



# A MAX K-CUT IMPLEMENTATION FOR QAOA IN THE MEASUREMENT BASED QUANTUM COMPUTING FORMALISM

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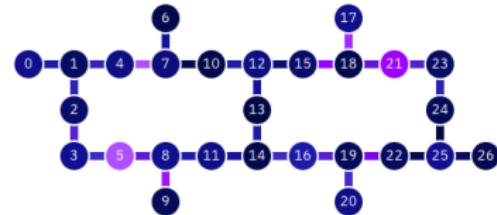
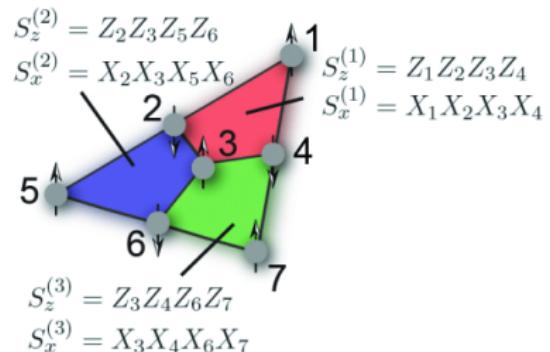
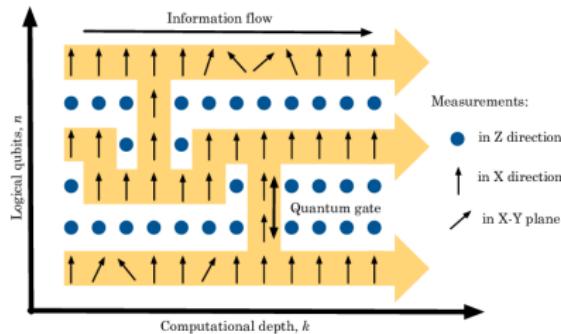




# Graph Theory in Quantum Computing

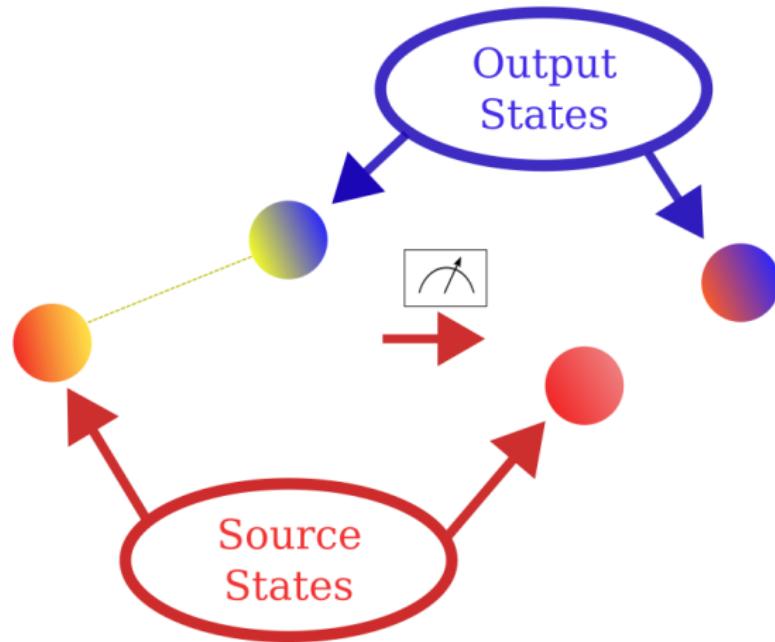
## Fields of interest

- ① Error correction codes
- ② Qubit connectivity
- ③ One-Way Quantum Computing



A. Bermudez et al. "Assessing the progress of trapped-ion processors towards fault-tolerant quantum computation." Physical Review X 7.4 (2017): 041061.

# Measurement Based Quantum Computing (MBQC)



## MBQC States

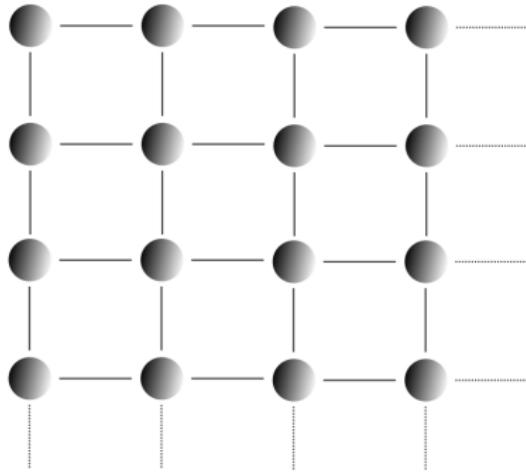
- ① Source states
  - inputs
  - ancillae
- ② Output states

