

# Quantum Amplitude Estimation – Applications to Derivative Pricing

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# Classical Monte Carlo

We have seen how classical Monte Carlo can be used to compute the expected payoff  $\mu = \mathbb{E}[X]$  of a simple option  $X = f(S_T)$ :

- ▶ Sample  $x_1, \dots, x_N \stackrel{i.i.d}{\sim} S_T$ ,
- ▶ Set  $\tilde{\mu}_N = \frac{1}{N} \sum_{i=1}^N f(x_i)$ .

By the law of large numbers,

$$\tilde{\mu}_N \xrightarrow{\text{a.s.}} \mu, \quad \text{as } N \rightarrow \infty.$$

Moreover, by the central limit theorem,

$$\|\tilde{\mu}_N - \mu\|_2 = \sqrt{\mathbb{V}(\tilde{\mu}_N)} = \frac{\mathbb{V}(f(S_T))}{\sqrt{N}} \sim \mathcal{O}\left(\frac{1}{\sqrt{N}}\right).$$

Can a quantum computer do better?

# Encoding the problem

Björn showed us how to construct a unitary operator  $\mathcal{U}$  encoding the desired quantity  $\mu$  as

$$\mathcal{U} |0\rangle_{n+1} = \sqrt{1-a} |\psi_0\rangle_n |0\rangle + \sqrt{a} |\psi_1\rangle_n |1\rangle.$$

I.e.,  $a \in [0, 1]$  is the probability of measuring  $|1\rangle$  in the  $(n+1)$ -th qubit.

Can we estimate  $a$ ?

- ▶ Quantum Amplitude Estimation (QAE) does just that.
- ▶ But first: Phase estimation.

# Phase estimation

Given:

- ▶ Unitary operator  $\mathcal{U}$ ,
- ▶ Known eigenvector  $|\psi\rangle$ , ...
- ▶ ... with associated **unknown** eigenvalue  $\lambda = e^{i2\pi q/2^n}$ ,  $q \in [0, 2^n)$ .

Task: Estimate  $\lambda$ .

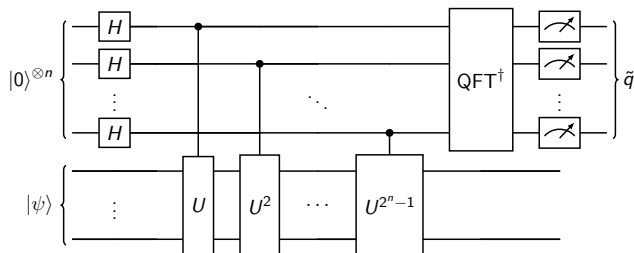


Figure: Phase estimation circuit.

## Theorem (Quantum phase estimation)

*With probability  $1 - \delta$ , the estimate  $\tilde{q}$  provided by QPE satisfies*

$$|\tilde{q} - q| \leq \frac{2}{N} \left( 2 + \frac{1}{2\delta} \right) \sim \mathcal{O} \left( \frac{1}{N} \right),$$

*where  $N = 2^n$ .*

# Phase estimation

“Proof idea” (integer case).

Assume  $q \in \{0, 1, \dots, 2^n - 1\}$ . Then,

$$\begin{aligned} |0\rangle^{\otimes n} |\psi\rangle &\xrightarrow{H^{\otimes n}} \frac{1}{\sqrt{2^n}} \sum_{j=0}^{2^n-1} |j\rangle |\psi\rangle \xrightarrow{\text{Controlled } \mathcal{U}'\text{'s}} \\ &\frac{1}{\sqrt{2^n}} \sum_{j=0}^{2^n-1} |j\rangle e^{2\pi i j q / 2^n} |\psi\rangle = \\ \text{Phase kickback} &\left( \frac{1}{\sqrt{2^n}} \sum_{j=0}^{2^n-1} e^{2\pi i j q / 2^n} |j\rangle \right) |\psi\rangle = \\ &= \text{QFT } |q\rangle |\psi\rangle \xrightarrow{\text{QFT}^\dagger} |q\rangle |\psi\rangle. \end{aligned}$$





# Phase estimation

Quantum Fourier Transform acts as

$$\text{QFT} |q\rangle = \frac{1}{\sqrt{2^n}} \sum_{j=0}^{2^n-1} e^{2\pi i j q / 2^n} |j\rangle.$$

Therefore,

$$|0\rangle^{\otimes n} |\psi\rangle \xrightarrow{\text{Phase estimation}} |q\rangle |\psi\rangle$$



The eigenvalue can be recovered as  $\lambda = e^{2\pi i j q / 2^n}$ .

**Note.** For  $q \in \{0, 1, \dots, 2^n - 1\}$ , phase estimation introduces no error.

**Note'**. For the general case  $q \in [0, 2^n)$ , see [[Kit95](#)], [[MW23](#)].

# Amplitude estimation

Given: unitary operator  $\mathcal{U}$  acting on  $n + 1$  qubits as

$$\mathcal{U} |0\rangle_{n+1} = \sqrt{1-a} |\psi_0\rangle_n |0\rangle + \sqrt{a} |\psi_1\rangle_n |1\rangle.$$

Task: estimate  $a$ , i.e., the probability of measuring 1 in the  $(n + 1)$ -th qubit.

