Quantum Amplitude Amplification and Estimation (COM-611) Quantum Information Theory and Computation

Alexandre Carlier

EPFL, Lausanne

January 24, 2019



Introduction

- Given a set $X = \{0, 1, \dots, N-1\}$ and a boolean function $\chi : X \longrightarrow \{0, 1\}$, we want to find a *good* element, i.e. an $x \in X$ such that $\chi(x) = 1$.
- If there is only one good element, a classical search algorithm has an average complexity of $\sum_{i=1}^{N} i \times \frac{1}{N} = \frac{N+1}{2}$.
- Quantum approach: given an equal superposition of states $|\Psi\rangle = \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x\rangle$, if we measure $|\Psi\rangle$, we get the correct $|x\rangle$ with probability 1/N: so, the average number of iterations is N.
- Grover's algorithm [Grover, 1996]: we can transform $|\Psi\rangle$ in $\mathcal{O}(\sqrt{N})$ iterations so that performing a measurement on it gives the correct $|x\rangle$ with high probability.

Introduction

- Amplitude amplification [Brassard et al., 2002] is a generalization of Grover's algorithm where the input is given as an arbitrary superposition of elements of X: $|\Psi\rangle = \mathcal{A} |0\rangle = \sum_{x \in X} \alpha_x |x\rangle$ and more than one element may be good elements.
- We can write:

$$|\Psi\rangle = \sum_{x:\chi(x)=1} \alpha_x |x\rangle + \sum_{x:\chi(x)=0} \alpha_x |x\rangle = |\Psi_1\rangle + |\Psi_0\rangle$$

with $a=\langle \Psi_1|\Psi_1\rangle\ll 1$ is the probability that measuring $|\Psi\rangle$ produces a good state.

• The standard approach would thus need to iterate 1/a times to find a good state. Amplitude amplification enables a **quadratic speed-up** in $O(1/\sqrt{a})$.

Outline

1 Quantum amplitude amplification

- The amplitude amplification operator
- Amplitude amplification when a is not known
- Quantum de-randomization

2 Quantum amplitude estimation

3 Applications

Outline

1 Quantum amplitude amplification

- The amplitude amplification operator
- Amplitude amplification when a is not known
- Quantum de-randomization

2 Quantum amplitude estimation

3 Applications

The amplitude amplification operator

- $|\Psi\rangle = \mathcal{A} |0\rangle = |\Psi_1\rangle + |\Psi_0\rangle.$
- S_{χ} is the oracle function:

$$|x\rangle \longmapsto \left\{ \begin{array}{c} -|x\rangle & \text{if } \chi(x)=1\\ |x\rangle & \text{otherwise} \end{array} \right.$$

$$S_{\chi} = \frac{2}{1-a} \left| \Psi_0 \right\rangle \left\langle \Psi_0 \right| - I$$

- $S_0 = I 2 |0\rangle \langle 0|.$
- The amplitude amplification operator is:

$$Q = -\mathcal{A}S_0\mathcal{A}^{\dagger}S_{\chi}$$

= $(\mathcal{A}(2|0\rangle \langle 0| - I)\mathcal{A}^{\dagger}) \times S_{\chi}$
= $(2|\Psi\rangle \langle \Psi| - I)(\frac{2}{1-a}|\Psi_0\rangle \langle \Psi_0| - I)$

Geometrical representation of Q

• We can rewrite $Q = U_{\Psi}U_{\Psi_0}$, where $U_{\Psi} = 2 |\Psi\rangle \langle \Psi| - I$ and $U_{\Psi_0} = \frac{2}{1-a} |\Psi_0\rangle \langle \Psi_0| - I$.



Figure 1: Operator Q as the composition of two reflections.

