



VARIATIONAL QUANTUM ALGORITHM FOR QUANTUM SPIN GLASS MODELS

Master Thesis by
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5th Year Integrated M.Sc.

Quantum science and technology

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Collaborators



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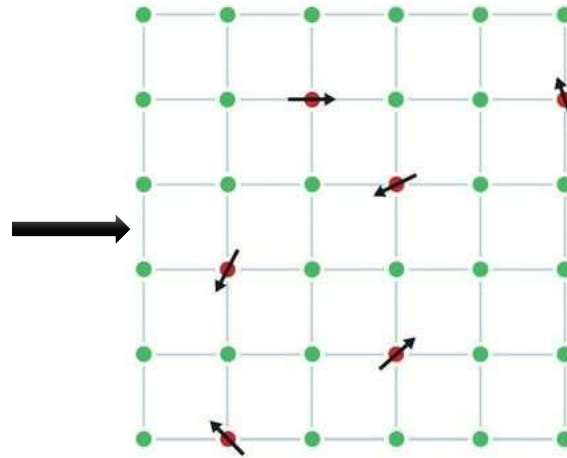


OUTLOOK:

- The Spin glass problem and corresponding Hamiltonian models
- Intro to Variational Quantum Algorithms and ansatz
- Results
- Expressibility, overparameterization and Barren Plateaus phenomenon
- Entanglement spectrum of ansatz
- Future work

QUANTUM SPIN GLASS:

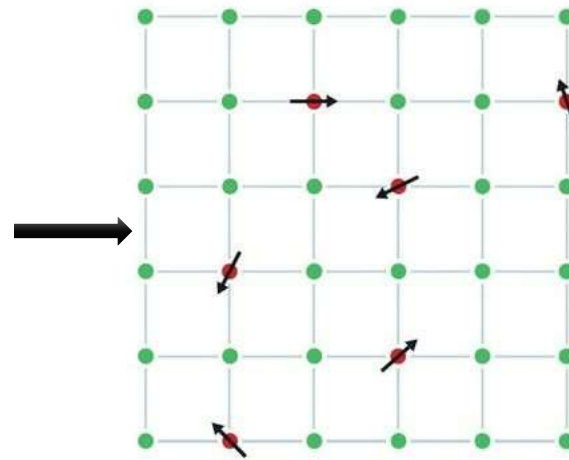
A spin glass is a metal alloy where the iron atoms (red), for example, are randomly mixed into a grid of copper atoms (green). Each iron atom behaves like a small magnet, or spin, which is affected by other magnets around it. However, in a spin glass they are frustrated and have difficulty choosing which direction to point.





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$$H = - \sum_{i < j} J_{ij} \sigma_i \sigma_j$$

S-K model

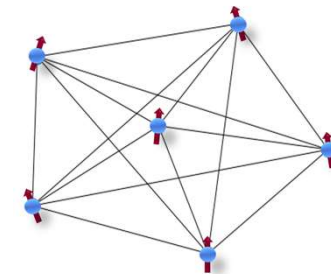
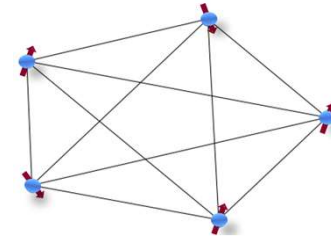
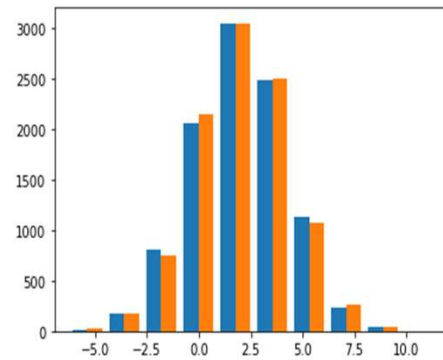
$$H = - \sum_{\langle ij \rangle} J_{ij} \sigma_i \sigma_j$$

E-D model

S-K HAMILTONIAN:

$$H = - \sum_{i < j} J_{ij} \sigma_i \sigma_j$$

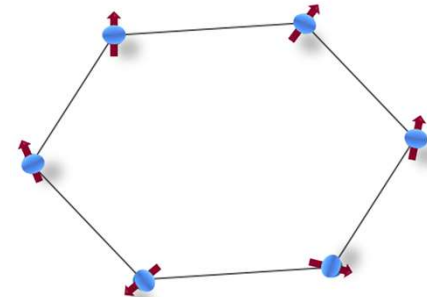
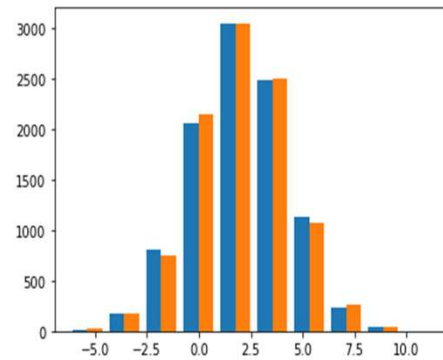
$N(\mu, \sigma^2_{SK})$



