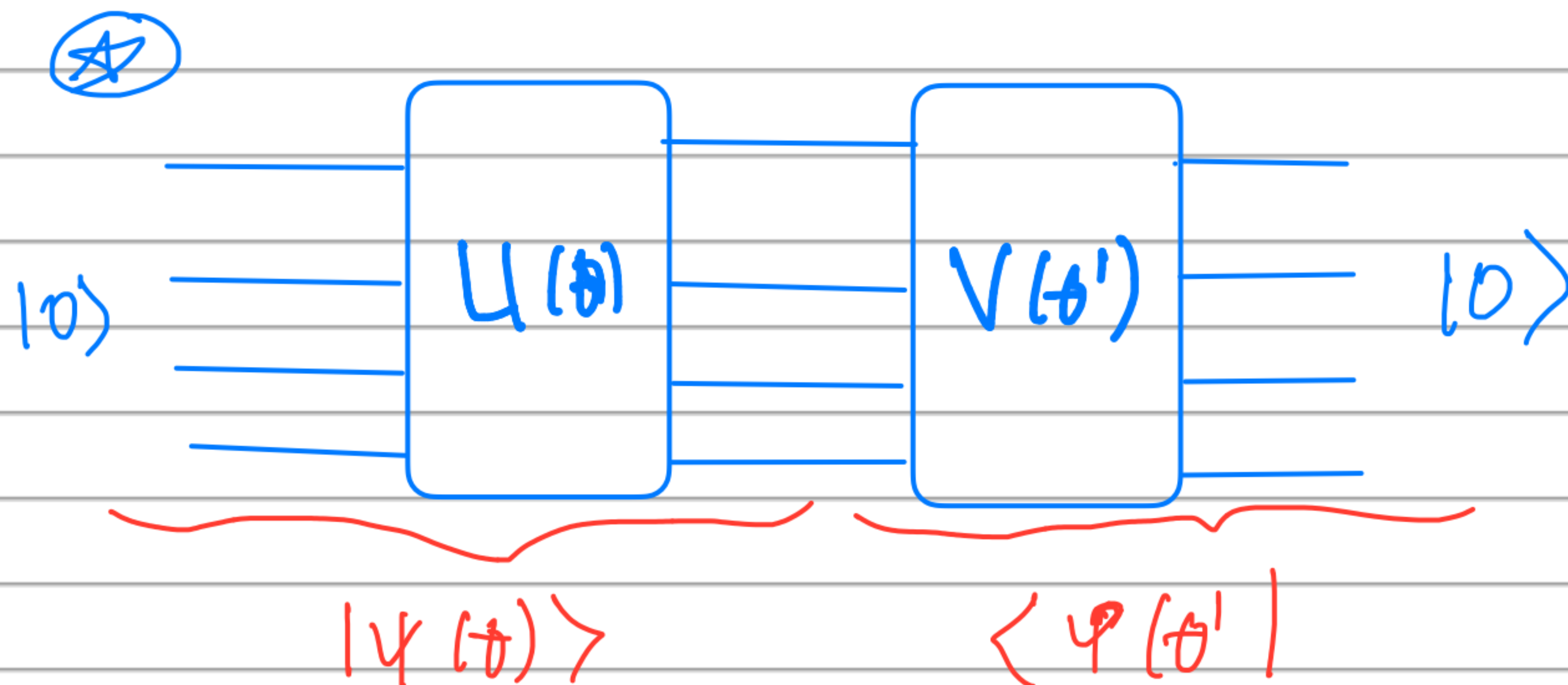


Quantum circuit learning



- We introduce a fidelity landscape:

$$K(\theta, \theta') = \langle \phi(\theta') | \psi(\theta) \rangle$$

and define a cost function as:

$$C(\theta, \theta') = 1 - K(\theta, \theta')$$

- In quantum learning circuit, we vary θ

to learn $|\psi(\theta)\rangle$ to be a target state $|\phi(\theta')\rangle$

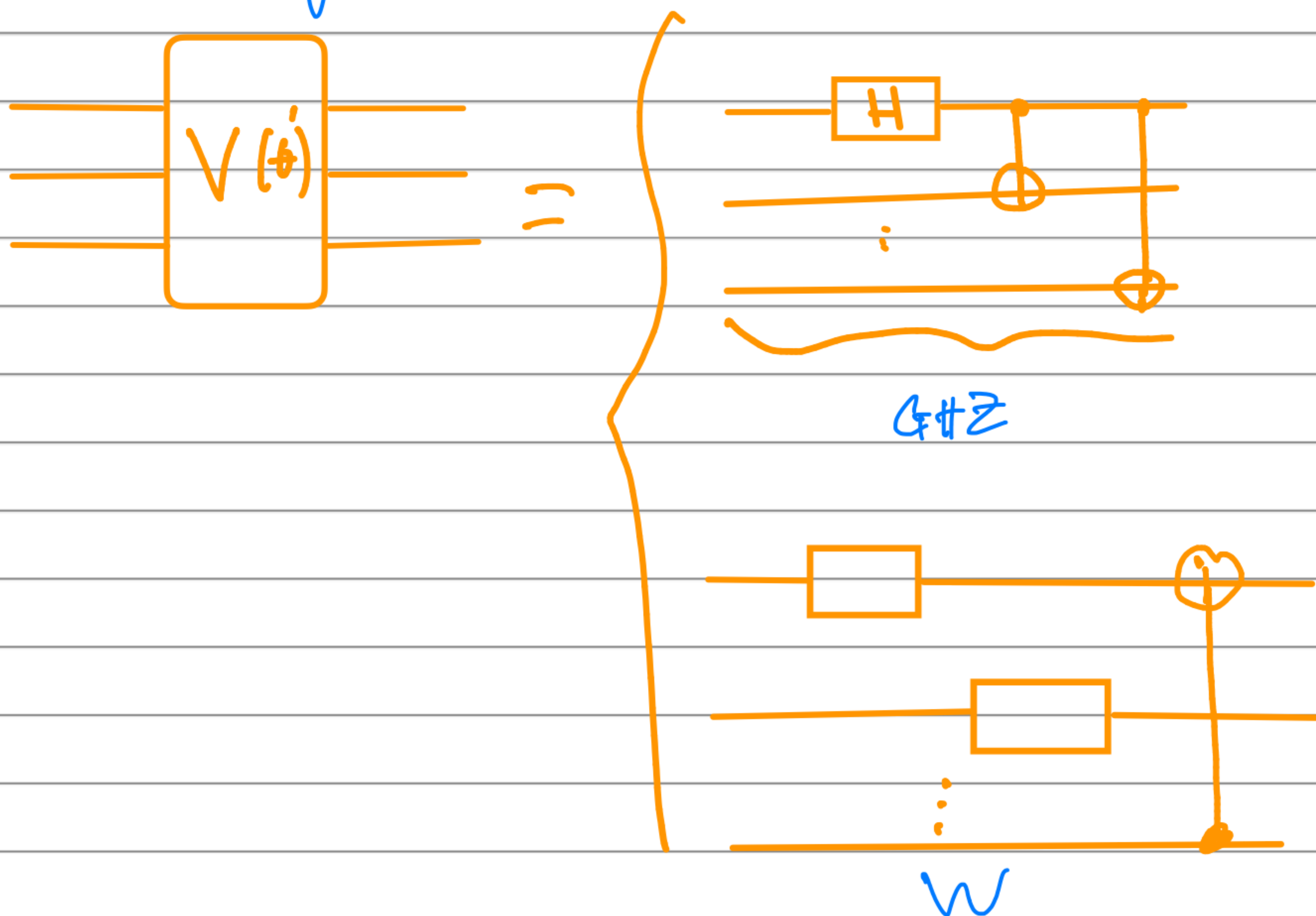
$\underset{\text{learn}}{\psi(\theta)}$ $\underset{\text{var}}{\theta}$ $\underset{\text{target}}{\phi(\theta')}$ $\underset{\text{fix}}{\theta'}$

when the learning state becomes the target state we have the minimum cost function

$$\min_{\theta_{\text{var}}} C(\theta_{\text{var}}, \theta'_{\text{fix}}) = \min_{\theta_{\text{var}}} \left[1 - \langle \varphi(\theta'_{\text{fix}}) | \psi(\theta_{\text{var}}) \rangle \right]$$

- We consider a learning ansatz as before
- We consider several entangled target states, such as GHZ , W , and also a random state.

The circuits for these target states as in Fig. 1 :



- Quantum state tomography.

