

# 量子信息与量子计算习题解答

quantum computation and quantum information 是量子计算领域的圣书，但书中的习题没有官方答案，互联网上的答案也是众说纷纭，缺乏自我纠错的特性。现将该书中的基础部分（第 1-2 章）习题解答如下。

## 1 Chapter 1

### 1.1

经典最好情形 **2** 次。当且仅当测量两次结果相同时无法确定平衡函数和常函数，其容错概率

$$p = \frac{\binom{2^{n-1}}{2} \cdot 2}{\binom{2^n}{2}} = \frac{2^{n-1} - 1}{2^n - 1} < \frac{1}{2}$$

而量子算法仅需 **1** 次完全确定结果（ $\epsilon = 0$ ），说明量子计算在该问题上的优越性。

### 1.2

若状态可区分，发送的状态可以确定，设计合适的哈密顿设计器，在相同的状态下建立第二个系统；相反，准备量子态的多个副本，计算可观测值的平均值加以区分量子态。

## 2 Chapter 2

### 2.1

$$\begin{bmatrix} 1 \\ -1 \end{bmatrix} + \begin{bmatrix} 1 \\ 2 \end{bmatrix} - \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

### 2.2

$$A = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, \text{input} : \{|0\rangle, |1\rangle\}, \text{output} : \{|1\rangle, |0\rangle\}, A = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

### 2.3

$$\begin{aligned} BA|v_i\rangle &= B\left(\sum_j A_{ji}|w_j\rangle\right) = \sum_j A_{ji}B|w_j\rangle = \sum_{j,k} A_{ji}B_{kj}|x_k\rangle \\ &= \sum_k \left(\sum_j B_{kj}A_{ji}\right)|x_k\rangle = \sum_k (BA)_{ki}|x_k\rangle \end{aligned}$$

### 2.4

$$I_{ij} = \delta_{ij}$$

## 2.5

$$\begin{aligned} (1) \left( (y_1, \dots, y_n), \sum_i \lambda_i (z_{i1}, \dots, z_{in}) \right) &= \sum_{i,j} y_i^* \lambda_j z_{ji} = \sum_j \lambda_j \left( \sum_i y_i^* z_{ji} \right) \\ &= \sum_j \lambda_j ((y_1, \dots, y_n), (z_{j1}, \dots, z_{jn})) = \sum_i \lambda_i ((y_1, \dots, y_n), (z_{i1}, \dots, z_{in})) \end{aligned}$$

$$(2) ((y_1, \dots, y_n), (z_1, \dots, z_n))^* = \left( \sum_i z_i^* y_i \right) = ((z_1, \dots, z_n), (y_1, \dots, y_n))$$

$$\begin{aligned} (3) ((y_1, \dots, y_n), (y_1, \dots, y_n)) &= \sum_i y_i^* y_i = \sum_i |y_i|^2 \geq 0 \\ ((y_1, \dots, y_n), (y_1, \dots, y_n)) &= 0 \iff (y_1, \dots, y_n) = 0 \end{aligned}$$

## 2.6

$$\left( \sum_i \lambda_i |w_i\rangle, |v\rangle \right) = \left( |v\rangle, \sum_i \lambda_i |w_i\rangle \right)^* = \left( \sum_i \lambda_i (|v\rangle, |w_i\rangle) \right)^* = \sum_i \lambda_i^* (|w_i\rangle, |v\rangle)$$

## 2.7

$$\langle w | v \rangle = [1 \quad 1] \begin{bmatrix} 1 \\ -1 \end{bmatrix} = 0, \frac{|w\rangle}{\| |w\rangle \|} = \frac{|w\rangle}{\sqrt{\langle w | w \rangle}} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}, \frac{|v\rangle}{\| |v\rangle \|} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \end{bmatrix}$$

## 2.8

数学归纳法,  $k=1$  时  $\langle v_1 | v_2 \rangle = \langle v_1 | \left( \frac{|w_2\rangle - \langle v_1 | w_2 \rangle |v_1\rangle}{\| |w_2\rangle - \langle v_1 | w_2 \rangle |v_1\rangle \|} \right) = 0$