E-D HAMILTONIAN:

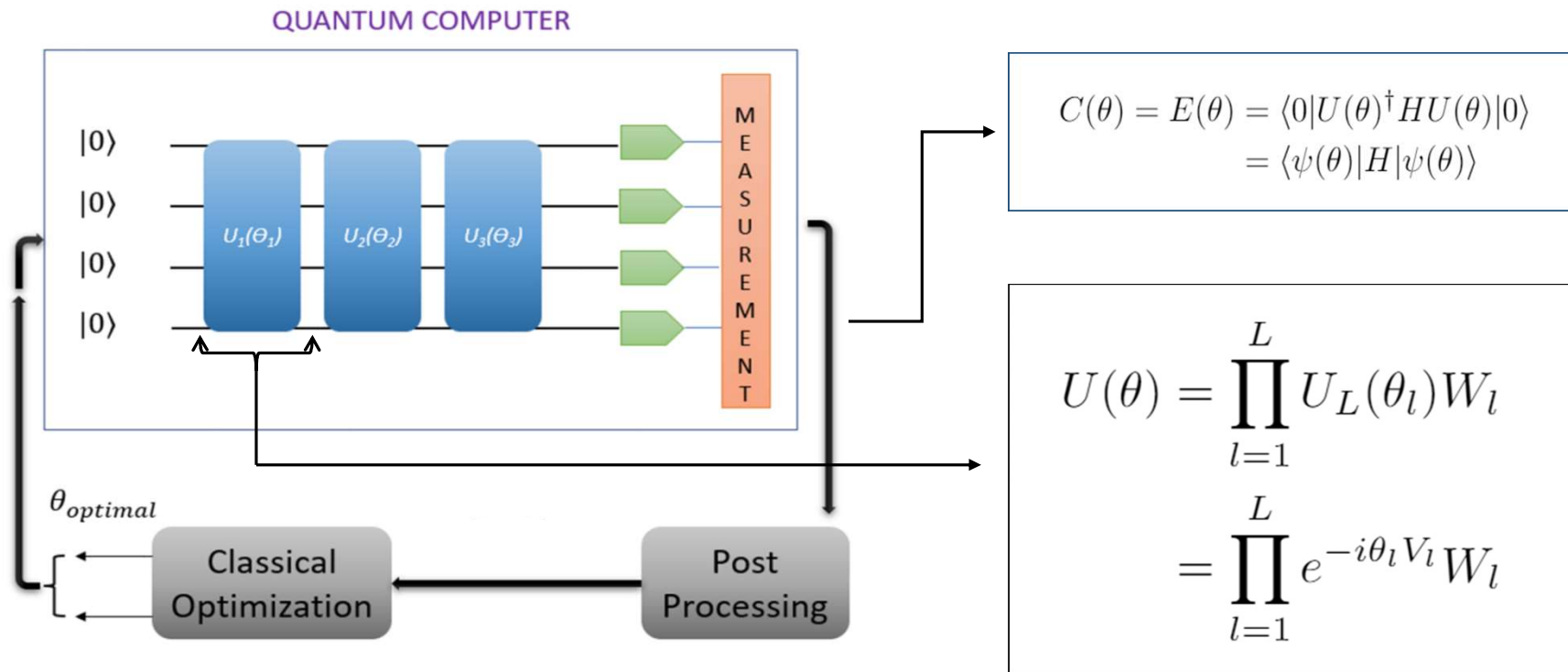
$$H = - \sum_{\langle ij \rangle} J_{ij} \sigma_i \sigma_j$$

$N(\mu, \sigma^2_{ED})$



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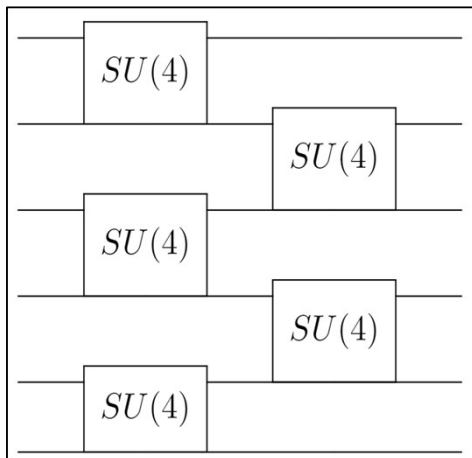
VARIATIONAL QUANTUM ALGO.



