

## Problem 1

Consider a 2-dimensional Hamiltonian  $\mathcal{H}$ , where  $|+\frac{\pi}{4}\rangle$  is the eigenvector with 0 eigenvalue and  $|-\frac{\pi}{4}\rangle$  is the other eigenvector with eigenvalue 1. Recall that you can always write an operator as the sum of the product of an eigenvalue with the projection to the corresponding eigenspace.

**a.** Write the Hamiltonian in terms of the eigenvectors and in matrix form (in the computational basis).

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**Solution:**

$$H = \mathbf{0} |+\frac{\pi}{4}\rangle \langle +\frac{\pi}{4}| + \mathbf{1} |-\frac{\pi}{4}\rangle \langle -\frac{\pi}{4}| = |-\frac{\pi}{4}\rangle \langle -\frac{\pi}{4}| = \frac{1}{2} \begin{pmatrix} 1 & -e^{-i\frac{\pi}{4}} \\ -e^{i\frac{\pi}{4}} & 1 \end{pmatrix}$$

**b.** Express  $\mathcal{H} = \sum_i c_i P_i$  in terms of Pauli matrices ( $P_i \in \{I, X, Y, Z\}$ ) using the formula for calculating the coefficients  $c_i$  given in the lecture, and for simplicity here too:  $c_i = \langle P_i, \mathcal{H} \rangle$ , where  $\langle A, B \rangle := \frac{\text{Tr}(A^\dagger B)}{2}$ .

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**Solution:**

$$\mathcal{H} = \sum_{i=0}^3 \alpha_i P_i = \alpha_0 I + \alpha_1 X + \alpha_2 Y + \alpha_3 Z$$

To calculate the  $\alpha_i$  coefficients, we use the above formula:

$$\alpha_0 = \langle I, \mathcal{H} \rangle = \frac{\text{Tr}(\mathcal{H})}{2} = \frac{1}{2}$$

$$\alpha_1 = \langle X, \mathcal{H} \rangle = \frac{\text{Tr}(X |-\frac{\pi}{4}\rangle \langle -\frac{\pi}{4}|)}{2} = \frac{1}{2} (\langle -\frac{\pi}{4}| X |-\frac{\pi}{4}\rangle)$$

$$\langle -\frac{\pi}{4}| X |-\frac{\pi}{4}\rangle = \frac{1}{2} (\langle 0| - e^{-i\frac{\pi}{4}} \langle 1|) (\langle 1| - e^{i\frac{\pi}{4}} |0\rangle) = \frac{1}{2} (-e^{i\frac{\pi}{4}} - e^{-i\frac{\pi}{4}}) = -\cos \frac{\pi}{4} = -\frac{\sqrt{2}}{2}$$

And the coefficient is:

$$\alpha_1 = -\frac{\sqrt{2}}{4}$$

For the  $Z$  operator we have:

$$Z |-\frac{\pi}{4}\rangle = |+\frac{\pi}{4}\rangle$$

We can calculate the fourth coefficient:

$$\alpha_3 = \langle Z, \mathcal{H} \rangle = \frac{\text{Tr}(Z |-\frac{\pi}{4}\rangle \langle -\frac{\pi}{4}|)}{2} = \frac{1}{2} (\langle -\frac{\pi}{4}| Z |-\frac{\pi}{4}\rangle) = \frac{1}{2} (\langle -\frac{\pi}{4}| +\frac{\pi}{4}\rangle) = 0$$

Finally, for Pauli  $Y = -iZX$  we have:

$$\alpha_2 = \langle Y, \mathcal{H} \rangle = \frac{\text{Tr}(-iZX |-\frac{\pi}{4}\rangle \langle-\frac{\pi}{4}|)}{2} = \frac{-i}{2} (\langle-\frac{\pi}{4}| ZX |-\frac{\pi}{4}\rangle) = \frac{-i}{2\sqrt{2}} (\langle-\frac{\pi}{4}| Z(|1\rangle - e^{i\frac{\pi}{4}}|0\rangle))$$

$$\alpha_2 = \frac{-i}{4} (\langle 0| - e^{-i\frac{\pi}{4}} \langle 1|)(-|1\rangle - e^{i\frac{\pi}{4}}|0\rangle) = \frac{i}{4} (e^{i\frac{\pi}{4}} - e^{-i\frac{\pi}{4}}) = \frac{i}{4} (2i \sin \frac{\pi}{4}) = -\frac{\sin \frac{\pi}{4}}{2} = -\frac{\sqrt{2}}{4}$$

