

# QUANTUM ALGORITHMS

## HOMEWORK 7 SELECTED SOLUTIONS

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**8.7.** Prove the properties (8.10)–(8.13) of the operator norm.

**Solution:** The properties in question are

$$(8.10) \quad \|XY\| \leq \|X\|\|Y\|,$$

$$(8.11) \quad \|X^\dagger\| = \|X\|,$$

$$(8.12) \quad \|X \otimes Y\| = \|X\|\|Y\|,$$

$$(8.13) \quad \|U\| = 1 \quad \text{if } U \text{ is unitary.}$$

We will prove each of these in turn.

*Proof.* We have

$$\begin{aligned} \|XY\| &= \sup_{|\xi\rangle \neq 0} \frac{\|XY|\xi\rangle\|}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \left( \frac{\|XY|\xi\rangle\|}{\|Y|\xi\rangle\|} \right) \left( \frac{\|Y|\xi\rangle\|}{\| |\xi\rangle \|} \right) \\ &\leq \left( \sup_{|\xi\rangle \neq 0} \frac{\|XY|\xi\rangle\|}{\|Y|\xi\rangle\|} \right) \left( \sup_{|\xi\rangle \neq 0} \frac{\|Y|\xi\rangle\|}{\| |\xi\rangle \|} \right) \leq \left( \sup_{|\zeta\rangle \neq 0} \frac{\|X|\zeta\rangle\|}{\| |\zeta\rangle \|} \right) \left( \sup_{|\xi\rangle \neq 0} \frac{\|Y|\xi\rangle\|}{\| |\xi\rangle \|} \right) \\ &= \|X\|\|Y\|, \end{aligned}$$

proving (8.10).

Next, we begin by noting that  $\|\langle \xi | \cdot \| = \|\cdot | \xi\rangle\|$ . Therefore

$$\|X^\dagger\| = \sup_{|\xi\rangle \neq 0} \frac{\|X^\dagger|\xi\rangle\|}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \frac{\| |X^\dagger\xi\rangle \|}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \frac{\|\langle \xi | X\|}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \frac{\|\langle \xi | X\|}{\| |\xi\rangle \|} = \|X\|,$$

proving (8.11).

Next, we have

$$\| |\alpha\rangle \otimes |\beta\rangle \| = \sqrt{\langle \alpha \otimes \beta | \alpha \otimes \beta \rangle} = \sqrt{\langle \alpha | \alpha \rangle \langle \beta | \beta \rangle} = \|\alpha\|\|\beta\|.$$

It follows that

$$\begin{aligned} \|X \otimes Y\| &= \sup_{|\xi\rangle \otimes |\zeta\rangle \neq 0} \frac{\|(X \otimes Y)(|\xi\rangle \otimes |\zeta\rangle)\|}{\| |\xi\rangle \otimes |\zeta\rangle \|} = \sup_{|\xi\rangle \otimes |\zeta\rangle \neq 0} \frac{\| |X\xi\rangle \otimes |Y\zeta\rangle \|}{\| |\xi\rangle \otimes |\zeta\rangle \|} \\ &= \sup_{|\xi\rangle \otimes |\zeta\rangle \neq 0} \frac{\| |X\xi\rangle \| \| |Y\zeta\rangle \|}{\| |\xi\rangle \| \| |\zeta\rangle \|} = \sup_{|\xi\rangle \otimes |\zeta\rangle \neq 0} \left( \frac{\| |X\xi\rangle \|}{\| |\xi\rangle \|} \right) \left( \frac{\| |Y\zeta\rangle \|}{\| |\zeta\rangle \|} \right) \\ &= \left( \sup_{|\xi\rangle \neq 0} \frac{\| |X\xi\rangle \|}{\| |\xi\rangle \|} \right) \left( \sup_{|\zeta\rangle \neq 0} \frac{\| |Y\zeta\rangle \|}{\| |\zeta\rangle \|} \right) = \|X\|\|Y\|, \end{aligned}$$

proving (8.12).

Finally, if  $U$  is unitary then

$$\begin{aligned}\|U\| &= \sup_{|\xi\rangle \neq 0} \frac{\|U|\xi\rangle\|}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \frac{\sqrt{\langle U\xi | U\xi \rangle}}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \frac{\sqrt{\langle \xi | \xi \rangle}}{\| |\xi\rangle \|} = \sup_{|\xi\rangle \neq 0} \frac{\| |\xi\rangle \|}{\| |\xi\rangle \|} \\ &= \sup_{|\xi\rangle \neq 0} 1 = 1,\end{aligned}$$

proving (8.13).  $\square$

**8.8.** Prove the two basic properties of approximation with ancillas:

- If  $\tilde{U}$  approximates  $U$  with precision  $\delta$ , then  $\tilde{U}^{-1}$  approximates  $U^{-1}$  with the same precision  $\delta$ .
- If unitary operators  $\tilde{U}_k$  approximate unitary operators  $U_k$  ( $1 \leq k \leq L$ ) with precision  $\delta_k$ , then  $\tilde{U}_L \cdots \tilde{U}_1$  approximates  $U_L \cdots U_1$  with precision  $\sum_k \delta_k$ .

