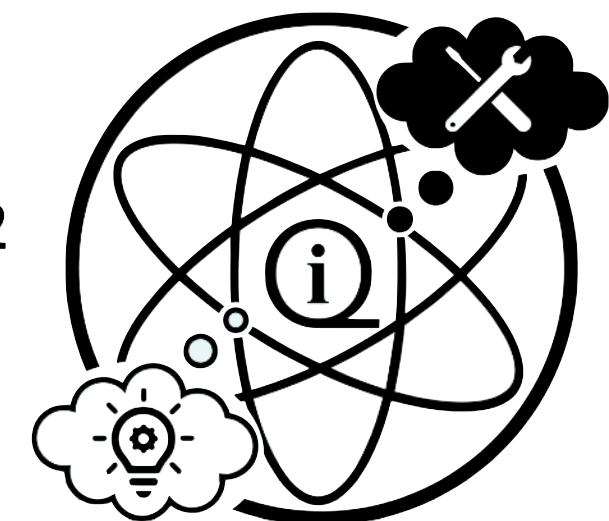


Quantum variational learning for entanglement witnessing

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VCQ and
AppQInfo 2022
Summer
School



Concepts &
Applications of
Quantum
Information



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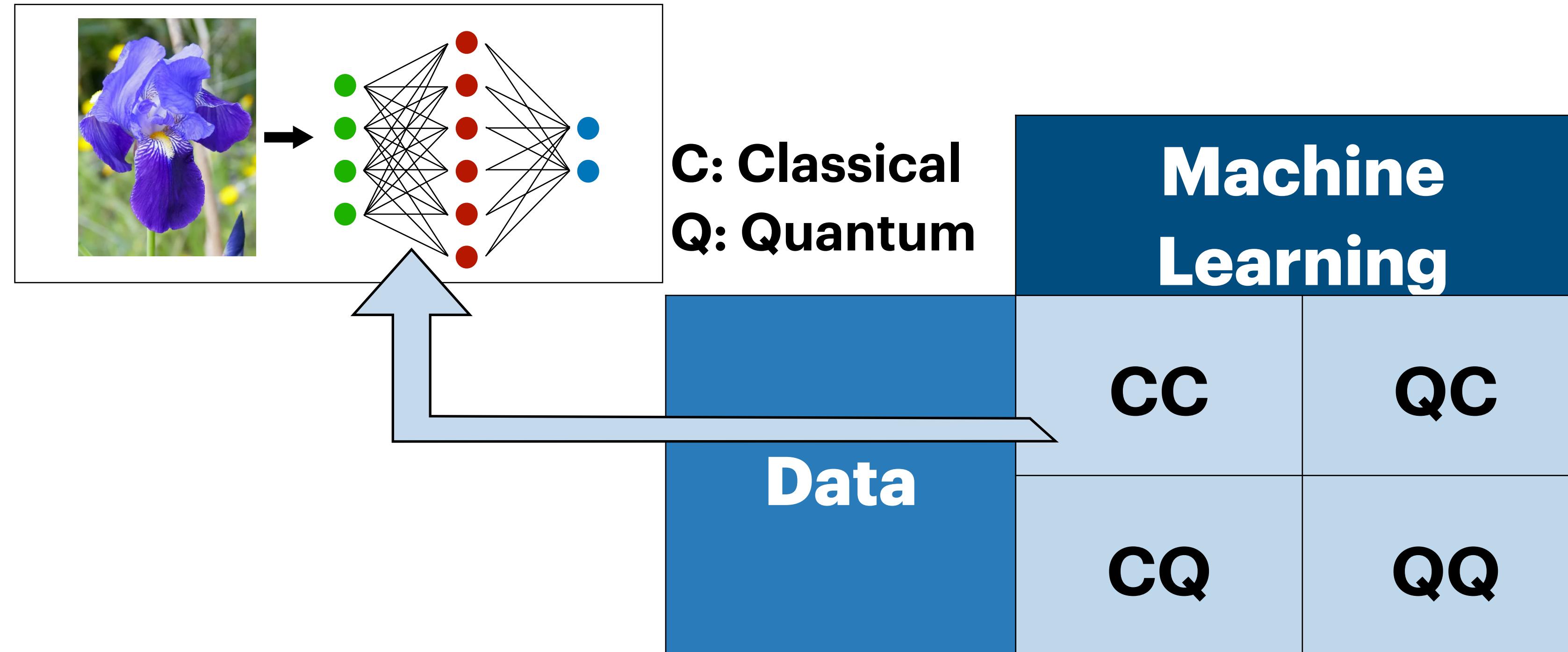
Machine Learning and Data