$n=k$  成立, 对于  $n=k+1$  有,  $\forall 1 \leq j \leq n$

$$\begin{aligned} \langle v_j | v_{n+1} \rangle &= \langle v_j | \left( \frac{|w_{n+1}\rangle - \sum_{i=1}^n \langle v_i | w_{n+1} \rangle |v_i\rangle}{\| |w_{n+1}\rangle - \sum_{i=1}^n \langle v_i | w_{n+1} \rangle |v_i\rangle \|} \right) \\ &= \frac{\langle v_j | w_{n+1} \rangle - \sum_{i=1}^n \langle v_i | w_{n+1} \rangle \langle v_j | v_i \rangle}{\| |w_{n+1}\rangle - \sum_{i=1}^n \langle v_i | w_{n+1} \rangle |v_i\rangle \|} = 0 \end{aligned}$$

## 2.9

根据完备性关系,  $A = I_W A I_V = \sum_{i,j} |\omega_j\rangle \langle \omega_j | A | v_i \rangle \langle v_i | = \sum_{i,j} \langle \omega_j | A | v_i \rangle |\omega_j\rangle \langle v_i |$

$$X = |0\rangle\langle 1| + |1\rangle\langle 0|, Y = -i|0\rangle\langle 1| + i|1\rangle\langle 0|, Z = |0\rangle\langle 0| - |1\rangle\langle 1|$$

## 2.10

$$(|v_j\rangle\langle v_k|)_{pq} = \delta_{pj}\delta_{kq}$$

## 2.11

$$\lambda_{X,Y,Z} = \pm 1, \begin{cases} X: |\lambda = 1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ 1 \end{bmatrix}, |\lambda = -1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} -1 \\ 1 \end{bmatrix} \\ Y: |\lambda = 1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ i \end{bmatrix}, |\lambda = -1\rangle = \frac{1}{\sqrt{2}} \begin{bmatrix} i \\ 1 \end{bmatrix} \\ Z: |\lambda = 1\rangle = \begin{bmatrix} 1 \\ 0 \end{bmatrix}, |\lambda = -1\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \end{cases}$$

$$\text{对角表示为} \begin{cases} X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \end{pmatrix} - \frac{1}{2} \begin{pmatrix} 1 \\ -1 \end{pmatrix} \begin{pmatrix} 1 & -1 \end{pmatrix} \\ Y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 \\ i \end{pmatrix} \begin{pmatrix} 1 & -i \end{pmatrix} - \frac{1}{2} \begin{pmatrix} i \\ 1 \end{pmatrix} \begin{pmatrix} -i & 1 \end{pmatrix} \\ Z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \end{pmatrix} - \begin{pmatrix} 0 \\ 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \end{pmatrix} \end{cases}$$

## 2.12

$$\lambda = 1, |\lambda = 1\rangle = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, A \neq a \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

## 2.13

伴随算子的抽象定义:  $\forall |v\rangle, |\omega\rangle, (|v\rangle, A|\omega\rangle) = (A^\dagger|v\rangle, |\omega\rangle)$

$$\begin{aligned} \forall |x\rangle, |y\rangle, \langle x|(|\omega\rangle\langle v|)^\dagger|y\rangle &= (|x\rangle, (|\omega\rangle\langle v|)^\dagger|y\rangle) = ((|\omega\rangle\langle v|)^\dagger|y\rangle, |x\rangle)^* \\ &= (|y\rangle, |\omega\rangle\langle v|x\rangle)^* = (\langle y|\omega\rangle\langle v|x\rangle)^* = \langle v|x\rangle^*\langle y|\omega\rangle^* = \langle x|v\rangle\langle\omega|y \end{aligned}$$

$$\langle x|(|\omega\rangle\langle v|)^\dagger|y\rangle = \langle x|v\rangle\langle\omega|y\rangle \implies (|w\rangle\langle v|)^\dagger = |v\rangle\langle w|$$

## 2.14

$$(a_i A_i)^\dagger = (a_i I A_i)^\dagger = A_i^\dagger (a_i I)^\dagger = a_i^* A_i^\dagger$$

## 2.15

$$\left((A^\dagger)^\dagger|\psi\rangle, |\phi\rangle\right) = (|\psi\rangle, A^\dagger|\phi\rangle) = (A^\dagger|\phi\rangle, |\psi\rangle)^* = (|\phi\rangle, A|\psi\rangle)^* = (A|\psi\rangle, |\phi\rangle)$$

## 2.16

$$P = \sum_{i=1}^k |i\rangle\langle i|, P^2 = \left( \sum_{i=1}^k |i\rangle\langle i| \right)^2 = \sum_{i=1}^k |i\rangle\langle i|i\rangle\langle i| + 2 \sum_{i<j} |i\rangle\langle i|j\rangle\langle j| = \sum_{i=1}^k |i\rangle\langle i| = P$$