So the decomposition of the Hamiltonian will be:

$$\mathcal{H} = \frac{1}{2}(I - \frac{\sqrt{2}}{2}(X + Y))$$

c. Evaluate the output state  $|\psi(\theta)\rangle$  given by the following parametrised circuit :

$$|0\rangle \text{ --- } \boxed{H} \text{ --- } \boxed{R(\theta)} \text{ --- } |\psi(\theta)\rangle$$

**Solution:** The output of the circuit can be evaluated as follows:

After the first Hadamard gate, the  $|0\rangle$  state will be transformed to  $|+\rangle$ . Then, by performing the  $R(\theta)$  gate, we will have:

$$|\psi(\theta)\rangle = R(\theta) |+\rangle = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 0 \\ 0 & e^{i\theta} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ e^{i\theta} \end{pmatrix} = \frac{1}{\sqrt{2}}(|0\rangle + e^{i\theta}|1\rangle)$$

d. For any state express  $E(\theta) := \langle \psi(\theta) | \mathcal{H} | \psi(\theta) \rangle$  by using part b, in terms of  $\langle \psi(\theta) | P_i | \psi(\theta) \rangle$ .

**Solution:**

$$E(\theta) = \langle \psi(\theta) | \mathcal{H} | \psi(\theta) \rangle = \frac{1}{4} \langle \psi(\theta) | (2I + (-\sqrt{2})(X + Y)) | \psi(\theta) \rangle = \frac{1}{4} (2 + (-\sqrt{2})(\langle \psi(\theta) | X | \psi(\theta) \rangle + \langle \psi(\theta) | Y | \psi(\theta) \rangle))$$

First, note that  $\sqrt{2}$  can be written as:

$$\sqrt{2} = 2 \cos \frac{\pi}{4} = 2 \sin \frac{\pi}{4}$$

Then, the expectation values of  $X$  and  $Y$  are:

$$\begin{aligned} \langle \psi(\theta) | X | \psi(\theta) \rangle &= \frac{1}{2} (e^{-i\theta} + e^{i\theta}) = \cos \theta \\ \langle \psi(\theta) | Y | \psi(\theta) \rangle &= \frac{-i}{2} (e^{i\theta} - e^{-i\theta}) = \sin \theta \end{aligned}$$

If we replace the above equations in the expectation value  $E(\theta)$ , we get:

$$E(\theta) = \frac{1}{4}[2 - 2(\cos \theta \cos \frac{\pi}{4} + \sin \theta \sin \frac{\pi}{4})] = \frac{1 - \cos(\theta - \frac{\pi}{4})}{2}$$

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e. Now find the ground state with the following steps:

- Start from  $\theta_0 = 0$
- Estimate the gradient in each step using  $\Delta(\theta_i) = E(\theta_i + \delta\theta) - E(\theta_i - \delta\theta)$  where  $\delta\theta = \frac{\pi}{8}$
- Update  $\theta$  accordingly by moving in the opposite direction of the gradient by a step  $\delta\theta$ , i.e. set  $\theta_1 = \theta_0 \pm \delta\theta$  with the sign determined by the computed gradient
- Continue this for another two steps and find the value of  $\theta$  which minimises  $\langle \mathcal{H} \rangle$ .