Different approaches

C: Classical	Machine Learning	
Q: Quantum	CC	QC
Data	CQ	QQ

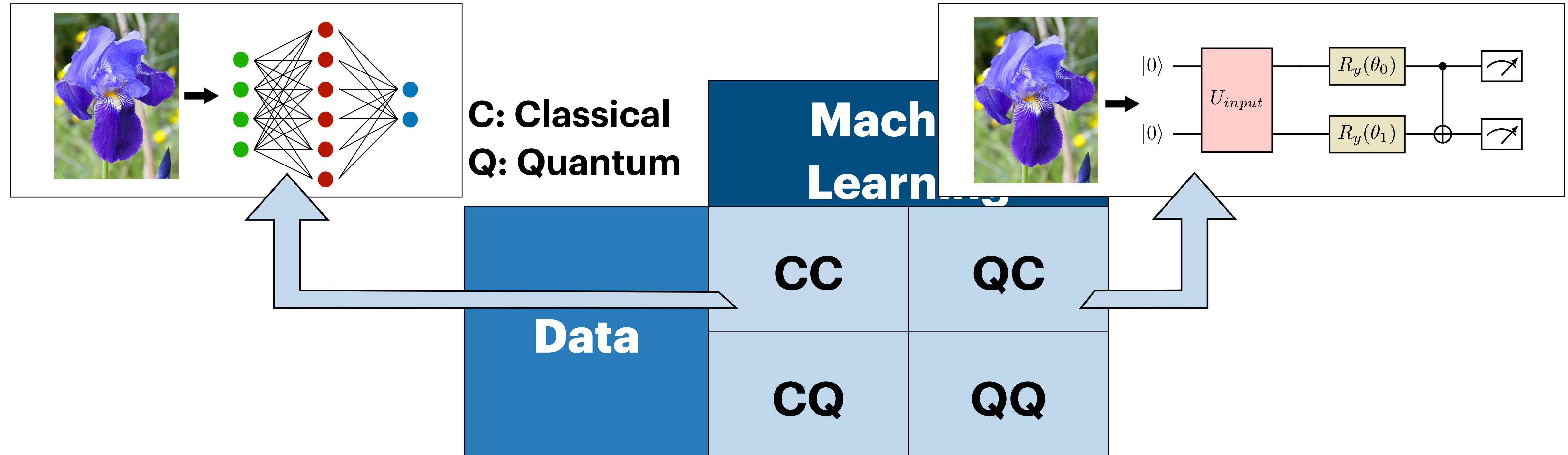
Machine Learning and Data

Different approaches



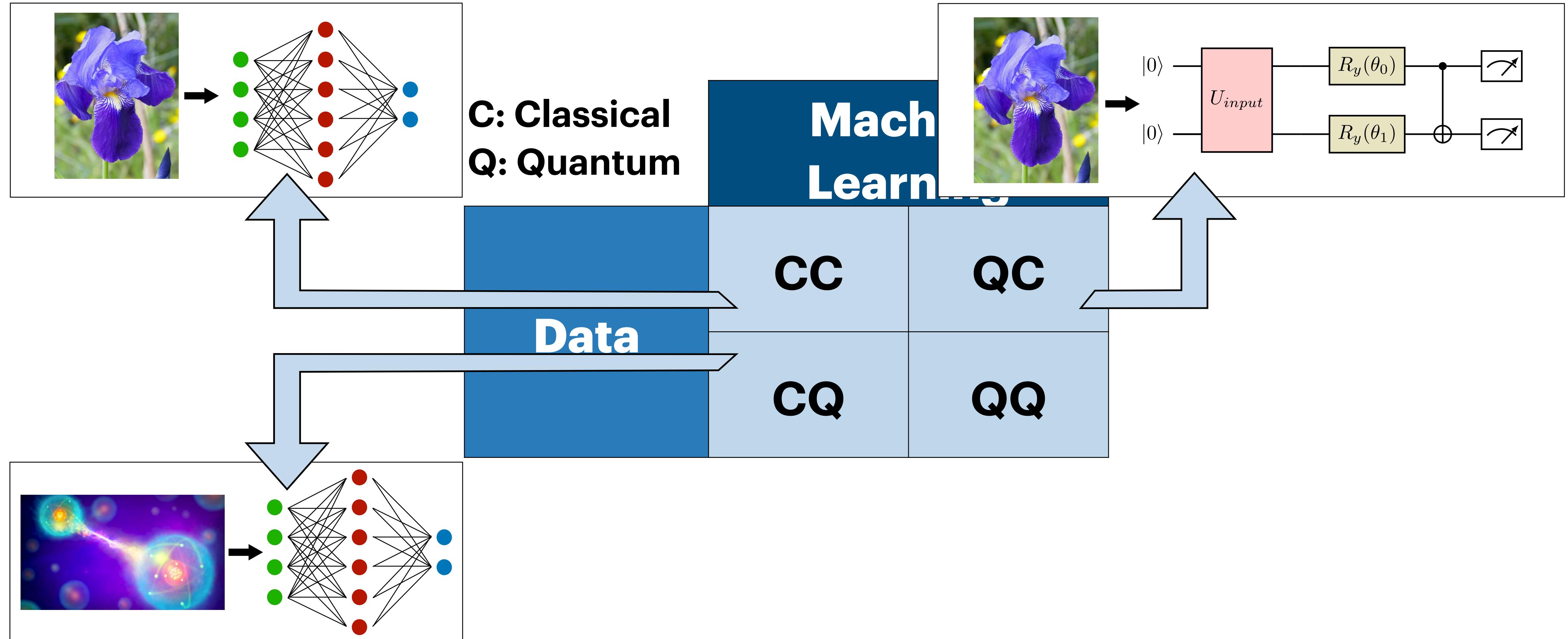
Machine Learning and Data

Different approaches



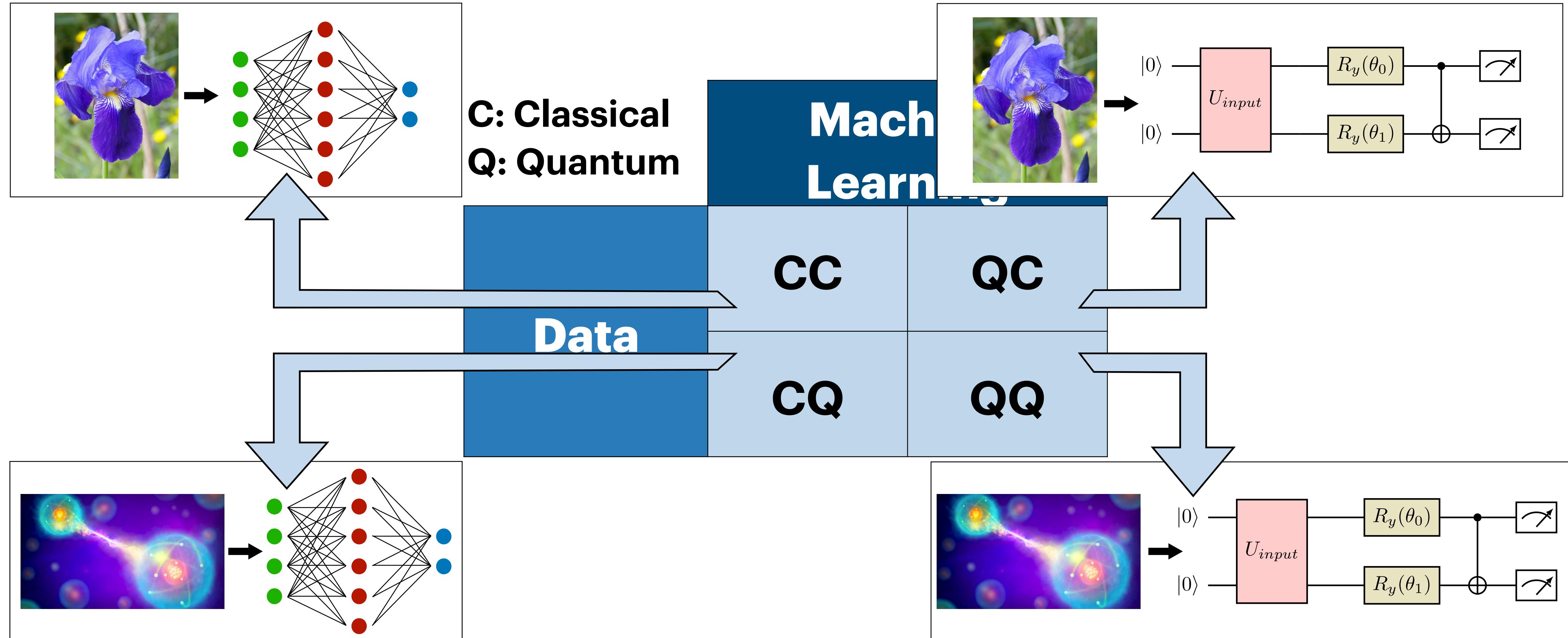
Machine Learning and Data

Different approaches



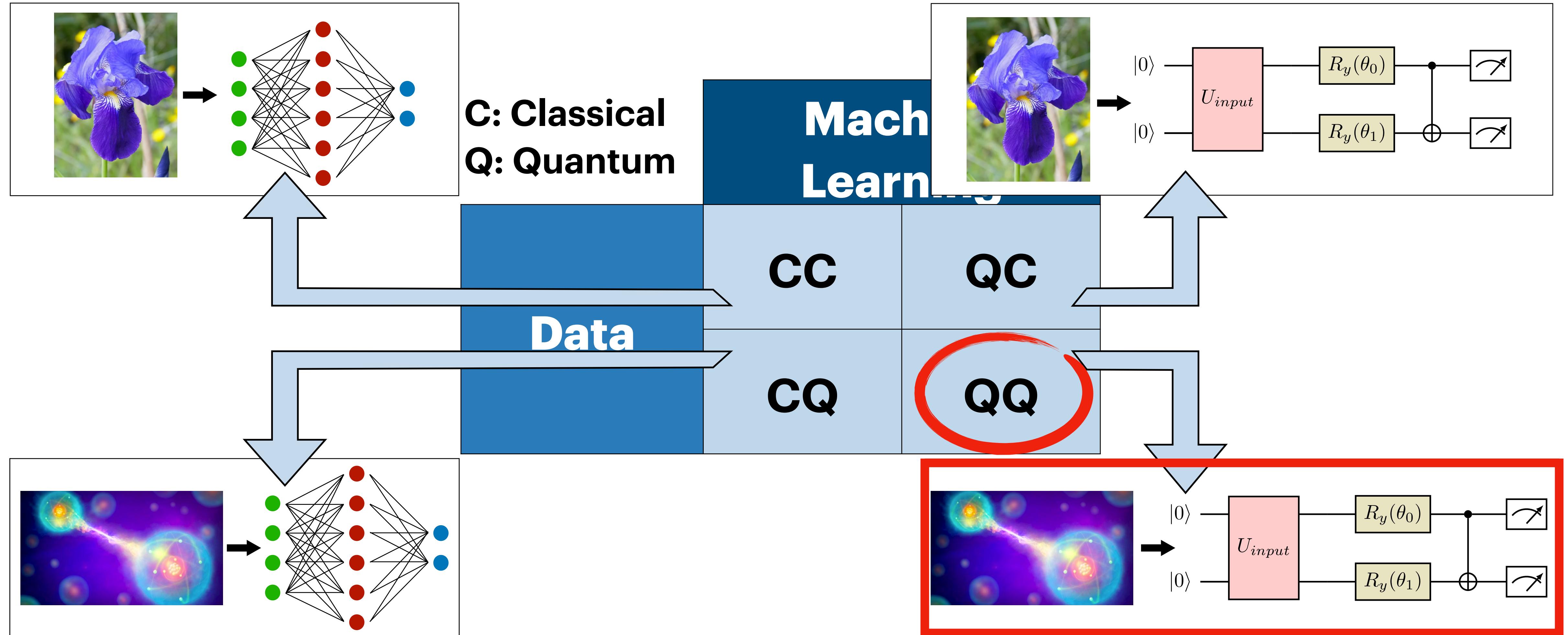