## MBQC Operators

- Entanglement
- Measurements
- Corrections



# Graph States



## Encoding

- Vertices: qubits
- Edges: entanglements

## Representation

- Vertices:  
 $V = \{1, 2, \dots, n\}$
- Edges:  
 $E = \{(1, 2), (2, 3), \dots, (n - 1, n)\}$
- Neighborhood of  $a$ :  
 $N_a = \{b \in V | (a, b) \in E\}$

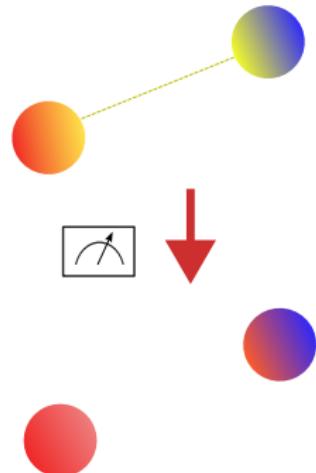
## ! Caveat !

- Increasing number of logical qubits



# Measurement-Based Quantum Computing

## MBQC Pipeline



- ① Input state:  $|\psi\rangle = a|0\rangle + b|1\rangle$
- ② Projection:  $|\theta\rangle = |0\rangle + e^{i\theta}|1\rangle$
- ③ Entanglement:  $\hat{CZ}_{12}|\psi\rangle_1|+\rangle_2 = |\Psi\rangle$   
 $|\Psi\rangle = a|0\rangle_1|+\rangle_2 + b|1\rangle_1|-\rangle_2$
- ④ Measurement:  $|\theta\rangle_1 \langle \theta|\Psi\rangle = |\theta\rangle_1 |\psi'\rangle$
- ⑤ Output:  $|\psi'\rangle = a|+\rangle + b|-\rangle$



# The Max k-Cut Problem

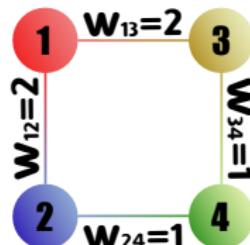
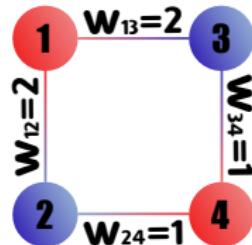
## Features

- Combinatorial optimization
- Encoding: Ising Hamiltonian

$$w(S, \bar{S}) = \frac{1}{2} \sum_{i \in S, j \in \bar{S}} w_{ij}$$

$$\max \frac{1}{2} \sum_{1 \leq i < j \leq n} w_{ij}(1 - z_i z_j)$$

$$\hat{H}_C = \sum_{ij} \frac{w_{ij}}{2} (I - \hat{Z}_i \hat{Z}_j)$$





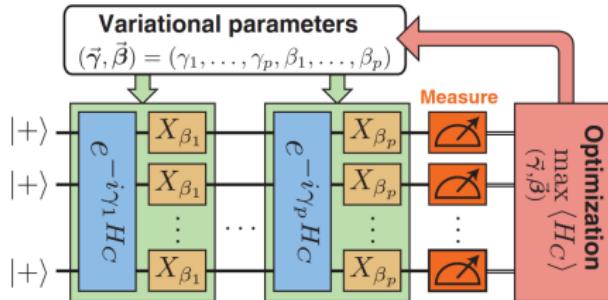
# QAOA algorithm

## Features

- Combinatorial optimization
- Variational algorithm:  
parameters  $\{\beta_i, \gamma_i\}_{i=1}^p$
- Hybrid algorithm
  - ➊ Backpropagation: classical
  - ➋ State evolution: quantum

## Quantum features

- Mixing Hamiltonian:  
 $\hat{H}_M |+\rangle^{\otimes n} = |+\rangle^{\otimes n}$
- Cost Hamiltonian:  
 $\hat{H}_C |\psi\rangle = E |\psi\rangle$
- Goal state:  $\hat{H}_C |\tilde{\psi}\rangle = E_{max} |\tilde{\psi}\rangle$
- Approx. ratio:  $r = E/E_{max}$



L. Zhou et al., *Quantum Approximate Optimization Algorithm: Performance, Mechanism, and Implementation on Near-Term Devices*, PHYSICAL REVIEW X, 2020