## 2.17

$$\begin{aligned} \implies: A|\lambda\rangle &= \lambda|\lambda\rangle, \langle\lambda|A|\lambda\rangle = \lambda\langle\lambda|\lambda\rangle = \langle\lambda|A^\dagger|\lambda\rangle = (A|\lambda\rangle)^\dagger|\lambda\rangle = \lambda^*\langle\lambda|\lambda\rangle, \lambda = \lambda^* \\ \Longleftarrow: AA^\dagger &= A^\dagger A, A = \sum_i \lambda_i |\lambda_i\rangle\langle\lambda_i|, A^\dagger = \sum_i \lambda_i^* |\lambda_i\rangle\langle\lambda_i| = A^\dagger \end{aligned}$$

## 2.18

设  $\lambda$  为  $U$  特征值,  $|\lambda\rangle$  为  $U$  在  $\lambda$  本征空间的特征向量,  $U|\lambda\rangle = \lambda|\lambda\rangle, \langle\lambda|U^\dagger = \langle\lambda|\lambda^*$   
相乘得  $\langle\lambda|U^\dagger U|\lambda\rangle = \lambda\lambda^*\langle\lambda|\lambda\rangle = \|\lambda\|^2\langle\lambda|\lambda\rangle = \langle\lambda|\lambda\rangle \implies \|\lambda\| = 1$

## 2.19

$$X = X^\dagger, Y = Y^\dagger, Z = Z^\dagger, XX^\dagger = X^2 = I, YY^\dagger = Y^2 = I, ZZ^\dagger = Z^2 = I,$$

## 2.20

$$\langle v_i | A | v_j \rangle = \sum_{kl} \langle v_i | w_k \rangle \langle w_k | A | w_l \rangle \langle w_l | v_j \rangle$$

## 2.21

厄米算子  $A$  的对角化: 与正规算子相似, 其中  $QMQ$  的正规性易得

$$A = IAI = (P + Q)A(P + Q) = PAP + QAP + PAQ + QAQ = \lambda P + QAQ$$

$$QAQ(QAQ)^\dagger = QAQQA^\dagger Q = QA^\dagger QQAQ = (QAQ)^\dagger QAQ$$

将  $A$  的对角化化解为  $P$  和  $Q$  的, 从而由归纳法, 任何厄米算子在标准正交基下可对角化

## 2.22

$$\begin{aligned} A|\lambda_1\rangle &= \lambda_1|\lambda_1\rangle, A|\lambda_2\rangle = \lambda_2|\lambda_2\rangle, \langle\lambda_2|A^\dagger = \langle\lambda_2|\lambda_2^* \implies \langle\lambda_2|A = \langle\lambda_2|\lambda_2 \\ \langle\lambda_2|A|\lambda_1\rangle &= \lambda_2\langle\lambda_2|\lambda_1\rangle = \lambda_1\langle\lambda_2|\lambda_1\rangle, \lambda_1 \neq \lambda_2 \implies \langle\lambda_2|\lambda_1\rangle = 0 \end{aligned}$$

## 2.23

$$P|\lambda\rangle = \lambda|\lambda\rangle = P^2|\lambda\rangle = P\lambda|\lambda\rangle = \lambda^2|\lambda\rangle, |\lambda\rangle \neq 0, \lambda^2 = \lambda, \lambda = 0, 1$$

## 2.24

仿造实矩阵可分解为实对称矩阵和实反对称矩阵, 将  $A$  拆解为

$$A = \frac{A + A^\dagger}{2} + i \frac{A - A^\dagger}{2i} = B + iC$$

其中  $B$  和  $C$  都为厄米矩阵, 则  $\langle v|A|v\rangle = \langle v|B|v\rangle + i\langle v|C|v\rangle \geq 0$

而由厄米矩阵可以谱分解为  $\sum_i \lambda_i |i\rangle\langle i|$ , 其中  $|i\rangle$  为标准正交基, 且  $\lambda_i$  为实数

故  $i\langle v|C|v\rangle$  为虚数, 从而  $\langle v|C|v\rangle = 0 \implies C = O$ , 故  $A = B = \frac{A + A^\dagger}{2}$  为厄米算子

引理:  $\forall |v\rangle, \langle v|C|v\rangle = 0 \iff C = O$ .