Machine Learning and Data

Different approaches



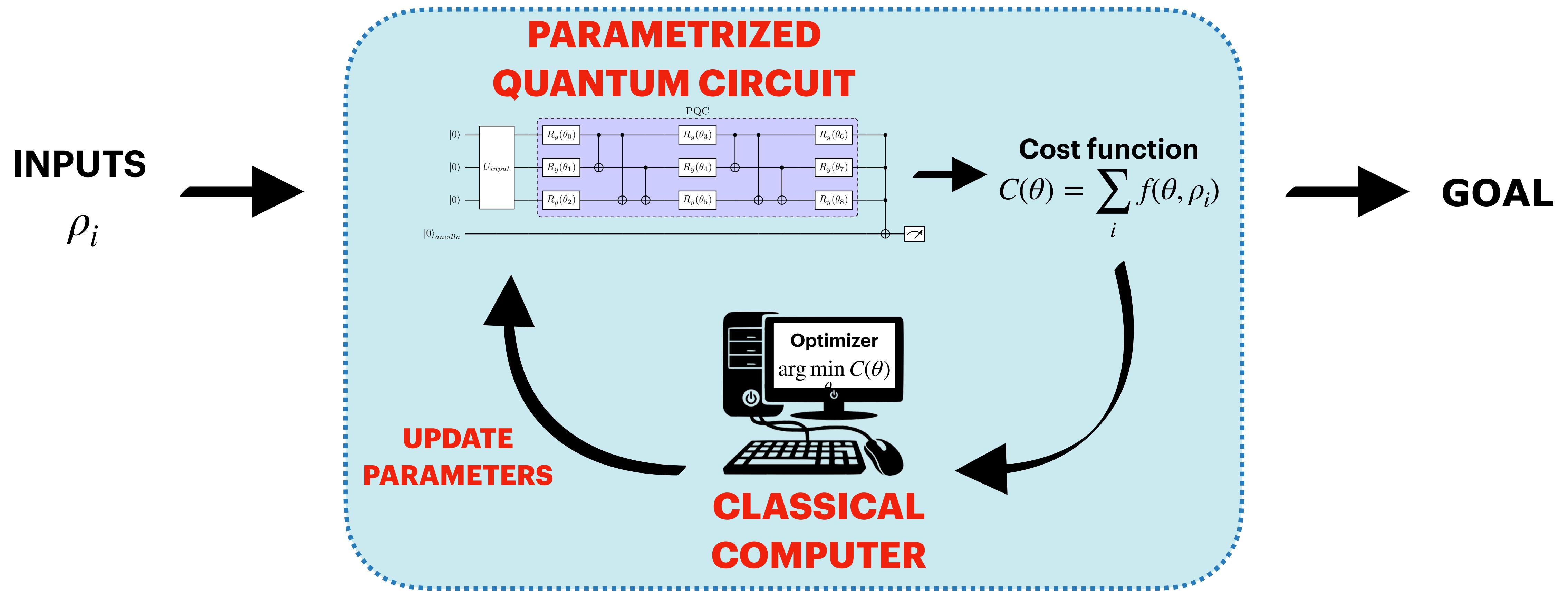
Machine Learning and Data

Different approaches



Variational Quantum Algorithms¹ (VQAs)

Are hybrid

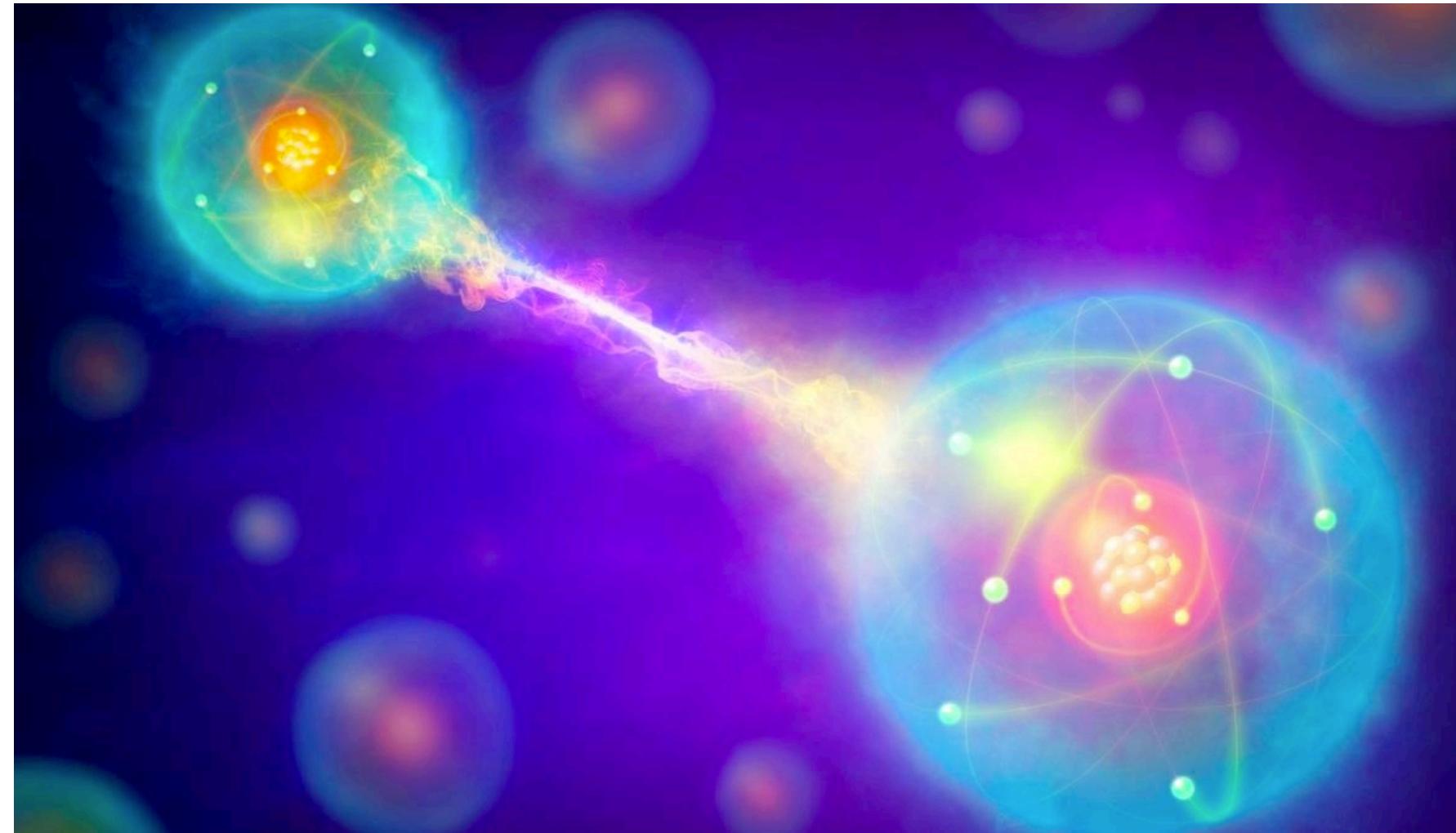


Quantum Entanglement

A **non-classical** correlation

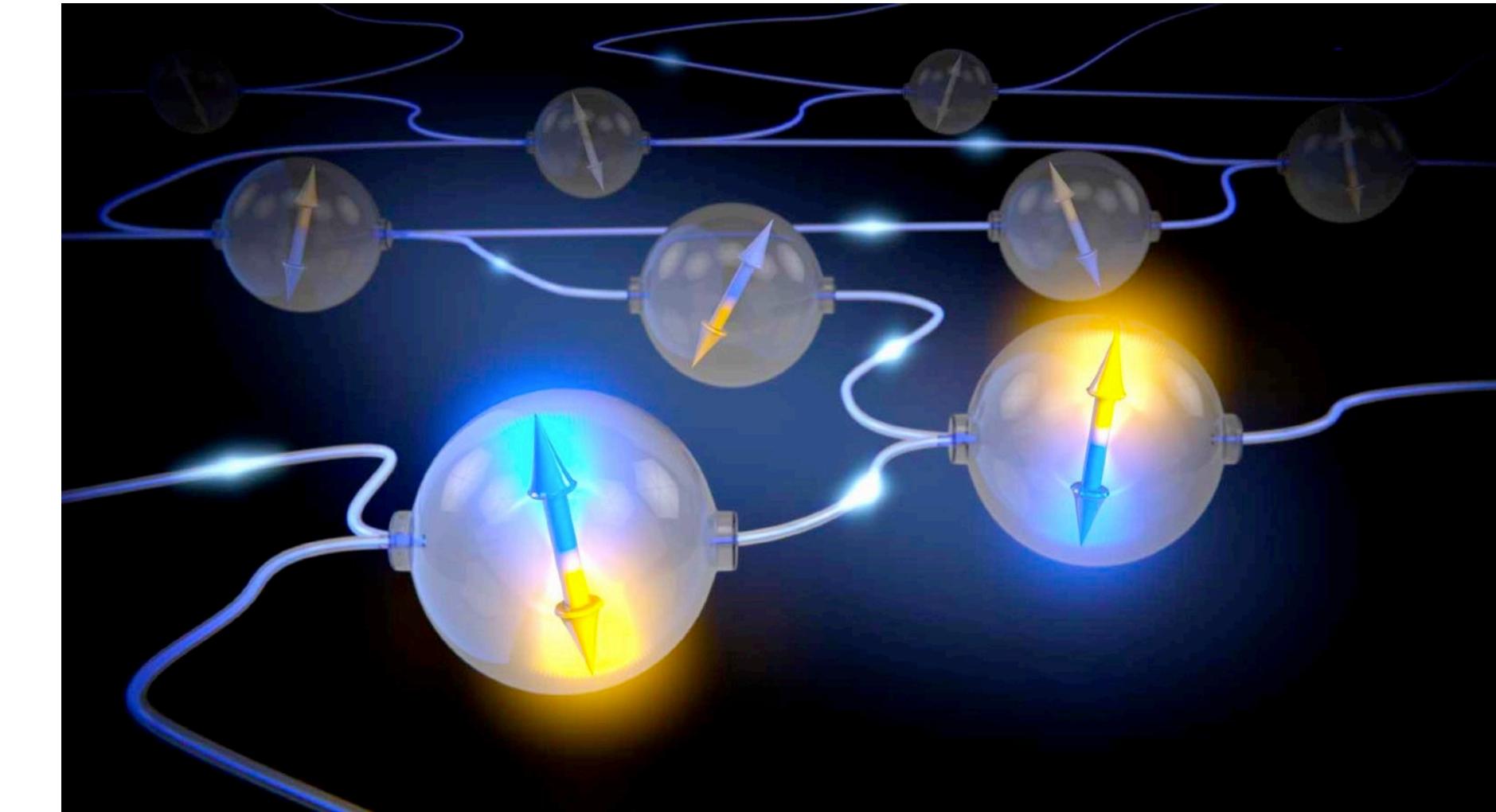
A genuine quantum mechanical property of physical systems

