

3.1 a) $\{\sigma_1, \sigma_2\} = \sigma_1 \sigma_2 + \sigma_2 \sigma_1 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} + \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} i & 0 \\ 0 & -i \end{pmatrix} + \begin{pmatrix} -i & 0 \\ 0 & i \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$

$$\{\sigma_2, \sigma_3\} = \sigma_2 \sigma_3 + \sigma_3 \sigma_2 = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} = \begin{pmatrix} 0 & i \\ i & 0 \end{pmatrix} + \begin{pmatrix} 0 & -i \\ -i & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

$$\{\sigma_3, \sigma_1\} = \sigma_3 \sigma_1 + \sigma_1 \sigma_3 = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} + \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ -1 & 0 \end{pmatrix} + \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

b) $[\sigma_1, \sigma_2] = \sigma_1 \sigma_2 - \sigma_2 \sigma_1 = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} - \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} = 2i \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} = 2i \sigma_3$

$$[\sigma_2, \sigma_3] = \sigma_2 \sigma_3 - \sigma_3 \sigma_2 = \begin{pmatrix} 0 & i \\ i & 0 \end{pmatrix} - \begin{pmatrix} 0 & -1 \\ -1 & 0 \end{pmatrix} = 2i \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = 2i \sigma_1$$

$$[\sigma_3, \sigma_1] = \sigma_3 \sigma_1 - \sigma_1 \sigma_3 = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} - \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = 2 \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} = 2i \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} = 2i \sigma_2$$

c) $e^{U^\dagger A U} = \sum_{k=0}^{\infty} \frac{(U^\dagger A U)^k}{k!} \stackrel{①}{=} \sum_{k=0}^{\infty} \frac{U^\dagger A^k U}{k!} = U^\dagger \left(\sum_{k=0}^{\infty} \frac{A^k}{k!} \right) U = U^\dagger e^A U$

① Proof by induction: $(U^\dagger A U)^k = U^\dagger A^k U$

Base case ($k=0$):

$$(U^\dagger A U)^0 = 1$$

$$U^\dagger A^0 U = U^\dagger I U = U^\dagger U = 1$$

Inductive step (proof $k+1$ when know the case for k):

$$(U^\dagger A U)^{k+1} = (U^\dagger A U)^k (U^\dagger A U) = U^\dagger A^k U \overset{\text{Unitary}}{U^\dagger U} A U = U^\dagger A^{k+1} U$$

d) $H \times H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 2 & 0 \\ 0 & 2 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = I$

$$H \times H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 0 & 2 \\ 2 & 0 \end{pmatrix} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} = X$$

e) $H R_X(\theta) H = H e^{-i\frac{\theta}{2} X} H \stackrel{①}{=} H^\dagger e^{-i\frac{\theta}{2} X} H \stackrel{②}{=} e^{H^\dagger (-i\frac{\theta}{2} X) H} = e^{-i\frac{\theta}{2} H^\dagger X H} \stackrel{③}{=} e^{-i\frac{\theta}{2} Z} = R_Z(\theta)$

① H is hermitian

② showed in part c (H is self-inverse)

③ showed in part d

3.2 a) $\bullet X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad |\lambda I - X| = \begin{vmatrix} \lambda & -1 \\ -1 & \lambda \end{vmatrix} = \lambda^2 - 1 \stackrel{!}{=} 0 \Rightarrow \lambda = \pm 1$

$$\lambda_1 = 1 \Rightarrow X v_1 = \lambda_1 v_1 = v_1 : \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow \begin{pmatrix} b \\ a \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix}$$

with the normalization constraint we have $1 = \sqrt{2a^2} \Rightarrow a = b = \frac{1}{\sqrt{2}} \Rightarrow v_1 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$

$$\lambda_2 = -1 \Rightarrow X v_2 = \lambda_2 v_2 = -v_2 : \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = -\begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow \begin{pmatrix} b \\ a \end{pmatrix} = -\begin{pmatrix} a \\ b \end{pmatrix} \rightarrow v_2 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$\bullet Y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \quad |\lambda I - Y| = \begin{vmatrix} \lambda & i \\ -i & \lambda \end{vmatrix} = \lambda^2 + i^2 = \lambda^2 - 1 \stackrel{!}{=} 0 \Rightarrow \lambda = \pm i$

$$\lambda_1 = i \Rightarrow Y v_1 = v_1 : \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow \begin{pmatrix} -bi \\ ai \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow v_1 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\lambda_2 = -1 \Rightarrow \forall v_2 = -v_2 : \begin{pmatrix} 0 & -1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = -\begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow \begin{pmatrix} -b \\ a \end{pmatrix} = -\begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow v_2 = \begin{pmatrix} \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} \end{pmatrix}$$

$$\bullet Z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \quad |\lambda_2 - Z| = \begin{vmatrix} \lambda - 1 & 0 \\ 0 & \lambda + 1 \end{vmatrix} = \lambda^2 - 1 \stackrel{!}{=} 0 \Rightarrow \lambda = \pm 1$$

$$\lambda_1 = 1 \Rightarrow Z v_1 = v_1 : \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow \begin{pmatrix} a \\ -b \end{pmatrix} = \begin{pmatrix} a \\ b \end{pmatrix} \Rightarrow v_1 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\lambda_2 = -1 \Rightarrow Z v_2 = -v_2 : \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} a \\ b \end{pmatrix} = -\begin{pmatrix} a \\ b \end{pmatrix} \Leftrightarrow \begin{pmatrix} a \\ -b \end{pmatrix} = \begin{pmatrix} -a \\ -b \end{pmatrix} \Rightarrow v_2 = \begin{pmatrix} 0 \\ 1 \end{pmatrix}$$

$$b) H|0\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \Rightarrow \text{both probs are } \left(\frac{1}{\sqrt{2}}\right)^2 = \frac{1}{2}.$$

Similarly we can see that with the eigenvectors of X :

$$|0\rangle = (|+\rangle + |-\rangle) \frac{1}{\sqrt{2}} \quad |+\rangle = v_1, |-\rangle = v_2 \quad \text{eigenvectors of Pauli-}X$$

This matches the second figure, after exiting the X magnetic field, there are $\frac{1}{2}$ prob. for $|+\rangle$ and $|-\rangle$.

- c) Since it corresponds to a standard Z -axis measurement and the output state of the previous stage consists of $|+\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$, then the output probability of the third stage will be $\frac{1}{2}$ for both $|1\rangle$ and $|0\rangle$.

If we replace it with X -axis measurement, then the output from last stage remains unchanged, as it's just a repeated consequent measurement.

In conclusion we can say that measuring the spin over an axis somehow "erases" the information over the previous axes.