


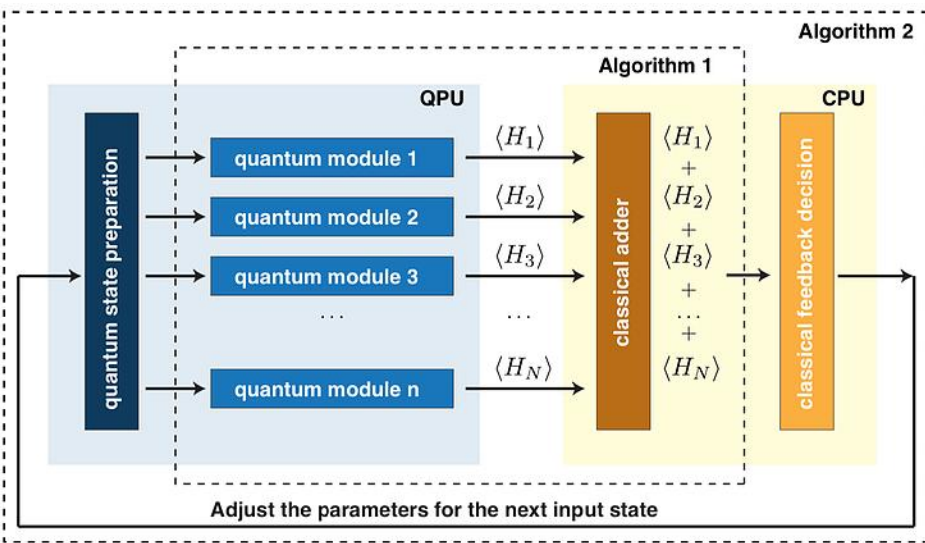
VQE on quantum magnetism

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A dark blue diagonal gradient bar that starts from the bottom left and extends towards the top right, covering the lower half of the slide.

Presentation contents

- ❖ VQE outline and the variational principle
- ❖ The ansatz
- ❖ Optimization strategies
- ❖ VQE applied on a cube
- ❖ 1D quantum phase transition
- ❖ VQE for a Kagome lattice



QPU \rightarrow evaluates

CPU \rightarrow optimizes

Similar to neural networks?

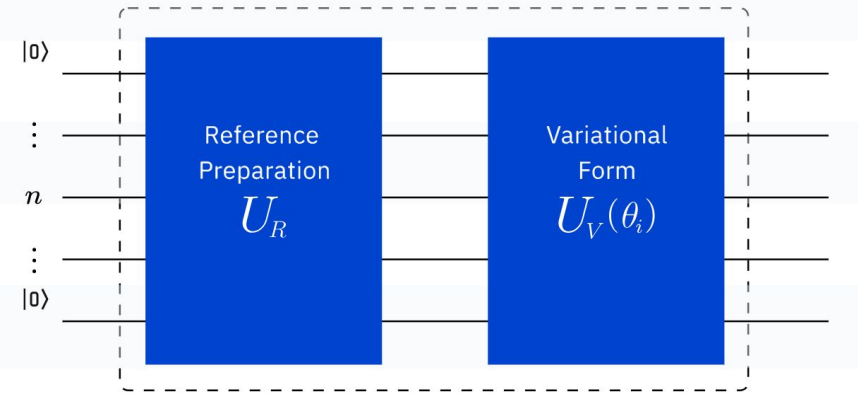
The algorithm follows an iterative process by evaluating some expectation value on a parameterized quantum circuit. The parameters are then to be optimized by a classical subroutine.

$$\langle \hat{H} \rangle \equiv \frac{\langle \psi | \hat{H} | \psi \rangle}{\langle \psi | \psi \rangle} \geq E_0$$

It can be proven easily that we can't go wrong by taking the expectation value.

Since it will be greater or equal to the exact ground energy.

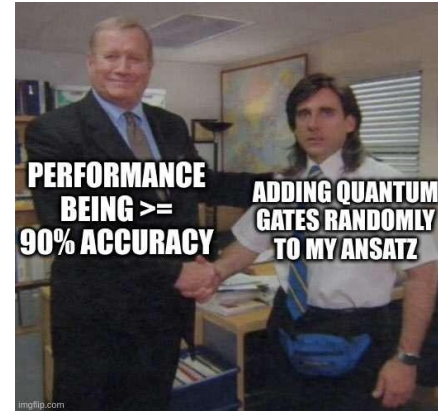
Ansatz $U_A(\theta_i)$



Defines the optimization landscape!

