An Introduction to Quantum Computing

Lecture 19:

Variational Quantum Algorithms

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Agenda

- Variational Algorithms and the NISQ Era
- Variational Quantum Eigensolver and Optimization

Variational Algorithms

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- the same circuit is used for inputs of some maximum length;
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Variational Quantum Algorithms:

- quantum circuits are updated as a way to solve an optimization problem;
- the circuits are usually small and not exceedingly deep;
- the circuits can be run on NISQ (Noisy Intermediate-Scale Quantum) computers.

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Given a generic optimization problem where $C(\cdot)$ is a real cost function and S is a set representing some constraints

$$\max / \min C(x)$$
 subject to $x \in S$

can be reduced to an extremal eigenvalue problem.

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Therefore:

- $\min_{x} C(x) = \min \text{ minimal eigenvalue of } H_C;$
- $\max_{x} C(x) = \max_{x} C(x) = \min_{x} C(x) = \min$

Variational Quantum Eigensolver

How to find the minimal eigenvalue of an Hermitian operator?

Theorem

Let A be an Hermitian operator/matrix on an Hilbert space $\mathcal H$ and λ_{min} its least eigenvalue. Then:

$$\forall |\psi\rangle \in \mathcal{H} \quad \langle \psi | A\psi \rangle \geqslant \lambda_{\min}$$

with equality iff $|\psi\rangle = |\psi\rangle_{\rm min}$ (an eigenvector associated to $\lambda_{\rm min}$).

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Note that f is a well-behaved function. Only problem: the state space is usually huge!

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We "simply" measure A (it is an Hermitian operator, hence a valid observable) multiple times in order to estimate the amplitudes of the basis states.

Recall that given a cost function $C(\cdot)$ we have

$$\langle \psi | \mathcal{H}_C \psi \rangle = \sum_y |\alpha_y|^2 C(y)$$
 where $|\psi\rangle = \sum_y \alpha_y |y\rangle$

The quantum circuits are built from an **ansatz** (a 'template' or 'educated guess') circuit containing single-qubit parameterized rotations and 2-qubit gates.

An example of ansatz on two qubits:

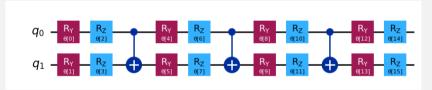


Figure: from https://learning.quantum.ibm.com/tutorial/variational-quantum-eigensolver

The ansatz should be able to reach much of the Hilbert space by an appropriate choice of parameters. Choosing the right ansatz is quite an art.

Variational Quantum Optimization

Now that we can estimate $f(|\psi_{\theta}\rangle) = \langle \psi_{\theta} | A \psi_{\theta} \rangle$, and since $f : \mathbb{R}^p \to \mathbb{R}$ is a classical function, we can use any standard (classical) optimization technique to minimize f.

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Algorithm 2: Optimization by VQE

```
Input: Cost function C, number of circuit evaluations N
   Output: An approximation of min C
   \theta = \theta_0
2 done = false
   while not done do
         generate circuit Q_{\theta} from ansatz with parameters \theta
         for N times do
               |\psi_{\theta}\rangle = Q_{\theta} |00\dots 0\rangle
               measure H_C on |\psi_{\theta}\rangle
 7
         I_{\theta} = \text{estimate } \langle \psi_{\theta} | A \psi_{\theta} \rangle \text{ from measurements}
         if classical optimization algorithm decides I_{\theta} is OK then
               done = true
10
         else
11
               update \theta
12
```