Matrix representation of \boldsymbol{Q}

$$\begin{split} Q \left| \Psi_{1} \right\rangle &= U_{\Psi} U_{\Psi_{0}} \left| \Psi_{1} \right\rangle = -U_{\Psi} \left| \Psi_{1} \right\rangle = \left(I - 2 \left| \Psi \right\rangle \left\langle \Psi \right| \right) \left| \Psi_{1} \right\rangle \\ &= \left| \Psi_{1} \right\rangle - 2a \left| \Psi \right\rangle = \left(1 - 2a \right) \left| \Psi_{1} \right\rangle - 2a \left| \Psi_{0} \right\rangle \\ Q \left| \Psi_{0} \right\rangle &= U_{\Psi} \left| \Psi_{0} \right\rangle = \left(2 \left| \Psi \right\rangle \left\langle \Psi \right| - I \right) \left| \Psi_{0} \right\rangle \\ &= 2(1 - a) \left| \Psi \right\rangle - \left| \Psi_{0} \right\rangle = 2(1 - a) \left| \Psi_{1} \right\rangle + (1 - 2a) \left| \Psi_{0} \right\rangle \\ \\ \text{Using } \sin^{2}(\theta_{a}) = a \text{ and } \cos^{2}(\theta_{a}) = 1 - a, \text{ we get:} \end{split}$$

$$Q\frac{|\Psi_1\rangle}{\sqrt{a}} = (1-2a)\frac{|\Psi_1\rangle}{\sqrt{a}} - 2\sqrt{a(1-a)}\frac{|\Psi_0\rangle}{\sqrt{1-a}}$$
$$= (1-2\sin^2(\theta_a))\frac{|\Psi_1\rangle}{\sqrt{a}} - 2\cos(\theta_a)\sin(\theta_a)\frac{|\Psi_0\rangle}{\sqrt{1-a}}$$
$$= \cos(2\theta_a)\frac{|\Psi_1\rangle}{\sqrt{a}} - \sin(2\theta_a)\frac{|\Psi_0\rangle}{\sqrt{1-a}}$$
$$Q\frac{|\Psi_0\rangle}{\sqrt{1-a}} = \sin(2\theta_a)\frac{|\Psi_1\rangle}{\sqrt{a}} + \cos(2\theta_a)\frac{|\Psi_0\rangle}{\sqrt{1-a}}$$

Alexandre Carlier (EPFL, Lausanne)

Matrix representation of Q

• Thus, Q is a rotation matrix in the basis $\{\frac{1}{\sqrt{a}} |\Psi_1\rangle, \frac{1}{\sqrt{1-a}} |\Psi_0\rangle\}$:

$$Q = \begin{pmatrix} \cos 2\theta_a & \sin 2\theta_a \\ -\sin 2\theta_a & \cos 2\theta_a \end{pmatrix}$$

• It has eigenvalues $e^{2i\theta_a}, e^{-2i\theta_a}$ with corresponding eigenvectors $\frac{1}{2} \begin{pmatrix} 1 \\ i \end{pmatrix}, \frac{1}{2} \begin{pmatrix} 1 \\ -i \end{pmatrix}$, noted $|\Psi_+\rangle$ and $|\Psi_-\rangle$.

Quantum amplitude amplification

 ${\, \bullet \, }$ We can now write $|\Psi\rangle$ in the Q-eigenvector basis:

$$|\Psi\rangle = \frac{-i}{2} (e^{i\theta_a} |\Psi_+\rangle - e^{-i\theta_a} |\Psi_-\rangle)$$

and it follows that:

$$Q^{j} |\Psi\rangle = \frac{-i}{2} (e^{(2j+1)i\theta_{a}} |\Psi_{+}\rangle - e^{-(2j+1)i\theta_{a}} |\Psi_{-}\rangle)$$

• By writing it back in the original $\{\frac{1}{\sqrt{a}} \ket{\Psi_1}, \frac{1}{\sqrt{1-a}} \ket{\Psi_0}\}$ basis:

$$Q^{j} |\Psi\rangle = \sin((2j+1)\theta_{a}) \frac{1}{\sqrt{a}} |\Psi_{1}\rangle + \cos((2j+1)\theta_{a}) \frac{1}{\sqrt{1-a}} |\Psi_{0}\rangle$$

Quantum amplitude amplification

- After m applications of the operator Q, measuring the state $|\Psi\rangle$ produces a good state with probability equal to $\sin^2((2m+1)\theta_a)$.
- $x \mapsto \sin^2((2x+1)\theta_a)$ is maximized for $x = \frac{\pi}{4\theta} \frac{1}{2}$.
- Thus the probability is maximized for $m = \lfloor \pi/(4\theta_a) \rfloor$ (when the value of a is known).
- We can show that $\sin^2((2m+1)\theta_a) \ge 1-a$.

