# CS 860 Topics in Coding Theory Lecture 1 Lecturer: Elena Grigorescu Scribe: Tiger Wu

# 1 Random Error

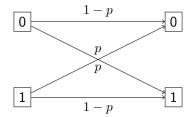
We consider two types of random errors:

- Binary Symmetric Channel(BSC)
- Binary Eraser Channel(BEC)

#### Binary Symmetric Channel (BSC)

Binary symmetric channel has an input space equal to the output space,  $\{0,1\}$ . A BSC with crossover probability  $0 \le p \le \frac{1}{2}$ , denoted as BSC<sub>p</sub>, flips an input with probability p.

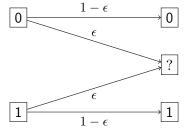
Figure 1:  $BSC_p$ 



#### Binary Erasure Channel (BEC)

Binary erasure channel has input space  $\{0,1\}$  and output space  $\{0,1,?\}$ . Binary erasure channel with parameter  $0 \le \epsilon \le \frac{1}{2}$ , denotes as  $\mathrm{BEC}_{\epsilon}$  erase an input with probability  $\epsilon$ .

Figure 2: BEC $_{\epsilon}$ 



## 2 Basic Probability Facts

Let D be a probability distribution over U. A random variable  $X \leftarrow D$  is drawn from U according to U. Let  $f: U \to \mathbb{R}$  be a function.

Variance. The variance is defined as

$$\operatorname{Var}_{X \leftarrow D}[f(X)] = \mathbb{E}[f^2(x)] - \mathbb{E}[f(X)]^2.$$

**Expectaion.** For random variables  $X_1$  and  $X_1$ ,

$$\mathbb{E}(X_1 + X_2) = \mathbb{E}(X_1) + \mathbb{E}(X_2)$$

and

$$\mathbb{E}(X_1X_2) = \mathbb{E}(X_1)\mathbb{E}(X_2)$$
 if and only  $X_1$  and  $X_2$  are independent.

**Expection.** For events  $\mathcal{E}_1$  and  $\mathcal{E}_2$ , the union bound states that

$$\Pr[\mathcal{E}_1 \cup \mathcal{E}_2] \leq \Pr[\mathcal{E}_1] + \Pr[\mathcal{E}_2].$$

Indicator Random Variable. For event  $\mathcal{E}$ , the indicator variable  $I_{\epsilon}$  is a 0/1 random variable defind as

$$I_{\mathcal{E}}(X) = \begin{cases} 1 & \text{if } X \in \epsilon \\ 0 & \text{otherwise.} \end{cases}$$

Therefore

$$\mathbb{E}[I_{\epsilon}(X)] = \Pr[\mathcal{E}].$$

Tail Bound/Markov Inequality. Let X be a non-negative random variable and  $\alpha > 0$ . Then

$$\Pr[X \ge \alpha] \le \frac{\mathbb{E}[X]}{\alpha}.$$

E.g.

- If  $\alpha = 2\mathbb{E}[X]$ ,  $\Pr[X \ge \alpha] \le \frac{1}{2}$ .
- If  $\alpha = \sqrt{\mathbb{E}[X]}$ ,  $\Pr[X \ge \sqrt{\mathbb{E}[X]}] \le \sqrt{\mathbb{E}[X]}$ .

Chernoff Bounds for Bernoulli Random Variable. If  $X_1, ..., X_n$  are i.i.d.  $\{0, 1\}$ -valued random variables with  $\Pr[X_i = 1] = p$ . For  $\epsilon > 0$ , for large enought n,

$$\Pr\left[\sum X_i \ge (p+\epsilon)n\right] \le 2^{-\frac{\epsilon^2}{2}n}$$

and

$$\Pr\left[\sum X_i \le (p-\epsilon)n\right] \le 2^{-\frac{\epsilon^2}{3}n}.$$

# 3 Shannon Capacity

The capacity of a  $BSC_p$  channel is

$$\lim_{n\to\infty}\frac{r}{n}$$

which depends only on the channel. We will show that the capcity of BSC<sub>p</sub> is  $1 - H(p)^{-1}$ . That is, we will show that there exists encoding and decoding maps  $E: \{0,1\}^k \to \{\alpha k\}$  and  $D: \{0,1\}^{\alpha k} \to \{0,1\}$  for  $\alpha > 1$  such that with high probability over noise vector  $\eta$  decoding is successful.

Recall that  $\eta \in \{0,1\}^n$  is a random variable with each word being 0 with probability p and 1 with probability 1-p.

**Theorem 1 (Shannon's Capacity Theorem)** For all  $p \in [0, \frac{1}{2})$ ,  $0 \le \gamma \le \frac{1}{2} - p$ , for large enough n, there exists  $\beta = \beta(\gamma, p)$ , let  $\alpha = \frac{1}{1 - H(p + \gamma)}$ , and  $E : \{0, 1\}^k \to \{0, 1\}^{\alpha k}$ ,  $D : \{0, 1\}^{\alpha k} \to \{0, 1\}^k \cup \{\text{'fail'}\}$  such that

$$\Pr_{\eta}[D(E(m) + \eta) = m] \ge 1 - 2^{-\Omega(\epsilon^2)n}.$$

**Proof** By the probabilistic method. We look at the random encoding functions. Let  $\ell = k+1$ . Let  $E: \{0,1\}^k \to \{0,1\}^\ell$  be randomly chosen from all such functions. This means that for a fixed m, E(m) is uniformly random in  $\{0,1\}^n$ . Let  $\epsilon = \epsilon(\gamma) > 0$  be small enough. The decoding function is deifned as:

$$D(y) = \begin{cases} m & \text{if } m \text{ is unique codeword such that } \Delta(y, E(m)) \leq (p + \epsilon)n \\ \text{fail} & \text{otherwise.} \end{cases}$$

(D is not efficient).