It's difficult to come up with or design one...

The variational form is the main part. The reference preparation is optional but might give a significant boost.



The ansatz

Optimizers

Local: they focus on a singular point of the landscape

Global: opposite from local but take more resources

Gradient based: they rely on derivative evaluations in order to progress

Gradient free: distinguishes function values in different points

COBYLA: global and gradient free

QNGD: local and gradient based

Barren plateaus: gradients vanish exponentially to the circuits complexity.

It is proven that this affects also gradient free optimizers.

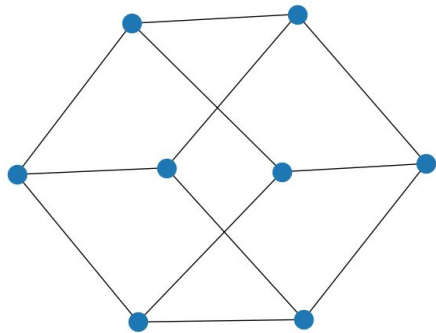
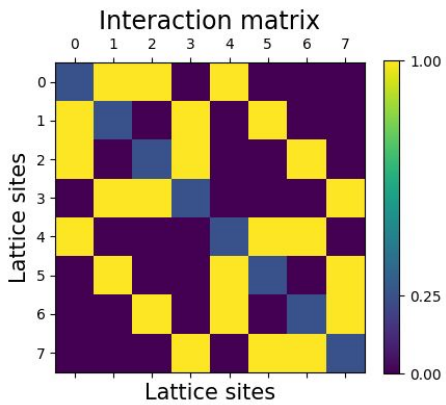
Bootstrapping: utilizing previous statistical knowledge in order to head start the optimizer.

$$\mathbf{a}_{n+1} = \mathbf{a}_n - \gamma \nabla F(\mathbf{a}_n)$$

← that's not QNGD

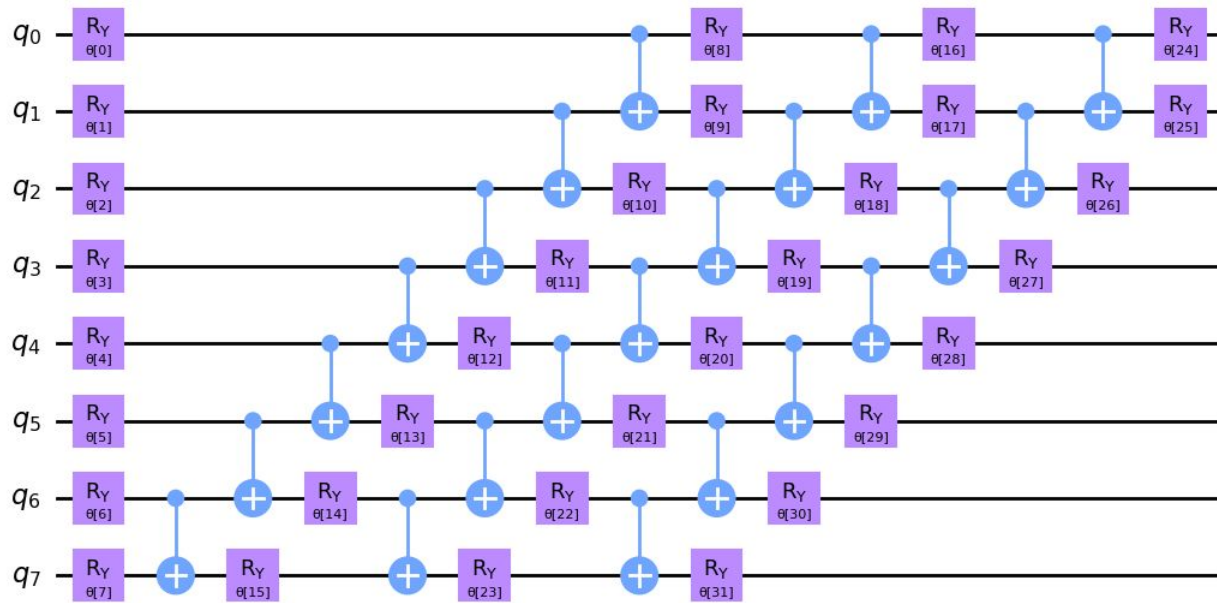
QNGD is a form of gradient that is independent of the parameterization (not to be confused with the ansatz).