Complexity of the algorithm

• We use 2m + 1 applications of \mathcal{A} and \mathcal{A}^{\dagger} .

• Since
$$\theta_a \approx \sin(\theta_a) = \sqrt{a}$$
, we get:

$$2m + 1 = 2 \lfloor \pi/(4\theta_a) \rfloor + 1$$
$$\approx 2 \lfloor \pi/(4\sqrt{a}) \rfloor + 1$$
$$= \mathcal{O}(\frac{1}{\sqrt{a}})$$

• And the success probability is $1 - a \approx 1$.

























Figure 2: Visualization of the Quantum amplitude amplification algorithm

And indeed $m = \lfloor \pi/4\theta_a \rfloor = 11$.



Figure 2: Visualization of the Quantum amplitude amplification algorithm

And indeed $m = \lfloor \pi/4\theta_a \rfloor = 11$.

Grover's algorithm

Example

$$|\Psi\rangle=\frac{1}{\sqrt{N}}\sum_{x=0}^{N-1}|x\rangle$$
 and $\chi=\mathbb{1}_{x=0}.$ Then $a=1/N\ll1$,

$$m = \left\lfloor \frac{\pi}{4\theta_a} \right\rfloor \approx \left\lfloor \frac{\pi}{4\sin\theta_a} \right\rfloor = \left\lfloor \frac{\pi\sqrt{N}}{4} \right\rfloor = \mathcal{O}(\sqrt{N})$$

and we get the state $|0\rangle$ with probability $\sin^2((2m+1)\theta_a) \ge 1 - a \approx 1$.

Alexandre Carlier

A special case

Example

 $|\Psi\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) |x\rangle$ and $\chi = \mathbb{1}_{x=0}$. We have a = 1/2, $\theta_a = \frac{\pi}{4}$. Then, m = 1 and $\sin^2((2m+1)\theta_a) = \sin^2\frac{3\pi}{4} = \frac{1}{2} = a$. Amplitude amplification has no effect.



Outline



Quantum amplitude amplification

- The amplitude amplification operator
- Amplitude amplification when a is not known
- Quantum de-randomization

2 Quantum amplitude estimation

3 Applications

Amplitude amplification when a is not known

- When *a* is not known, we can first estimate it using quantum amplitude estimation (see section 2) and then run the previous algorithm by replacing the exact *a* by its estimate.
- Another approach is to use QSearch. The intuition is the following: for θ ~ U[0, 2π], E [sin² θ] = ½. By choosing M sufficiently large, Mθ_a is large and by picking j ∈_U [[1, M]], jθ_a mod 2π follows a good approximation of U[0, 2π] (and so does (2j + 1)θ_a mod 2π).
- Then, the probability $\sin^2((2j+1)\theta_a)$ that the measurement produces a good state is in average $\frac{1}{2}$.
- Since we don't know θ_a , we use an exponential search space for $M = c^l$ by iteratively incrementing the value of l for a constant c.

The QSearch algorithm

Initialization: l = 0. Repeat: (while $|z\rangle$ is not a good state)



Figure 3: The QSearch algorithm

Outline

1

Quantum amplitude amplification

- The amplitude amplification operator
- Amplitude amplification when a is not known
- Quantum de-randomization

2 Quantum amplitude estimation

3 Applications

Quantum de-randomization when a is known

The success probability of the Quantum Amplitude Amplification algorithm is 1 - a. It turns out we can actually find a good solution with *certainty*.