$\forall |u\rangle, |v\rangle$ , 可以构造出下列等式使得  $\langle u|C|v\rangle = 0$ :

$$\begin{aligned} \langle u|C|v\rangle = \frac{1}{4} \left[ \langle u+v|C|u+v\rangle - \langle u-v|C|u-v\rangle + \frac{1}{i} \langle u+iv|C|u+iv\rangle \right. \\ \left. - \frac{1}{i} \langle u-iv|C|u-iv\rangle \right] \end{aligned}$$

$$\text{上式右侧化简 } \frac{2\langle v|C|u\rangle + 2\langle u|C|v\rangle}{4} + \frac{2\langle iv|C|u\rangle + 2\langle u|C|iv\rangle}{4i}$$

由复空间内积定义

$$\langle iv|C|u\rangle = \bar{i} \langle v|C|u\rangle = -i \langle v|C|u\rangle, \langle u|C|iv\rangle = i \langle u|C|v\rangle$$

$$RHS = \frac{\langle v|C|u\rangle + \langle u|C|v\rangle}{2} + \frac{-i\langle v|C|u\rangle + i\langle u|C|v\rangle}{2i} = \langle u|C|v\rangle$$

故  $\forall |v\rangle, \langle v|C|v\rangle = 0 \iff \forall |u\rangle, |v\rangle, \langle u|C|v\rangle = 0$ , 而取  $|u\rangle = C|v\rangle$ , 由内积正定性

$$\langle C|v\rangle, C|v\rangle = 0 \implies C|v\rangle = 0, \forall |v\rangle \implies C = O$$

## 2.25

$$\forall |v\rangle, \langle v|A^\dagger A|v\rangle = \langle A|v\rangle, A|v\rangle \geq 0$$

## 2.26

$$\begin{aligned} |\psi\rangle^{\otimes 2} &= \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) \otimes \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) = \frac{1}{2}(|00\rangle + |01\rangle + |10\rangle + |11\rangle) \\ |\psi\rangle^{\otimes 3} &= \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) \otimes \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) \otimes \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle) \\ &= \frac{\sqrt{2}}{4}(|000\rangle + |001\rangle + |010\rangle + |011\rangle + |100\rangle + |101\rangle + |110\rangle + |111\rangle) \end{aligned}$$

## 2.27

$$\begin{aligned}
X \otimes Z &= \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \otimes \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 \\ 0 & -1 & 0 & 0 \end{bmatrix} \\
I \otimes X &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \otimes \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \end{bmatrix} \\
X \otimes I &= \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \otimes \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \end{bmatrix}
\end{aligned}$$

一般情形下，张量积不可交换。

## 2.28

$$\begin{aligned}
(A \otimes B)^* &= \begin{bmatrix} A_{11}B & \cdots & A_{1n}B \\ \vdots & \ddots & \vdots \\ A_{m1}B & \cdots & A_{mn}B \end{bmatrix}^* = \begin{bmatrix} A_{11}^*B^* & \cdots & A_{1n}^*B^* \\ \vdots & \ddots & \vdots \\ A_{m1}^*B^* & \cdots & A_{mn}^*B^* \end{bmatrix} = A^* \otimes B^* \\
(A \otimes B)^T &= \begin{bmatrix} A_{11}B & \cdots & A_{1n}B \\ \vdots & \ddots & \vdots \\ A_{m1}B & \cdots & A_{mn}B \end{bmatrix}^T = \begin{bmatrix} A_{11}B^T & \cdots & A_{m1}B^T \\ \vdots & \ddots & \vdots \\ A_{1n}B^T & \cdots & A_{mn}B^T \end{bmatrix} = A^T \otimes B^T \\
(A \otimes B)^\dagger &= ((A \otimes B)^*)^T = (A^* \otimes B^*)^T = (A^*)^T \otimes (B^*)^T = A^\dagger \otimes B^\dagger
\end{aligned}$$

## 2.29

$$(U_1 \otimes U_2)(U_1 \otimes U_2)^\dagger = U_1 U_1^\dagger \otimes U_2 U_2^\dagger = I \otimes I = I$$

## 2.30

$$(A \otimes B)^\dagger = A^\dagger \otimes B^\dagger = A \otimes B$$

## 2.31

$$(\langle u| \otimes \langle v|)(A \otimes B)|u\rangle \otimes |v\rangle = \langle u|A|u\rangle \langle v|B|v\rangle \geq 0$$