Maps functions to probability distributions and computes their distance (via KL divergence).



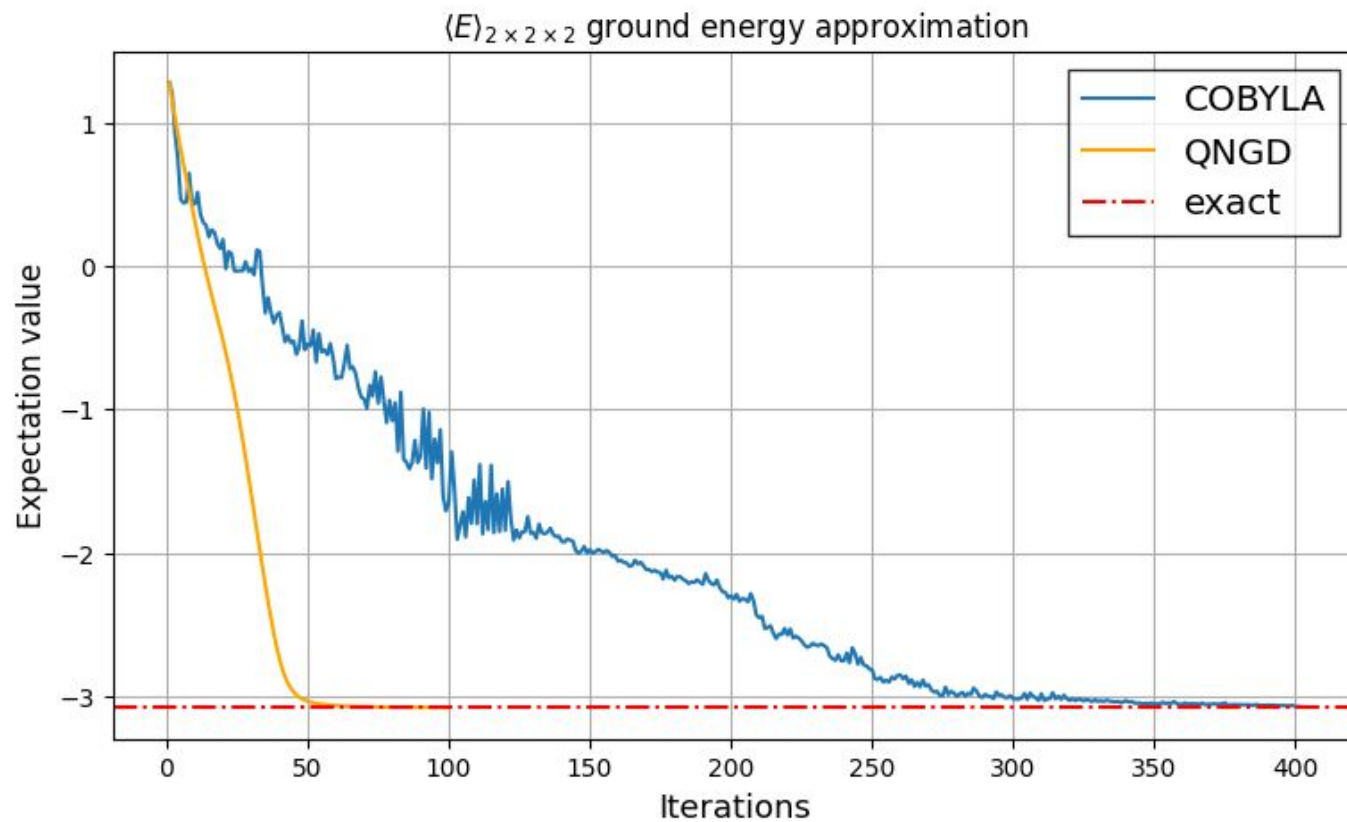
$$\hat{H} = - \sum_{\langle i,j \rangle} \hat{\sigma}_i^z \hat{\sigma}_j^z - \frac{1}{4} \sum_i \hat{\sigma}_i^x.$$

