

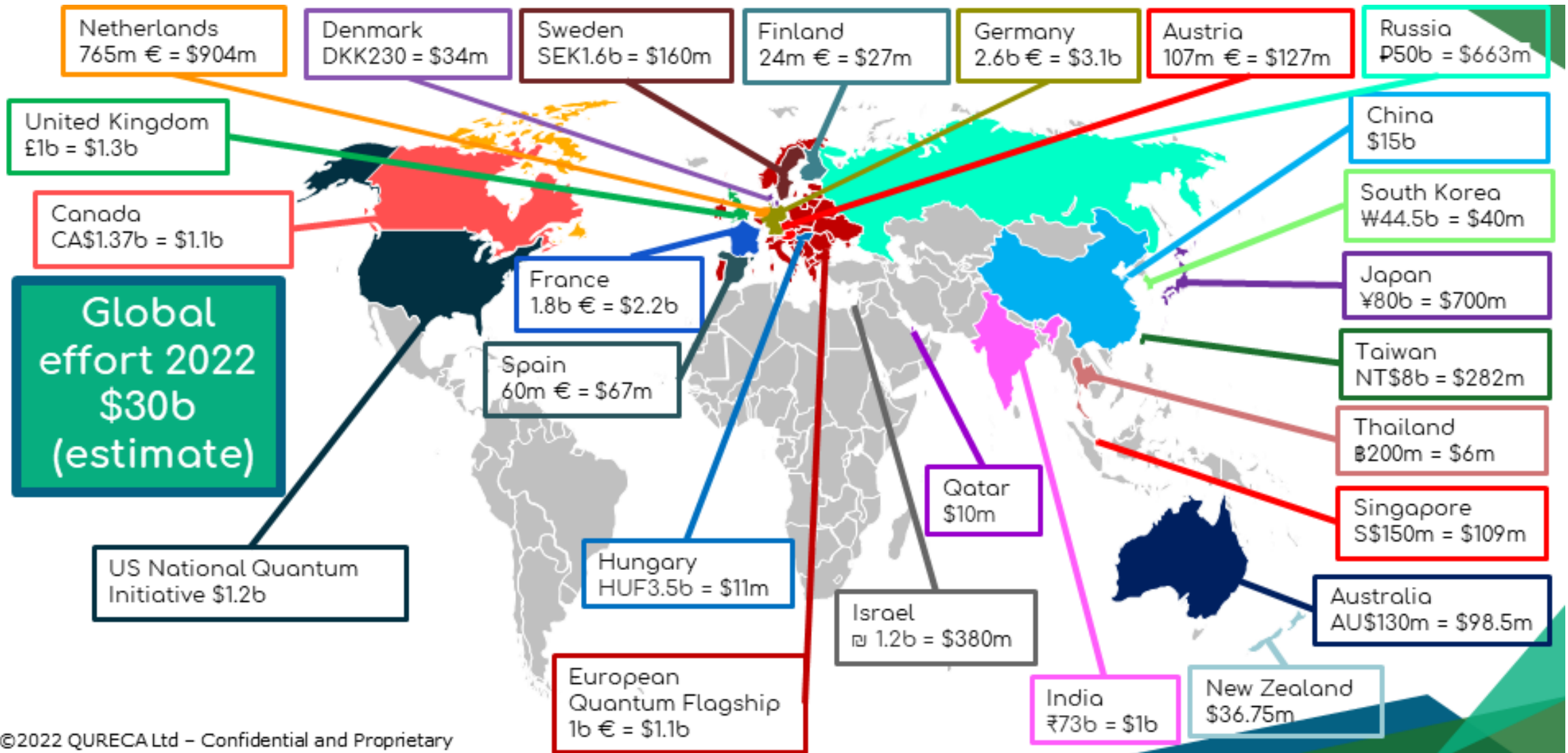


Alfred University
OUTSIDE of ORDINARY

Variational Quantum Search

Junpeng (Jim) Zhan, Ph.D.
Assistant Professor, Alfred University
zhanj@alfred.edu

Quantum Effort Worldwide



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Source: <https://qureca.com/overview-on-quantum-initiatives-worldwide-update-2022/>

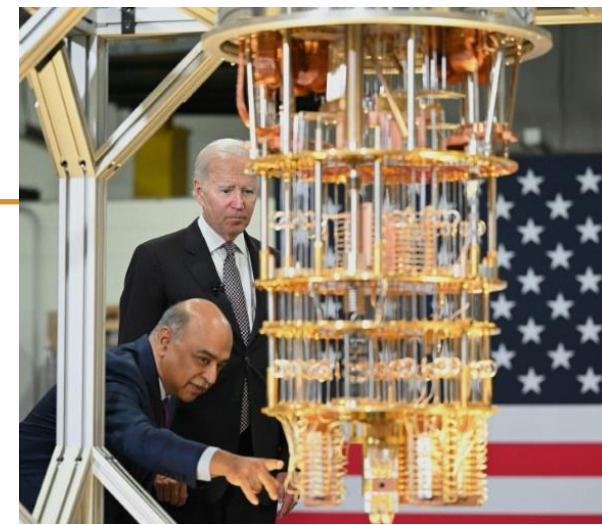
Quantum Computing Race - Corporations



Source: <https://hsrc.biz/reports/quantum-computing-market-technologies/>

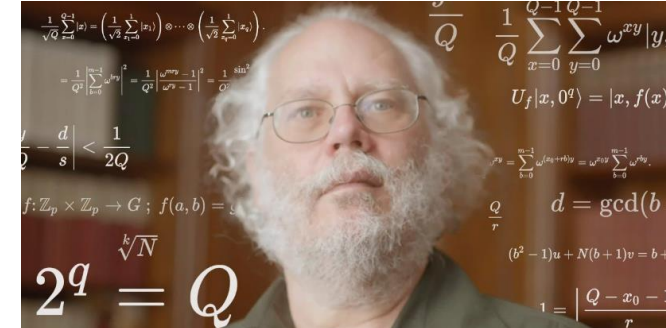
Introduction

- Quantum computers are developing rapidly.
 - IBM 433 (1121) qubits in Nov. 2022 (Dec. 2023)
 - IBM aims to reach 1386 qubits within two years
- Quantum Supremacy: Google quantum computer 200 seconds that would take the fastest supercomputers about 10,000 years?
- Variational quantum algorithms (VQAs) show great promise in helping to realize systematic *supremacy* of quantum computing
 - they typically require fewer qubits and lower circuit depth
- VQAs have been successfully used in many fields such as optimization, error correction, machine learning, physics and chemistry, etc.