# Amplitude estimation

Amplitude estimation uses a combination of phase estimation and amplitude amplification [Bra+02] in order to get an estimate of  $a$ .

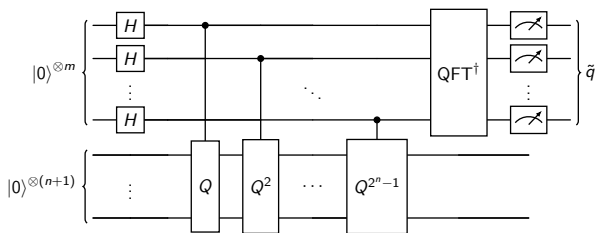


Figure: Amplitude estimation circuit.

Where  $Q$  is the Grover operator

$$Q = US_0U^\dagger S_{\psi_0},$$

and

$$S_{\psi_0} = \mathbb{I} - 2|\psi_0\rangle_n \langle \psi_0|_n \otimes |0\rangle \langle 0|, \quad S_0 = \mathbb{I} - 2|0\rangle_{n+1} \langle 0|_{n+1}.$$

# Amplitude estimation

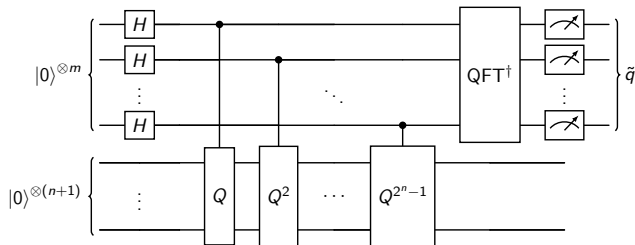


Figure: Amplitude estimation circuit.

The measured number  $\tilde{q} \in \{0, 1, \dots, 2^m - 1\}$  is then mapped to the estimate

$$\tilde{a} = \sin^2 \left( \tilde{\theta}_a \right),$$

where

$$\tilde{\theta}_a = \frac{\pi \tilde{q}}{M}, \quad M = 2^m.$$

## Theorem (Quantum amplitude estimation)

*With probability  $8/\pi^2 \approx 0.81$ , the estimate  $\tilde{a}$  provided by QAE satisfies*

$$|\tilde{a} - a| \leq \frac{2\pi\sqrt{a(1-a)}}{M} + \frac{\pi^2}{M^2} \sim \mathcal{O}\left(\frac{1}{M}\right).$$

Quadratic speedup over classical Monte Carlo!

- ▶ Monte Carlo:  $\mathcal{O}(1/\sqrt{N})$ ,
- ▶ Amplitude estimation:  $\mathcal{O}(1/M)$ .

In practice: run QAE multiple times and take the median value.  
This boost success probability close to 100%

*“Proof idea”*

Use the QPE theorem and the fact that the Grover operator acts as

$$Q^k \mathcal{U} |0\rangle^{\otimes(n+1)} = \cos((2k+1)\theta_a) |\psi_0\rangle_n |0\rangle + \sin((2k+1)\theta_a) |\psi_1\rangle_n |1\rangle,$$

which implies

$$\mathbb{P}(|1\rangle) = \sin^2((2k+1)\theta_a),$$

where  $\theta_a = \arcsin(\sqrt{a})$ .

## A toy example

Let  $\mathcal{U}$  describe a Bernoulli random variable with success probability  $p$ :

$$\mathcal{U}|0\rangle = \sqrt{1-p}|0\rangle + \sqrt{p}|1\rangle.$$

We can model  $\mathcal{U}$  with a rotation around the  $Y$ -axis:

$$\mathcal{U} = R_Y(\theta_p), \quad \theta_p = 2 \arcsin(\sqrt{p}).$$

The Grover operator for this case is particularly simple:

$$\mathcal{Q} = R_Y(2\theta_p),$$

whose powers are very easy to calculate:

$$\mathcal{Q}^k = R_Y(2k\theta_p).$$

# A toy example

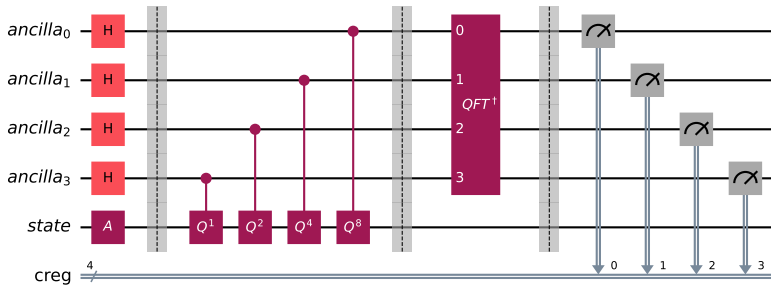


Figure: QAE circuit for binomial random variable, implemented in Qiskit.