## 2.32

$$(P_1 \otimes P_2)^2 = P_1^2 \otimes P_2^2 = P_1 \otimes P_2$$

## 2.33

$$\begin{aligned}
H &= \frac{1}{\sqrt{2}} [|0\rangle\langle 0| + |1\rangle\langle 0| + |0\rangle\langle 1| - |1\rangle\langle 1|] = \frac{1}{\sqrt{2}} \sum_{x,y} (-1)^{x \cdot y} |x\rangle\langle y| \\
H^{\otimes n} &= \frac{1}{\sqrt{2^n}} \sum_{x_1, y_1} (-1)^{x_1 \cdot y_1} |x_1\rangle\langle y_1| \otimes \cdots \otimes \sum_{x_n, y_n} (-1)^{x_n \cdot y_n} |x_n\rangle\langle y_n| \\
&= \frac{1}{\sqrt{2^n}} \sum_{x, y} (-1)^{x \cdot y} |x\rangle\langle y| \\
H^{\otimes 2} &= \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \otimes \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 & 1 & 1 & 1 \\ 1 & -1 & 1 & -1 \\ 1 & 1 & -1 & -1 \\ 1 & -1 & -1 & 1 \end{bmatrix}
\end{aligned}$$

### 2.34

正规算子  $A$  (正交) 谱分解  $A = \sum_i a |a\rangle\langle a|$ , 定义算子函数  $f(A) = \sum_i f(a) |a\rangle\langle a|$

将  $A = \begin{bmatrix} 4 & 3 \\ 3 & 4 \end{bmatrix}$  谱分解  $A = 1 \cdot \frac{1}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{bmatrix} 1 & -1 \end{bmatrix} + 7 \cdot \frac{1}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix}$ , 从而

$$\begin{aligned}
\sqrt{A} &= 1 \cdot \frac{1}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{bmatrix} 1 & -1 \end{bmatrix} + \sqrt{7} \cdot \frac{1}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 1 + \sqrt{7} & -1 + \sqrt{7} \\ -1 + \sqrt{7} & 1 + \sqrt{7} \end{bmatrix} \\
\log(A) &= 0 \cdot \frac{1}{2} \begin{bmatrix} 1 \\ -1 \end{bmatrix} \begin{bmatrix} 1 & -1 \end{bmatrix} + \log 7 \cdot \frac{1}{2} \begin{bmatrix} 1 \\ 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix} = \frac{\log(7)}{2} \begin{bmatrix} 1 & 1 \\ 1 & 1 \end{bmatrix}
\end{aligned}$$

### 2.35

$$\vec{v} \cdot \vec{\sigma} = \sum_{i=1}^3 v_i \sigma_i = v_1 \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} + v_2 \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} + v_3 \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} = \begin{bmatrix} v_3 & v_1 - iv_2 \\ v_1 + iv_2 & -v_3 \end{bmatrix}$$

计算特征值  $\det(\vec{v} \cdot \vec{\sigma} - \lambda I) = (v_3 - \lambda)(-v_3 - \lambda) - (v_1 - iv_2)(v_1 + iv_2) = \lambda^2 - 1$

由其厄米矩阵, 则可以 (正交) 谱分解  $\vec{v} \cdot \vec{\sigma} = |\lambda_1\rangle\langle\lambda_1| - |\lambda_{-1}\rangle\langle\lambda_{-1}|$

其中  $|\lambda_1\rangle\langle\lambda_1| + |\lambda_{-1}\rangle\langle\lambda_{-1}| = I$ , 而代入指数函数分别作用在  $\pm i\theta$  上, 展开得到

$$\begin{aligned}
e^{i\theta \vec{v} \cdot \vec{\sigma}} &= e^{i\theta} |\lambda_1\rangle\langle\lambda_1| + e^{-i\theta} |\lambda_{-1}\rangle\langle\lambda_{-1}| \\
&= (\cos \theta + i \sin \theta) |\lambda_1\rangle\langle\lambda_1| + (\cos \theta - i \sin \theta) |\lambda_{-1}\rangle\langle\lambda_{-1}| \\
&= \cos \theta (|\lambda_1\rangle\langle\lambda_1| + |\lambda_{-1}\rangle\langle\lambda_{-1}|) + i \sin \theta (|\lambda_1\rangle\langle\lambda_1| - |\lambda_{-1}\rangle\langle\lambda_{-1}|) \\
&= \cos(\theta) I + i \sin(\theta) \vec{v} \cdot \vec{\sigma}
\end{aligned}$$