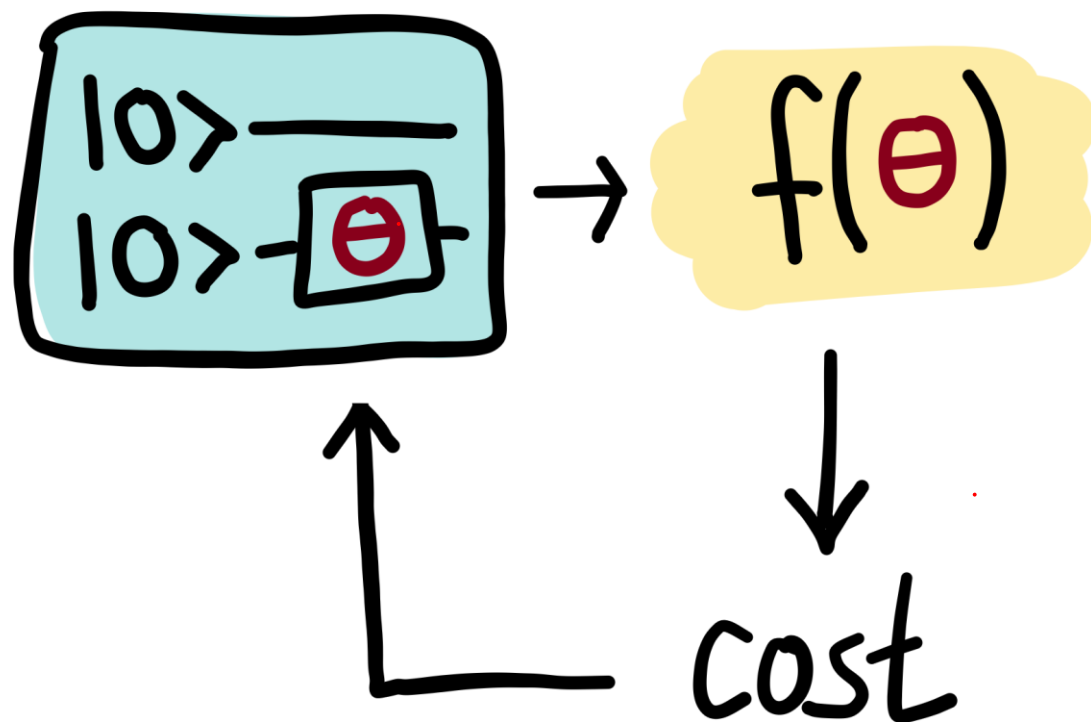
Introduction – Quantum Algorithms

- Pure quantum algorithms can be divided into two categories
 - based upon Shor's quantum Fourier transform
 - based upon Grover's algorithm to perform quantum search
- Grover's algorithm provides quadratic speedup, e.g., search unstructured database
 - The circuit depth increases exponentially with the number of qubits