Bipartite entanglement



Shared by two systems

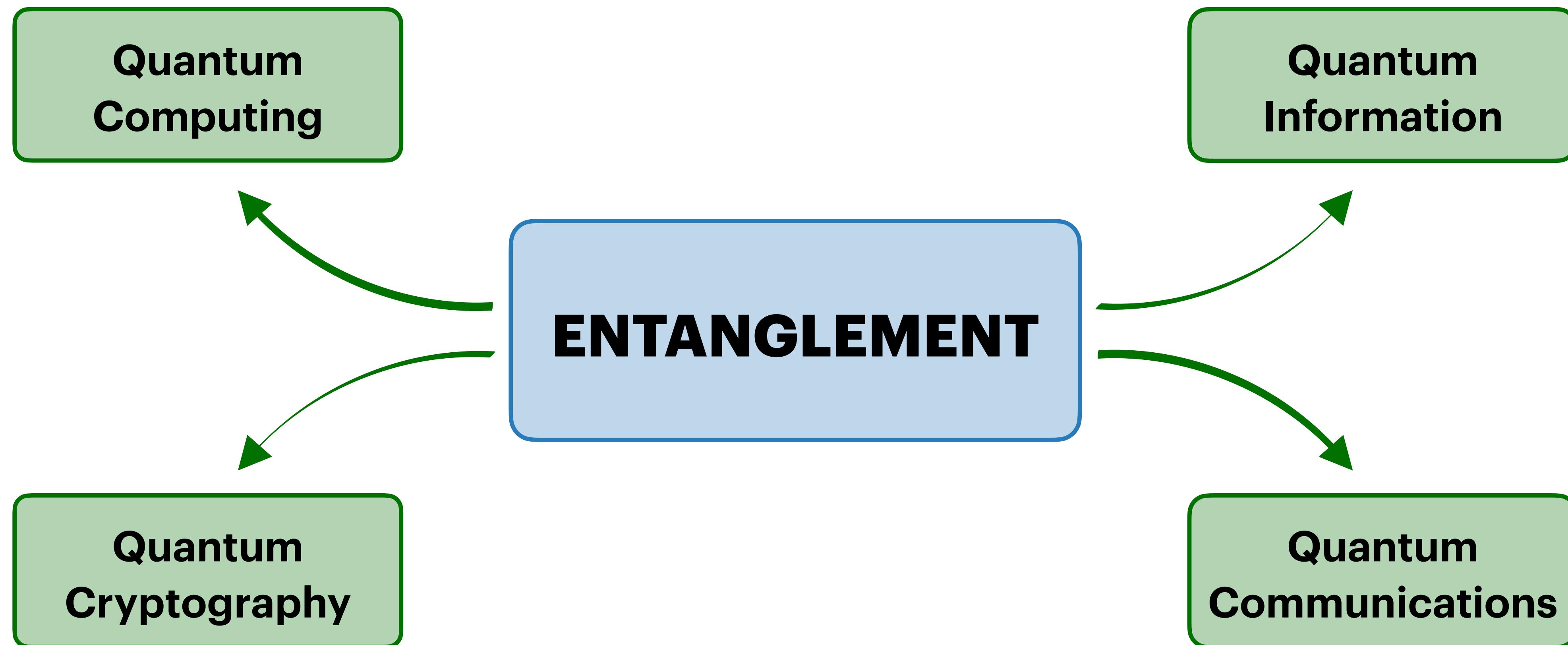
Multipartite entanglement



Shared by multiple systems

Why seeking for entanglement?

It is a fundamental **resource**



How to detect entanglement?

Exponential difficulties

Classical methods:

HARD to simulate large quantum systems



$\dim(\mathcal{H}) = 2^N$ for N qubits

Quantum methods:

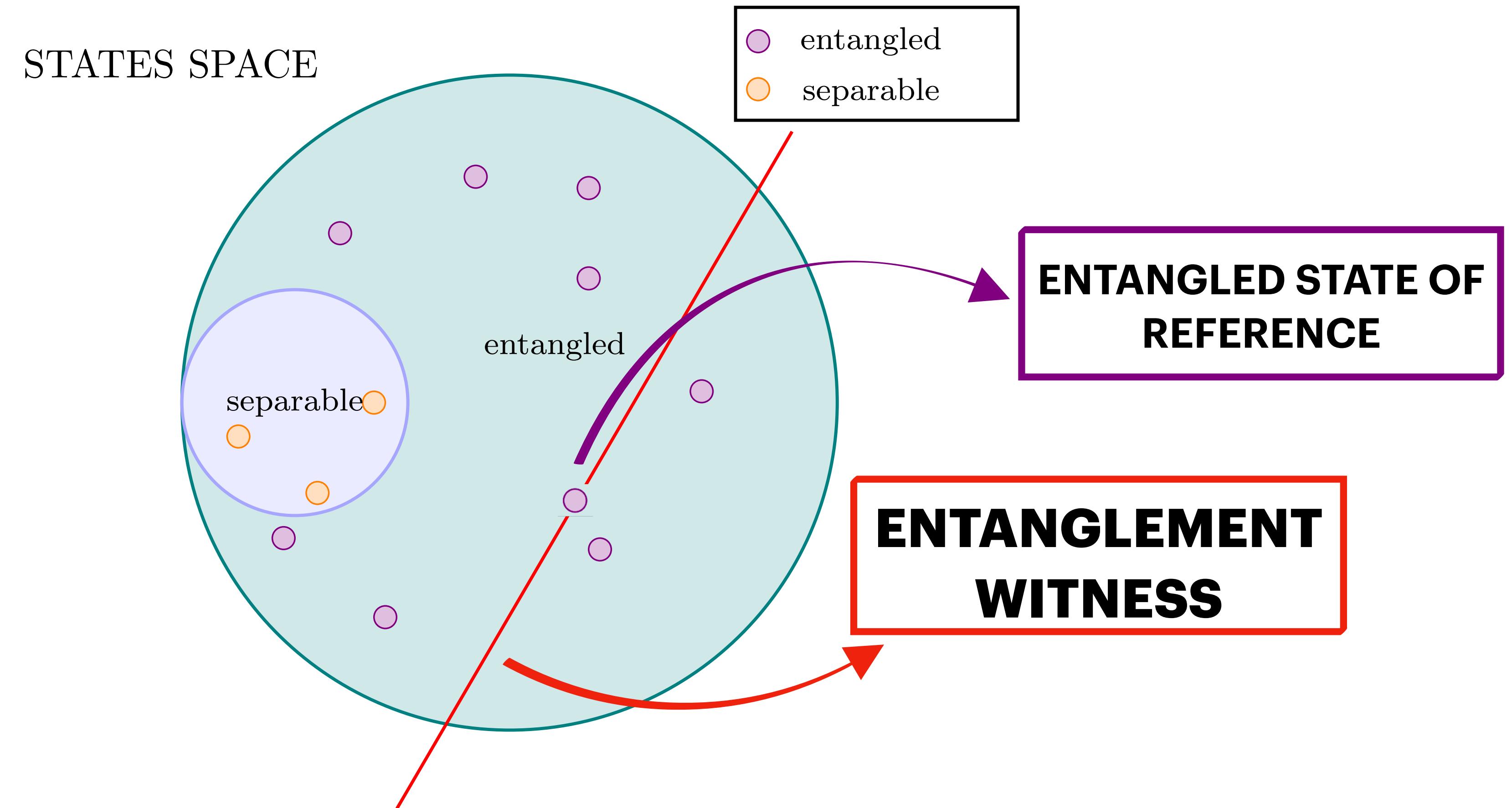
Quantum State Tomography



3^N quantum measurements for N qubits

How to detect entanglement?

Projective entanglement witness



How to detect entanglement?