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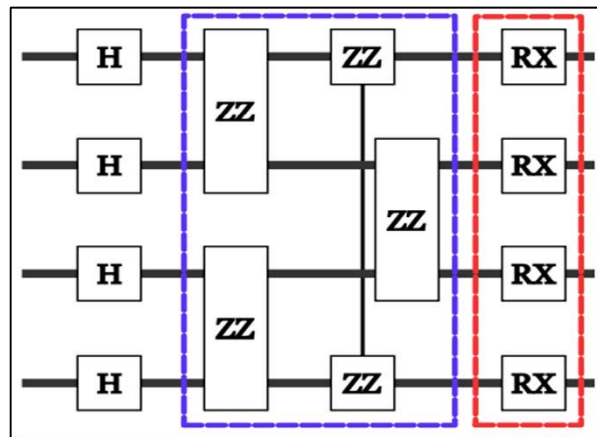


VARIATIONAL ANSATZ



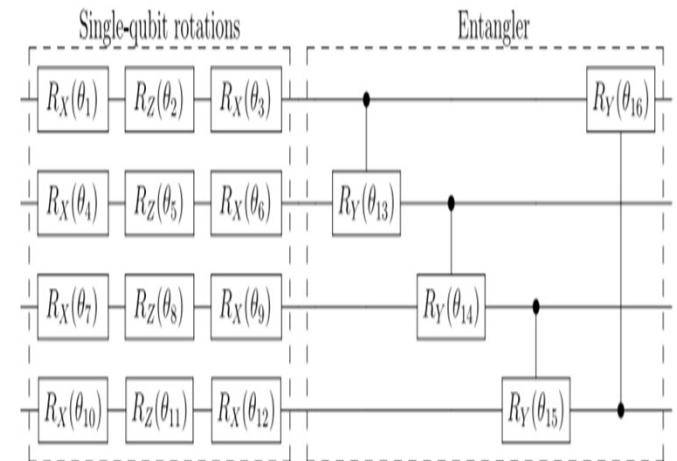
Type 1-ULA

Complexity – $O(n)$



Type 2-HVA

Complexity – $O(n)$ (Best)
or $O(n^2)$ (Worst)

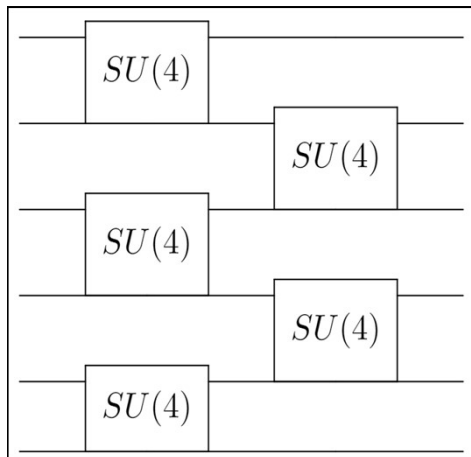


Type 3

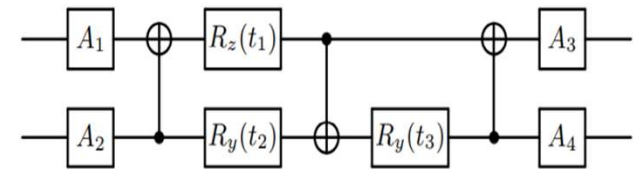
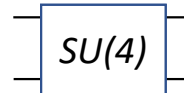
Complexity – $O(n)$

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VARIATIONAL ANSATZ



Type 1-ULA

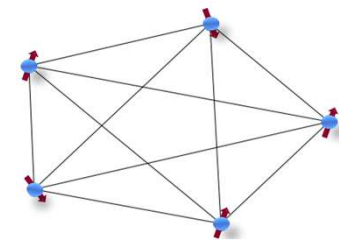
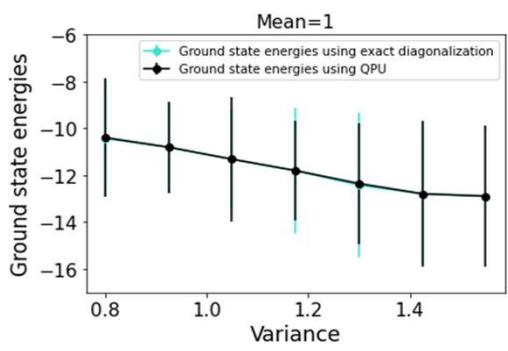
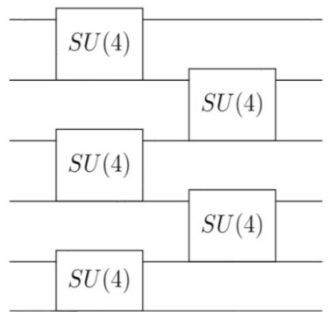


