# Final Recitation – Information Theory, Application for Quantum Fault Tolerance.

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- Classical no-go for reversible noisy computation at logarithmic depth (and polynomial space).

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- Classical no-go for reversible noisy computation at logarithmic depth (and polynomial space).
- Quantum case.
- ► Trading (local) entropy for space.

# Nosiy Circuit.



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#### Definition

p- Depolarizing Channel. The qubit depolarizing channel with parameter  $p \in [0,1]$  is the quantum channel  $\mathcal{D}_p$  defined by:

$$\mathcal{D}_{p}(\rho) = (1-p)\rho + p \cdot \frac{l}{2}$$

where  $\rho$  is a single-qubit density matrix and I is the identity matrix.

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#### Definition

p-Noisy Circuit. Given a circuit C (regardless of the model), its p-noisy version  $\tilde{C}$  is the circuit obtained by alternately taking layers from C and then passing each (qu)bit through a p-Depolarizing channel.

#### Informal:

We can't compute when subjected to noise, after a while, the bits become garbage.

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#### Claim

Denote b  $X = (X_1, X_2, ..., X_m)$  and  $Y = (Y_1, Y_2, ..., Y_m)$  the input and the output distrubtions of reversible p-noisy computation at widith m (bits) and depth d. Then:

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$$m - H(Y) \le \gamma^d (m - H(X))$$

In particular, for  $d = \Omega(\log m)$  we have  $H(Y) \to m$ .

#### Claim

Let Y be a bit given by moving X trough BSC(p), Then there is  $\gamma_p < 1$  such :

$$1 - H(Y) \le \gamma \left(1 - H(X)\right)$$

Denote by  $\delta$  the parameter for which X distributed as  $\sim Bin(\frac{1+\delta}{2})$ . First notice that:

$$\Pr(Y = 1) = \frac{1+\delta}{2}(1-p) + \frac{1-\delta}{2}p = \frac{1+\delta-2\delta p}{2}$$

So 
$$Y \sim \text{Bin}(\frac{1-\delta(1-2p)}{2})$$
, Or  $\delta \mapsto 1-2p\delta$ .

$$1 - H(X) = 1 - \frac{1}{2} \left( (1 + \delta) \log \left( \frac{1 + \delta}{2} \right) + (1 - \delta) \log \left( \frac{1 - \delta}{2} \right) \right)$$

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$$= -\frac{1}{2} \cdot (1+\delta) \sum_{i=1}^{\infty} \frac{(-1)^{n+1} \delta^n}{n} + (1-\delta) \sum_{i=1}^{\infty} \frac{(-1)^{n+1} (-\delta)^n}{n}$$

$$\begin{split} 1 - H(X) &= 1 - \frac{1}{2} \left( (1 + \delta) \log \left( \frac{1 + \delta}{2} \right) + (1 - \delta) \log \left( \frac{1 - \delta}{2} \right) \right) \\ &= -\frac{1}{2} \left( (1 + \delta) \log \left( \frac{1 + \delta}{2} \right) + (1 - \delta) \log \left( \frac{1 - \delta}{2} \right) \right) \\ &= -\frac{1}{2} \cdot (1 + \delta) \sum_{i=1}^{\infty} \frac{(-1)^{n+1} \delta^n}{n} + (1 - \delta) \sum_{i=1}^{\infty} \frac{(-1)^{n+1} (-\delta)^n}{n} \\ &= -\frac{1}{2} \cdot \sum_{i=1}^{\infty} 2 \frac{\delta^{2n}}{2n} - \sum_{i=1}^{\infty} 2 \frac{\delta^{2n}}{2n - 1} \end{split}$$

$$1 - H(X) = 1 - \frac{1}{2} \left( (1 + \delta) \log \left( \frac{1 + \delta}{2} \right) + (1 - \delta) \log \left( \frac{1 - \delta}{2} \right) \right)$$

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$$= \sum_{i=1}^{\infty} \frac{\delta^{2n}}{2n(2n-1)}$$