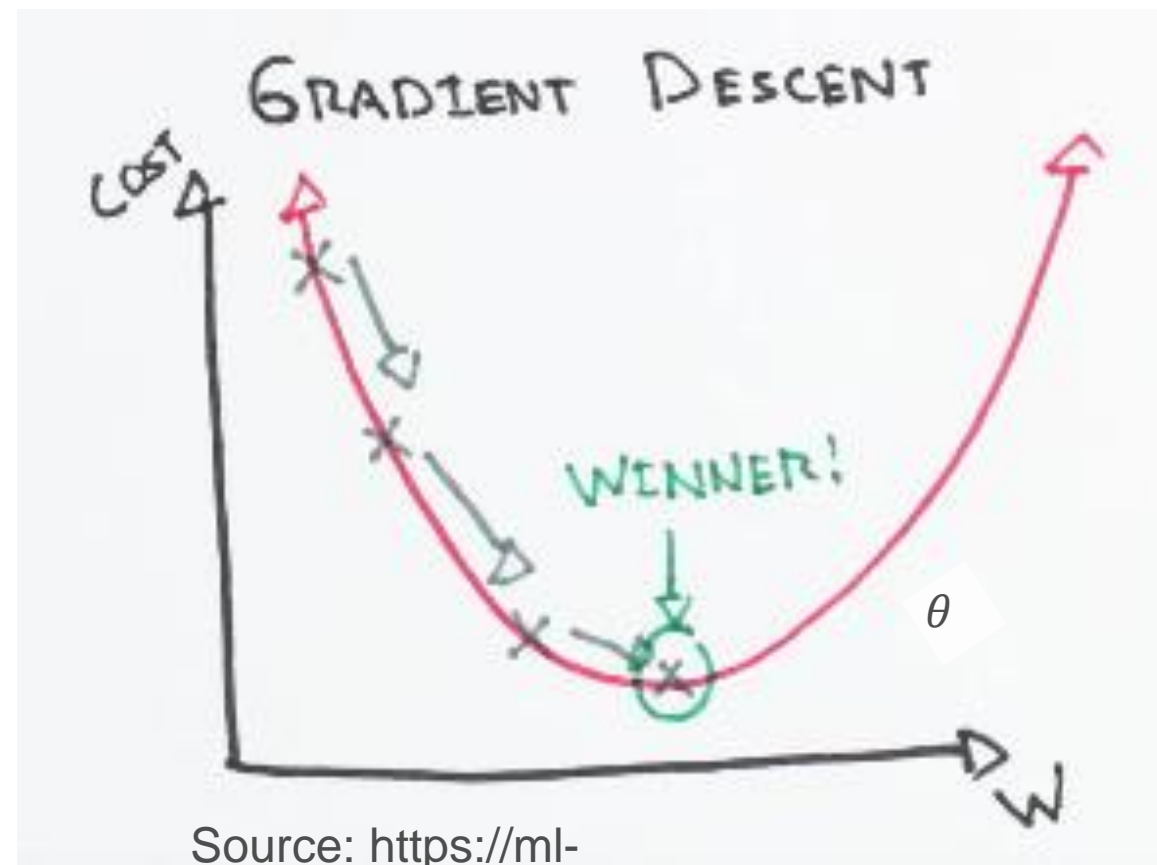


Introduction – Variational Quantum Algorithm

➤ Variational Quantum Algorithm



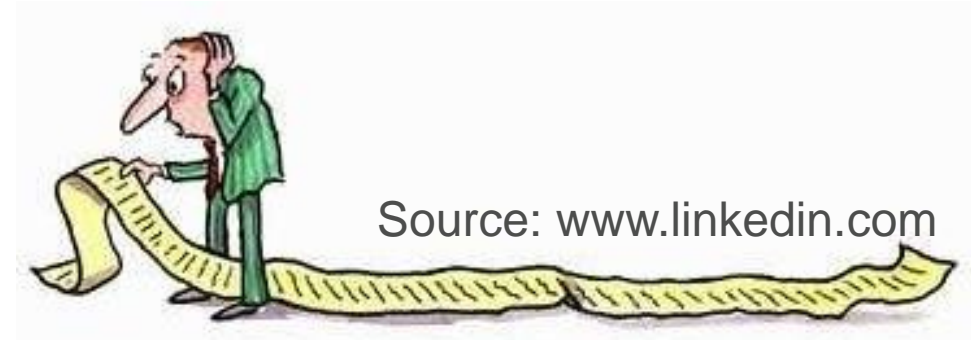
Source:
https://pennylane.ai/qml/glossary/variational_circuit.html



Source: https://ml-cheatsheet.readthedocs.io/en/latest/gradient_descent.html

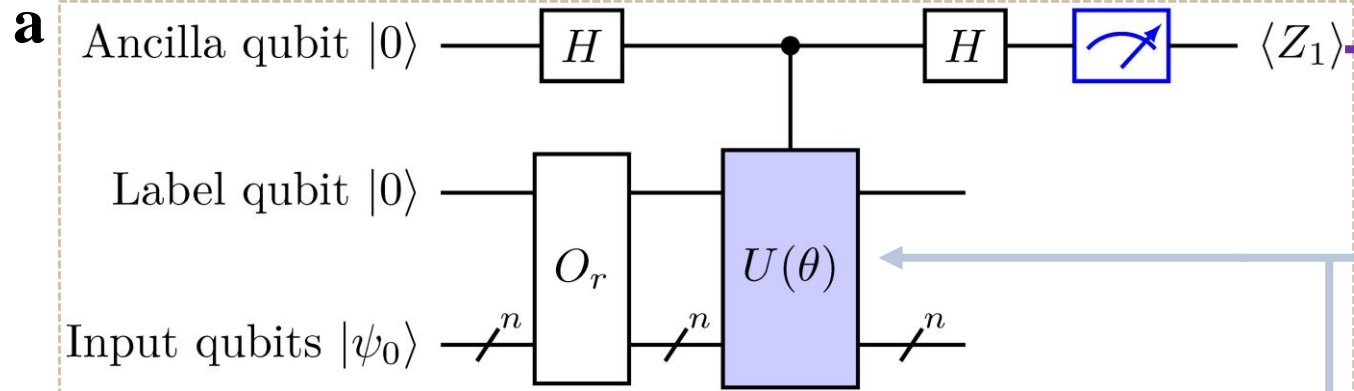
Search an Unstructured Database

- A database has 2^n elements
 - 266 qubits: $2^{266} = 10^{80}$