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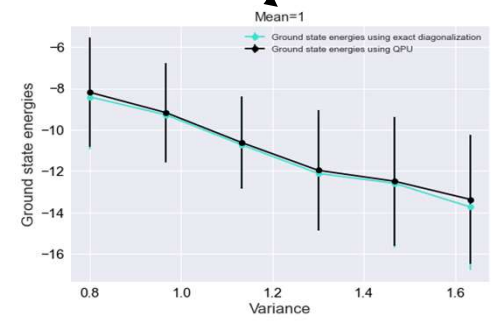
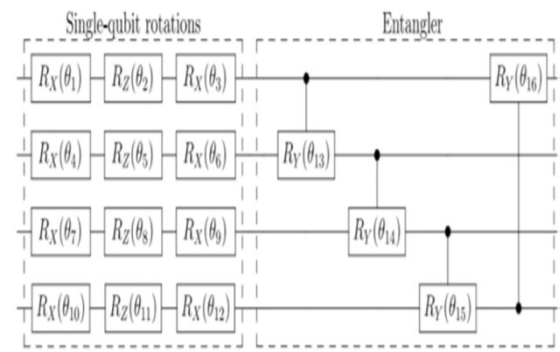
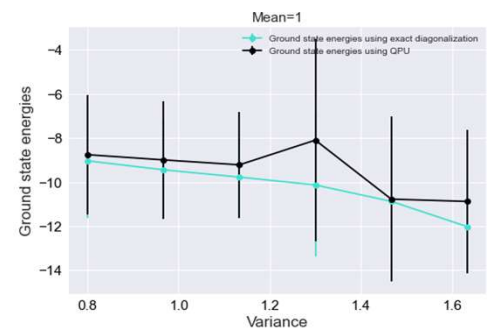
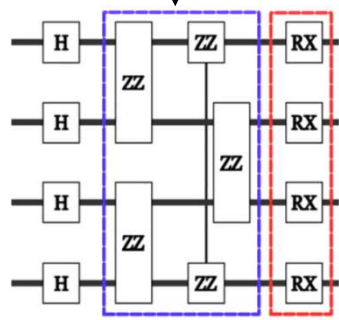
RESULTS



Minimizing $C(\theta)$



[S-K]

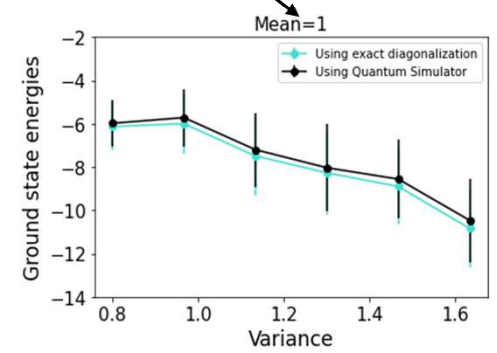
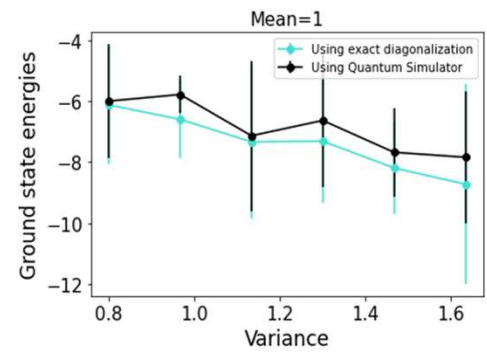
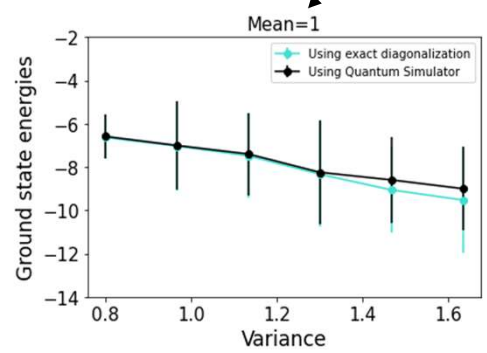
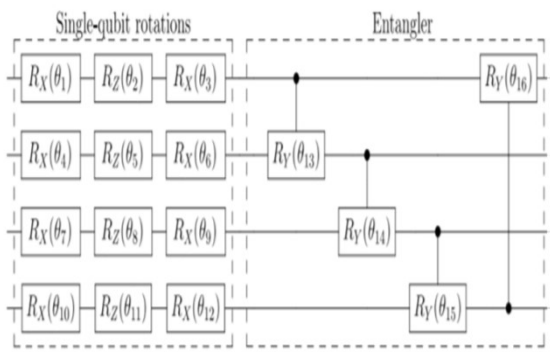
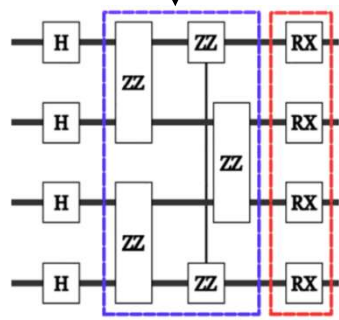
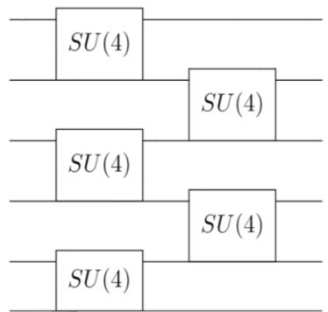


RESULTS



[E-D]

Minimizing $C(\theta)$



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OVERPARAMETERIZATION & UNDERPARAMETERIZATION



Theorem:

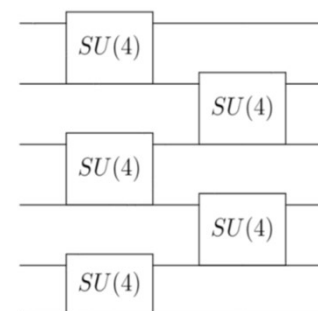
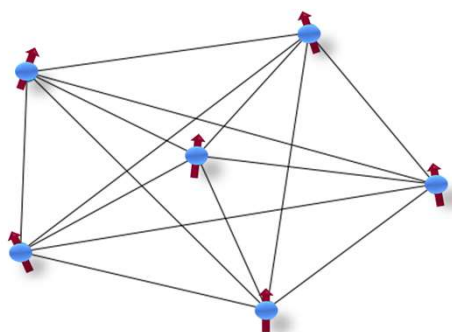
By having a low($M < M_c$) (high($M > M_c$)) number of parameters in Quantum circuit one is not able (is able) to explore all relevant directions in the Hilbert space, and thus the VQA is underparametrized (overparametrized).

M =Total no. of parameters

M_c =Critical no. of parameters

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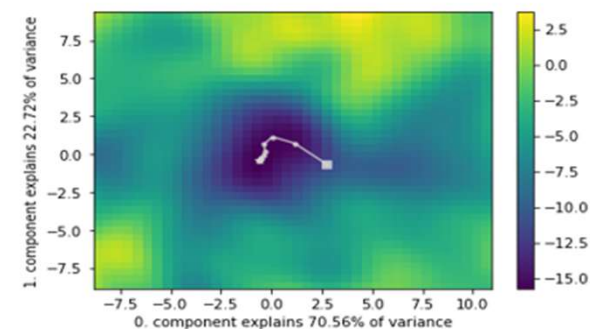
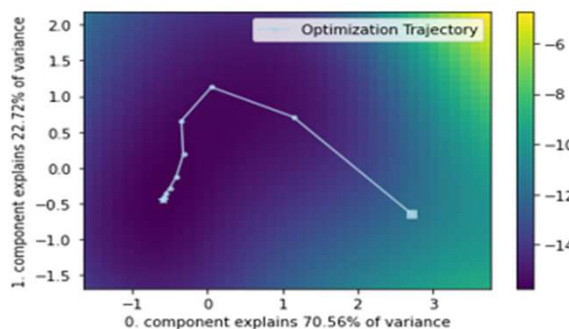
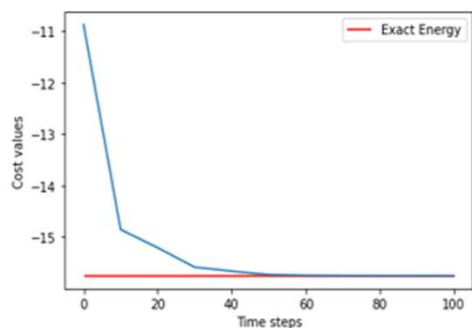
OVERPARAMETERIZATION & UNDERPARAMETERIZATION



Depth=1

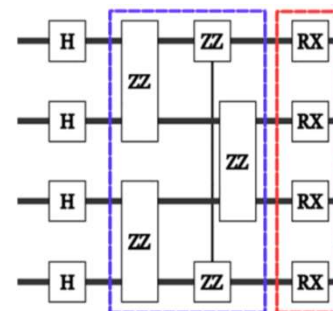
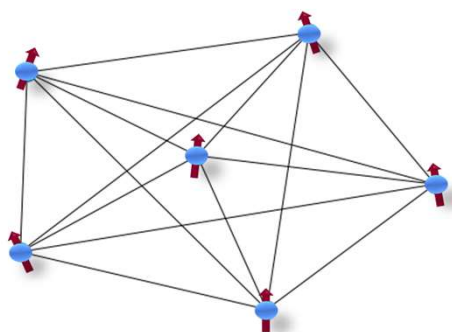
Total number of parameters $M \geq M_c$ (Critical no. of parameters)

OVERPARAMETERIZATION



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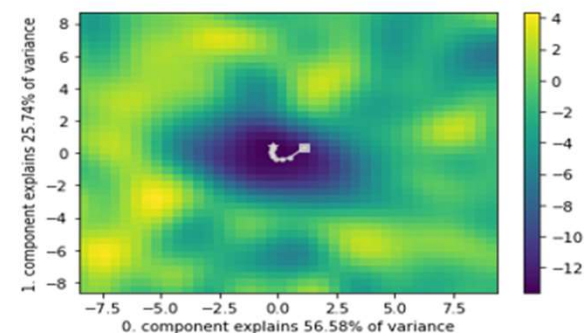
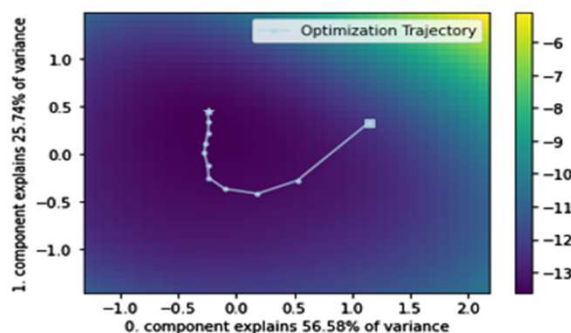
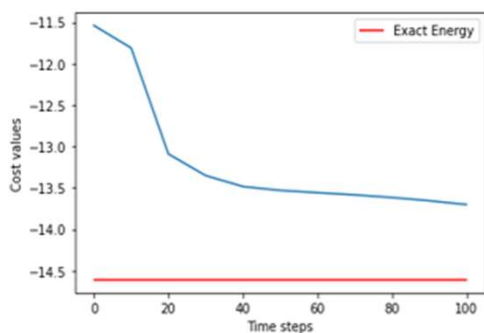
OVERPARAMETERIZATION & UNDERPARAMETERIZATION