- $m \mapsto \sin^2((2m+1)\theta_a)$ is maximized for $\tilde{m} = \frac{\pi}{4\theta} \frac{1}{2}$.
- If \tilde{m} is an integer, $\sin^2((2\tilde{m}+1)\theta_a) = 1$.
- Else we use $m=\lceil \tilde{m}\rceil=\lfloor \pi/(4\theta_a)\rfloor$ iterations, which is slightly too much.
- The de-randomization approach is the following:
 - Apply Q only $\lfloor \tilde{m} \rfloor$ times. The resulting state is:

$$\sin((2\lfloor \tilde{m} \rfloor + 1)\theta_a)\frac{1}{\sqrt{a}} |\Psi_1\rangle + \cos((2\lfloor \tilde{m} \rfloor + 1)\theta_a)\frac{1}{\sqrt{1-a}} |\Psi_0\rangle$$

 \bullet We further define $Q'(\phi,\varphi)=-\mathcal{A}S_0(\phi)\mathcal{A}^\dagger S_\chi(\varphi)$

where
$$\left\{ \begin{array}{l} S_{0}(\phi)=e^{i\phi}\left|0\right\rangle\left\langle 0\right|+\left|1\right\rangle\left\langle 1\right|\\ S_{\chi}(\varphi)=\frac{e^{i\varphi}}{\sqrt{a}}\left|\Psi_{1}\right\rangle\left\langle \Psi_{1}\right|+\frac{1}{\sqrt{1-a}}\left|\Psi_{0}\right\rangle\left\langle \Psi_{0}\right| \end{array} \right.$$

Quantum de-randomization when a is known

•
$$Q = Q'(\phi = \pi, \varphi = \pi)$$

 \bullet By applying one final $Q'(\phi,\varphi),$ we obtain:

$$\star |\Psi_1\rangle + \left(e^{i\varphi}(1-e^{i\phi})\sqrt{a}\sin((2\lfloor\tilde{m}\rfloor+1)\theta_a) - ((1-e^{i\phi})a + e^{i\phi}) \\ \frac{1}{\sqrt{1-a}}\cos((2\lfloor\tilde{m}\rfloor+1)\theta_a)\right)|\Psi_0\rangle$$

• We can choose ϕ and φ so that the coefficient in front of $|\Psi_0
angle=$ 0:

$$\iff \cot((2\lfloor \tilde{m} \rfloor + 1)\theta_a) = e^{i\varphi} 2\sqrt{a(1-a)} \frac{1 - e^{i\phi}}{2((1-e^{i\phi})a + e^{i\phi})}$$
$$= e^{i\varphi} \sin(2\theta_a) (2 \underbrace{a}_{=1-\cos(2\theta_a)} + \frac{2e^{i\phi}}{1-e^{i\phi}})^{-1}$$
$$= e^{i\varphi} \sin(2\theta_a) (-\cos(2\theta_a) + \underbrace{\frac{1+e^{i\phi}}{1-e^{i\phi}}}_{=i\cot(\phi/2)})^{-1}$$

Outline

Quantum amplitude amplification

- The amplitude amplification operator
- Amplitude amplification when a is not known
- Quantum de-randomization

2 Quantum amplitude estimation

3 Applications

Quantum amplitude estimation

- Amplitude amplification: find $x \in X$ such that $\chi(x) = 1$.
- Amplitude estimation: estimate $a = \langle \Psi_1 | \Psi_1 \rangle$.
- By $a = \sin^2(\theta_a)$, an estimate for a translates into an estimate for θ_a .
- The eigenvalues of Q are $\lambda_+ = e^{2i\theta_a}$ and $\lambda_- = e^{-2i\theta_a}$, so we can instead estimate one of these eigenvalues.

Quantum amplitude estimation

• Let us define the operator

$$\Lambda_M(Q): |j\rangle |y\rangle \mapsto |j\rangle Q^j |y\rangle$$

so that e.g:

$$\Lambda_M(Q) \left| j \right\rangle \left| \Psi_+ \right\rangle = e^{2i\theta_a j} \left| j \right\rangle \left| \Psi_+ \right\rangle$$

• We recall the quantum Fourier transform (for $x \in \{0, \dots, M-1\}$):

$$F_M: |x\rangle \mapsto \frac{1}{\sqrt{M}} \sum_{y=0}^{M-1} e^{2\pi i x y/M} |y\rangle$$

• And we define (for a real $0 \le \omega < 1$):

$$|S_M(\omega)\rangle = \frac{1}{\sqrt{M}} \sum_{y=0}^{M-1} e^{2\pi i \omega y} |y\rangle$$

so that, for $x \in \{0, \dots, M-1\}$: $|S_M(x/M)\rangle = F_M |x\rangle$.