 - The known universe has 10^{80} atoms
- *Good element*: the element needs to be found
- **Classical** method need an average of $2^n/2$ operations
- Grover's search algorithm: $\sqrt{2^n}$ layers of **quantum** circuits
- Can we do better?
 - VQS algorithm: $O(n)$ layers of quantum circuits



Variational Quantum Search (VQS)

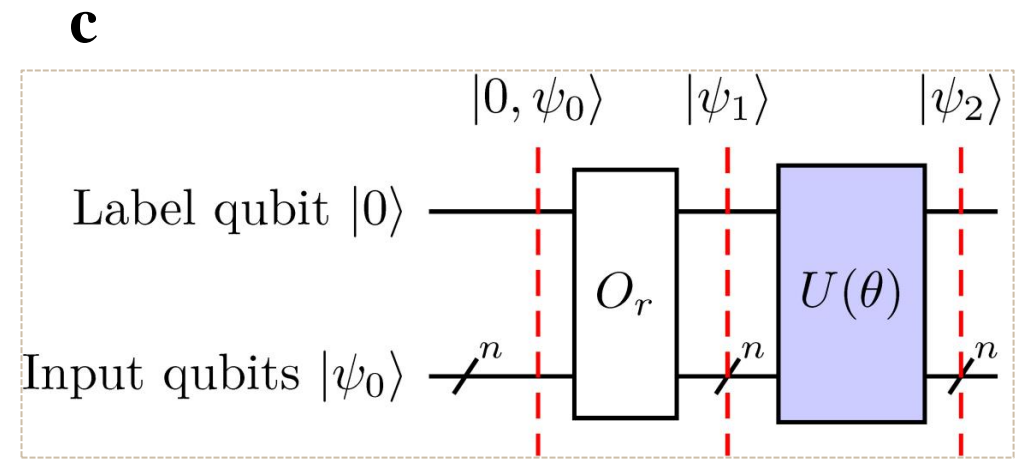
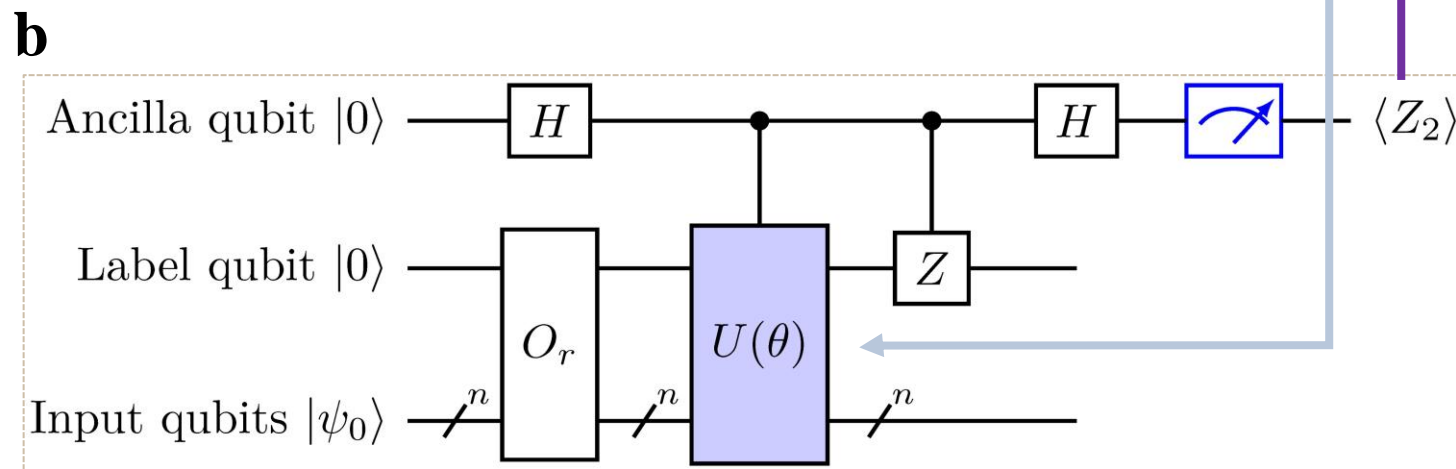
Quantum Computer