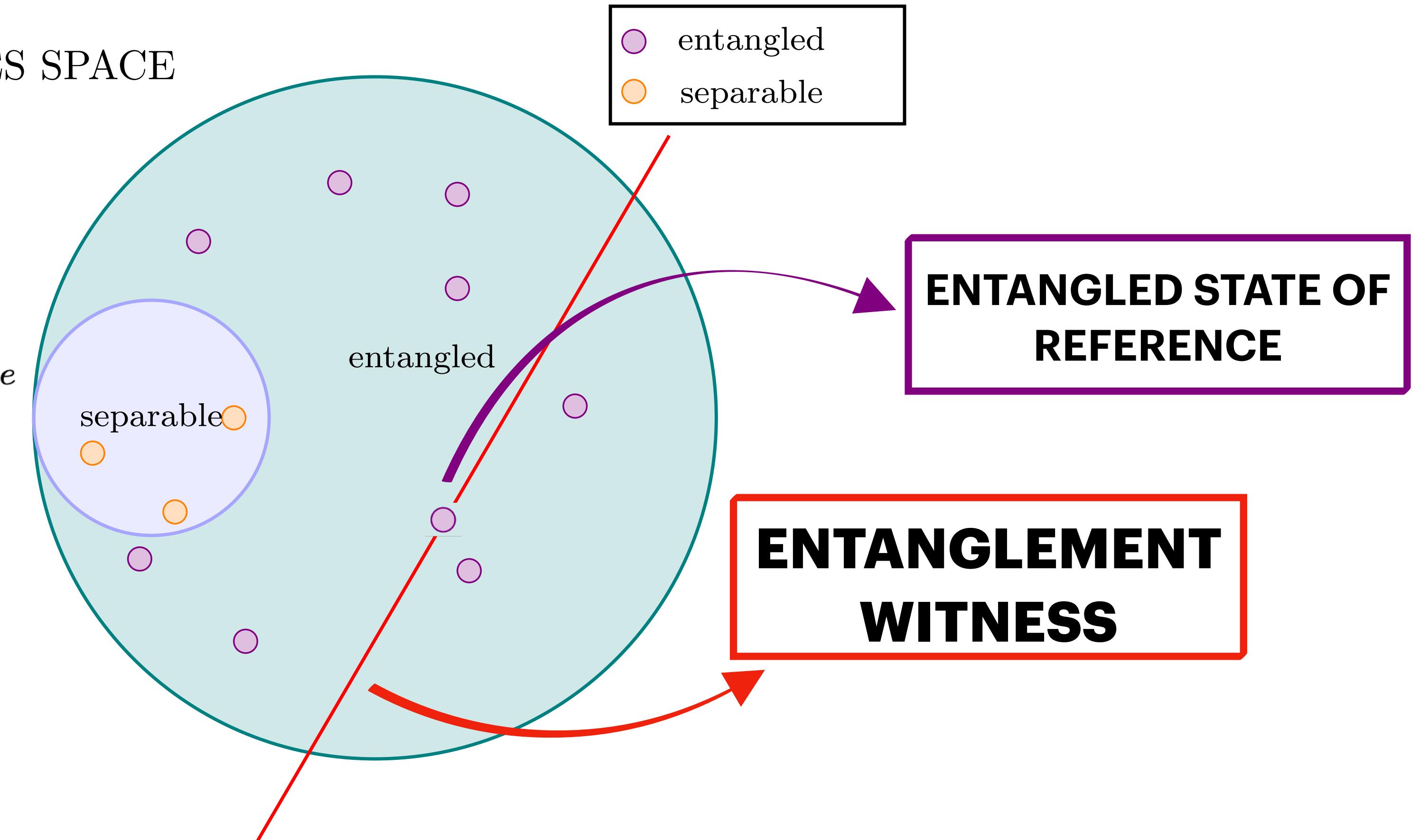
Projective entanglement witness

$$W = \alpha(|(H)\rangle\langle H| - |H\rangle\langle H|)$$

$$\begin{cases} \text{Tr}[W\rho_s] \geq 0 & \forall \rho_s \text{ separable}, \\ \text{Tr}[W\rho_e] < 0 & \text{for at least one entangled } \rho_e \end{cases}$$

$$\begin{aligned} \text{Tr}[\rho W] &= \text{Tr}[\rho(\alpha(|H\rangle\langle H|) - |H\rangle\langle H|)] = \\ &= \alpha(\text{Tr}[\rho|H\rangle\langle H|]) - \alpha(\text{Tr}[\rho|H\rangle\langle H|]) \end{aligned}$$

STATES SPACE



Witness Optimization

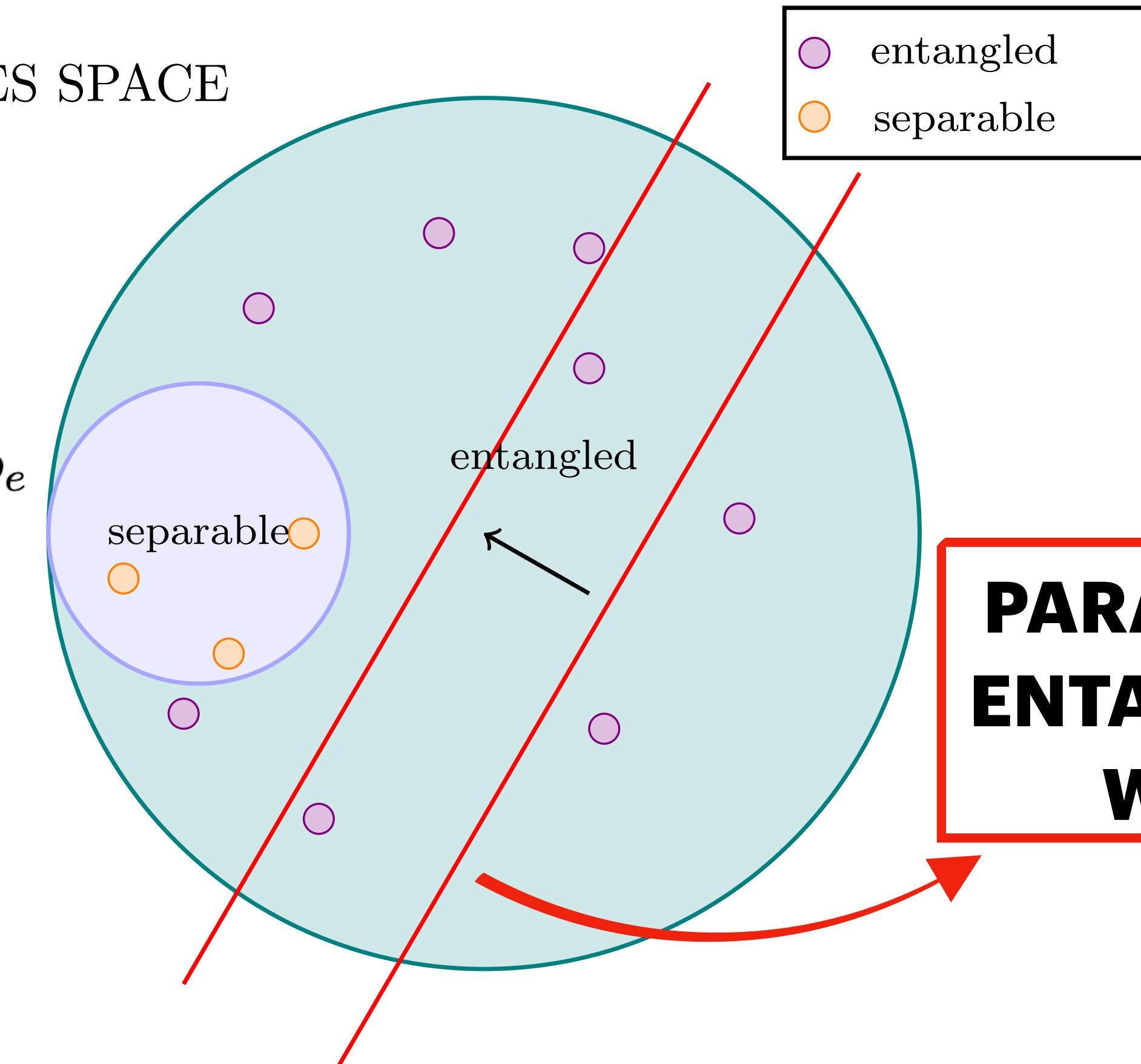
No state of reference

$$W = \alpha I - |V(\theta)\rangle\langle V(\theta)|$$

$$\begin{cases} \text{Tr}[W\rho_s] \geq 0 & \forall \rho_s \text{ separable}, \\ \text{Tr}[W\rho_e] < 0 & \text{for at least one entangled } \rho_e \end{cases}$$

$$\text{Tr}[\rho W] = \alpha - \text{Tr}[\rho |V(\theta)\rangle\langle V(\theta)|]$$

STATES SPACE



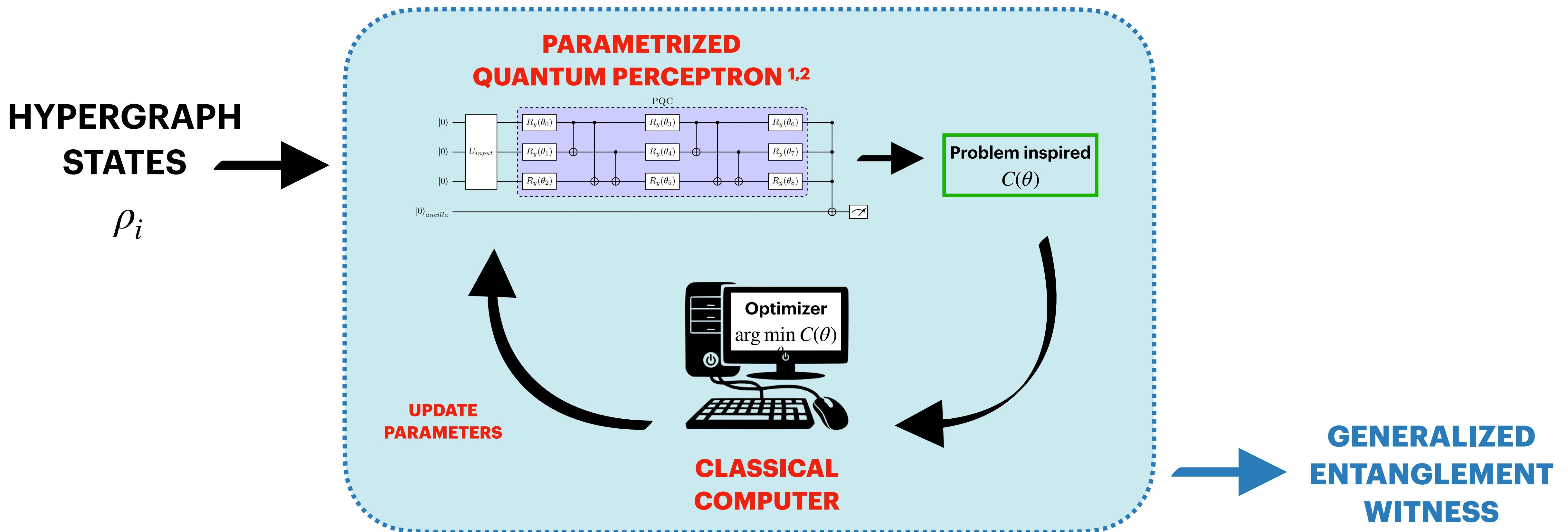
**PARAMETRIZED
ENTANGLEMENT
WITNESS**

J. Roik, et al., Phys. Rev. Applied, vol. 15, no. 5, 2021.

L. T. Wu, et al., in Conference on Lasers and Electro-Optics, Optical Society of America, 2021, FW3N.1

Variational Quantum Algorithm (VQA)