- We generate an unknown state $\left(\theta_{\text{unk}} : \text{random and unknown} \right)$
$$|\psi(\theta_{\text{unk}})\rangle$$

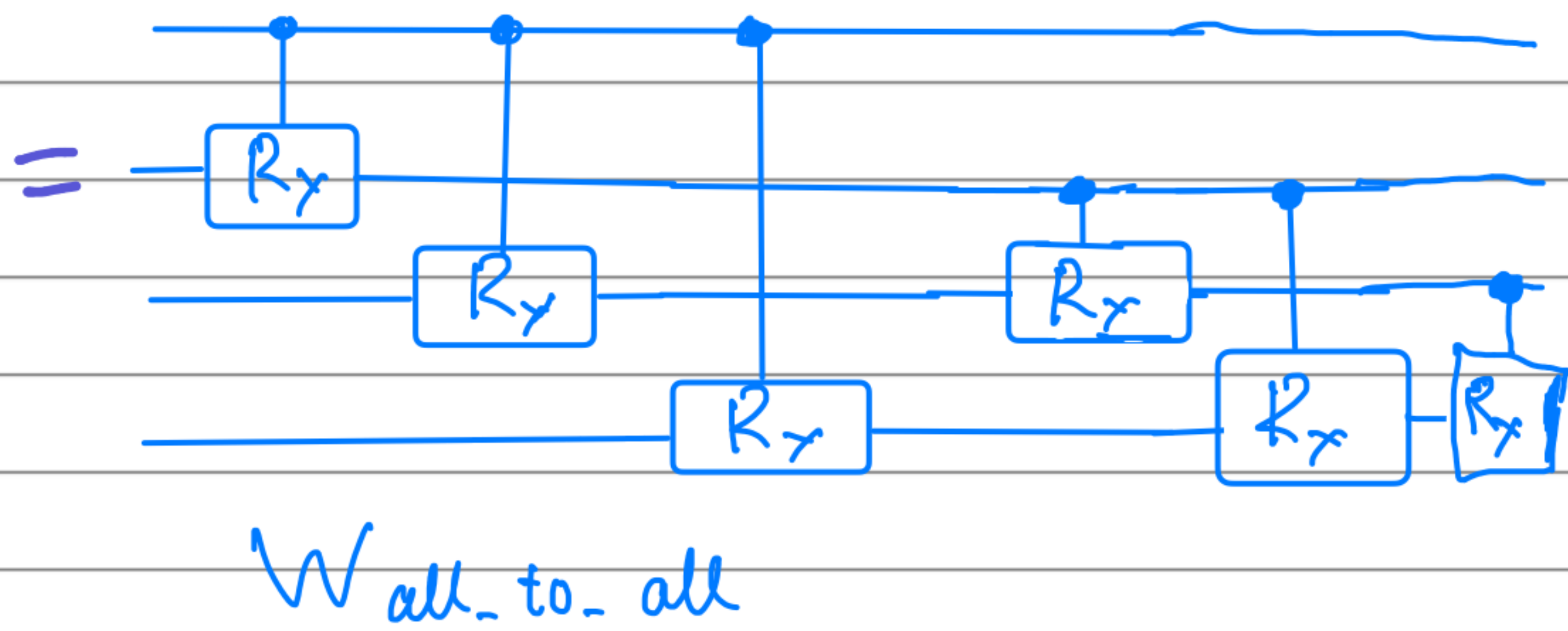