### 2.36

$$\text{Tr}(\sigma_1) = 0 + 0 = \text{Tr}(\sigma_2) = 0 + 0 = \text{Tr}(\sigma_3) = 1 + (-1) = 0$$

### 2.37

$$\begin{aligned}\text{Tr}(AB) &= \sum_i \langle i|AB|i\rangle = \sum_{i,j} \langle i|A|j\rangle \langle j|B|i\rangle = \sum_{i,j} \langle j|B|i\rangle \langle i|A|j\rangle \\ &= \sum_j \langle j|BA|j\rangle = \text{Tr}(BA)\end{aligned}$$

## 2.38

$$\begin{aligned}\text{Tr}(A+B) &= \sum_i \langle i|A+B|i\rangle = \sum_i (\langle i|A|i\rangle + \langle i|B|i\rangle) = \text{Tr}(A) + \text{Tr}(B) \\ \text{Tr}(zA) &= \sum_i \langle i|zA|i\rangle = z \sum_i \langle i|A|i\rangle = z \text{Tr}(A)\end{aligned}$$

## 2.39

(1) 分别验证对第二个参数线性、交换共轭、自身正定性：

$$\begin{aligned}(i) \left( A, \sum_i \lambda_i B_i \right) &= \text{Tr} \left[ A^\dagger \left( \sum_i \lambda_i B_i \right) \right] = \sum_j \langle j|A^\dagger \left( \sum_i \lambda_i B_i \right)|j\rangle \\ &= \sum_i \lambda_i \sum_j \langle j|A^\dagger B_i|j\rangle = \sum_i \lambda_i \text{Tr}(A^\dagger B_i)\end{aligned}$$

$$(ii) (A, B)^* = \left( \sum_j \langle j|A^\dagger B|j\rangle \right)^* = \left( \sum_j \langle j|A^\dagger B|j\rangle \right)^\dagger = \sum_j \langle j|B^\dagger A|j\rangle = (B, A)$$

$$(iii) (A, A) = \text{tr}(A^\dagger A) = \sum_j \langle j|A^\dagger A|j\rangle \geq 0, (A, A) = 0 \iff \forall |j\rangle, A|j\rangle = 0, A = O$$

(2)  $V \mapsto V$  中所有算子都可以用矩阵表示，维数为  $d \times d$ ，即  $\dim(L_V) = d^2$

(3) 对所有算子  $A$  的标准正交基为  $A_{ij} = |v_i\rangle\langle v_j|$ ， $|v_i\rangle$  为标准正交基

由于对所有厄米算子  $B$  都可表为  $\frac{A+A^\dagger}{2}$ ，则构造正交基

$$A_{ij} = \frac{A_{ij} + A_{ij}^\dagger}{2} = \frac{|v_i\rangle\langle v_j| + |v_j\rangle\langle v_i|}{2}$$

对  $i=j$  时为实数， $i \neq j$  时可为虚数，两项相差负号，共  $d+2 \cdot \frac{d(d-1)}{2} = d^2$  项

## 2.40

$$[X, Y] = XY - YX = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} - \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 2i & 0 \\ 0 & -2i \end{bmatrix} = 2iZ$$

$$[Y, Z] = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} - \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} = \begin{bmatrix} 0 & 2i \\ 2i & 0 \end{bmatrix} = 2iX$$

$$[Z, X] = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} - \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} = 2i \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix} = 2iY$$



可简写为  $[\sigma_j, \sigma_k] = 2i \sum_{l=1}^3 \epsilon_{jkl} \sigma_l$

**2.41**

$$\{\sigma_1, \sigma_2\} = \{\sigma_2, \sigma_3\} = \{\sigma_3, \sigma_1\} = 0, \sigma_0^2 = \sigma_1^2 = \sigma_2^2 = \sigma_3^2 = I$$

**2.42**

$$\frac{[A, B] + \{A, B\}}{2} = \frac{AB - BA + AB + BA}{2} = AB$$

**2.43**

结合 2.40 – 2.41 的计算结果可得

$$\sigma_j \sigma_k = \frac{[\sigma_j, \sigma_k] + \{\sigma_j, \sigma_k\}}{2} = \frac{2i \sum_{l=1}^3 \epsilon_{jkl} \sigma_l + 2\delta_{jk} I}{2} = \delta_{jk} I + i \sum_{l=1}^3 \epsilon_{jkl} \sigma_l$$