For entanglement witnessing



[1] F. Tacchino et al., npj Quantum Inf 5, 26 (2019)

[2] F. Tacchino et al., IEEE Transactions on Quantum Engineering, vol. 2, pp. 1-10 (2021)

Input states

Separable-entangled proportions

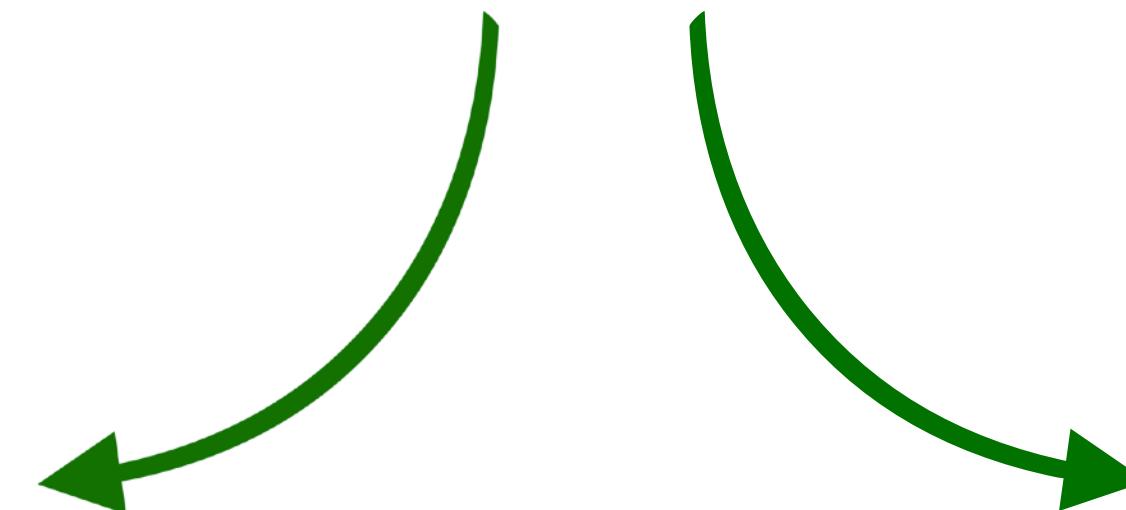
3 qubits hypergraph states¹:

$$|\psi_f\rangle = \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n-1} (-1)^{f(x)} |x\rangle$$

256 states
 (scales as 2^{2^n})

64 bi-separable

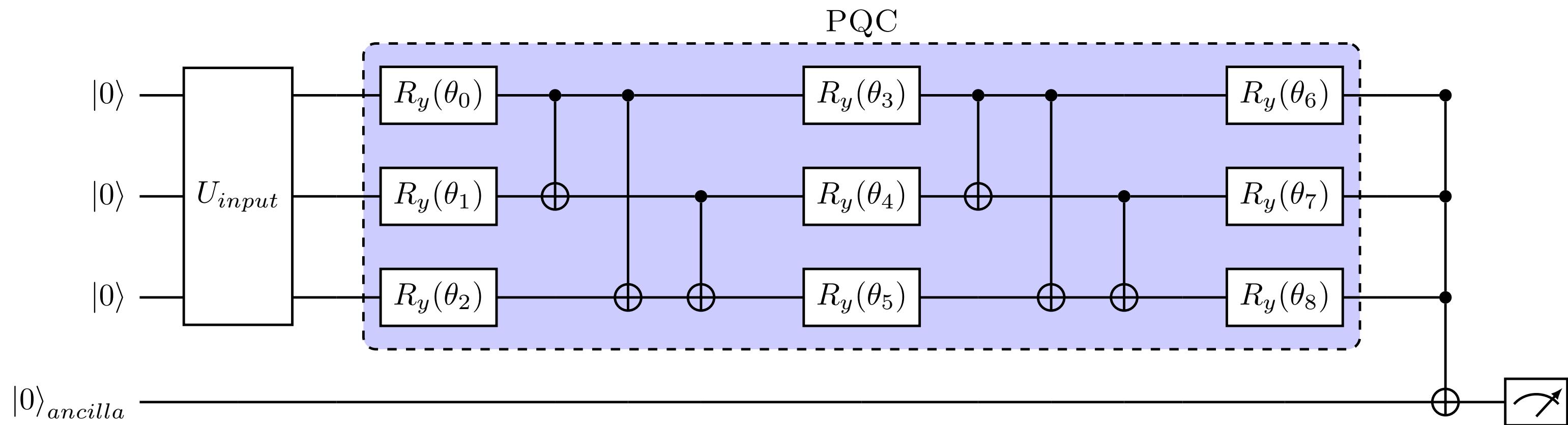
192 entangled
 (64 maximally entangled)



[1] M. Rossi et al., New J. Phys. 15 113022 (2013)

Parametrized Quantum Model

Parametrized Quantum Perceptron¹



Only one measurement setting required to estimate

$$\text{Tr}[\rho | V(\theta)\rangle\langle V(\theta) |]$$

- **2 layers Parametrized Quantum Model (R_y rotations)**
- **COBYLA optimizer**
- **Activation threshold $\alpha = 0.5$**

[1] F. Tacchino et al. , IEEE Transactions on Quantum Engineering, vol. 2, pp. 1–10 (2021)

Problem inspired cost function

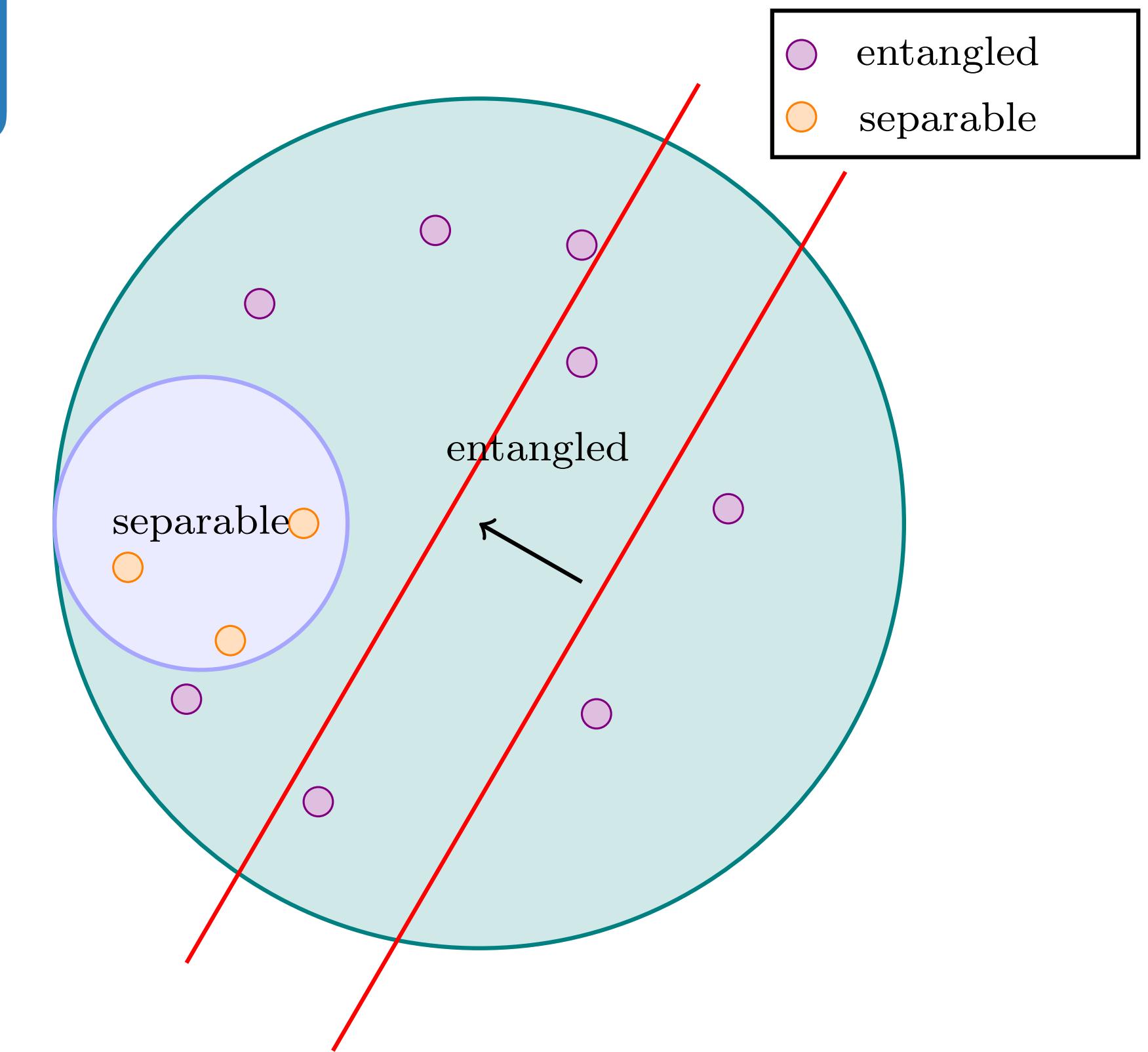
$$\text{Precision} = \frac{\text{True Positive}}{\text{All Positive}}$$