f Classical Computer

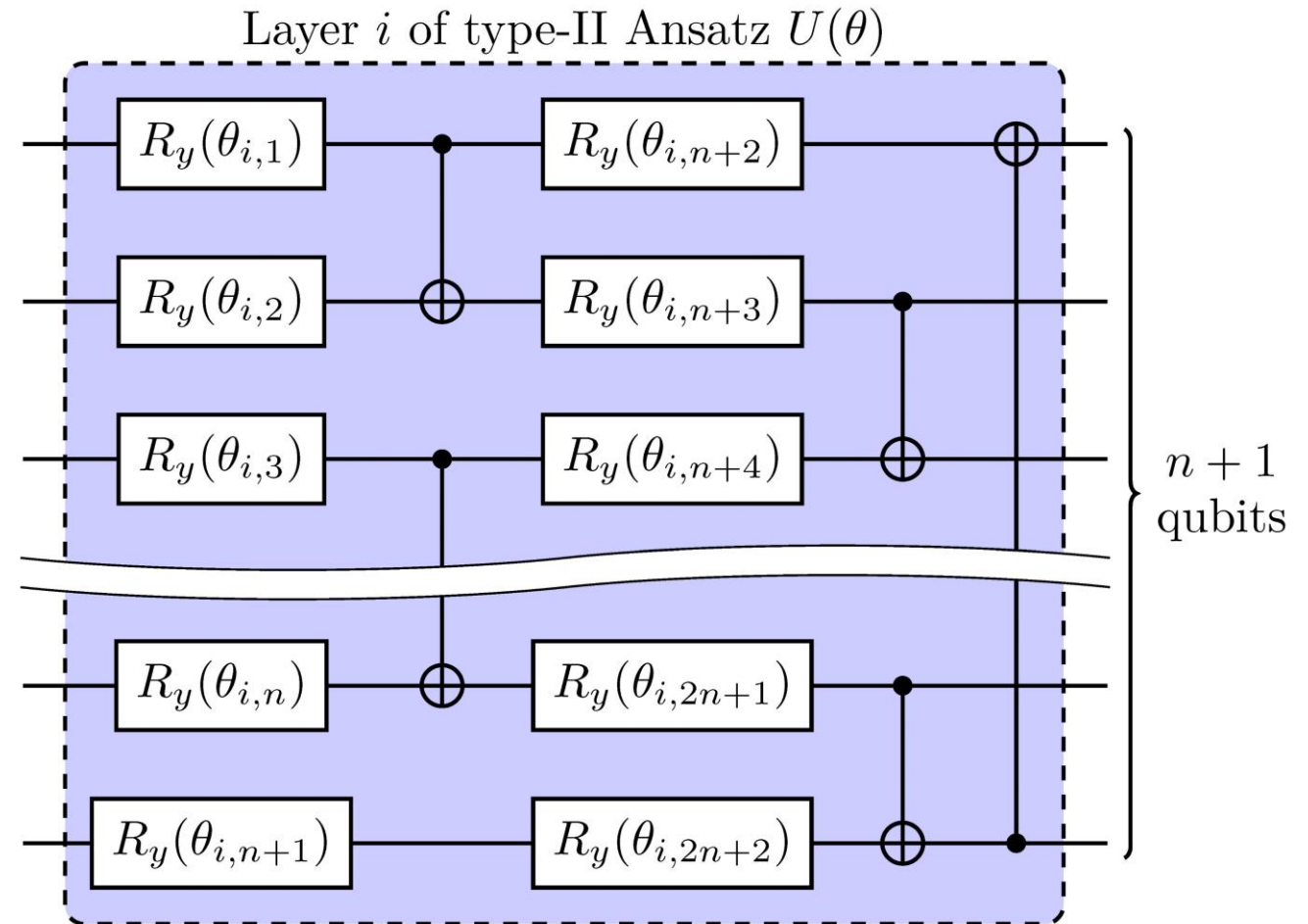
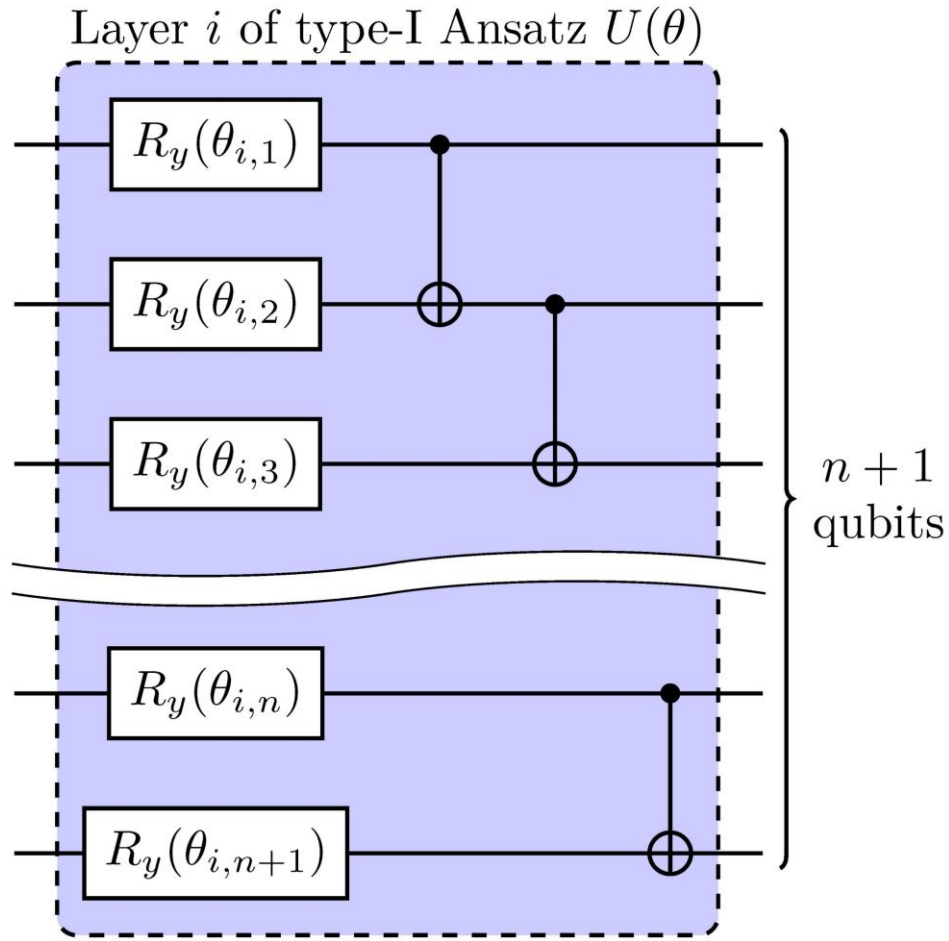
$$f(\theta^j) = -0.5(\langle Z_1 \rangle - \langle Z_2 \rangle)$$