VQE applied on a cube

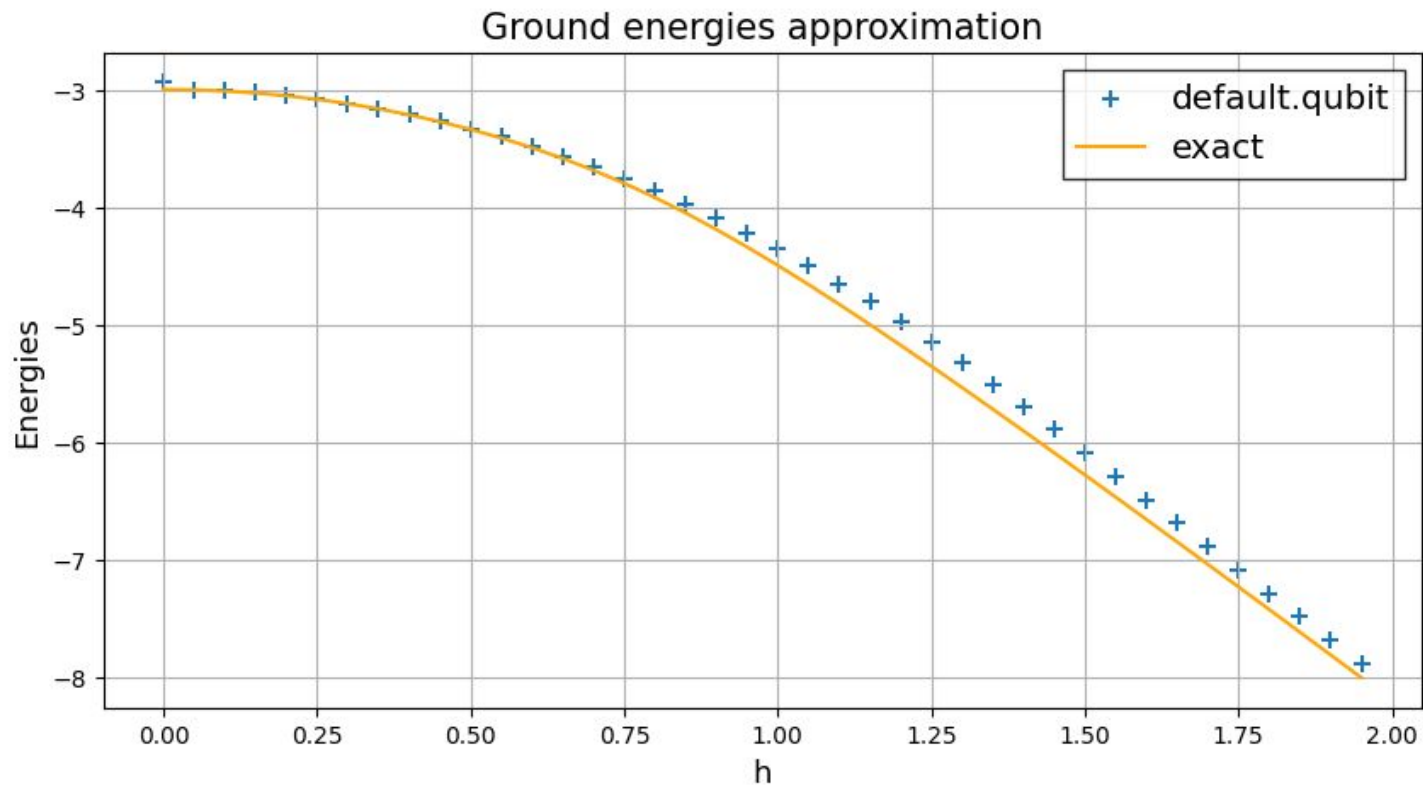


The ansatz preserves only real amplitudes while the reverse linear entanglement is equivalent to full.