$$\text{Recall} = \frac{\text{True Positive}}{\text{Positive Class}}$$

$$F_\beta = (1 + \beta^2) \cdot \frac{\text{Precision} \cdot \text{Recall}}{\beta^2 \cdot \text{Precision} + \text{Recall}}$$

Favor precision over recall $\downarrow 0 < \beta < 1$

$$C(\theta) = 1 - F_\beta$$



Simulation results

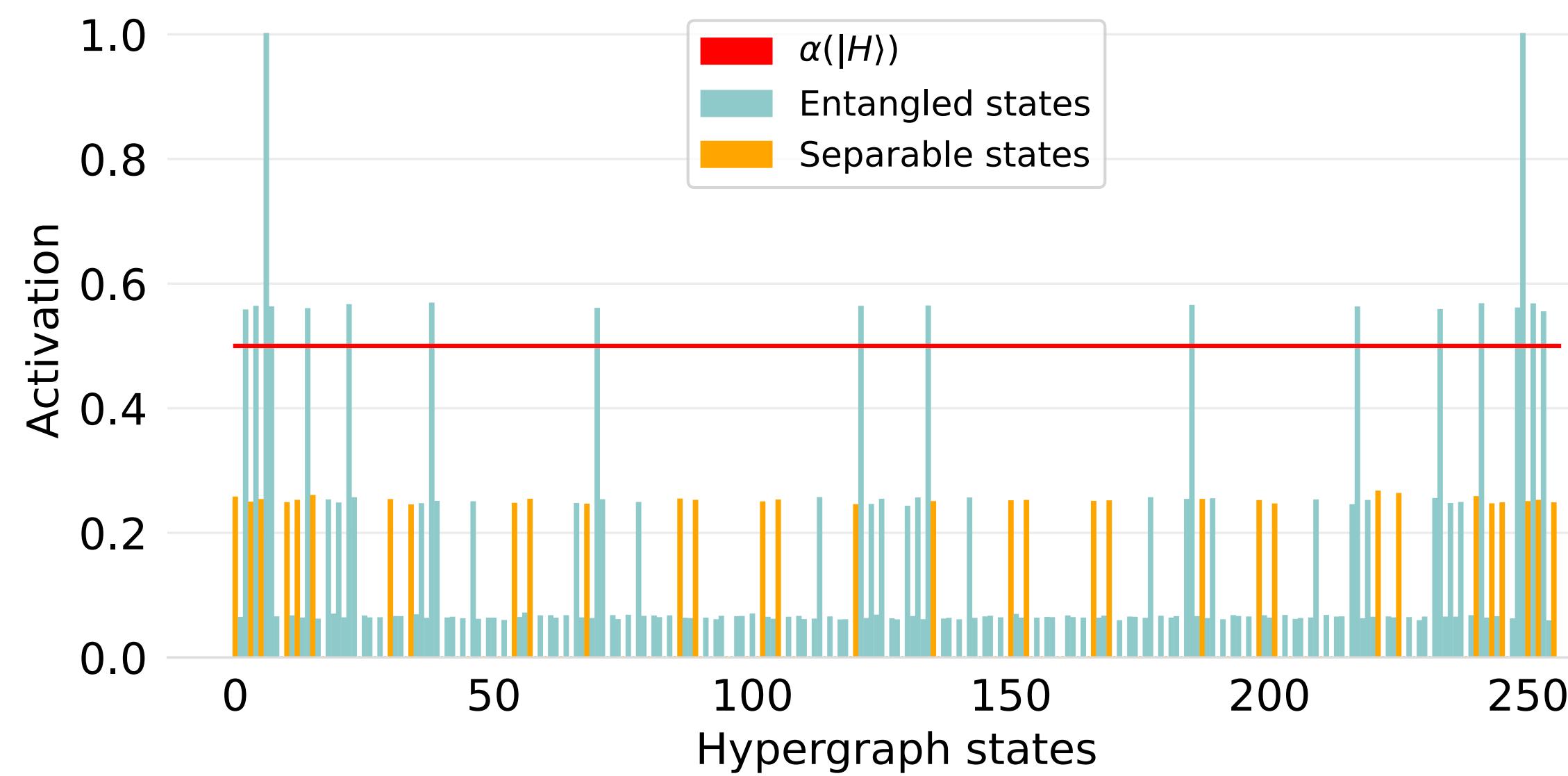
python™

Qiskit

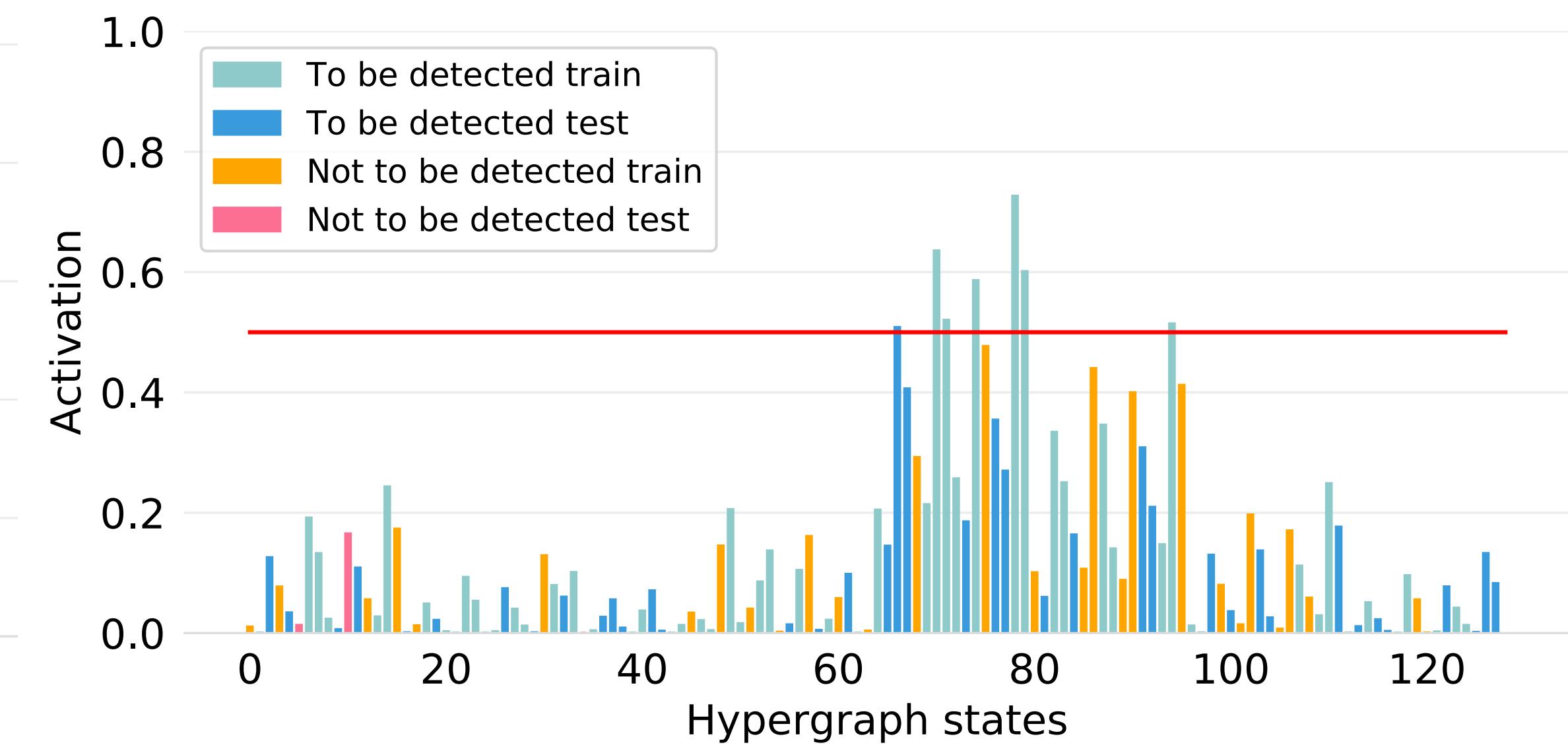


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Theoretical with reference state



Learned Entanglement Witness



Summary and outlooks

Entanglement detection is hard in general

- VQAs are able to learn an **entanglement witness**
- VQAs are more suited to work with **quantum data**
 - direct state manipulation

Future works

Generalization:

- Inputs and model
- Optimizer
- Entanglement measures
- Real quantum hardware

Collaborators



Stefano Mangini



Chiara Macchiavello



Daniele Bajoni



Dario Gerace



UNIVERSITÀ
DI PAVIA

Dipartimento di Fisica
Dipartimento di
Ingegneria Industriale
e dell'Informazione

For further details:
arXiv:2205.10429 (2022)



Quantum Entanglement

A **non-classical** correlation

DEFINITION

A n-parties **pure state** $|\psi\rangle$ is called **fully entangled** if and only if it is not bi-separable with respect to any bipartition:

$$|\psi\rangle = |A\rangle \otimes |B\rangle,$$

where $|A\rangle$ is an m parties state and $|B\rangle$ is an $n-m$ parties state.

