# Quantum Algorithms Homework 7 Solutions

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## 1 Book Problems

1. Exercise 8.7

Prove Properties 8.10-8.13 of the operator norm.

### Solution:

(i)  $||XY|| \le ||X|| ||Y||$ 

*Proof.* We know that ||X|| is given by

$$||X|| = \sup_{|\xi\rangle \neq 0} \frac{||X|\xi\rangle||}{||\xi\rangle||}$$

So it follows that ||X|| ||Y|| can be given by

$$||X|| \ ||Y|| = \sup_{|\xi\rangle \neq 0} \frac{||X|\xi\rangle||}{||\ |\xi\rangle||} \ \sup_{|\xi\rangle \neq 0} \frac{||Y|\xi\rangle||}{||\ |\xi\rangle||}$$

but ||XY|| is given by

$$||XY|| = \sup_{|\xi\rangle \neq 0} \frac{||XY|\xi\rangle||}{||\,|\xi\rangle\,||}$$

Take  $|\xi\rangle$  to be the vector that maximizes the norm, i.e.  $||\xi\rangle||=1$ . The following inequality can be established:

$$\begin{split} ||XY|| &\leq ||XY|\xi\rangle|| \leq ||X(Y|\xi\rangle)|| \\ &\leq \frac{||X(Y|\xi\rangle)||}{||Y|\xi\rangle||} \leq ||X|| \\ &\leq ||X|| \, ||Y|\xi\rangle|| \end{split}$$

At each step we utilized a constant factor for  $\xi$ , whose norm will be the same in each computation. So, we can establish the following chain of inequalities:

$$||XY|| \le ||XY|| ||\xi|| ||X|| ||Y|| ||XY|| |$$

(ii) 
$$||X^{\dagger}|| = ||X||$$

*Proof.* It is known that the largest eigenvalue of  $X^{\dagger}X = ||X||^2$  (page 72 in the text). Define  $M = X^{\dagger}$ . It then follows that

$$\begin{split} ||M||^2 &= \max\{\lambda \mid \! \lambda \text{ is an eigenvalue of } M^\dagger M\} \\ \Longrightarrow ||X^\dagger||^2 &= \max\{\lambda \mid \! \lambda \text{ is an eigenvalue of } X^{\dagger^\dagger} X^\dagger\} \\ &= \max\{\lambda \mid \! \lambda \text{ is an eigenvalue of } XX^\dagger\} \end{split}$$

so it holds that  $||X||^2 = ||X^{\dagger}||^2$  if  $X^{\dagger}X = XX^{\dagger}$ , or that  $X^{\dagger}X, XX^{\dagger}$  share the same nonzero set of eigenvalues (since we picked  $\lambda$  to be the largest of the set of these eigenvalues)

Book problem 6.4 asserts that the eigenvalues of  $X^{\dagger}X$  are the same as the eigenvalues of  $XX^{\dagger}$ , so the two sets must have the same maximum. The assertion of problem 6.4 has a proof that can be seen on page 204 of the text. Since this proves the assertion above, it proves item (ii).

#### (iii) $||X \otimes Y|| = ||X|| ||Y||$

*Proof.* First, define operator  $\Gamma$  to be  $X \otimes Y$ . Note that  $X: A \to V, Y: B \to W$ . Take K to be in  $A \otimes B$ . Then:

$$||X \otimes Y|| = \sup_{|\xi\rangle \neq 0} \frac{||\Gamma|K\rangle||}{||K||}$$

$$= \frac{\sqrt{\langle \xi \mid \xi \rangle}}{\sqrt{\langle K \mid K \rangle}} \quad \text{where } \xi \text{ is the result of } \Gamma |K\rangle \text{ that is sup}$$

$$= \frac{\sqrt{\langle \sum_{j,k} \lambda^* \mu^* \langle e_j \otimes f_k | \mid \sum_{l,m} \lambda \mu \mid e_l \otimes f_m \rangle \rangle}}{\sqrt{\langle K \mid K \rangle}} \quad e_{j,l} \in \mathcal{B}_V, f_{k,m} \in \mathcal{B}_W$$

