

LECTURE 05 Quantum Mechanics II: Quantum Machine Learning

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Lecture Outline

- 1) Two key building blocks of quantum machine learning (QML)
 - A. Quantum Phase Estimation (QPE)
 - B. Harrow-Hassidim-Lloyd (HHL)
- 2) a brief introduction to
 - A. Quantum SVM (QSVM)

Quantum Machine Learning: why QPE and HHL?

- □ QPCA and QSVM are the quantum counterparts of PCA and SVM.
- □ The HHL algorithm is essential for comprehending QSVM.
- The QPE algorithm is essential for comprehending both the HHL algorithm and QPCA.
- These are versatile and powerful tools in quantum computing, enabling a range of applications.



Quantum Phase Estimation (QPE)

Quantum Phase Estimation: beginning

"Quantum Phase Estimation (QPE)" is an algorithm for estimating the eigenvalues λ of a **unitary matrix U** using a quantum computer.

$$\nabla |\Psi\rangle = \Lambda |\Psi\rangle$$

Assuming the matrix U is <u>a unitary matrix</u>,

$$\langle \psi | \mathbf{U}^{\dagger} \mathbf{U} | \psi \rangle = \langle \psi | \lambda * \lambda | \psi \rangle$$

$$\langle \psi | \psi \rangle = |\lambda|^{2} \langle \psi | \psi \rangle$$

$$|\lambda|^{2} = 1$$

$$\lambda = \cos(2\pi\phi) + i\sin(2\pi\phi) = e^{2\pi i\phi}$$

Euler's formula



Quantum Phase Estimation: beginning

□ Thus, the previous expression can be written as follows:

$$U|\psi\rangle = \lambda|\psi\rangle \implies U|\psi\rangle = e^{2\pi i\emptyset}|\psi\rangle$$

The phase, Ø, is in the range between 0 and 1, which has <u>a decimal format (Ø)</u>.
 QPE is to estimate the phase, Ø, using qubits which has <u>a binary format(Øn)</u>.
 Thus, it would be convenient to express a decimal format as <u>a binary format</u>.

In the above, when $\phi_1 \phi_2 \dots \phi_n \{0,1\}$ are known, the phase ϕ can be obtained.

 0.11_{2}

Toy example



• We want to know how much phase is made due to this unitary matrix "U", given the eigen vector $|\psi\rangle$.

$$\begin{array}{ccc} U|\psi\rangle = \lambda|\psi\rangle & \longrightarrow & U|\psi\rangle = e^{2\pi i 0 / |\psi\rangle} \\ & & & & \\ \hline \\ \\ \\ & & & \\ \hline \\ \\ \hline \\ \\ & & & \\ \hline \\ \\ & & & \\ \hline \\ \\ & & & \\ \hline \\ \\ \hline \\ \\ \\ & & & \\ \hline \\ \\ \hline \\ \\ \hline \\ \\ \\ \hline \\ \\ \\ \hline \\ \\ \\ \hline \\ \hline \\ \hline \\ \\ \hline \\$$

We are going to use it.

Quantum Phase Estimation: toy example





11

$$= (1 + e^{2\pi i 0.\phi_1}) |0\rangle + (1 - e^{2\pi i 0.\phi_1}) |1\rangle$$

- $= (|0\rangle + |1\rangle) + (|0\rangle |1\rangle) e^{2\pi i 0.\phi_1}$
- $= H|0\rangle + H|1\rangle e^{2\pi i 0.\phi_1}$
- $H(|0\rangle + e^{2\pi i 0.\phi_1}|1\rangle)$



Quantum Phase Estimation: toy example

Quantum Phase Estimation: toy example



- ✤ The phase becomes either 0 or ½,
- The eigen value becomes either 1 or -1

- 1) If we measure $|0\rangle$
 - ϕ_1 must be 0,
 - $\Rightarrow (1 + e^{2\pi i 0.0}) |0\rangle + (1 e^{2\pi i 0.0}) |1\rangle$
 - = $2|0\rangle => |0\rangle$

<u>Please, remember that the coefficient $\frac{1}{2}$ </u> in front of each term is being omitted.

