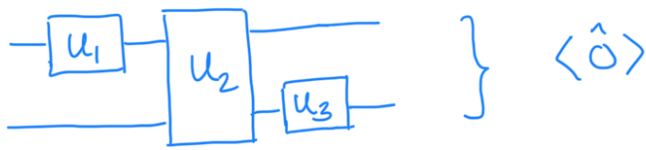
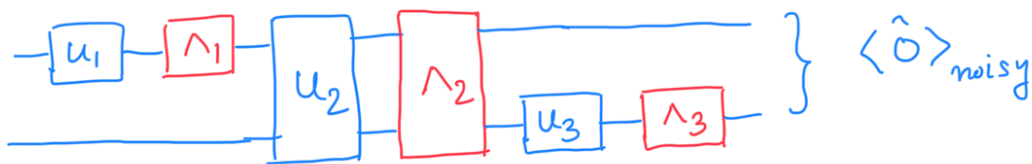


ZERO NOISE EXTRAPOLATION (ZNE)



But each gate is noisy

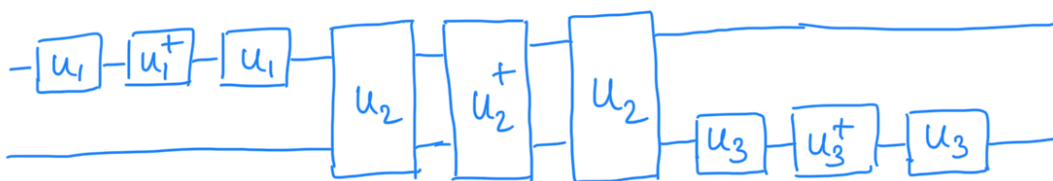


Q> Can we get an estimate of $\langle \hat{O} \rangle$ from $\langle \hat{O} \rangle_{\text{noisy}}$?

$U \rightarrow U(U^\dagger U)$ These two are functionally equivalent.

$$\rightarrow U(U^\dagger U)^{\lambda-1} \text{ for any } \lambda \geq 1 \quad \left| \begin{array}{l} \lambda = 1 \text{ is the} \\ \text{original circuit} \end{array} \right.$$

In ideal circuit:



But in noisy circuit? The noise gets amplified!

Assume depolarizing noise

N qubit circuit, density matrix ρ

p.r. can be a 2 qubit operation U between qubits k & l .

Let there be noise

$$\rho \rightarrow (1 - \epsilon_{kl}) U_{kl} \rho U_{kl}^\dagger + \epsilon_{kl} \left(\frac{I_{kl}}{4} \otimes \rho_{kl} \right)$$

when there is no noise, $\epsilon_{kl} = 0 \Rightarrow \rho \rightarrow U_{kl} \rho U_{kl}^\dagger$

If $U_{kl} \rightarrow U_{kl} (U_{kl}^\dagger U_{kl})$ i.e. $\lambda = 3$

$$\rho \rightarrow (1 - \epsilon_{kl})^3 U_{kl} \rho U_{kl}^\dagger + [1 - (1 - \epsilon_{kl})^3] \left(\frac{I_{kl}}{4} \otimes \rho_{kl} \right)$$

$$= [1 - 3\epsilon_{kl} + O(\epsilon_{kl}^2)] U_{kl} \rho U_{kl}^\dagger + [1 - 1 + 3\epsilon_{kl} - O(\epsilon_{kl}^2)] \left(\frac{I_{kl}}{4} \otimes \rho_{kl} \right)$$

$$= (1 - 3\epsilon_{kl}) \underbrace{U_{kl} \rho U_{kl}^\dagger}_{\text{gives ideal expectation value}} + 3\epsilon_{kl} \underbrace{\left(\frac{I_{kl}}{4} \otimes \rho_{kl} \right)}_{\text{gives noisy expectation value}} + O(\epsilon_{kl}^2)$$

$$\Rightarrow (1 - 3\epsilon_{kl}) \langle \hat{O} \rangle + 3\epsilon_{kl} \langle \hat{O} \rangle_{\text{noisy}} + O(\epsilon_{kl}^2)$$

$$\text{Then for any } \lambda \rightarrow (1 - \lambda \epsilon_{kl}) \langle \hat{O} \rangle + \lambda \epsilon_{kl} \langle \hat{O} \rangle_{\text{noisy}} + O(\epsilon_{kl}^2)$$