$$= \frac{\sqrt{\langle X\alpha \mid X\alpha \rangle \langle Y\beta \mid Y\beta \rangle}}{\sqrt{\langle \alpha \mid \alpha \rangle \langle \beta \mid \beta \rangle}} \quad \alpha \in A, \beta \in B$$

$$= \frac{\sqrt{\langle X\alpha \mid X\alpha \rangle \sqrt{\langle Y\beta \mid Y\beta \rangle}}}{\sqrt{\langle \alpha \mid \alpha \rangle \sqrt{\langle \beta \mid \beta \rangle}}}$$

$$= \frac{||X \mid \alpha \rangle || ||Y \mid b \rangle ||}{||\alpha || ||\beta ||}$$

$$= ||X|| ||Y||$$

So, this proves item (iii).

(iv) 
$$||U|| = 1$$

(over)

*Proof.* First, introduce vector  $|\xi\rangle$  as the vector that maximizes the norm:

$$||U|| = \frac{||U|\xi\rangle||}{|\xi\rangle}$$

$$= \frac{\sqrt{\langle U\xi \mid U\xi\rangle}}{\sqrt{\langle \xi \mid \xi\rangle}}$$

$$= \frac{\sqrt{\langle \xi \mid U^{\dagger}U \mid \xi\rangle}}{\sqrt{\langle \xi \mid \xi\rangle}}$$

$$= \frac{\sqrt{\langle \xi \mid \xi\rangle}}{\sqrt{\langle \xi \mid \xi\rangle}}$$

$$= 1$$

Note that the property  $U^{\dagger}U = I$  is utilized in this proof, this fact has been previously utilized and proven (specifically Quiz 5). This proves item (iv).

#### 2. Exercise 8.8

Prove the two basic properties of approximation i with ancillas:

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

#### Solution:

(i) *Proof.* We can use the fact that

$$||\tilde{U}V - UV|| \le \delta$$

In order to show that this can hold for  $\tilde{U}^{-1}$ , the above expression must be recreated using U and  $\tilde{U}^{-1}$ :

$$||U^{-1}V - V\tilde{U}^{-1}|| \le \delta$$

so multiply the original expression on either side by the norms of the two inverse unitary matrices.

$$\begin{split} ||\tilde{U}V - UV|| &= ||\tilde{U}^{-1}|| \; ||\tilde{U}V - VU|| \; ||U^{-1}|| \\ &= ||\tilde{U}^{-1}V\tilde{U}U^{-1} - \tilde{U}^{-1}UU^{-1}|| \\ &= ||IU^{-1}V - V\tilde{U}^{-1}I|| \\ &= ||U^{-1}V - V\tilde{U}^{-1}|| \leq \delta \end{split}$$

(ii) *Proof.* To show that the errors propagate linearly, we can consider a concrete example of the propagation since it is a fact that the product of unitary matrices is also unitary. Take L=2:

$$\begin{split} ||\tilde{U}_{2}\tilde{U}_{1} - U_{2}U_{1}|| &= ||\tilde{U}_{2}(\tilde{U}_{1}V - VU_{1}) + (\tilde{U}_{2}V - VU_{2})U_{1}|| \\ &\leq ||\tilde{U}_{2}(\tilde{U}_{1}V - VU_{1})|| + ||(\tilde{U}_{2}V - VU_{2})U_{1}|| \\ &\leq ||\tilde{U}_{2}|| ||(\tilde{U}_{1}V - U_{1})|| + ||(\tilde{U}_{2}V - VU_{2})|| ||U_{1}|| \\ &= ||(\tilde{U}_{1}V - VU_{1})|| + ||(\tilde{U}_{2}V - VU_{2})|| \leq \delta_{1} + \delta_{2} \end{split}$$