, \dots, d_s\}$, where $d_0 = 0$, $d \leq d_i \leq N$ and $i = 1, 2, \dots, s$. The number of codewords with weight d_i is denoted as A_{d_i} . Hence, weight spectrum is $\{A_{d_i}, \forall i\}$.
- Union upper bound on a linear block code's ML decoding frame error rate (FER) over the AWGN channel is

$$P_{\text{ML,e}} \leq \sum_{d_i=d}^{d_s} A_{d_i} Q\left(\frac{\sqrt{d_i}}{\sigma}\right) = \sum_{d_i=d}^{d_s} A_{d_i} Q\left(\frac{\sqrt{r d_i E_b}}{\sigma}\right).$$

Diagram labels for the equation above:

- code rate (points to r)
- energy of each info. bit (points to E_b)
- noise standard deviation (points to σ)



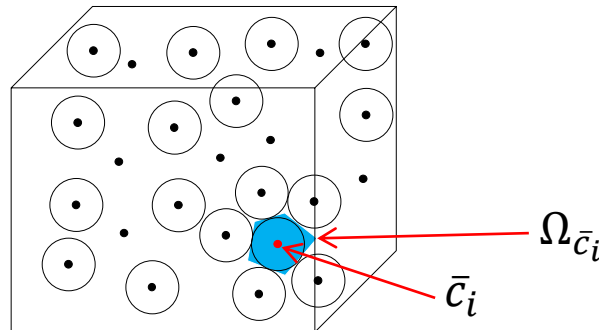
§ 4.5* Maximum Likelihood (ML) Decoding

- Proof:

The function $\mathbb{1}(\cdot)$ denotes the indicator function, where $\mathbb{1}(\text{true}) = 1$ and $\mathbb{1}(\text{false}) = 0$. Then,

$$P_{\text{ML,e}} = \sum_{\bar{c} \in \mathcal{C}} \sum_{\bar{y} \in \mathcal{Y}} \Pr(\bar{c}, \bar{y}) \cdot \mathbb{1}(\text{Decoder}_{\text{ML}}(\bar{y}) \neq \bar{c}).$$

The set $\Omega_{\bar{c}_i}$ is defined as $\Omega_{\bar{c}_i} = \{\bar{y} \mid \Pr(\bar{y} | \bar{c}_i) > \Pr(\bar{y} | \bar{c}_{i'}), \forall i \neq i'\}$, which represents the space of all received signals \bar{y} that will be decoded as the codeword \bar{c}_i under the ML decision rule. The set \mathcal{Y} is n -dimensional real vector space, and $\mathcal{Y} = \cup_{\bar{c}_i \in \mathcal{C}} \Omega_{\bar{c}_i}$.





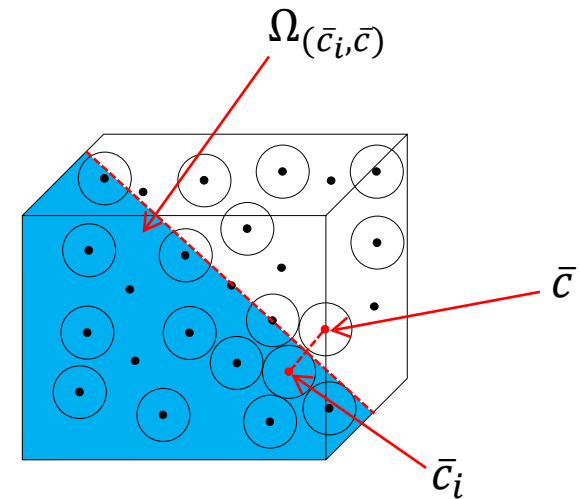
§ 4.5* Maximum Likelihood (ML) Decoding

By symmetry, let \bar{c} be $\bar{0}$. We have

$$P_{\text{ML},e} = \sum_{\bar{c} \in \mathcal{C}} \sum_{\bar{y} \in \mathcal{Y} \setminus \Omega_{\bar{c}}} \Pr(\bar{c}, \bar{y}) = \Pr(\bar{c}) \sum_{\bar{c} \in \mathcal{C}} \sum_{\bar{y} \in \cup_{\bar{c}_i \neq \bar{c}} \Omega_{\bar{c}_i}} \Pr(\bar{y}|\bar{c}) = \sum_{\bar{y} \in \cup_{\bar{c}_i \neq \bar{c}} \Omega_{\bar{c}_i}} \Pr(\bar{y}|\bar{c}).$$

The set $\Omega_{(\bar{c}_i, \bar{c})}$ is further defined as $\Omega_{(\bar{c}_i, \bar{c})} = \{\bar{y} | \Pr(\bar{y}|\bar{c}_i) > \Pr(\bar{y}|\bar{c}), \bar{c}_i \neq \bar{c}\} \supset \Omega_{\bar{c}_i}$, which represents the space of \bar{y} that will be decoded as \bar{c}_i instead of \bar{c} . Hence

$$\begin{aligned} P_{\text{ML},e} &= \sum_{\bar{y} \in \cup_{\bar{c}_i \neq \bar{c}} \Omega_{\bar{c}_i}} \Pr(\bar{y}|\bar{c}) \\ &= \sum_{\bar{y} \in \cup_{\bar{c}_i \neq \bar{c}} \Omega_{(\bar{c}_i, \bar{c})}} \Pr(\bar{y}|\bar{c}) \\ &\leq \sum_{i=1}^{2^K-1} \sum_{\bar{y} \in \Omega_{(\bar{c}_i, \bar{c})}} \Pr(\bar{y}|\bar{c}). \end{aligned}$$





§ 4.5* Maximum Likelihood (ML) Decoding

Under AWGN channel and BPSK modulation,

$$\sum_{\bar{y} \in \Omega(\bar{c}_i, \bar{c})} \Pr(\bar{y}|\bar{c}) = Q\left(\sqrt{\frac{2E_c d_{\text{Ham}}(\bar{c}_i, \bar{c})}{N_0}}\right) = Q\left(\frac{\sqrt{d_{\text{Ham}}(\bar{c}_i, \bar{c})}}{\sigma}\right) = Q\left(\frac{\sqrt{d_{\text{Ham}}(\bar{c}_i, \bar{0})}}{\sigma}\right).$$

Then,

pairwise error probability

assuming $E_c = 1$

$$\sum_{i=1}^{2^K-1} \sum_{\bar{y} \in \Omega(\bar{c}_i, \bar{c})} \Pr(\bar{y}|\bar{c}) = \sum_{\bar{y} \in \Omega(\bar{c}_1, \bar{c})} \Pr(\bar{y}|\bar{c}) + \sum_{\bar{y} \in \Omega(\bar{c}_2, \bar{c})} \Pr(\bar{y}|\bar{c}) + \cdots + \sum_{\bar{y} \in \Omega(\bar{c}_{2^K-1}, \bar{c})} \Pr(\bar{y}|\bar{c}).$$

$$E_c = 1 = rE_b \\ \Rightarrow$$

$$P_{\text{ML,e}} \leq \sum_{d_i=d}^{d_s} A_{d_i} Q\left(\frac{\sqrt{rd_i E_b}}{\sigma}\right).$$

□