If there are M gates:

$$\langle O \rangle_\lambda = \left(1 - \lambda \sum_{i=1}^M \epsilon_i \right) \langle \hat{O} \rangle + \left(\lambda \sum_{i=1}^M \epsilon_i \right) \langle \hat{O} \rangle_{\text{noisy}} + O\left(\sum_{i=1}^M \epsilon_i^2 \right)$$

\downarrow
 dominated by the

maximum error
probability

$$= \left(1 - \lambda \sum_{i=1}^M \epsilon_i\right) \langle \hat{O} \rangle + \left(\lambda \sum_{i=1}^M \epsilon_i\right) \langle \hat{O} \rangle_{\text{noisy}} + O(\epsilon_{\max}^2) \dots (i)$$

If we can set $\lambda=0 \Rightarrow \left(1 - \lambda \sum_{i=1}^M \epsilon_i\right) \langle \hat{O} \rangle + O(\epsilon_{\max}^2)$

1) we get an improved estimate of $\langle \hat{O} \rangle$

2) ZNE is a biased estimator

But how to physically set $\lambda=0$?

For simplicity, take $\epsilon_i = \epsilon \quad \forall i$

$$\langle \hat{O} \rangle_{\lambda} = (1 - \lambda \cdot M \cdot \epsilon) \langle \hat{O} \rangle + \lambda \cdot M \cdot \epsilon \cdot \langle \hat{O} \rangle_{\text{noisy}} + O(\epsilon^2)$$

$$\begin{aligned} \lambda = 1 \Rightarrow \langle \hat{O} \rangle_1 &= (1 - M \cdot \epsilon) \langle \hat{O} \rangle + M \cdot \epsilon \cdot \langle \hat{O} \rangle_{\text{noisy}} + O(\epsilon^2) \\ &= \langle \hat{O} \rangle + M \cdot \epsilon \cdot [\langle \hat{O} \rangle_{\text{noisy}} - \langle \hat{O} \rangle] + O(\epsilon^2) \dots (ii) \end{aligned}$$

$$\begin{aligned} \lambda = 3 \Rightarrow \langle \hat{O} \rangle_3 &= (1 - 3 \cdot M \cdot \epsilon) \cdot \langle \hat{O} \rangle + 3 \cdot M \cdot \epsilon \langle \hat{O} \rangle_{\text{noisy}} + O(\epsilon^2) \\ &= \langle \hat{O} \rangle + 3 M \cdot \epsilon [\langle \hat{O} \rangle_{\text{noisy}} - \langle \hat{O} \rangle] + O(\epsilon^2) \dots (iii) \end{aligned}$$

Take

$$\begin{aligned} \frac{3}{2} \langle \hat{O} \rangle_1 - \frac{1}{2} \langle \hat{O} \rangle_3 &= \frac{3}{2} \langle \hat{O} \rangle + \frac{3}{2} M \cdot \epsilon \cdot \langle \hat{O} \rangle_{\text{noisy}} - \frac{3}{2} M \cdot \epsilon \cdot \langle \hat{O} \rangle + O(\epsilon^2) \\ &\quad - \frac{1}{2} \langle \hat{O} \rangle - \frac{3}{2} M \cdot \epsilon \cdot \langle \hat{O} \rangle_{\text{noisy}} + \frac{3}{2} M \cdot \epsilon \cdot \langle \hat{O} \rangle + O(\epsilon^2) \\ &= \langle \hat{O} \rangle + O(\epsilon^2) \end{aligned}$$

Richardson Extrapolation

$$\left[\begin{array}{c} 2 \dots 1 \\ \hline \frac{1}{2} \langle \hat{O} \rangle_1 - \frac{1}{2} \langle \hat{O} \rangle_3 + O(\epsilon^2) \end{array} \right]$$

$$\left| \frac{2}{2} \langle \hat{O} \rangle_1 - \frac{1}{2} \langle \hat{O} \rangle_3 - \dots \right|$$

Summarize :

1. Prepare the circuit at different values of λ and calculate the expectation values $\langle \hat{O} \rangle_\lambda$ for each λ .
 2. Extrapolate to $\lambda = 0$
 3. Can use higher order extrapolators
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