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**Solution:** For  $\theta_0 = 0$  and  $\delta\theta = \frac{\pi}{8}$

$$\Delta(\theta_0) = E(\frac{\pi}{8}) - E(-\frac{\pi}{8}) = \frac{1}{2}(\cos(\frac{\pi}{8} - \frac{\pi}{4}) - \cos(-\frac{\pi}{8} - \frac{\pi}{4})) = \frac{1}{2}(\cos(\frac{3\pi}{8}) - \cos(\frac{\pi}{8})) \approx -0.27$$

The gradient is negative. We move in the opposite direction of the gradient, so we need to go to larger values of  $\theta$ . We update the  $\theta$  as:

$$\theta_1 = \theta_0 + \delta\theta = \frac{\pi}{8}$$

Now we repeat the last step for  $\theta_1$  and we calculate the gradient:

$$\Delta(\theta_1) = E(\frac{\pi}{8} + \frac{\pi}{8}) - E(\frac{\pi}{8} - \frac{\pi}{8}) = E(\frac{\pi}{4}) - E(0) = \frac{1}{2}(-\cos(0) + \cos(-\frac{\pi}{4})) = \frac{1}{2}(-1 + \frac{\sqrt{2}}{2})$$

Again, the gradient is negative, so we update the  $\theta$  with the larger value. The next  $\theta$  is:

$$\theta_2 = \theta_1 + \delta\theta = \frac{\pi}{8} + \frac{\pi}{8} = \frac{\pi}{4}$$

And the gradient is:

$$\Delta(\theta_2) = E(\frac{\pi}{4} + \frac{\pi}{8}) - E(\frac{\pi}{4} - \frac{\pi}{8}) = E(\frac{3\pi}{8}) - E(\frac{\pi}{8}) = \frac{1}{2}(-\cos(\frac{\pi}{8}) + \cos(-\frac{\pi}{8})) = 0$$

The gradient is zero and we have found the minimum.  $\theta = \frac{\pi}{4}$  minimizes the energy. This can also be checked by taking the derivation of  $E(\theta)$  and finding the minimum value of the function.

## Problem 2

Consider the cost function of a Binary Optimisation problem with up to cubic terms:

$$C(x) = \sum_i a_i x_i + \sum_{i,j} b_{ij} x_i x_j + \sum_{i,j,k} c_{ijk} x_i x_j x_k$$

Each of  $x_i$  is a binary variable  $x_i \in \{0, 1\}$ , and the index  $i$  runs from 1 to  $n$ . Finding the configuration with the smallest cost gives a general cubic,  $n$ -binary-variables optimisation problem. For a number of reasons, one may be interested in turning this problem to a Quadratic Unconstrained Binary Optimisation (QUBO) problem, i.e. restricting the cost function to a cost function that has at most quadratic terms (but no cubic terms).

**a.** Let  $g(x, y) := (2 - x_i - x_j - x_k)y = 2y - x_i y - x_j y - x_k y$ , where  $y \in \{0, 1\}$  is another, new, binary variable. Show that the following functions are the same:

$$\begin{aligned} f(x) &= -x_i x_j x_k \\ &= \min_y g(x, y) \end{aligned}$$

This proves that we can replace cubic terms  $-x_i x_j x_k$  with quadratic ones  $(2 - x_i - x_j - x_k)y$  with the cost of introducing (in this case a single) extra variable  $y$ . Note, that we are interested in the minimum value of the cost function (i.e. we are taking the minimum of this expression over all binary variables, including the newly introduced  $y$ ).

**Solution:** We will prove that for every possible configuration  $x_i x_j x_k$  with  $x_i \in \{0, 1\}$  the two functions above are equivalent. Let  $x = (x_i, x_j, x_k)$ , then:

$$\begin{aligned} x = (0, 0, 0) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y 2y = 0 \\ x = (0, 0, 1) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y (2y - y) = 0 \\ x = (0, 1, 0) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y (2y - y) = 0 \\ x = (0, 1, 1) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y (2y - y - y) = 0 \\ x = (1, 0, 0) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y (2y - y) = 0 \\ x = (1, 0, 1) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y (2y - y - y) = 0 \\ x = (1, 1, 0) &\implies f(x) = 0 \text{ and } \min_y g(x, y) = \min_y (2y - y - y) = 0 \\ x = (1, 1, 1) &\implies f(x) = -1 \text{ and } \min_y g(x, y) = \min_y (2y - y - y - y) = -1 \end{aligned}$$

Intuitively, we see that unless  $x_i = x_j = x_k = 1$  the term  $(2 - x_i - x_j - x_k)$  is non-negative, thus it is minimised when we multiply it with  $y = 0$  giving cost 0 in agreement with the product

of  $-x_i x_j x_k$  (since at least one of the variables is zero). In the case that  $x_i = x_j = x_k = 1$ , then  $(2 - x_i - x_j - x_k)$  is  $-1$  and thus we get minimum when  $y = 1$ , leading to total cost of  $-1$ , again in agreement with  $-x_i x_j x_k$  for the same case.