Depth=1

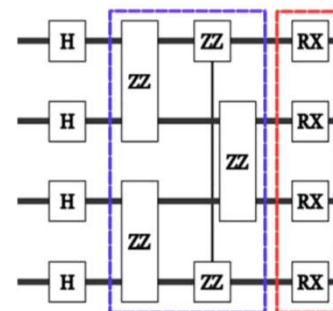
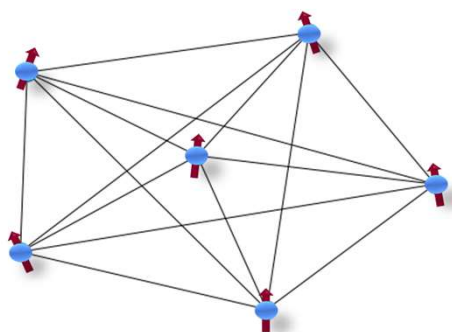
Total number of parameters $M < M_c$ (Critical no. of parameters)

UNDERPARAMETERIZATION



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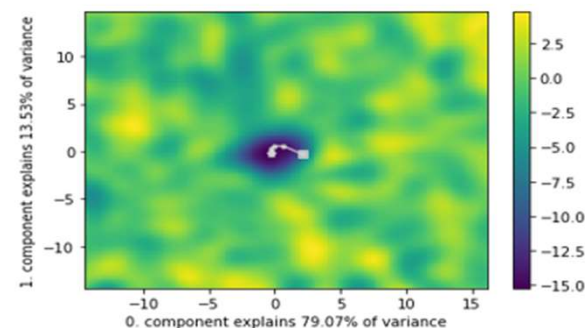
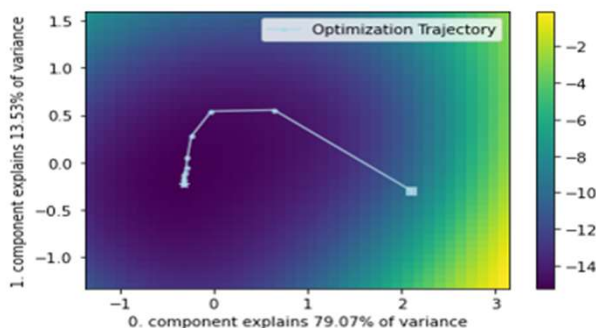
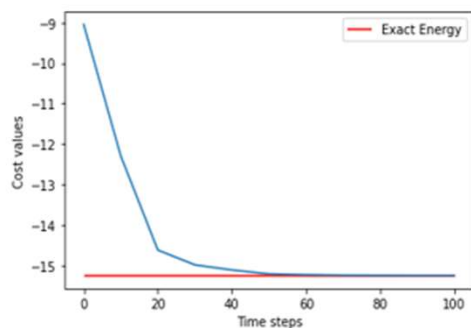
OVERPARAMETERIZATION & UNDERPARAMETERIZATION