# Derivation from Adiabatic Theorem

## Adiabatic Theorem

- Adiabatic evolution,  $t \in [0, T]$

$$\hat{H}(t) = \left[1 - \frac{t}{T}\right] \hat{H}_M - \frac{t}{T} \hat{H}_C$$

- Time evolution

$$\hat{U}(t) = e^{-i \int_0^T dt \hat{H}(t)}$$

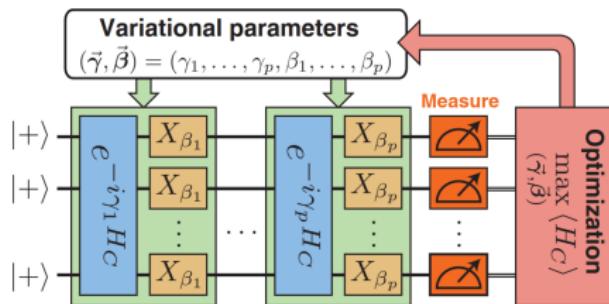
## QAOA algorithm

- QAOA approximation on  $\hat{H}(t)$

$$\prod_{i=1}^p e^{-i\gamma_i \hat{H}_C} e^{-i\beta_i \hat{H}_M}$$

- Backpropagation on angles

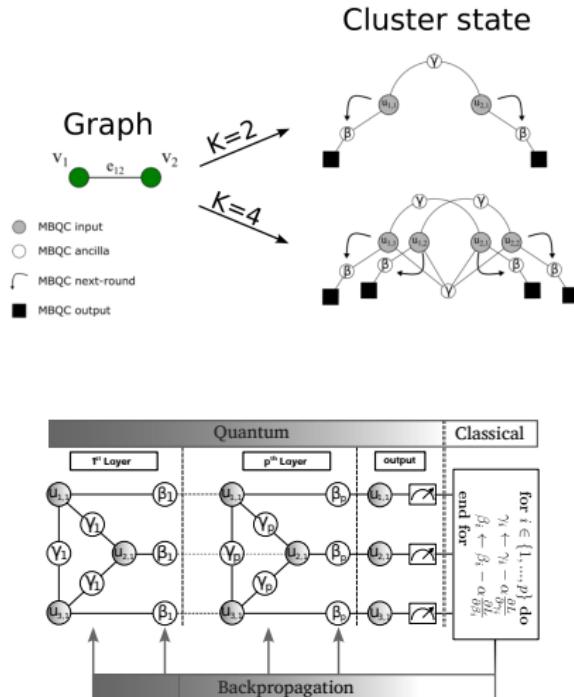
$$\beta_i \rightarrow \beta_i - \nabla_{\beta_i} E$$



L. Zhou et al., *Quantum Approximate Optimization Algorithm: Performance, Mechanism, and Implementation on Near-Term Devices*, PHYSICAL REVIEW X, 2020