**Solution:** Before we begin, we build some tools. Define  $V : \mathbb{B}^{\otimes n} \rightarrow \mathbb{B}^{\otimes N}$  by  $V|\xi\rangle = |\xi\rangle \otimes |0^{N-n}\rangle$ . The statement “ $\tilde{U}$  approximates  $U$  with ancillas with precision  $\delta$ ” is equivalent to

$$\|\tilde{U}V - VU\| \leq \delta.$$

We are now ready to do the proofs.

**a): Proof.** Note that  $\tilde{U}$  is always taken to be *unitary*. Using the fact that  $\|T\| = 1$  and  $\|T\|\|X\| = \|TX\|$  for unitary  $T$ , we have

$$\begin{aligned}\|\tilde{U}^{-1}V - VU^{-1}\| &= \|\tilde{U}\|\|\tilde{U}^{-1}V - VU^{-1}\|\|U\| = \|\tilde{U}(\tilde{U}^{-1}V - VU^{-1})U\| \\ &= \|\tilde{U}\tilde{U}^{-1}VU - \tilde{U}VU^{-1}U\| = \|VU - \tilde{U}V\| = \|\tilde{U}V - VU\| \leq \delta.\end{aligned}$$

Hence  $\tilde{U}^{-1}$  approximates  $U^{-1}$  with precision  $\delta$ .  $\square$

**b): Proof.** We will proceed by induction on  $L$ . The base case of  $L = 1$  is included in the assumptions, so it certainly holds. Assume now that the claim holds for a product of  $L - 1$  matrices. Let  $\tilde{W} = \tilde{U}_2 \cdots \tilde{U}_L$  and  $W = U_2 \cdots U_L$ . These are products of  $L - 1$  matrices, so by the inductive hypothesis we have that  $\tilde{W}$  approximates  $W$  with ancillas with precision  $\sum_{k=2}^L \delta_k$ . We now have

$$\begin{aligned}\|\tilde{U}_1 \cdots \tilde{U}_L V - VU_1 \cdots U_L\| &= \|\tilde{U}_1 \tilde{W} V - VU_1 W\| \\ &= \|\tilde{U}_1(\tilde{W}V - VW) + (\tilde{U}_1 V - VU_1)W\| \\ &\leq \|\tilde{U}_1(\tilde{W}V - VW)\| + \|(\tilde{U}_1 V - VU_1)W\| \\ &= \|\tilde{U}_1\|\|\tilde{W}V - VW\| + \|\tilde{U}_1 V - VU_1\|\|W\| \\ &= \|\tilde{W}V - VW\| + \|\tilde{U}_1 V - VU_1\| \\ &\leq \sum_{k=2}^L \delta_k + \delta_1 = \sum_{k=1}^L \delta_k\end{aligned}$$

(we use the triangle inequality on lines 2 – 3). This completes the induction and proves the claim.  $\square$

**AP 1.** After  $k$  iterations of  $\mathcal{G}$  in Grover’s algorithm, we obtained

$$\mathcal{G}^k |\Psi(1, 1)\rangle = \left| \Psi \left( a^{-1/2} \sin((2k+1)\theta), b^{-1/2} \cos((2k+1)\theta) \right) \right\rangle$$

where  $\theta$  is such that  $\sin(\theta) = \sqrt{a}$ . Show that when  $k = \lfloor \pi/(4\theta) \rfloor$ , upon measuring this state the probability of observing a state in  $|A\rangle$  is  $\geq 1 - a$ .

**Solution:** Given a state vectors  $|\alpha\rangle$  and  $|\beta\rangle$ , the probability of observing  $|\alpha\rangle$  to be among the vectors in  $|\beta\rangle$  is  $|\langle\alpha|\beta\rangle|^2$ . The vector  $|A\rangle$  is not a state vector since it doesn't have norm 1, but we can normalize it. Therefore, the probability of measuring a state  $|\alpha\rangle$  and observing it to be among the vectors in  $|A\rangle$  is

$$\left| \frac{1}{\| |A\rangle \|} \langle A | \alpha \rangle \right|^2.$$

For the state vector given in the problem, this is

$$\begin{aligned} & \frac{1}{a} \left| \left\langle A \mid \Psi \left( a^{-1/2} \sin((2k+1)\theta), b^{-1/2} \cos((2k+1)\theta) \right) \right\rangle \right|^2 \\ &= \frac{1}{a} \left| \left\langle A \mid a^{-1/2} \sin((2k+1)\theta) |A\rangle + b^{-1/2} \cos((2k+1)\theta) |B\rangle \right\rangle \right|^2 \\ &= \frac{1}{a} \left| a^{-1/2} \sin((2k+1)\theta) \langle A | A \rangle \right|^2 = \frac{1}{a} \left| a^{-1/2} \sin((2k+1)\theta) a \right|^2 \\ &= \sin^2((2k+1)\theta). \end{aligned}$$

Thus the probability of observing a correct answer is  $\sin^2((2k+1)\theta)$ .