Now expand 1 - H(X) to it's Taylor Seryias at  $\delta$  gives:

$$1 - H(X) = 1 - \frac{1}{2} \left( (1 + \delta) \log \left( \frac{1 + \delta}{2} \right) + (1 - \delta) \log \left( \frac{1 - \delta}{2} \right) \right)$$

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Denote the above by  $K(\delta)$ 

Now, observes that:

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And notice that since p<1 we have  $\gamma<1$ , noitce also that inequlity is symmetric to  $p\mapsto 1-p$ , in paritcular the entropy is not increase if either p=0 or p=1.

#### Claim

Let  $Y = (Y_1, Y_2, ..., Y_m)$  be a bit given by moving each of  $X_i \in X = (X_1, X_2, ..., X_m)$  trough BSC(p). Then:

$$m - H(Y) \le \gamma (m - H(X))$$

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Let's continue to the quantum case. From now on we will denote by  $I(\rho) = \dim \rho - S(\rho)$ .

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#### Quantum no-go.

#### Claim

Let  $\rho_1$  be a reduce density matrix of  $\rho$  Then:

$$-\mathsf{Tr}\ \rho\log\left(\rho_1\otimes I\right)=S(\rho_1)$$

First consider the case in which  $\rho$  is a tensor of  $\rho_1$  namely  $\rho=\rho_1\otimes\rho_2$ , Then clearly  $\rho$  and  $\log\rho_1\otimes I$  commute. Denote by  $\lambda_1,...\lambda_n$  and  $\mu_1,...\mu_m$  the eigen values of  $\rho_1$  and  $\rho_2$ . So the trace equals:

$$\sum \lambda_i \mu_j \log(\lambda_i \cdot 1) = \left(\sum \mu_j\right) \left(\sum_i \lambda_i \log \lambda_i\right)$$

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Then consider the case where  $\rho$  is entangle. Let's span  $\rho$  over base at the form  $\{\rho_1 \otimes \rho_2^i\}$  We get:

$$\mathsf{Tr} \; \rho \log \left( \rho_1 \otimes I \right) = \mathsf{Tr} \; \sum_i \mathsf{a}_i \rho_1 \otimes \rho_2^i \log \left( \rho_1 \otimes I \right)$$
$$= - \sum_i \mathsf{a}_i S(\rho_1) = - S(\rho_1)$$

Let's use the notation  $\sum_{A_k} \rho|_{A_k}$  to denote the sum over all the reduced matrices over k qubits.

#### Claim

Let  $\rho$  be a density matrix over n qubits then:

$$\binom{n}{k}^{-1} \sum_{A_k} I(\rho|_{A_k}) \le \frac{k}{n} I(\rho)$$

#### Proof

Let 
$$\rho_1$$
 be  $\rho^{\otimes k\binom{n}{k}}$  and let  $\rho_2 = \left(\prod_{A_k} \rho|_{A_k}\right)^{\otimes n}$ .

$$0 \leq S(\rho_2||\rho_1) = \mathsf{Tr} \; \left( \rho_1 \left( \log \rho_1 - \log \rho_2 \right) \right) = -S(\rho_1) - \mathsf{Tr} \; \left( \rho_1 \log \rho_2 \right)$$

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$$0 \le -k \binom{n}{k} S(\rho) + \sum_{A_k} n S(\rho|_{A_k})$$

$$\Rightarrow \sum_{A_k} I(\rho|_{A_k}) \le \frac{k}{n} \binom{n}{k} I(\rho)$$

### Claim

Let  $\rho$  be a density matrix of n qubits. Let each qubit be replaced with independent probability  $\rho$  by a fully mixed qubit denoted by v, to give the density matrix  $\sigma$ . Then  $I(\sigma) \leq (1-\rho) \, I(\rho)$ .