2) If we measure $|1\rangle$

•
$$\phi_1$$
 must be 1.

$$\Rightarrow (1 + e^{2\pi i 0.1_2}) |0\rangle + (1 - e^{2\pi i 0.1_2}) |1\rangle$$

$$= (1 + e^{2\pi i 0.5_{10}}) |0\rangle + (1 - e^{2\pi i 0.5_{10}}) |1\rangle$$

$$\Rightarrow = 2|1\rangle => |1\rangle$$

 $e^{2\pi i \emptyset} = \cos(2\pi \emptyset) + i \sin(2\pi \emptyset)$

General Quantum Circuit for QPE



□ The quantum phase estimation algorithm uses phase $\frac{\text{kickback}}{\text{kickback}}$ to write the phase of *U* to the *t* qubits in the counting register.



□ <u>Inverse Quantum Fourier transform (QFT)</u>: (QFT⁺), which is a process to read out the output and produce **|0** or **|1**







- □ Measurement is carried out, and the phase is encoded in binary format, e.g., $0.11..10_2$
- \Box Next, it is converted to decimal format, e.g., 0.78 10

Implementation

Quantum Phase Estimation: Qiskit implementation (t=2)



Creating the circuit above to estimate the phase of a unitary operator U which is equivalent to estimate θ below.

$$U|\psi\rangle = e^{2\pi i\theta}|\psi\rangle$$

Assume that the phase of the unitary matrix U is $4\pi/3$. In other words, the quantum phase estimation algorithm will find the value θ below.

$$2\pi\theta = 4\pi/3$$
 \longrightarrow $\theta = 4/6$ \longleftarrow The answer we expect

Quantum Phase Estimation: Qiskit implementation (t=2)



The exact solution is " $\theta = 4/6 (4\pi/3)''$ in decimal number.

 \Box The result is approximately "11", which corresponds to 3/4 in decimal form.



$$\sum_{k=1}^{n} \emptyset_k 2^{-k}$$

$$\theta = 0.11_2 = 0.75_{10}$$

1*2⁻¹ + 1*2⁻² = 3/4

Quantum Phase Estimation: Qiskit implementation (t=3)



□ What is the reason for using more qubits in the counting register?

Quantum Phase Estimation: Comparison (t=2) vs (t=3)



Harrow-Hassidim-Lloyd (HHL)

The HHL is for solving linear systems of equations, developed by Aram Harrow, Avinatan Hassidim, and Seth Lloyd (HHL) in 2009.

$$\mathbf{A}\vec{\mathbf{x}} = \vec{\mathbf{b}} \qquad \begin{bmatrix} 1 & -1/3 \\ -1/3 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

- Theoretically speaking, the HHL algorithm achieves an exponential improvement compared to the classical algorithm to solve the problem.
 - Quantum algorithm (HHL) **

- *s* is the sparsity: the maximum number of non-zero elements in any row
- $O(\log(N) s^2 \kappa^2 / \epsilon)$ • κ is the condition number: the ratio of the largest and the smallest eigenvalues
 - ε the precision (error)
- Classical algorithm (Gaussian elimination) • $O(N^{3})$

Harrow-Hassidim-Lloyd (HHL)

□ The linear system of equation is presented in the quantum domain through a process called "state preparation".

 $A \vec{x} = \vec{b} \implies A|x\rangle = |b\rangle$

 \Box We would like to obtain $|x\rangle$, so

Harrow-Hassidim-Lloyd (HHL)

□ So what does it mean?

$$|\mathbf{x}\rangle = \sum_{j=0}^{N-1} \frac{1}{\lambda_j} \mathbf{b}_j |u_j\rangle$$

 \Box It means that the solution $|x\rangle$ is the amplitudes of the eigen vector $|u_j\rangle$.

- □ The things that you need to remember to understand the following slides.
 - 1) We need " $1/\lambda_i$ ": inverse of the eigenvalue,
 - 2) $|b\rangle = \sum_{j=0}^{N-1} b_j |u_j\rangle$: the vector $|b\rangle$ can be represented using eigen vectors $|u_j\rangle$,
 - 3) The sum of the squares of the coefficients of $|x\rangle$ equals 1.

$$\sum_{j=0}^{N-1} \left[\frac{b_j}{\lambda_j} \right]^2 = 1$$

General Quantum Circuit for HHL

- ☐ There are three registers
 - R1) Ancilla qubit
 - > a single qubit
 - R2) the eigenvalues of A; $(N \times N)$
 - \succ n_c = N qubits
 - R3) the encoded value of $|{\rm b}\rangle$
 - \succ n_c = log₂N qubits



U: <u>state preparation</u>

 amplitude encoding: encoding a vector as a quantum state, e.g.,

$$\succ$$
 $\vec{b} => |b\rangle$



 $|\psi_1\rangle = |\mathbf{b}\rangle^{\otimes n_b} |0\rangle^{\otimes n_c} |0\rangle$

After QPE, we obtain the eigen value $\tilde{\lambda}_j$ which is encoded in R2 register.