$$\theta^{j+1} \leftarrow \text{Optimizer}(\theta^j, f(\theta^j))$$

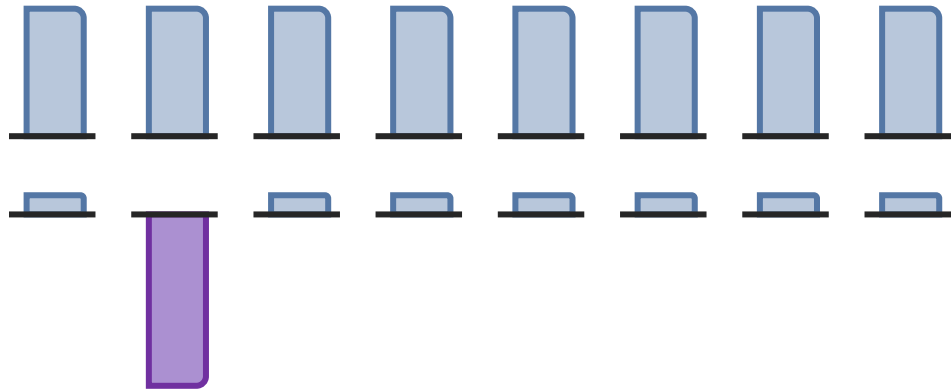


VQS - Ansatz

Hands-on, <https://quantum-computing.ibm.com/composer>



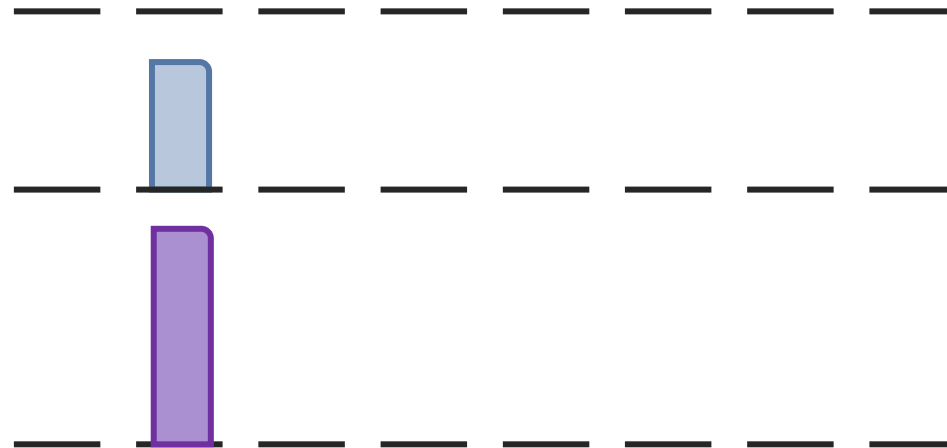
Search an Unstructured Database - 2



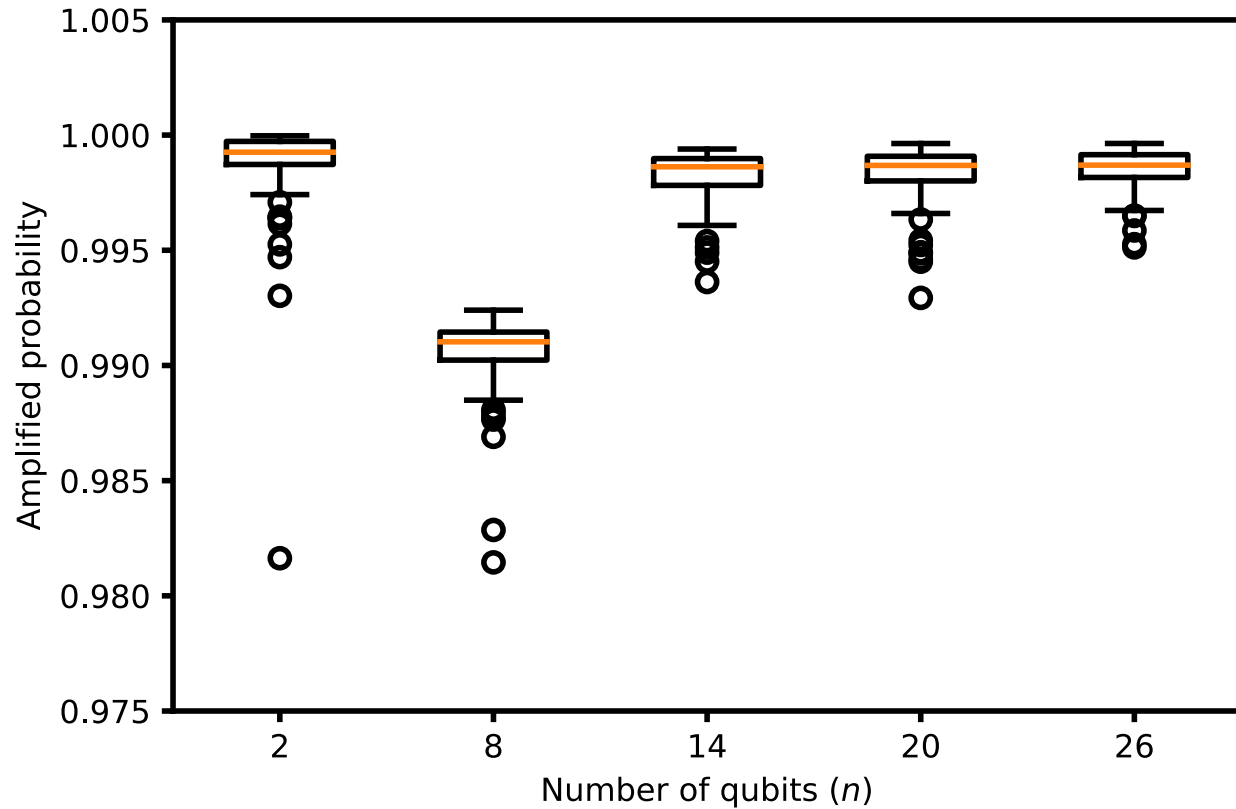
probability = amplitude²

Grover's search algorithm

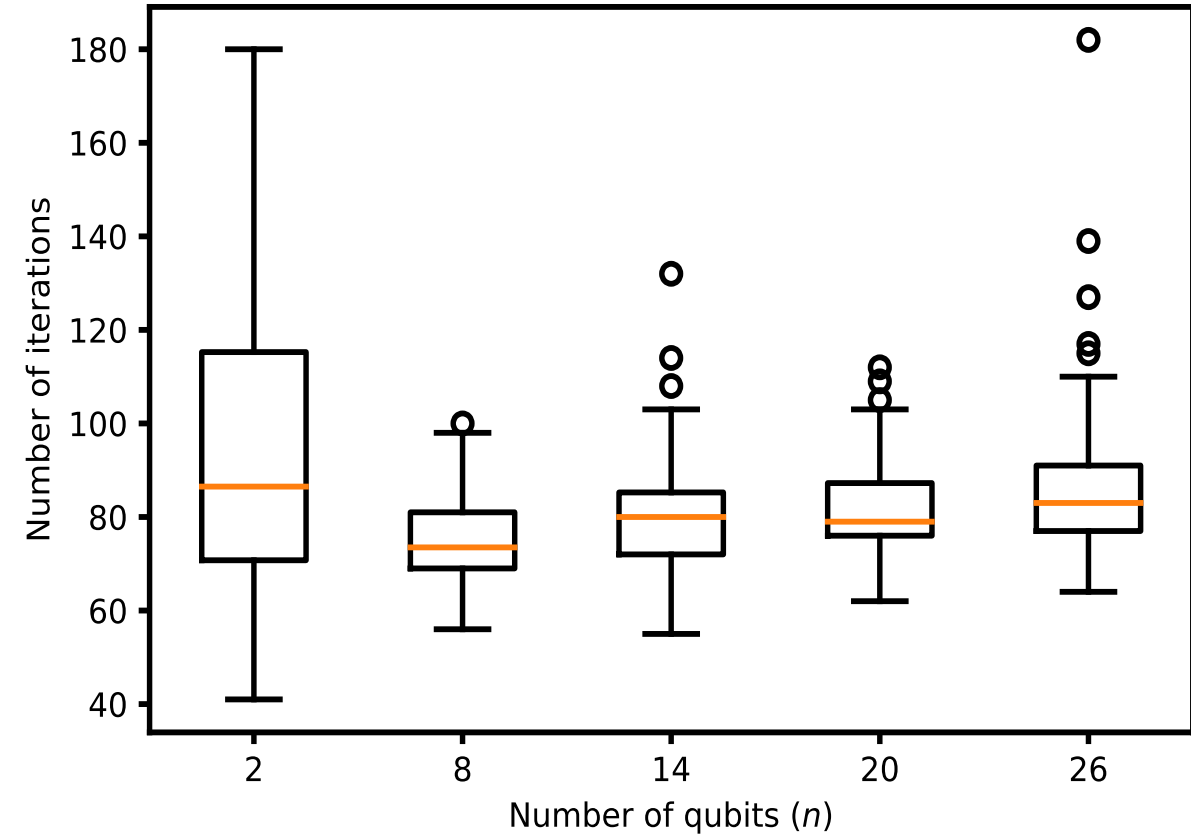
Variational Quantum Search (VQS) algorithm