Multipartite Entanglement measure

$$\begin{aligned}
 E(|\psi_n\rangle) &:= \min_{AB} E^{AB}(|\psi_n\rangle) = \\
 &= 1 - \max_{|\phi\rangle_A |\phi\rangle_B, AB} |\langle \phi|_A \langle \phi|_B \psi_n \rangle|^2 = \\
 &= 1 - \alpha(|\psi_n\rangle)
 \end{aligned}$$

↗

the max is taken over ALL the possible biseparable states $|\phi_k\rangle_A, |\phi_{n-k}\rangle_B$

$$\alpha^{AB}(|\psi\rangle) = \max_{AB} \alpha^{AB}(|\psi\rangle)$$

→ 2

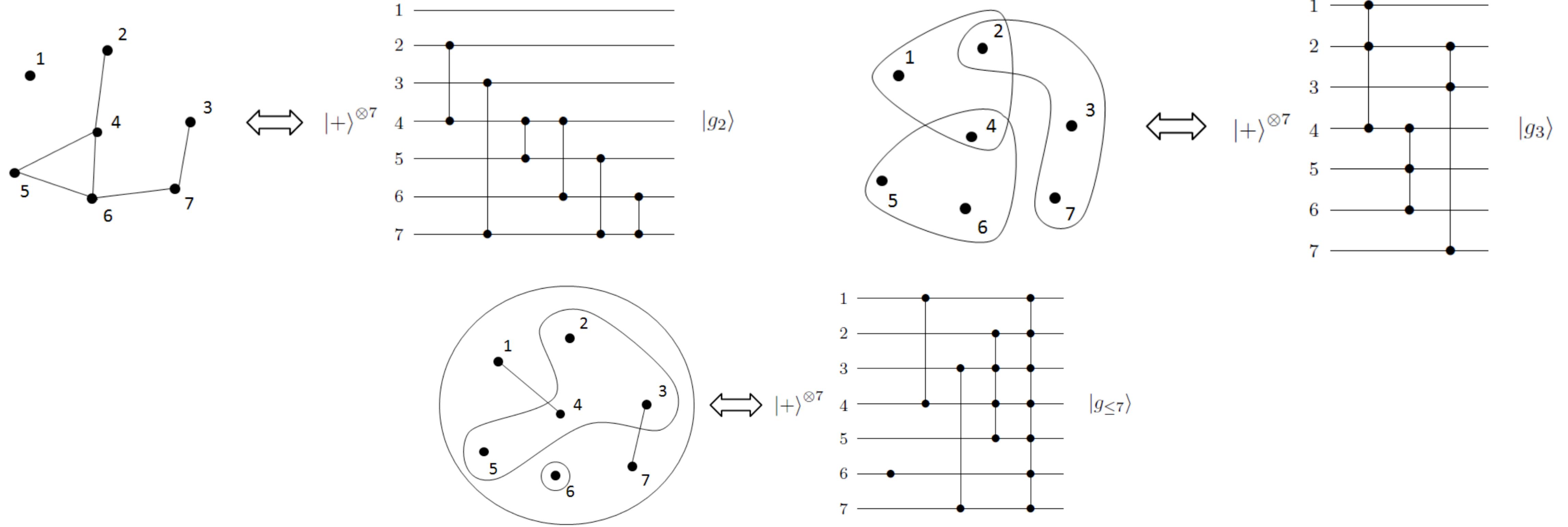
$$\alpha^{AB}(|\psi\rangle) = \max_{k=1,\dots,R} s_k^{AB}(|\psi\rangle)^2$$

where s_k^{AB} are the Schmidt coefficients,
and R is the Schmidt rank

[1] M. Ghio et al., J. Phys. A: Math. Theor. 51 045302 (2018)

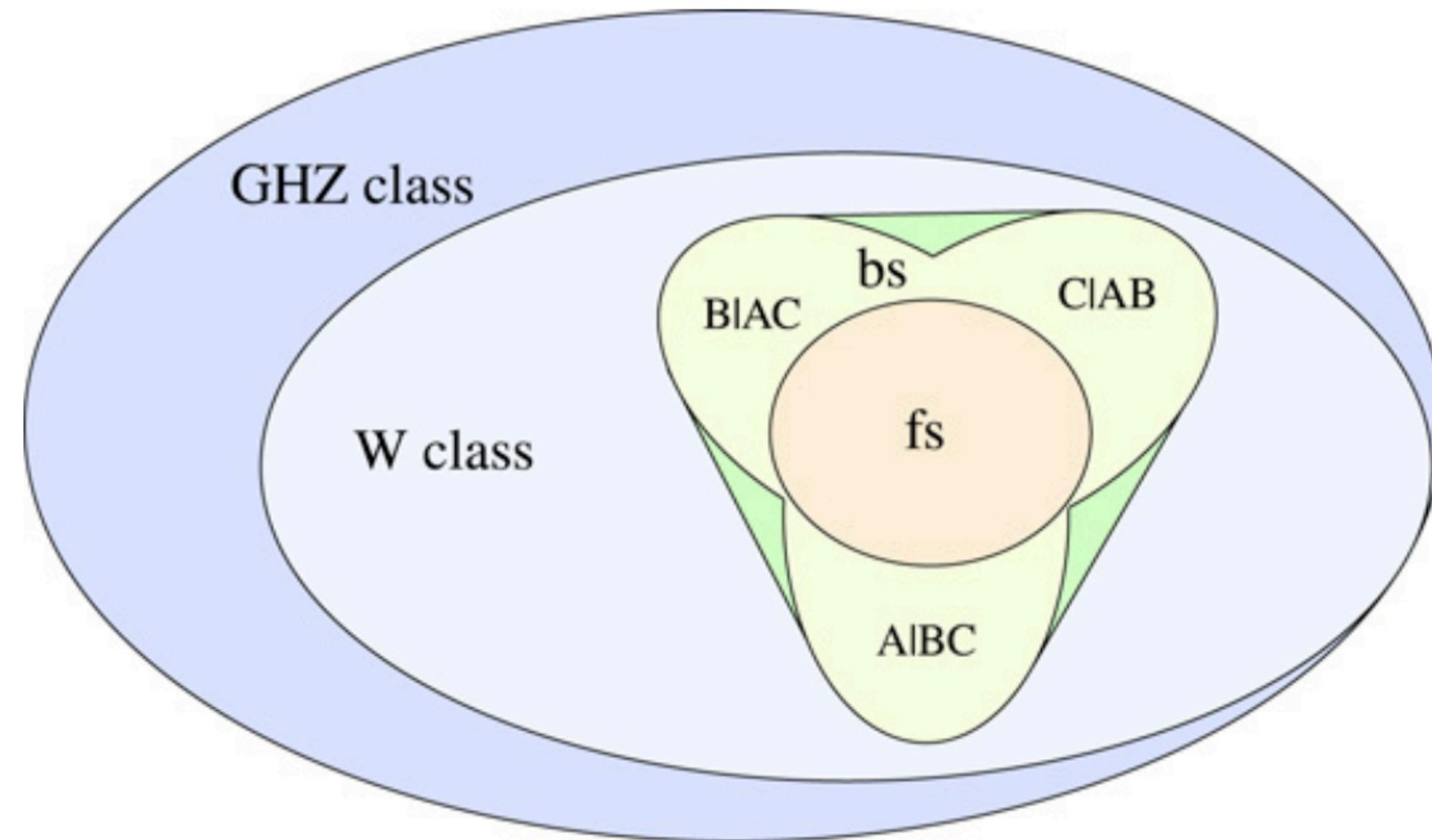
[2] M. Bourennane, et al., Phys. Rev. Lett., vol. 92, p. 087 902, 8 (2004)

Hypegraph states



States space structure

3 qubits



$$|GHZ_3\rangle = \frac{1}{\sqrt{2}}(|000\rangle + |111\rangle)$$

$$|W_3\rangle = \frac{1}{\sqrt{3}}(|100\rangle + |010\rangle + |001\rangle)$$

$$|H\rangle = \frac{1}{\sqrt{8}}(|000\rangle + |001\rangle + |010\rangle + |011\rangle + |100\rangle + |101\rangle + |110\rangle - |111\rangle)$$