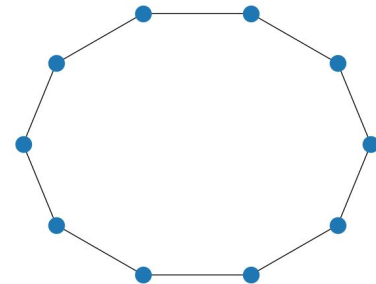
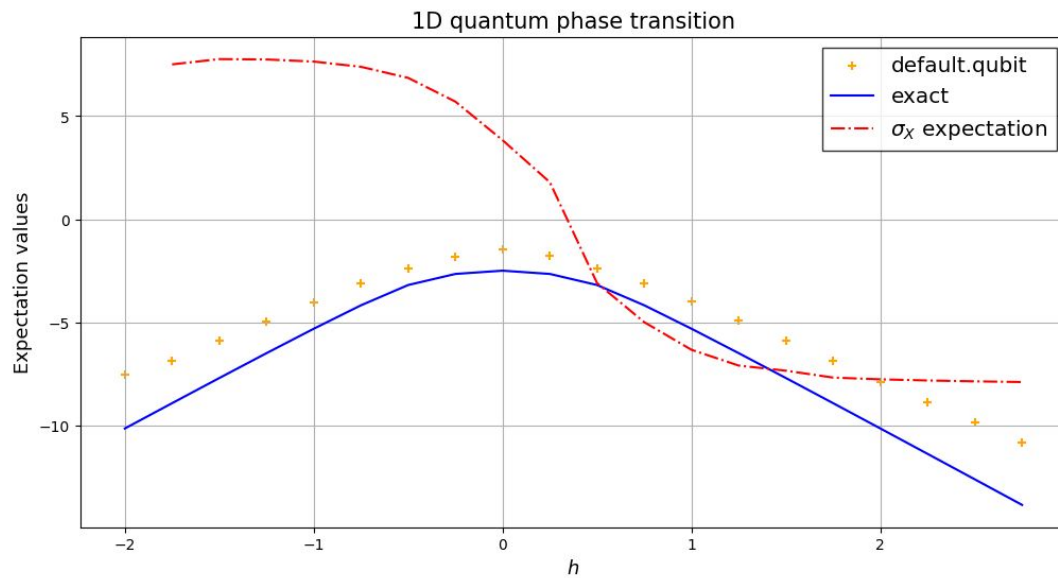
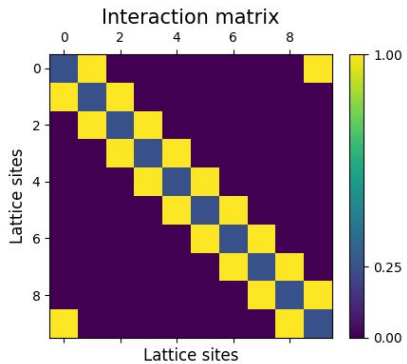
VQE applied on a cube



VQE applied on a cube

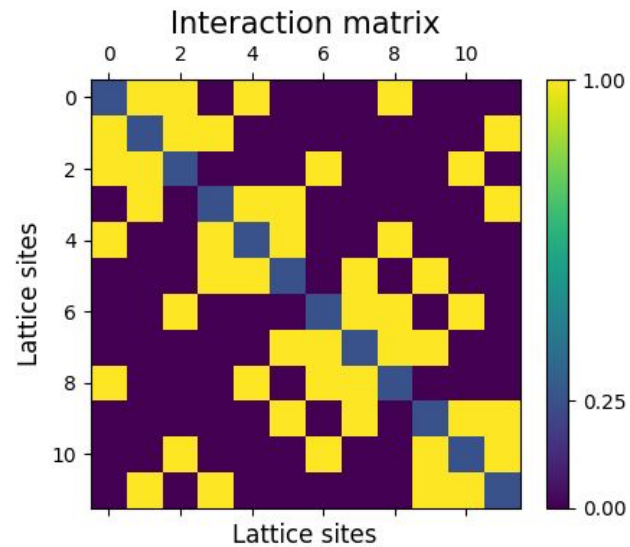
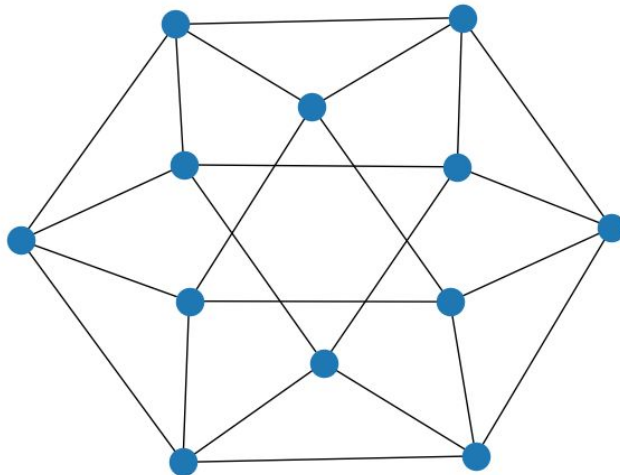
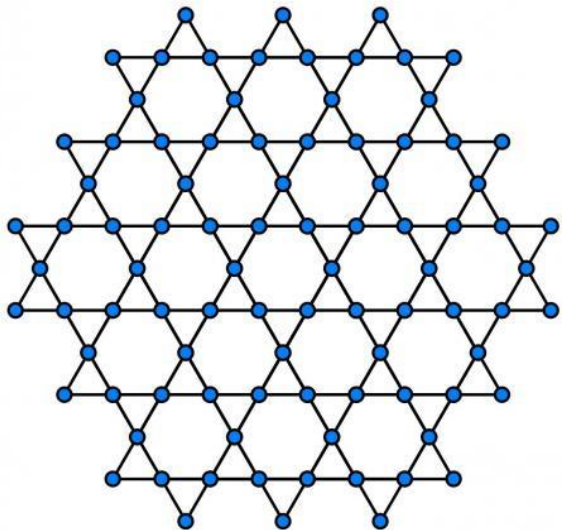