Quantum circuit for amplitude estimation

 $(F_M^{-1}\otimes I)(\Lambda_M(Q))(F_M\otimes I)$ applied on the state $|0
angle\otimes \mathcal{A}\,|0
angle$



(If M is a power of 2, we can replace the Quantum Fourier transforms by Hadamard gates)

Proof of correctness

The quantum circuit corresponds to the unitary transformation $(F_M^{-1} \otimes I)(\Lambda_M(Q))(F_M \otimes I)$ applied on the state $|0\rangle \otimes \mathcal{A} |0\rangle$, with

$$\mathcal{A} \left| 0 \right\rangle = -\frac{i}{\sqrt{2}} (e^{i\theta_a} \left| \Psi_+ \right\rangle - e^{-i\theta_a} \left| \Psi_- \right\rangle)$$

By applying $F_M \otimes I$:

$$\frac{1}{\sqrt{2M}}\sum_{j=0}^{M-1}|j\rangle\otimes\left(e^{i\theta_{a}}|\Psi_{+}\rangle-e^{-i\theta_{a}}|\Psi_{-}\rangle\right)$$

After applying $\Lambda_M(Q)$:

$$\frac{e^{i\theta_a}}{\sqrt{2}} \left| S_M(\theta_a/\pi) \right\rangle \otimes \left| \Psi_+ \right\rangle - \frac{e^{-i\theta_a}}{\sqrt{2}} \left| S_M(1 - \theta_a/\pi) \right\rangle \otimes \left| \Psi_- \right\rangle$$

Proof of correctness

• Finally, after $F_M^{-1} \otimes I$, we have:

$$rac{e^{i heta_a}}{\sqrt{2}}F_M^{-1}\ket{S_M(heta_a/\pi)}\otimes\ket{\Psi_+}-rac{e^{-i heta_a}}{\sqrt{2}}F_M^{-1}\ket{S_M(1- heta_a/\pi)}\otimes\ket{\Psi_-}$$

- By tracing out the second register in the eigenvector basis $\{|\Psi_+\rangle, |\Psi_-\rangle\}$, we obtain a $\frac{1}{2}$ - $\frac{1}{2}$ mixture of $F_M^{-1} |S_M(\theta_a/\pi)\rangle$ and $F_M^{-1} |S_M(1 \theta_a/\pi)\rangle$.
- By symmetry (since $\sin^2(\pi \frac{y}{M}) = \sin^2(\pi(1-\frac{y}{M}))$), we can assume the measured $|y\rangle$ is the result of measuring $F_M^{-1} |S_M(\theta_a/\pi)\rangle$.

• We thus have $\tilde{\theta_a} = \pi \frac{y}{M}$ is a good estimate of θ_a (see next slide).

Bounding the error of the estimate (1/6)

 $\frac{1}{M}F_M^{-1}|S_M(\omega)\rangle$ is a good estimate of ω . Indeed, if $\omega = x/M$ for some $0 \le x < M$, then $F_M^{-1}|S_M(x/M)\rangle = |x\rangle$. Otherwise:

Theorem

Let X be the r.v. corresponding to the result of measuring $F_M^{-1} |S_M(\omega)\rangle$. Then:

$$\mathbb{P}\left(\left|\frac{1}{M}X - \omega\right| \le \frac{1}{M}\right) \ge \frac{8}{\pi^2} \approx 0.81$$

Lemma

Letting
$$\Delta = \left|\frac{1}{M}x - \omega\right|$$
 for some $x \in \{0, \dots, M-1\}$, we have:

$$\mathbb{P}[X = x] = \frac{\sin^2(M\Delta\pi)}{M^2\sin^2(\Delta\pi)}$$

Bounding the error of the estimate (2/6)

Proof of the Lemma.

$$\begin{aligned} \mathbb{P}[X=x] &= |\langle x| F_M^{-1} | S_M(\omega) \rangle |^2 \\ &= |(F_M | x \rangle)^{\dagger} | S_M(\omega) \rangle |^2 \\ &= |\langle S_M(x/M) | S_M(\omega) \rangle |^2 \\ &= \left| (\frac{1}{\sqrt{M}} \sum_{y=0}^{M-1} e^{2\pi i x/My} \langle y|) (\frac{1}{\sqrt{M}} \sum_{y=0}^{M-1} e^{2\pi i \omega y} | y \rangle) \right|^2 \\ &= \frac{1}{M^2} \left| \sum_{y=0}^{M-1} e^{2\pi i \Delta y} \right|^2 = \frac{\sin^2(M\Delta\pi)}{M^2 \sin^2(\Delta\pi)} \end{aligned}$$