Decoding is not successful when one of the two following events will happen:

- 1. Too many errors have been incurred. i.e.,  $\Delta(y, E(m)) > (p+\epsilon)n$ , where  $y = E(m) + \eta$ .
- 2. More than 2 messages have encodings in  $B(y, (p + \epsilon)n)$ .

For fixed m, consider all E.

Case 1. Consider the first event.

$$\Pr_{\eta}[\Delta(y, E(m)) > (p + \epsilon)n] = \Pr[\text{wt}(y - E(m)) > (p + \epsilon)n]$$

$$= 2^{-\Omega(\epsilon^2)n}$$
 (by Chernoff bound)

Case 2.. Consider the second event. Fix a y, (which fixes  $\eta$  and  $m' \neq m$ .

$$\Pr_{E}[E(m') \in B(y, (p+\epsilon)n)] \le \frac{|B(y, p+\epsilon)|}{2^{n}}$$

$$\le 2^{(H(p+\epsilon)+o(1))n}$$

<sup>&</sup>lt;sup>1</sup>binary entopy:  $H(x) = -x \lg(x) - (1-x) \lg(1-x)$ 

By union bound over all m',

$$\Pr_{E} \left[ \exists m', E(m') \in B(y, (p+\epsilon)n) \right] \le 2^{k+1} \cdot 2^{n(H(p+\epsilon)-1+o(1))} \\
= 2 \cdot 2^{(1-H(p+\gamma))n+n(H(p+\epsilon)-1+o(1))} \\
= 2 \cdot 2^{(-H(p+\gamma)+H(p+\epsilon)+o(1))n}$$

By lineraity of expectation,

$$\mathbb{E}_{E}\left[\Pr_{\eta}[D(E(m)+\eta)\neq m]\right] \leq 2^{-\Omega(\epsilon^{2})n} + 2 \cdot 2^{(H(p+\epsilon)-H(p+\gamma)+o(1))n}$$

$$< \frac{1}{2}2^{-\beta n}$$

where  $\beta$  is a function of p and  $\gamma$ . This implies there exists an encoding function that is decoded correctly with high probability. Our goal is to say there exists such an E that is good for all messages m, and the union bound is not strong enough for this.

Since we have the above for each fixed m,

$$\mathbb{E}_{m} \left[ \mathbb{E}_{E} \left[ \Pr_{\eta} [D(E(m) + \eta) \neq m] \right] \right] < \frac{1}{2} 2^{-\beta n}$$

$$\implies \mathbb{E}_{E} \left[ \mathbb{E}_{m} \left[ \Pr_{\eta} [D(E(m) + \eta) \neq m] \right] \right] < \frac{1}{2} 2^{-\beta n}$$

Therefore there exsits  $E^*$  such that

$$\mathbb{E}_m \left[ \Pr_{\eta} \left[ D(E^*(m) + \eta) \neq m \right] \right] < \frac{1}{2} 2^{-\beta n}$$

Applying Markov's inequality,

$$\Pr_{m}[\Pr_{\eta}[D(E^{*}(m) + \eta) \neq m] > 2^{-\beta n}] < \frac{1}{2}.$$

Hence, for at least 1/2 fraction of the messages, m,  $E^*$  fails on m with probability  $< 2^{-\beta n}$ . Removing all the other messages, which are fewer than  $\frac{1}{2}e^{\ell}$ , we are left with a code with  $> 2^{\ell} \cdot \frac{1}{2} = 2^k$  many messages for which  $E^*$  decodes correctly with probability  $1 - 2^{-\beta n}$ .

**Observation 2** Neither  $E^*$  nor  $D^*$  are efficiently computable.

### 3.1 Connection between Shannon and Hamming

Claim 3 If  $C \subseteq \{0,1\}^n$  and  $\Delta(C) = 2p + \epsilon$ , then we can communicate over  $BSC_p$  decoding correctly with  $1 - 2^{-\Omega(\epsilon^2)}$ .

**Proof** By Chernoff bound, with high probability, the number of errors is

$$\leq \frac{\Delta(c)}{2} = \frac{p+\epsilon}{2}$$

So we have unique closest codeword.

# 4 Shannon's Converse Theorem

**Theorem 4 (Shannon's Converse Theorem)** For all  $p \in [0, \frac{1}{2})$ , for all  $\epsilon, \delta > 0$ , for large enough n, for  $k \ge (1 - H(p) + \epsilon)n$ , for all  $E : \{0, 1\}^k \to \{0, 1\}^n$ , for all  $D : \{0, 1\}^n \to \{0, 1\}^k$ 

$$\Pr_{\eta, m \leftarrow \{0,1\}^k} [D(E(m) + \eta) \neq m] \ge 1 - \delta.$$

In other words, if we are sending information at a rate  $> 1 - H(p) + \epsilon$ , decoding is on average erroneous for any encoding/decoding.

#### Intuition

For 1-H(p) being the upper bound, suppose  $D: \{0,1\}^n \to \{0,1\}^k \cup \{\text{'fail'}\}$  decodes with negligible error probability. Typical noise has weight  $\in [(p-\epsilon)n, (p+\epsilon)n]$ . So  $E(m)+\eta$  takes approximately  $2^{H(p)n}$  many possibilities. To decode correctly, D should map most of these  $2^{H(p)}$  strings back to m. So  $|C| \leq 2^n/2^{H(p)n}$ , therefore R < 1 - H(p).