VQE applied on a cube

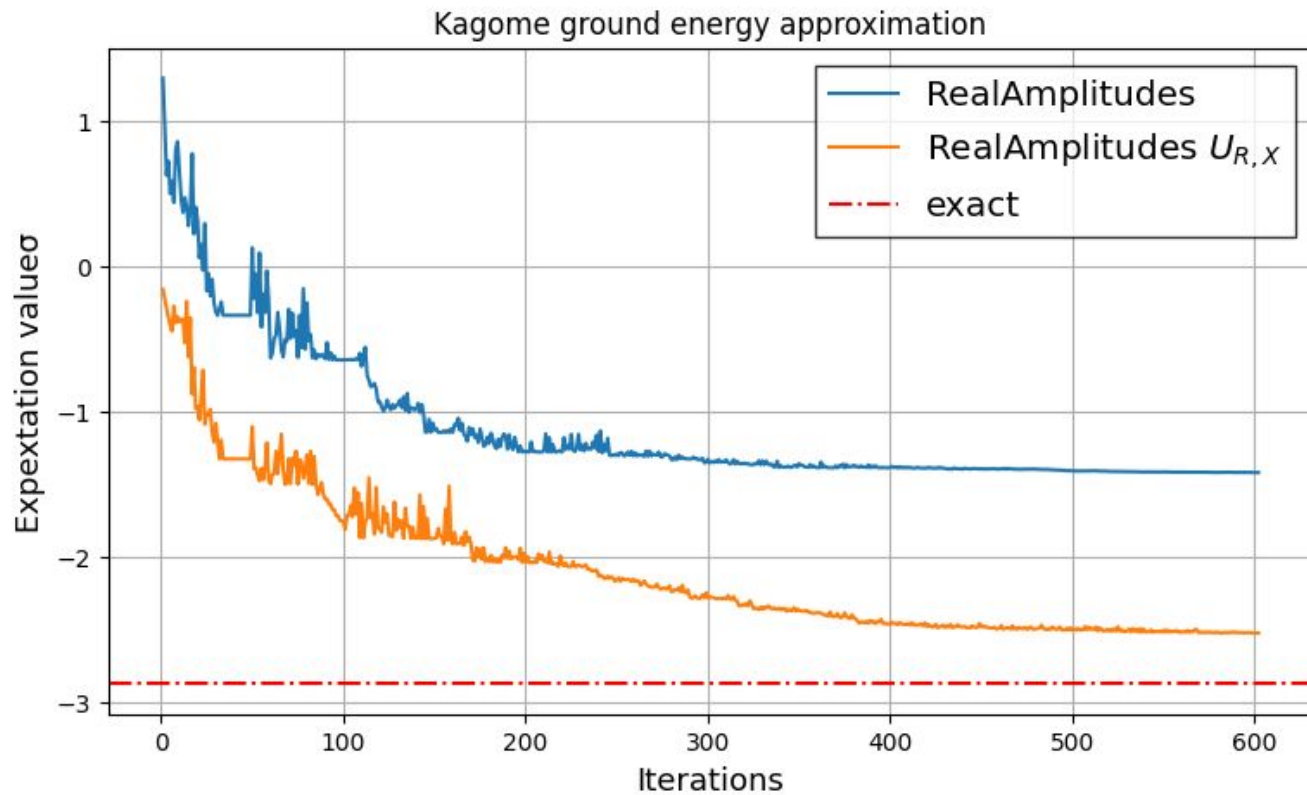


1D quantum phase transition

The Kagome's lattice intriguing geometric is includes triangles arranged in a hexagonal pattern.



VQE for a Kagome lattice



VQE for a Kagome lattice

Questions?

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Ever thought of Ising sounds?



<https://www.youtube.com/watch?v=mD-0VpNSJA0>

References

- <https://arxiv.org/abs/1304.3061>