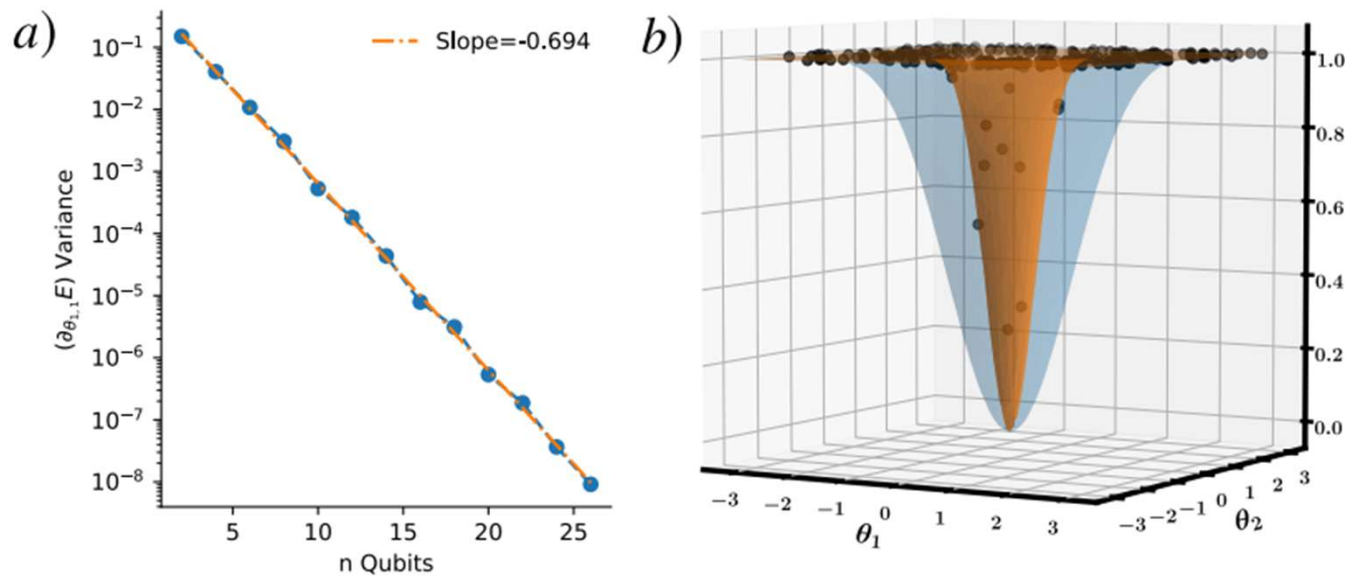
Depth=2

Total number of parameters $M \geq M_c$ (Critical no. of parameters)

OVERPARAMETERIZATION



BARREN PLATEAUS

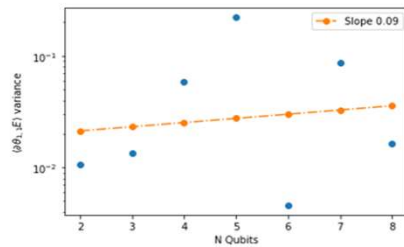


As the number of qubits increases, the landscape becomes flatter. Thus the cost values shrink **exponentially** with n

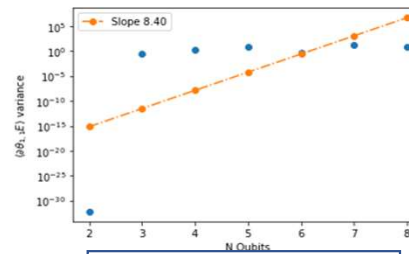
BARREN PLATEAUS

Gradient of Cost w.r.t the last variational parameter

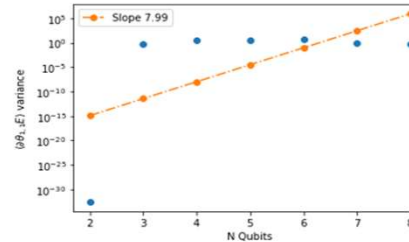
$$\partial_k C(\theta) = \frac{\partial_k C(\theta)}{\partial \theta_k}$$



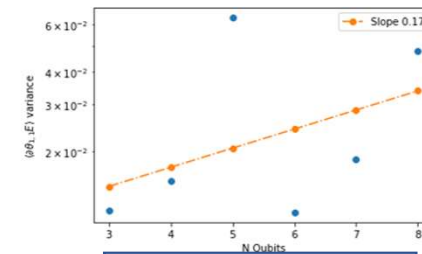
Depth=1, Type 1



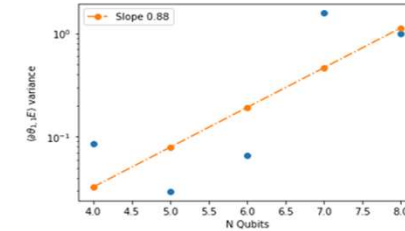
Depth=1, Type 2



Depth=2, Type 2



Depth=1, Type 3



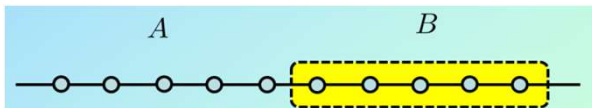
Depth=2, Type 3

N-qubit S-K model

As the number of qubits increases, the landscape becomes flatter. Thus the cost values shrink **exponentially** with n