So we can see that we can replace every term  $-x_i x_j x_k$  with quadratic ones  $\min_y (2 - x_i - x_j - x_k)y$  with the cost of introducing (in this case a single) extra variable  $y$ . By doing so, we can reduce any cubic binary optimisation problem to a quadratic one.

**b.** Find how we reduce the order of a cost function that has fourth order terms of the form  $x_i x_j x_k x_l$  and make it quadratics (i.e. an expression with at most quadratic terms).

**Solution:** We define a new function  $h(x, y) := (3 - x_i - x_j - x_k - x_l)y$ . It's easy to see that:

$$\min_y h(x, y) = -x_i x_j x_k x_l$$

In fact, we can generalise and replace every term with  $n$  binary variables  $-x_0 \dots x_{n-1}$  with the function:

$$h(x, y) = \left[ (n - 1) - \sum_{i=0}^{n-1} x_i \right] y$$

that has only up to quadratic terms. Then:

$$\min_y h(x, y) = -x_0 \dots x_{n-1}$$

**c.** Consider the cost function:

$$C(x) = -5x_1 x_2 x_3 x_4 + x_2 + 2x_3$$

Using the result of b. reduce the order to quadratic. Then change the variables to spins using this  $x_i = \frac{1-s_i}{2}$ . Construct a Hamiltonian  $\mathcal{H}_C$  by replacing each spin variable  $s_i$  with the Pauli  $Z_i$  gate. Finally, calculate the expectation value  $\langle \psi | \mathcal{H}_C | \psi \rangle$  of the state  $|\psi\rangle$  if  $|\psi\rangle = a |00000\rangle + b |01011\rangle$  where the last qubit denotes the extra qubit, i.e.  $|s_1 s_2 s_3 s_4 y\rangle$ .

**Solution:** We start by mapping the cost function  $C(x)$  to the cost function  $C(x, y)$  by introducing an extra bit  $y$  and replacing the term  $-5x_1 x_2 x_3 x_4$ :

$$\begin{aligned} C(x) &\rightarrow C(x, y) = 5(3 - x_1 - x_2 - x_3 - x_4)y + x_2 + 2x_3 \\ \implies C(x, y) &= 15 - 5x_1 y - 5x_2 y - 5x_3 y - 5x_4 y + x_2 + 2x_3 \end{aligned}$$

Now, we can change every binary variable  $x_i$  to a spin variable  $s_i$  where  $x_i = \frac{1-s_i}{2}$ , i.e.:

$$C(s, s_y) = \frac{23}{2} + 5s_y + \frac{5}{4}s_1 + \frac{3}{4}s_2 + \frac{1}{4}s_3 + \frac{5}{4}s_4 - \frac{5}{4}(s_1s_y + s_2s_y + s_3s_y + s_4s_y)$$

Now, we can construct a Hamiltonian  $\mathcal{H}_C$  by replacing each variable  $s_i$  with the Pauli  $Z_i$  gate. Thus:

$$\mathcal{H}_C = \frac{23}{2} + 5Z_y + \frac{5}{4}Z_1 + \frac{3}{4}Z_2 + \frac{1}{4}Z_3 + \frac{5}{4}Z_4 - \frac{5}{4}(Z_1Z_y + Z_2Z_y + Z_3Z_y + Z_4Z_y)$$

Recall that for a computational basis state  $|x\rangle$ ,  $Z|x\rangle = (-1)^x|x\rangle$ . Initially, we will find the expectation value of the state  $|00000\rangle$ :

$$F_1 = \langle 00000 | \mathcal{H}_C | 00000 \rangle = 15$$

Then, for the second state:

$$F_2 = \langle 01011 | \mathcal{H}_C | 01011 \rangle = 6$$

So the total expectation value is:

$$\langle \psi | \mathcal{H}_C | \psi \rangle = 15|a|^2 + 6|b|^2$$