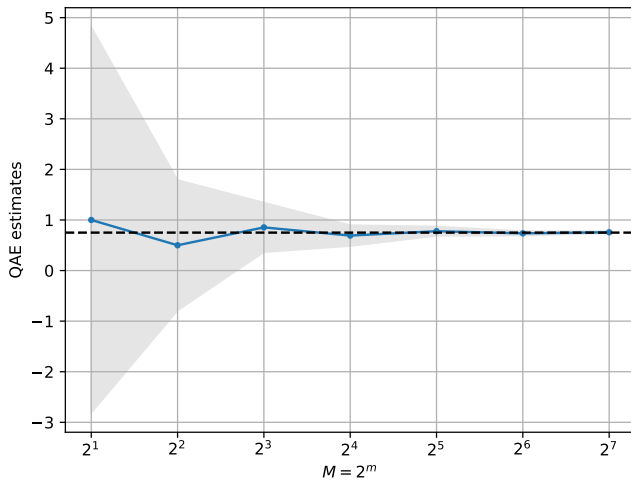


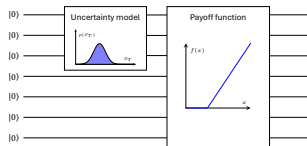
Figure: QAE estimates

# QAE for derivative pricing

We saw how to encode the expected payoff  $\mu$  of a european call option into an operator  $\mathcal{U}$

$$\mathcal{U} |0\rangle_{n+1} = \sqrt{1-a} |\psi_0\rangle_n |0\rangle + \sqrt{a} |\psi_1\rangle_n |1\rangle,$$

after an appropriate re-scaling to the interval  $[0, 1]$ .



Amplitude estimation can be used to approximate  $a$ .

# QAE for option pricing

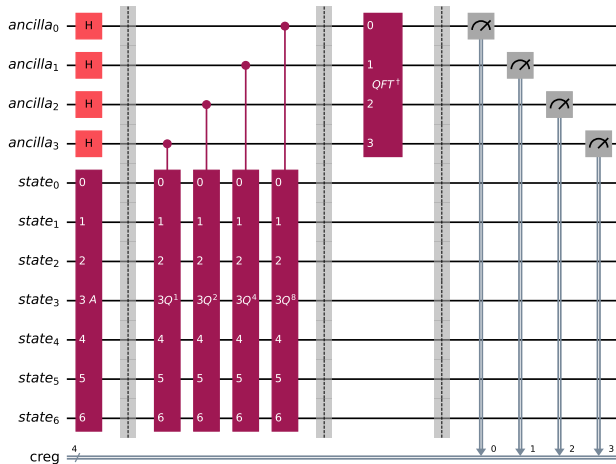


Figure: QAE circuit for option pricing. Implemented in Qiskit.

Notebook time!

# Is QAE NISQ-ready?

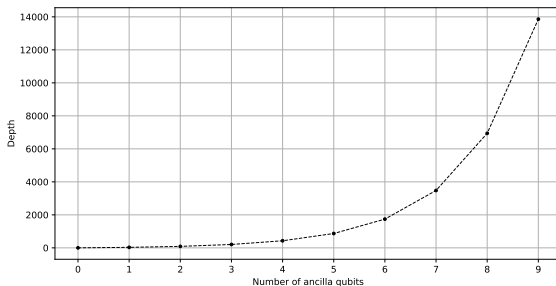


Figure: Depth vs ancillas.

IBM Heron can run  $\approx 1800$  gates within the coherence time of its qubits [IBM].

# QAE alternatives

- ▶ Iterative quantum amplitude estimation [[Gri+21](#)].
- ▶ Faster amplitude estimation [[Nak20](#)].
- ▶ Maximum likelihood amplitude estimation [[Suz+20](#)].

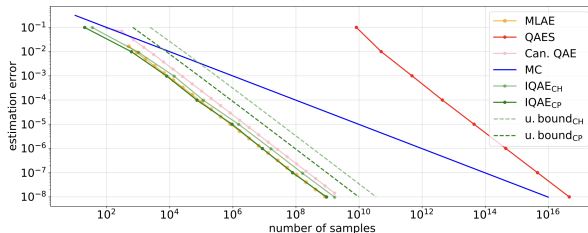


Figure: Convergence rates of different QAE algorithms (Figure 3 in [[Gri+21](#)]).

- ▶ Quadratic speedup over classical Monte Carlo,
- ▶ Useful in areas of mathematical finance where Monte Carlo is common practice, such as derivative pricing,
- ▶ Not NISQ-ready, but active area of research.

- ▶ [Sta+20]: Portfolios of options, basket options, path-dependent options.
- ▶ [WK24]: Asian and barrier options under Heston model.
- ▶ [ZLW19]: Option Pricing with qGANs.
- ▶ [DL21]: Quantum Support Vector Regression for Disability Insurance.



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