**2.44**

$$[A, B] = 0, \{A, B\} = 0 \implies AB = 0, A^{-1}AB = 0 = B$$

**2.45**

$$[A, B]^\dagger = (AB - BA)^\dagger = B^\dagger A^\dagger - A^\dagger B^\dagger = [B^\dagger, A^\dagger]$$

**2.46**

$$[A, B] = AB - BA = -(BA - AB) = -[B, A]$$

**2.47**

$$(i(BA - AB))^\dagger = -iA^\dagger B^\dagger + iB^\dagger A^\dagger = i(BA - AB)$$

**2.48**

$$P = UDU^\dagger (\lambda_i \geq 0), U = UII, H = U_1 \sqrt{H^2}, \sqrt{H^2} = U_2 DU_2^\dagger, H = U_1 U_2 DU_2^\dagger$$

**2.49**

$$A = \sum_i \lambda_i |i\rangle \langle i|, \sqrt{A^\dagger A} = \sqrt{\sum_i |i\rangle \langle i| \lambda_i^* \sum_j \lambda_j |j\rangle \langle j|} = \sqrt{\sum_i \|\lambda_i\|^2 |i\rangle \langle i|} = \sum_i \|\lambda_i\| |i\rangle \langle i|$$

构造标准正交基  $|e_i\rangle$ ，对  $\lambda \neq 0$  的所有  $\lambda_i$  取为  $\frac{\lambda_i|i\rangle}{\|\lambda_i\|}$ ，满足正交性，对  $\lambda = 0$  扩充至全空间

$$\text{令 } U = \sum_i |e_i\rangle\langle i|, \text{ 故 } A = U\sqrt{A^\dagger A} = \sum_i \frac{\lambda_i}{\|\lambda_i\|} \|\lambda_i\| |e_i\rangle\langle i| = A$$

## 2.50

$$A = \begin{bmatrix} 1 & 0 \\ 1 & 1 \end{bmatrix}, A^\dagger A = \begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$$

而对  $\begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix}$  计算特征值为  $\lambda_{1,2} = \frac{3 \pm \sqrt{5}}{2} = \left( \frac{\sqrt{5} \pm 1}{2} \right)^2$ ，对应特征向量为

$$|\lambda = \frac{3 + \sqrt{5}}{2}\rangle = \frac{1}{\sqrt{10 - 2\sqrt{5}}} \begin{bmatrix} 2 \\ -1 + \sqrt{5} \end{bmatrix}, |\lambda = \frac{3 - \sqrt{5}}{2}\rangle = \frac{1}{\sqrt{10 + 2\sqrt{5}}} \begin{bmatrix} 2 \\ -1 - \sqrt{5} \end{bmatrix}$$

$$\begin{aligned} J &= \sqrt{A^\dagger A} = \sqrt{\lambda_+} |\lambda_+\rangle \langle \lambda_+| + \sqrt{\lambda_-} |\lambda_-\rangle \langle \lambda_-| \\ &= \frac{\sqrt{5} + 1}{2(10 - 2\sqrt{5})} \begin{bmatrix} 2 \\ -1 + \sqrt{5} \end{bmatrix} \begin{bmatrix} 2 & -1 + \sqrt{5} \end{bmatrix} + \frac{\sqrt{5} - 1}{2(10 + 2\sqrt{5})} \begin{bmatrix} 2 \\ -1 - \sqrt{5} \end{bmatrix} \begin{bmatrix} 2 & -1 - \sqrt{5} \end{bmatrix} \\ &= \frac{1}{\sqrt{5}} \begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix} \end{aligned}$$

从而  $J^{-1} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ -1 & 3 \end{bmatrix}$ ，计算  $U = AJ^{-1} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}$ ，则左极式分解为

$$A = U\sqrt{A^\dagger A} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix} \sqrt{A^\dagger A}, \sqrt{A^\dagger A} = \frac{1}{\sqrt{5}} \begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix}$$

而  $AA^\dagger = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$ ，同样经过繁杂计算后得到  $\sqrt{AA^\dagger} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}$ ，则右极式分解为

$$A = \sqrt{AA^\dagger}U = \sqrt{AA^\dagger} \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & -1 \\ 1 & 2 \end{bmatrix}, \sqrt{AA^\dagger} = \frac{1}{\sqrt{5}} \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}$$