Bounding the error of the estimate (3/6)

Proof of the Theorem.

$$\mathbb{P}[d(X/M,\omega) \le 1/M] = \mathbb{P}[X = \lfloor M\omega \rfloor] + \mathbb{P}[X = \lceil M\omega \rceil]$$
$$= \frac{\sin^2(M\Delta\pi)}{M^2 \sin^2(\Delta\pi)} + \frac{\sin^2(M(\frac{1}{M} - \Delta)\pi)}{M^2 \sin^2((\frac{1}{M} - \Delta)\pi)}$$
$$\ge \frac{8}{\pi^2}$$

Since the minimum of this expression is reached at $\Delta = 1/(2M)$.

Bounding the error of the estimate (4/6)

A bounding error on $\tilde{\theta_a}$ translates into a bound on \tilde{a} .

Lemma

Let
$$a = \sin^2(\theta_a)$$
 and $\tilde{a} = \sin^2(\tilde{\theta_a})$ with $0 \le \theta_a, \tilde{\theta_a} \le \frac{\pi}{2}$. Then:

$$|\tilde{\theta_a} - \theta_a| \leq \epsilon \Longrightarrow |\tilde{a} - a| \leq 2\epsilon \sqrt{a(1-a)} + \epsilon^2$$

Bounding the error of the estimate (5/6)

A bounding error on $\tilde{\theta_a}$ translates into a bound on \tilde{a} .

Proof.

$$\begin{split} \tilde{a} - a &= \sin^2(\tilde{\theta_a}) - \sin^2(\theta_a) \leq \sin^2(\theta_a + \epsilon) - \sin^2(\theta_a) \\ &= (\sin(\theta_a)\cos(\epsilon) + \sin(\epsilon)\cos(\theta_a))^2 - \sin^2(\theta_a) \\ &= \sin^2(\theta_a)\cos(\epsilon) + \sin^2(\epsilon)\cos^2(\theta_a) + 2\cos(\theta_a)\sin(\theta_a)\cos(\epsilon)\sin(\epsilon) \\ &- \sin^2(\theta_a) \\ &= \sin^2(\epsilon)(\cos^2(\theta_a) - \sin^2(\theta_a)) + \sqrt{a(1-a)}\sin^2(\epsilon) \\ &= \sqrt{a(1-a)}\sin(2\epsilon) + (1-2a)\sin^2(\epsilon) \\ &\leq 2\epsilon\sqrt{a(1-a)} + \epsilon^2 \end{split}$$

Bounding the error of the estimate (6/6)

Combining those results, the Amplitude Estimation algorithm outputs $\hat{\theta_a}$ such that

$$\begin{split} |\tilde{\theta_a}/\pi - \theta_a/\pi| &\leq \frac{1}{M} \\ \iff |\tilde{\theta_a} - \theta_a| &\leq \frac{\pi}{M} \end{split}$$

with probability greater than $8/\pi^2$.

Thus, by setting $\epsilon = \frac{\pi}{M}$:

$$|\tilde{a}-a| \le 2\pi \frac{\sqrt{a(1-a)}}{M} + \frac{\pi^2}{M^2}$$

Outline

Quantum amplitude amplification

- The amplitude amplification operator
- Amplitude amplification when a is not known
- Quantum de-randomization

Quantum amplitude estimation

3 Applications

1. Application to counting

- The amplitude estimation algorithm can be used for counting the number of good elements t = |{x ∈ X s.t. χ(x) = 1}|.
- By choosing $\mathcal{A} = F_N$ the Quantum Fourier Transform:

$$F_N: |x\rangle \mapsto \frac{1}{\sqrt{N}} \sum_{y=0}^{N-1} e^{2\pi i x y/M} |y\rangle$$

we have:

$$\mathcal{A} \left| 0 \right\rangle = \frac{1}{\sqrt{N}} \sum_{y=0}^{N-1} \left| y \right\rangle = \underbrace{\frac{1}{\sqrt{N}} \sum_{\substack{y:\chi(y)=1\\ = |\Psi_1\rangle}} \left| y \right\rangle}_{= |\Psi_1\rangle} + \underbrace{\frac{1}{\sqrt{N}} \sum_{\substack{y:\chi(y)=0\\ = |\Psi_0\rangle}} \left| y \right\rangle}_{= |\Psi_0\rangle}$$