VQS – Results Using Type-I Ansatz



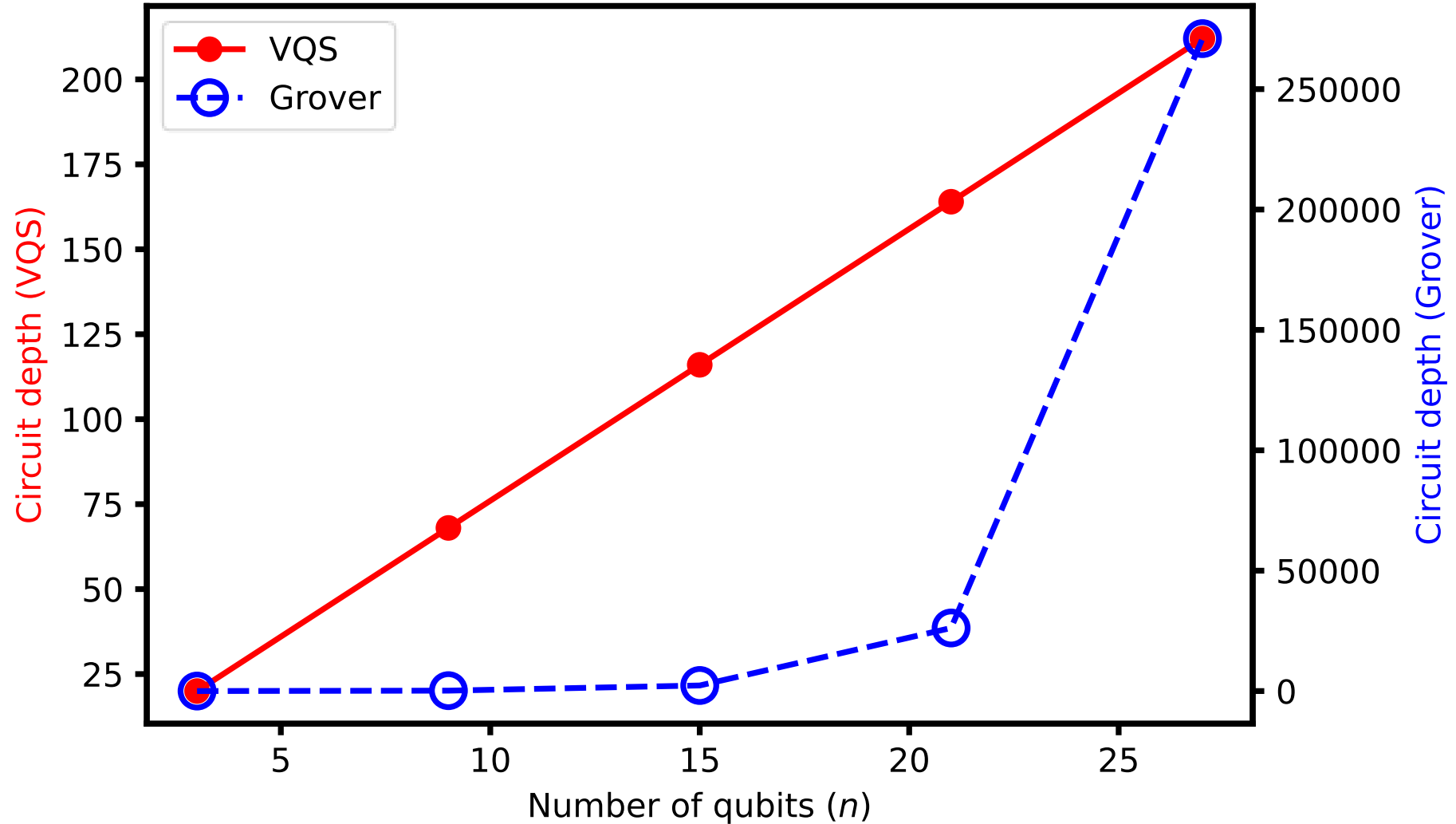
Probability of finding the good element



Number of iterations used to find the optimal parameters in Ansatz

VQS vs. Grover's Algorithm: Circuit Depth

- Exponential advantage!!
- Verified up to 26 qubits due to limitation of GPU memory
- Significant milestone if proved for any number of qubits!



Summary

- Proposed a Variational Quantum Search (VQS) algorithm [15]
 - The maximum circuit depth of VQS **increases linearly** with the qubit #
 - ✓ **Exponentially shallower** than Grover's algorithm
 - ✓ Verified up to 26 qubits, due to limitation of GPU memory
 - We demonstrate that a depth-56 circuit in VQS can replace a depth-270,989 circuit in **Grover's algorithm**
- Proposed Quantum Feasibility Labeling algorithm [17]
 - ✓ Together with VQS, can **efficiently solve** an NP-complete Vertex Coloring problem

Future Work

- Work on the objective function and **optimization** method
 - to determine the optimal parameters in the Ansatz (transfer learning)
 - to avoid barren plateau
 - to analyze the local optima and global optimum of the solution space
 - to propose more efficient optimization method
- Validate VQS on larger systems (circuit cutting, MPS, etc.)
- Apply VQS to solve other difficult & interesting problems
 - Optimization problems in power and energy systems, e.g., unit commitment, optimization of DER and energy storage systems, contingency analysis

Acknowledgement

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- ACCESS (advanced computing and data resource supported by the NSF)
- IBM Qiskit
- Xanadu PennyLane
- Visiting Ph.D. student: Yalin Liao, University of Delaware

Reference Books/Resources

- Michael A. Nielsen, Isaac L. Chuang. Quantum Computation and Quantum Information: 10th Anniversary Edition
- David Mermin. Quantum computer science: An Introduction
- Xanadu Quantum Codebook [<https://codebook.xanadu.ai/>]
- IBM Quantum Composer [<https://quantum-computing.ibm.com/composer>]

VQS - States in Vector Form

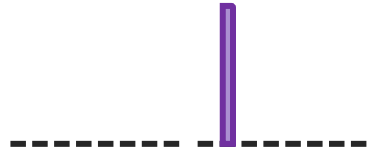
➤ $|0, \psi_0\rangle = \underbrace{[\alpha_{b_1}, \alpha_{b_2}, \dots, \alpha_{b_{N_b}}, \alpha_{g_1}, \alpha_{g_2}, \dots, \alpha_{g_{N_g}}]}_{\text{1st half: } N \text{ elements}}, \underbrace{[0, 0, \dots, 0]}_{\text{2nd half: } N \text{ elements}}^T$



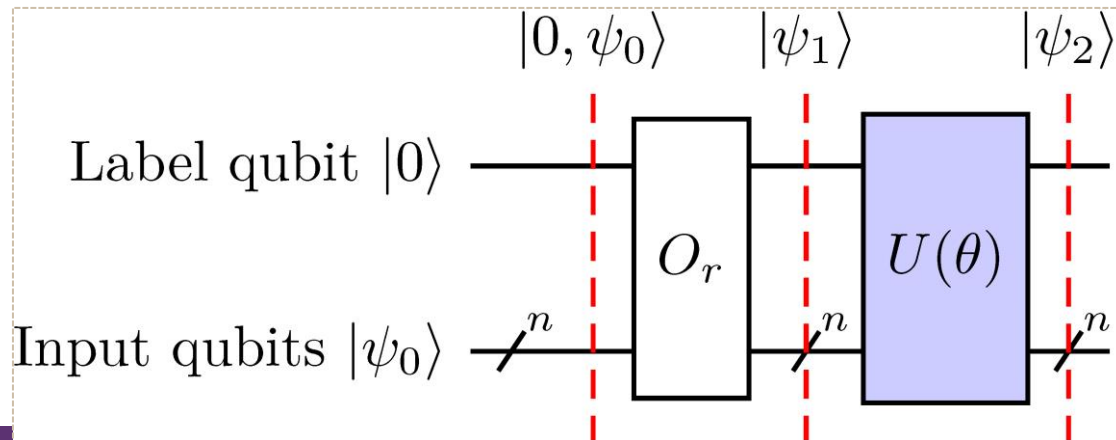
➤ $|\psi_1\rangle = \underbrace{[\alpha_{b_1}, \alpha_{b_2}, \dots, \alpha_{b_{N_b}}, 0, \dots, 0]}_{\text{1st half: } N \text{ elements}}, \underbrace{[0, \dots, 0, \alpha_{g_1}, \alpha_{g_2}, \dots, \alpha_{g_{N_g}}]}_{\text{2nd half: } N \text{ elements}}^T$