Let us write:

$$\sigma = \sum_{k=1}^{n} \sum_{A_k} p^{n-k} (1-p)^k \rho|_{A_k} \otimes v^{n-k}$$

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$$= \sum_{k=1}^{n} \sum_{A_k} p^{n-k} \left(1-p\right)^k I\left(\rho|_{A_k}\right)$$

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$$= \sum_{k=1}^{n} \sum_{A_{k}} p^{n-k} (1-p)^{k} I(\rho|_{A_{k}})$$

$$\leq \sum_{k=1}^{n} p^{n-k} (1-p)^{k} \frac{k}{n} {n \choose k} I(\rho)$$

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$$\leq \sum_{k=1}^{n} p^{n-k} (1-p)^{k} \frac{k}{n} \binom{n}{k} I(\rho)$$

$$= (1-p) I(\rho)$$

Zeros Distillation. Consider the unitary majority gate over 3 qubits, which on the computational basis sets the last 3rd bit to be the majority, and acts as follows:

$$M |0,0,1\rangle \mapsto |1,1,0\rangle$$
  
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And acts trivially on every other configuration.

#### Claim

There is a  $p_0$  such that for any  $p < p_0$ , it follows that if each qubit has a probability less than p of being at  $|1\rangle$ , then after a cycle of Noise  $\rightarrow$  Majority, the probability of the third qubit being  $|1\rangle$  is still less than p.

### Proof.

After the noise round, the density matrices of each qubit<sup>1</sup>:

$$\leq (1-
ho)\left((1-
ho)\ket{0}ra{0}+
ho\ket{1}ra{1}
ight)+
horac{1}{2}\left(\ket{0}ra{0}+\ket{1}ra{1}
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$$\leq (1-p)\left((1-p)\left|0\right\rangle\left\langle 0\right|+p\left|1\right\rangle\left\langle 1\right|\right)+p\frac{1}{2}\left(\left|0\right\rangle\left\langle 0\right|+\left|1\right\rangle\left\langle 1\right|\right)$$

$$=\left((1-p)^{2}+\frac{1}{2}p\right)\left|0\right\rangle\left\langle 0\right|+\left((1-p)p+\frac{1}{2}p\right)\left|1\right\rangle\left\langle 1\right|$$

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After the noise round, the density matrices of each qubit<sup>1</sup>:

$$\leq (1 - p) ((1 - p) |0\rangle \langle 0| + p |1\rangle \langle 1|) + p \frac{1}{2} (|0\rangle \langle 0| + |1\rangle \langle 1|)$$

$$= \left( (1 - p)^2 + \frac{1}{2}p \right) |0\rangle \langle 0| + \left( (1 - p) p + \frac{1}{2}p \right) |1\rangle \langle 1|$$

$$= " |0\rangle \langle 0| + (3/2p - p^2) |1\rangle \langle 1|$$

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$$= " |0\rangle \langle 0| + (3/2p - p^2) |1\rangle \langle 1|$$

And for convenience, let's bound it by looking at the following noisier state, where each qubit is in:

$$(1-2p)|0\rangle\langle 0|+2p|1\rangle\langle 1|$$

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Now, after applying the majority gate, the kets whose third bit is 1 are kets in  $\rho^{\otimes 3}$  that absorb at least two flips.

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Now, after applying the majority gate, the kets whose third bit is 1 are kets in  $\rho^{\otimes 3}$  that absorb at least two flips. Thus, after tracing out the first two qubits, the coefficient of  $|1\rangle \langle 1|$  would be less than:

$$(2p)^3 + 3(1-2p)(2p)^2$$

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$$\left(1-2p\right)\left|0\right\rangle \left\langle 0\right|+2p\left|1\right\rangle \left\langle 1\right|$$

Now, after applying the majority gate, the kets whose third bit is 1 are kets in  $\rho^{\otimes 3}$  that absorb at least two flips. Thus, after tracing out the first two qubits, the coefficient of  $|1\rangle \langle 1|$  would be less than:

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And for  $p_0 = \frac{1}{6}$  we have that this probability is less than p.

## The End.

June 25, 2025