

# Quantum info in TCS

## Lecture 6: Coding over quantum channel

Suppose a quantum channel  $\mathcal{N}_{A \rightarrow B}$  is connected Alice to Bob. Alice wants to transmit “information” to Bob over this channel. The information Alice can transmit can be either classical or quantum.

### 1 Sending classical information over classical channels

Consider a *classical* channel where Alice sends  $x \in \{0, 1\}$  and Bob receives  $x \oplus z$  where  $z$  is Bernoulli( $p$ ).

**Multiple independent use of channel** With one use of the channel, Alice cannot do anything non-trivial. So we consider  $n$  independent uses of the channel. Alice has  $m$  bits and encode it into a binary string of  $n$  bits and transmits it over the channel. Bob receives the noisy version and attempts to recover the original  $m$  bits

**Repetition code** As an example, suppose Alice has one bit  $b \in \{0, 1\}$ . She encodes this into  $n$  bits by repetition and transmits it over the channel. Bob does majority voting for decoding. If more than half of the received bits are 1 he decodes the bit as one, otherwise as zero.

**Rate vs probability of error** There are two important parameters of every code.

1. **Rate:** it is the number of bits transmitted per channel use, i.e.,  $m/n$ . For repetition code, the rate is  $1/n$ .
2. **Probability of error:** It is the probability that Bob cannot decode *any* transmitted bits. For repetition code, it is  $\sum_{i=0}^{(n-1)/2} (1-p)^{n-i} p^i$ , which converges to zero if  $p < \frac{1}{2}$ .

**Formal definition of a code** Fix the number of bits to be transmitted  $m$  and the number of channel uses  $n$ . The code consists of two functions  $f : \{0, 1\}^m \rightarrow \{0, 1\}^n$  for encoding and  $g : \{0, 1\}^n \rightarrow \{0, 1\}^m$  for decoding. Alice encodes  $m$  bits  $b_1, \dots, b_m$  into a codeword of length  $n$  using  $x = f(b_1, \dots, b_m)$  and transmits  $x$  over  $n$  uses of the channel. Bob receives  $y$  at the output of the channel and decodes the bits as  $\hat{b}_1, \dots, \hat{b}_m = g(y)$ .

#### Shannon channel coding

**Theorem 1.1.** *There exists a sequence of codes  $(f_n, g_n)_{n \geq 1}$  such that the probability of error goes to zero and the rate is converging to  $1 - h_2(p)$ .*

Here  $h_2(x) := -x \log(x) - (1-x) \log(1-x)$  is the binary entropy function. A few remarks about channel coding:

1.  $1 - h_2(p) = I(X : Y)$  where  $X$  is uniformly distributed bit and  $Y$  is the output of the channel when  $X$  is transmitted.
2. The optimal decoder for BSC is minimum distance decoder. It is not computationally efficient
3. We can generalize this theorem to any channel.

**Random coding** A code is characterized by the set of all codewords. We consider a random code where all codewords are chosen independently at random. Let  $x(1), \dots, x(2^{nR}) \in \{0, 1\}^n$  be the random codewords for  $R = 1 - h_2(p) - \delta$  for a fixed  $\delta > 0$ . The decoder works as follows. Fix  $\epsilon > 0$ . If there is a unique  $i$  such that  $|x(i) \oplus y|$ <sup>1</sup> is between  $(1 - \epsilon)np$  and  $(1 + \epsilon)np$  then the decoded value would be  $i$ . Otherwise, the decoder output 1. To analyze the probability of error, assume that the codeword  $x(1)$  is transmitted over the channel. Then, by law of large number  $|x(1) \oplus y|$  is between  $(1 - \epsilon)np$  and  $(1 + \epsilon)np$  with high probability. We need to show that with high probability there is not  $i \neq 1$  such that  $|x(i) \oplus y|$  is between  $(1 - \epsilon)np$  and  $(1 + \epsilon)np$ . By union bound, we have

$$\Pr[\exists i \neq 1 : (1 - \epsilon)np < |x(i) \oplus y| < (1 + \epsilon)np] \leq \sum_{i=2}^{2^{nR}} \Pr[(1 - \epsilon)np < |x(i) \oplus y| < (1 + \epsilon)np] \quad (1)$$

$$= (2^{nR} - 1) \Pr[(1 - \epsilon)np < |x(2) \oplus y| < (1 + \epsilon)np] \quad (2)$$

Note that  $x(2) \oplus y$  has uniform distribution. Therefore,

$$\Pr[(1 - \epsilon)np < |x(2) \oplus y| < (1 + \epsilon)np] = \frac{\#\{x : (1 - \epsilon)np \leq |x| \leq (1 + \epsilon)np\}}{2^n} \approx 2^{-n(1-h_2(p))} \quad (3)$$

By our choice of  $R$ , the second type of probability of error goes to zero as well.

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<sup>1</sup> $|x|$  is the number of 1 in a binary string  $x$