

Proof that VQC is a Linear Classifier

consider

that $\text{pred} = \langle \Phi(x) | W_\theta^\dagger f W_\theta | \Phi(x) \rangle$

$$= \text{Tr} [|\Phi(x)\rangle \langle \Phi(x)| \cdot W_\theta^\dagger Z W_\theta]$$

$\in \mathbb{R}^{2^n, 2^n} \quad \in \mathbb{R}^{2^n, 2^n}$

this works for arbitrary f .

← used: $\text{Tr}[AB] = \text{Tr}[BA]$

even if A, B are not square!
 $A \in \mathbb{C}^{n,k} \quad B \in \mathbb{C}^{k,n}$
 $n, k \in \mathbb{N}$

Now: decompose each of these in the operator basis (Pauli basis)

$$|\Phi(x)\rangle \langle \Phi(x)| = \frac{1}{2^n} \sum_{\alpha \in 4^n} \text{Tr} [|\Phi(x)\rangle \langle \Phi(x)| \cdot P_\alpha] \cdot P_\alpha$$

$$\Phi(x) = \frac{1}{2^n} \sum_{\alpha \in 4^n} \Phi_\alpha(x) \cdot P_\alpha$$

$$H_\theta = W_\theta^\dagger Z W_\theta = \frac{1}{2^n} \sum_{\alpha \in 4^n} \text{Tr} [W_\theta^\dagger Z W_\theta \cdot P_\alpha] \cdot P_\alpha$$

$$H_\theta = \frac{1}{2^n} \sum_{\alpha \in 4^n} \omega_\alpha(\theta) \cdot P_\alpha$$

← used: for a basis B and an operator O we write O in the basis B as

$$O = \sum_{\beta_1, \beta_2 \in B} \langle \beta_1 | O | \beta_2 \rangle \beta_1 \beta_2$$

inner product between basis element β_1 and O .

Which inner product depends on the space B . O live in

In that case, it's the Hilbert-Schmidt inner product.

$$\text{Tr} [|\Phi(x)\rangle \langle \Phi(x)| \cdot W_\theta^\dagger Z W_\theta]$$

$$\text{Tr} [\Phi(x) H_\theta]$$

$$= \text{Tr} \left[\left(\frac{1}{2^n} \sum_{\alpha} \Phi_\alpha(x) \cdot P_\alpha \right) \left(\frac{1}{2^n} \sum_{\beta} \omega_\beta(\theta) P_\beta \right) \right]$$

$$= \text{Tr} \left[\frac{1}{4^n} \sum_{\alpha, \beta} \Phi_\alpha(x) \omega_\beta(\theta) \cdot P_\alpha P_\beta \right]$$

$$= \frac{1}{4^n} \sum_{\alpha, \beta} \Phi_\alpha(x) \omega_\beta(\theta) \cdot \text{Tr} [P_\alpha P_\beta]$$

$$= \frac{1}{4^n} \sum_{\alpha} \Phi_\alpha(x) \omega_\alpha(\theta) \cdot 2^n$$

$$= \frac{1}{2^n} \sum_{\alpha} \Phi_\alpha(x) \omega_\alpha(\theta)$$

$$\in [-1, 1]$$

$$\text{Tr} [P_\alpha P_\beta] = \begin{cases} 2^n & \text{if } \alpha = \beta \\ 0 & \text{if } \alpha \neq \beta \end{cases}$$

The obvious decision rule would be to threshold the output on 0,

However, say we pick an arbitrary threshold $b \in [-1, 1]$

then our decision rule is:

$$\text{label}(x) = \text{sign} \left(\sum_{\alpha} \omega_\alpha(\theta) \Phi_\alpha(x) + b \right)$$

Which is the exact same equation as our linear classifier.

$$\text{sign} \left(\sum_i u_i x_i + b \right)$$