Then, $|\psi_2\rangle$ becomes ...



$$|\Psi_{2}\rangle = |\mathbf{b}\rangle^{\otimes n_{b}} |\widetilde{\lambda_{j}}\rangle^{\otimes n_{c}} |0\rangle$$

- RY gate
 - Encode the amplitudes of the quantum state based on the eigen values from QPE
- **Then**, $|\psi_3\rangle$ becomes as follows;

$$\begin{split} \Psi_{3} &\geq |\mathbf{b}\rangle^{\otimes n_{b}} |\widetilde{\lambda_{j}}\rangle^{\otimes n_{c}} \left(\sqrt{1 - \frac{C^{2}}{\lambda_{j}^{2}}} |0\rangle + \frac{C}{\lambda_{j}} |1\rangle \right) \\ &= \sum_{j=0}^{N-1} \mathbf{b}_{j} |u_{j}\rangle |\widetilde{\lambda_{j}}\rangle \left(\sqrt{1 - \frac{C^{2}}{\lambda_{j}^{2}}} |0\rangle + \frac{C}{\lambda_{j}} |1\rangle \right) \end{split}$$

 $|b\rangle$ is expressed using eigen values eigen vector $|u_i\rangle$ basis from QPE





□ IQPE is applied, which finally returns |x⟩ at R3) register



Then, $|\psi_5\rangle$ becomes as follows;

$$|\Psi_{4}\rangle = \sum_{j=0}^{N-1} 1/\lambda_{j} b_{j} |u_{j}\rangle |\tilde{\lambda}_{j}\rangle |1\rangle$$

Back to original qubit state; QPE => IQPE
$$|\Psi_{5}\rangle = \sum_{j=0}^{N-1} 1/\lambda_{j} b_{j} |u_{j}\rangle |0\rangle^{\otimes n_{c}} |1\rangle$$

X We confirm that it produces |x⟩ at the end.

Quantum Support Vector Machine (QSVM)

Quantum Support Vector Machine (QSVM) with HHL

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Quantum Support Vector Machine for Big Data Classification

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A key idea of the paper is to employ the least-squares reformulation of the SVM that avoids the quadratic programming and obtains the parameters from the solution of a linear equation system.

$$F\begin{pmatrix}b\\\vec{\alpha}\end{pmatrix} \equiv \begin{pmatrix}0 & \vec{1}^T\\\vec{1} & K+\gamma^{-1}\mathbb{1}\end{pmatrix}\begin{pmatrix}b\\\vec{\alpha}\end{pmatrix} = \begin{pmatrix}0\\\vec{y}\end{pmatrix}.$$
 (5)

Then, we can apply the Harrow-Hassidim-Lloyd (HHL) algorithm to handle SVM, which is called "QSVM"!! Quantum Support Vector Machine (QSVM) with QKernel

We formulated SVM as a quadratic programming before...

$$\min_{\lambda} L(\lambda) = \frac{1}{2} \sum_{n=1}^{N} \sum_{m=1}^{N} t_n t_m \lambda_n \lambda_m \mathbf{K}(\mathbf{x}_n^T \mathbf{x}_m) - \sum_{n=1}^{N} \lambda_n$$

s.t. $\lambda \ge 0$, $t^T \lambda = 0$



Quantum Support Vector Machine (QSVM)



Quantum Kernel result

RBF Kernel result

- Two main quantum algorithms—namely QPE and HHL—were explained in detail.
- □ These algorithms are the building blocks of many quantum applications, making it essential to understand their operations.
- QSVM was briefly introduced to demonstrate how QPE and HHL can serve as such building blocks.
- In the last two lectures, we have only just begun to explore the surface of quantum mechanics. I hope you have gained a solid foundation to explore this topic further on your own in the future.