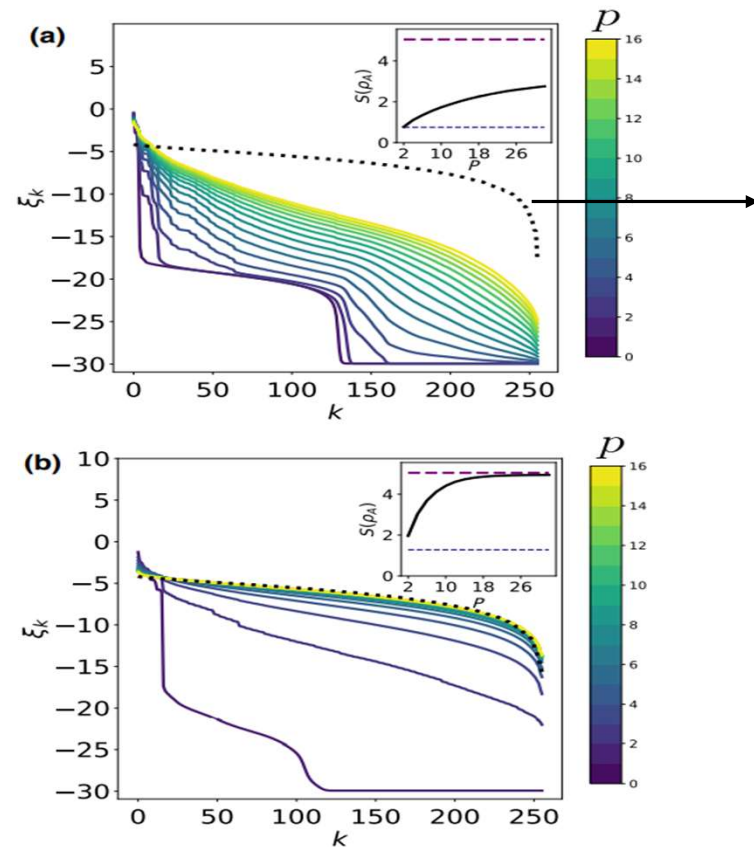
ENTANGLEMENT SPECTRUM



$$\rho_A(|\psi\rangle) = \text{Tr}_B(|\psi\rangle_{AB} \langle\psi|)$$

$$\{\xi_k\} = \{\log(p_k)\}$$

$$p_k = \text{Eigen}(\rho_A)$$



**MP
Distribution**

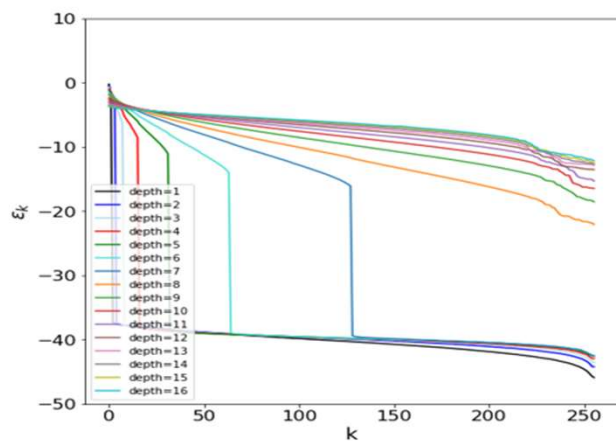
ENTANGLEMENT SPECTRUM



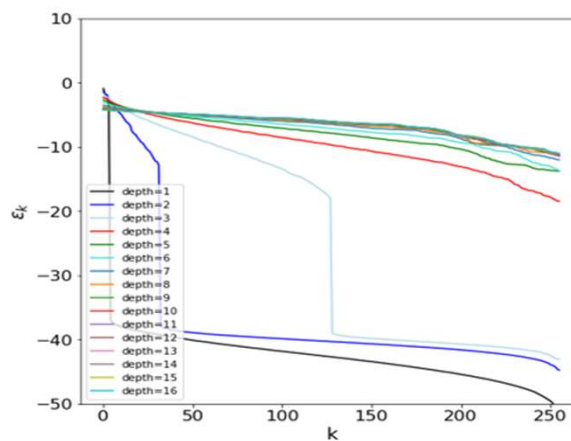
$$\rho_A(|\psi\rangle) = \text{Tr}_B(|\psi\rangle_{AB} \langle \psi|)$$

$$\{\xi_k\} = \{\log(p_k)\}$$

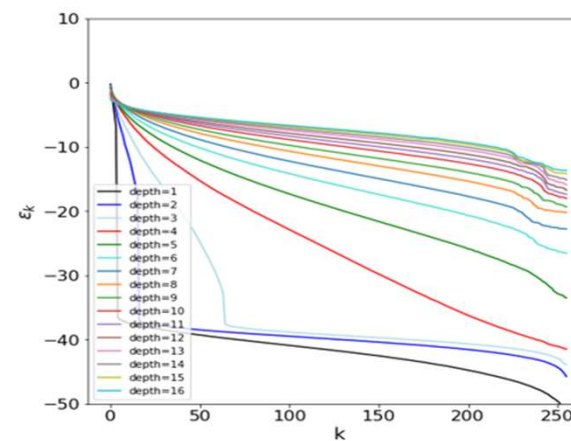
$$p_k = \text{Eigen}(\rho_A)$$



(a)



(b)



(c)

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FUTURE PLANS



- We would love to show, analytically, why the expressibility matters so much in some cases to have a more deeper understanding of such optimization problems.
- We would likely explore the concept of Quantum Fisher Information (QFI) in this context as QFI, in recent days, has emerged as a new frontier to explore the power of VQE.
- We would also explore some possible applications of Quantum Spin glass from graph theoretic approach.

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THANK YOU