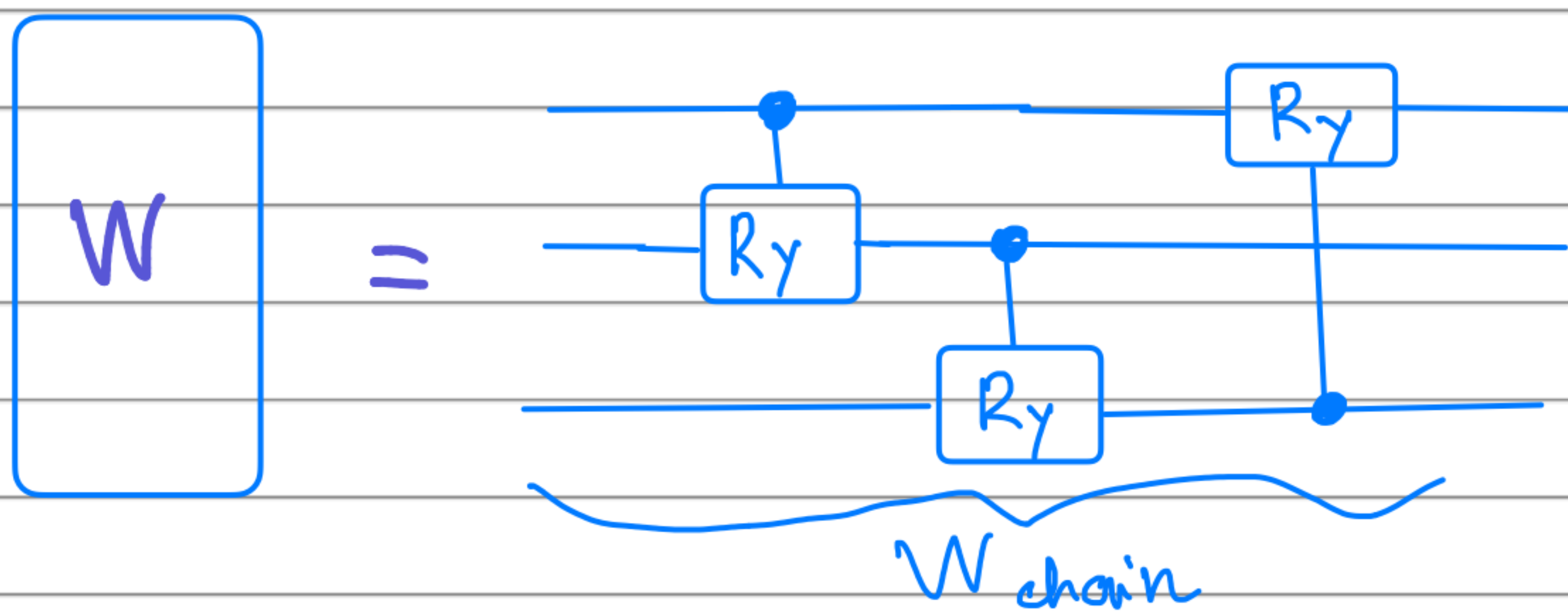
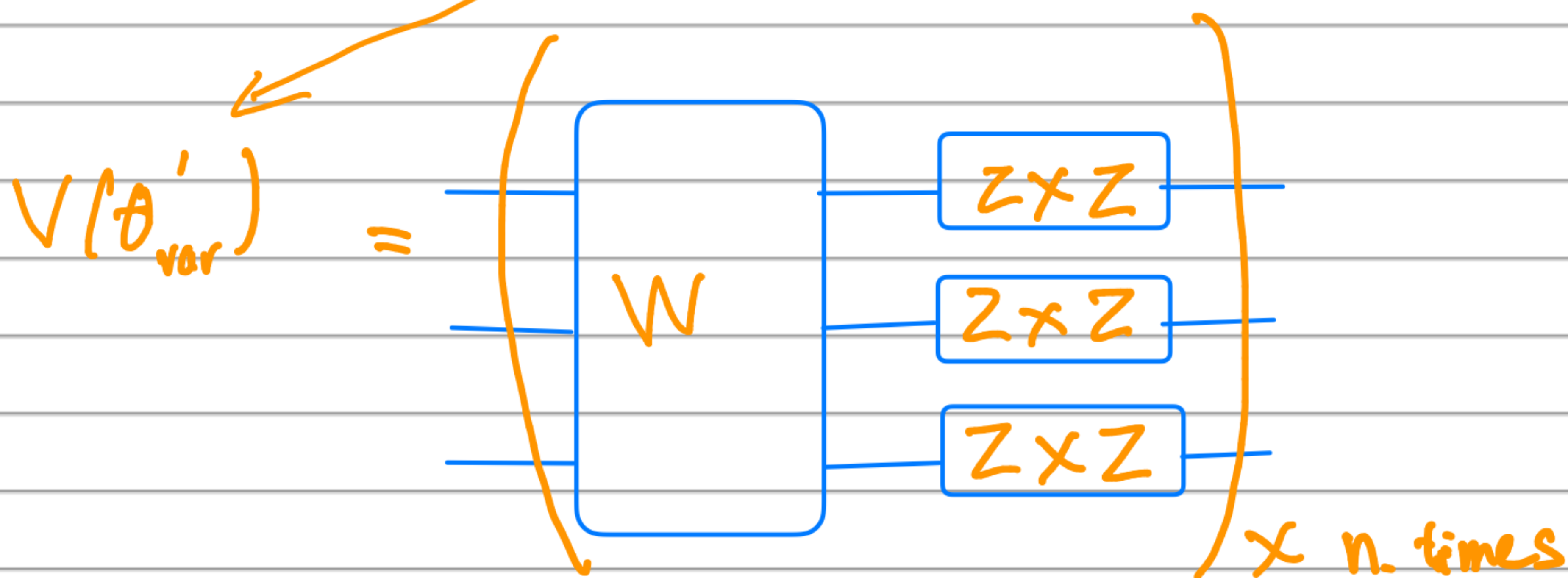
and using the variational ansatz θ'_{var} ,
such that $|\psi(\theta'_{\text{var}})\rangle$ to minimize

the cost function:

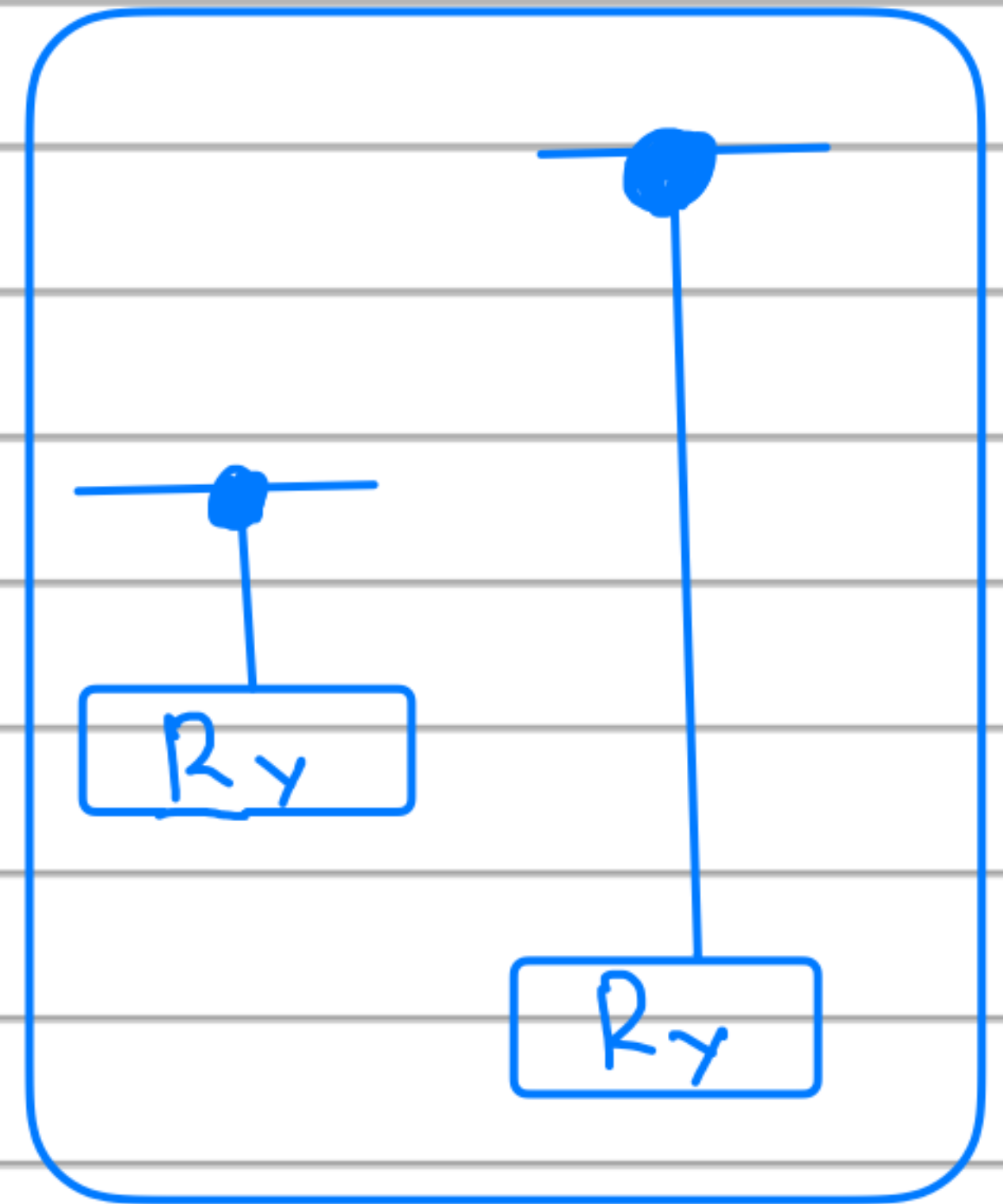
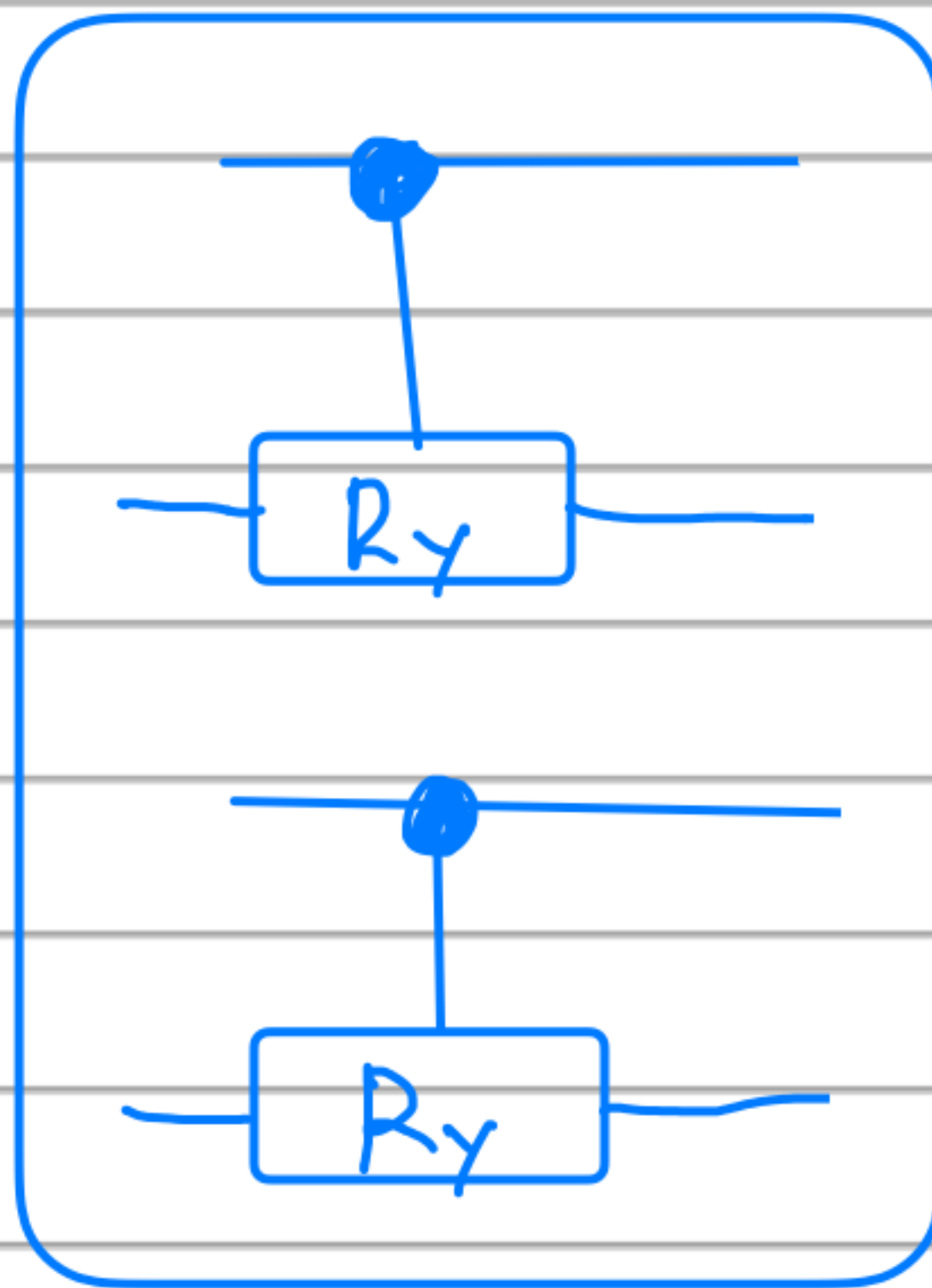
$$\min_{\theta'_{\text{var}}} C(\theta_{\text{unk}}, \theta'_{\text{var}}) = \min_{\theta'_{\text{var}}} \left[1 - \langle \psi(\theta'_{\text{var}}) | \psi(\theta_{\text{unk}}) \rangle \right]$$

- We use the Haar random ansatz to generate an unknown state $|\psi_{\text{Haar}}(\theta_{\text{unk}})\rangle$

- The structure of learning ansatz is as follow.



=



l odd

l even

W alternating

-) later, we can try CR_y by \sqrt{i} SWAP