Thus, $a = \langle \Psi_1 | \Psi_1 \rangle = \frac{1}{N}$, and so $t = a \times N$.

2. Application to Monte Carlo sampling

- Let X be a random variable taking values $\{0, \ldots, N\}$ with probability p_i . We want to compute $\mathbb{E}[f(X)]$.
- Using Monte Carlo sampling, with M evaluations of f, we get:

$$\frac{1}{M}\sum_{k=0}^{M} f(X_k) \approx \mathbb{E}[f(X)] \pm \frac{C}{\sqrt{M}}$$

• Quantum approach: define

$$\left|\Psi\right\rangle = \sum_{i=0}^{N-1} \sqrt{p_i} \left|i\right\rangle$$

and the operator

$$F:\left|i\right
angle \otimes \left|0
ight
angle \mapsto \left|i
ight
angle \otimes \left(\sqrt{1-f(i)}\left|0
ight
angle + \sqrt{f(i)}\left|1
ight
angle
ight)$$

Then:

$$F |\Psi\rangle \otimes |0\rangle = \sum_{i=0}^{N-1} \sqrt{1 - f(i)} \sqrt{p_i} |i\rangle \otimes |0\rangle + \sqrt{f(i)} \sqrt{p_i} |i\rangle \otimes |1\rangle$$

2. Application to Monte Carlo sampling

Using amplitude estimation, we estimate the probability to measure $|1\rangle$ in the last Qbit: $\tilde{a} = \sum_{i=0}^{N-1} p_i f(i) = \mathbb{E}[f(X)]$, and using M evaluations of f:

$$|\tilde{a}-a| \le 2\pi \frac{\sqrt{a(a-a)}}{M} + \frac{\pi^2}{M^2}$$

with a convergence rate of $\mathcal{O}(\frac{1}{M})$ to be compared to the classical $\mathcal{O}(\frac{1}{\sqrt{M}})$ rate.

3. Application to Quantum Risk Analysis

• *Quantum Risk Analysis* [Woerner and Egger, 2018] (IBM Research - Zurich):

In quantitative finance, VaR (Value at Risk) and CVaR (Conditional Value at Risk) are typically estimated using Monte Carlo sampling of the relevant probability distribution.

- For a confidence value $\alpha \in [0, 1]$, $VaR_{\alpha}(X)$ is the smallest l such that $\mathbb{P}[X \leq l] \geq (1 \alpha)$.
- By defining $f_l(x) = 1$ if $\mathbb{1}_{x \leq l}$, we thus want to approximate $\mathbb{P}[X \leq l] = \mathbb{E}[f_l(X)]$ of a random variable X taking values $\{0, \ldots, N\}$ with probability p_i .

Conclusion

- Quadratic speedup: this speedup is in fact the best we can attain [Bennett et al., 1997].
- Even if *amplitude amplification* and *estimation* doesn't solve NP-complete problems in polynomial time, we can apply it to more than just search problems, such as Monte Carlo sampling with a non-negligible speedup.

References

- Bennett, C. H., Bernstein, E., Brassard, G., and Vazirani, U. (1997).
 Strengths and weaknesses of quantum computing.
 SIAM journal on Computing, 26(5):1510–1523.
- Brassard, G., Hoyer, P., Mosca, M., and Tapp, A. (2002). Quantum amplitude amplification and estimation. Contemporary Mathematics, 305:53–74.
- Grover, L. K. (1996).

A fast quantum mechanical algorithm for database search. In Proceedings of the twenty-eighth annual ACM symposium on Theory of computing, pages 212–219. ACM.

Woerner, S. and Egger, D. J. (2018). Quantum risk analysis. arXiv preprint arXiv:1806.06893.