Let  $k = \lfloor \pi/(4\theta) \rfloor$  and recall that  $\theta$  is such that  $\sin(\theta) = \sqrt{a}$ . As a function in  $x$ ,  $\sin((2x+1)\theta)$  is decreasing near  $x = \pi/(4\theta)$ . Using this (at the marked inequality), we have

$$\sin^2((2k+1)\theta) \geq \sin^2((\pi/(2\theta) + 1)\theta) = \sin^2(\pi/2 + \theta) = \cos^2(\theta) = 1 - a.$$

Hence the probability of measuring a correct answer is  $\sin^2((2k+1)\theta) \geq 1 - a$ .

**AP 3.** Recall that the Fibonacci sequence  $(f_i)_{i \in \mathbb{N}}$  is defined

$$f_0 = 0, \quad f_1 = 1, \quad f_{n+1} = f_n + f_{n-1}.$$

(i) Show that

$$\begin{pmatrix} f_{n+1} \\ f_n \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} f_n \\ f_{n-1} \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^n \begin{pmatrix} 1 \\ 0 \end{pmatrix}.$$

(ii) Use the same technique that we used to find a closed form of the recurrence in Grover's algorithm to find a closed form for the Fibonacci sequence.

*Hint:*  $f_n = (1/\sqrt{5})(\varphi^n - \psi^n)$  where  $\varphi = (1/2)(1 + \sqrt{5})$  is the golden ratio and  $\psi = (1/2)(1 - \sqrt{5})$  is its conjugate.

**Solution:** Let  $F$  be the matrix mentioned in the problem. We begin by diagonalizing  $F$ . The eigenvalues are solutions to

$$\begin{aligned} \det(F - xI) &= \det \left( \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} - \begin{pmatrix} x & 0 \\ 0 & x \end{pmatrix} \right) = \det \begin{pmatrix} 1-x & 1 \\ 1 & -x \end{pmatrix} \\ &= (1-x)(-x) - 1 = x^2 - x - 1, \end{aligned}$$

so

$$\varphi = \frac{1 + \sqrt{5}}{2} \quad \text{and} \quad \psi = \frac{1 - \sqrt{5}}{2}$$

are the eigenvalues. Note that from  $x^2 - x - 1 = 0$ , we obtain  $x^2 = x + 1$  and  $x(x-1) = 1$ . Hence  $\varphi^2 = \varphi + 1$ ,  $\varphi^{-1} = \varphi - 1$ , and  $(\varphi - 1)^{-1} = \varphi$ . Similar identities hold for  $\psi$ . We also have  $\psi = 1 - \varphi$ .

Next, we find the eigenvectors by computing the nullspace of  $F - xI$  for  $x = \varphi$  and  $x = \psi$ . For  $x = \varphi$ , we row reduce  $F - \varphi I$ ,

$$\begin{pmatrix} 1-\varphi & 1 \\ 1 & -\varphi \end{pmatrix} \sim \begin{pmatrix} 1 & -\varphi \\ 0 & 0 \end{pmatrix},$$

yielding the eigenvector  $(\varphi, 1)^T$ . Similarly, for  $x = \psi$  we row reduce  $F - \psi I$  to obtain the eigenvector  $(\psi, 1)^T$ . It follows that

$$F = \begin{pmatrix} \varphi & \psi \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \varphi & 0 \\ 0 & \psi \end{pmatrix} \begin{pmatrix} \varphi & \psi \\ 1 & 1 \end{pmatrix}^{-1} = \frac{1}{\sqrt{5}} \begin{pmatrix} \varphi & \psi \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \varphi & 0 \\ 0 & \psi \end{pmatrix} \begin{pmatrix} 1 & \psi \\ -1 & \varphi \end{pmatrix}.$$

Finally, we have

$$\begin{aligned} \begin{pmatrix} f_{n+1} \\ f_n \end{pmatrix} &= \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^n \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \frac{1}{\sqrt{5}} \begin{pmatrix} \varphi & \psi \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \varphi^n & 0 \\ 0 & \psi^n \end{pmatrix} \begin{pmatrix} 1 & \psi \\ -1 & \varphi \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \\ &= \frac{1}{\sqrt{5}} \begin{pmatrix} \varphi & \psi \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \varphi^n & 0 \\ 0 & \psi^n \end{pmatrix} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = \frac{1}{\sqrt{5}} \begin{pmatrix} \varphi & \psi \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \varphi^n \\ -\psi^n \end{pmatrix} \\ &= \frac{1}{\sqrt{5}} \begin{pmatrix} \varphi^{n+1} - \psi^{n+1} \\ \varphi^n - \psi^n \end{pmatrix}. \end{aligned}$$

Therefore  $f_n = (1/\sqrt{5})(\varphi^n - \psi^n)$ .