➤ $|\psi_2\rangle = [\beta_1, \beta_2, \dots, \beta_N, \beta_{N+1}, \dots, \beta_{2N-1}, \beta_{2N}]^T$

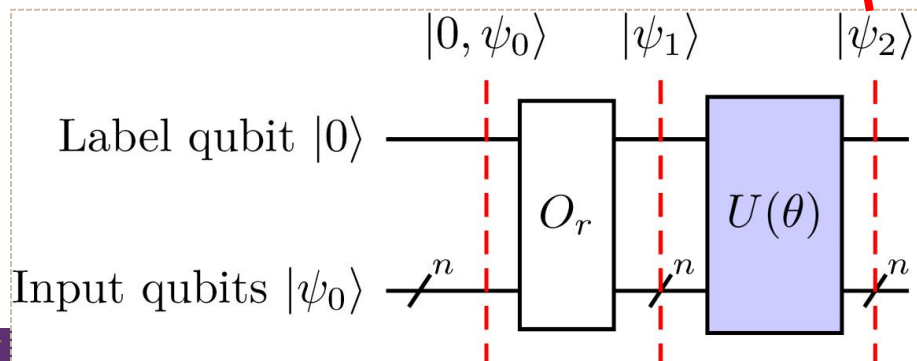



➤ $\sum_{i=1}^{2N} |\beta_i|^2 = 1$



VQS - Objective Function

- The **objective function** is set to $f(\boldsymbol{\theta}) = -0.5(\langle Z_1 \rangle - \langle Z_2 \rangle)$
- This function ensures that when $f(\boldsymbol{\theta})$ is minimized,
 - $|\psi_2\rangle$ is only a superposition of good elements
 - measuring $|\psi_2\rangle$ will obtain good elements with probabilities close to 1



f **Classical Computer** 

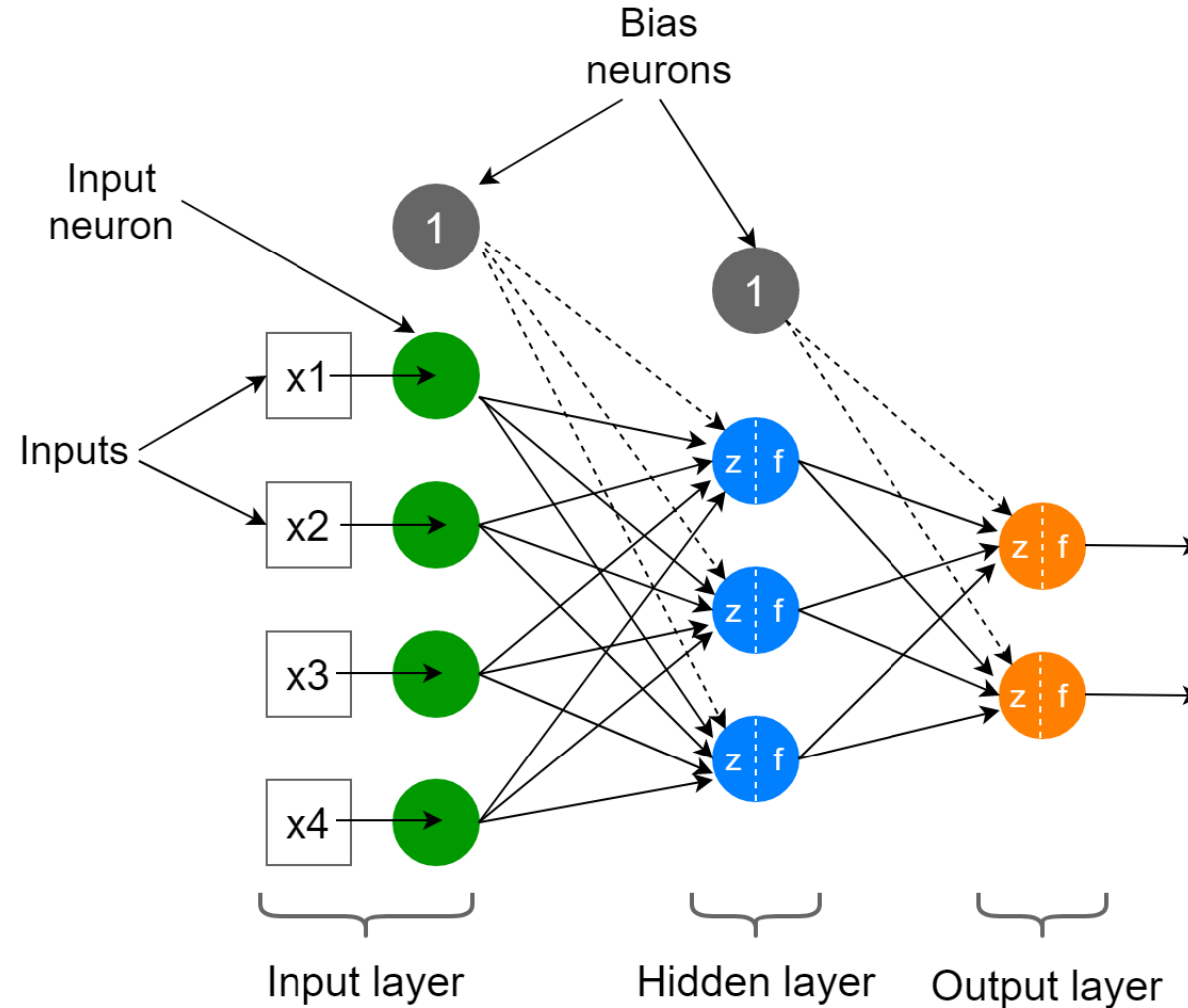
$$f(\boldsymbol{\theta}^j) = -0.5(\langle Z_1 \rangle - \langle Z_2 \rangle)$$

$\boldsymbol{\theta}^{j+1} \leftarrow \text{Optimizer}(\boldsymbol{\theta}^j, f(\boldsymbol{\theta}^j))$



Classical Neural Network

- Input layer
- Hidden layers
- Output layer



Source:
<https://rukshanpramoditha.medium.com/one-hidden-layer-shallow-neural-network-architecture-d45097f649e6>