# QAOA in the MBQC Frame




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## Algorithm MBQC-QAOA

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```

1: Set graph  $G[V, E]$ 
2: Set  $K, p$ 
3: for  $i$  in  $p$  do:
4:   Install graph state
5:   for  $|+\rangle_j \in \gamma$  do:
6:      $\langle \gamma_i | + \rangle_j$ 
7:   end for
8:   for  $|+\rangle_j \in \beta$  do:
9:      $\langle \beta_i | u_j \rangle$ 
10:     $\langle + | + \rangle_j$ 
11:   end for
12: end for

```

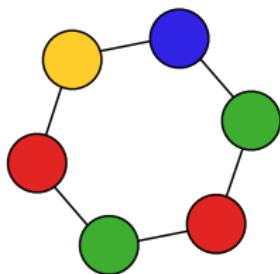
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M. Proietti, F. Cerocchi, M. Dispenza, *Native measurement-based quantum approximate optimization algorithm applied to the Max K-Cut problem*, PHYSICAL REVIEW A, 2022

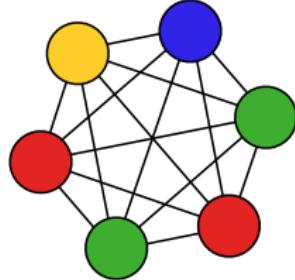


# Environment Settings

## Cyclic Graphs



## Fully Connected Graphs



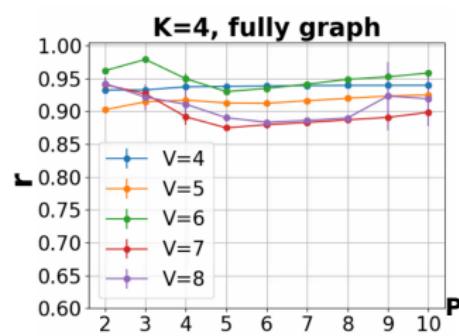
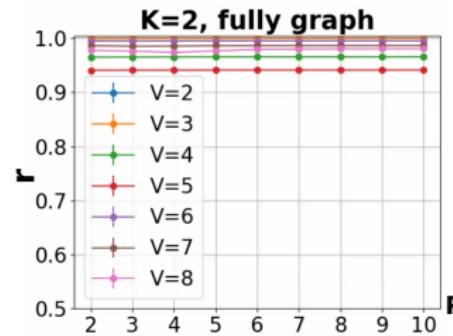
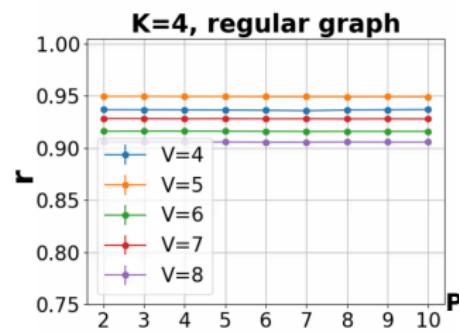
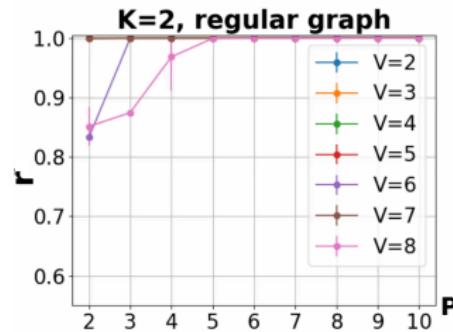
### Computational frame

- ① Software: python libraries
  - ▶ paddle (Baidu)
  - ▶ paddle-quantum
- ② Hardware: HPC system davinci-1 (Leonardo Company)
  - ▶ AMD Epyc CPUs (512GB)
  - ▶ A100 GPUs (40GB)





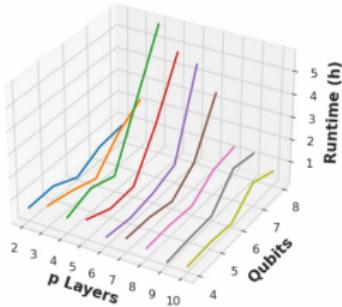
# Results: Approximation Ratio $r = E/E_{max}$



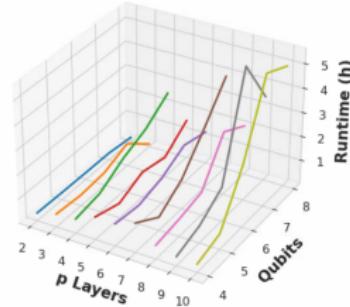


# Results: Global Runtime for Simulations

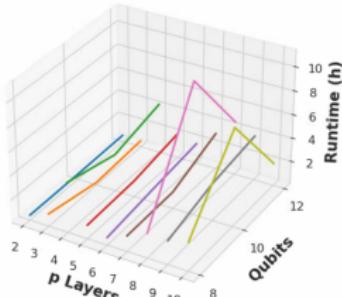
K=2, Regular Graph



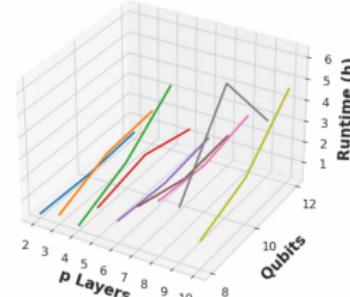
K=2, Fully Connected



K=4, Regular Graph



K=4, Fully Connected





# Acknowledgements and Perspectives



- Politecnico di Milano: actual institution
- Leonardo company: developers team & davinci-1 hardware provider
- On road: implementation on QuiX Quantum photonic hardware



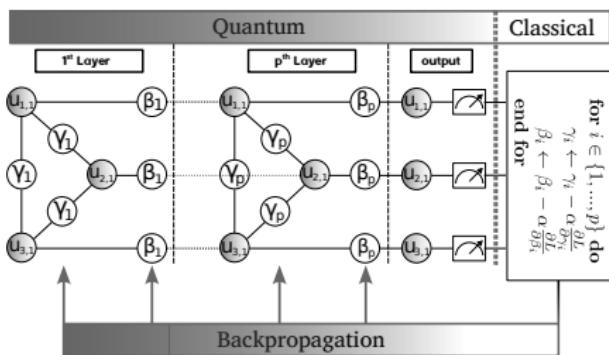


## Max k-Cut Hamiltonian

$$\hat{H}_C = \sum_{ij} \frac{w_{ij}}{2^q} \sum_{n=1}^q \left( C_n I - \hat{C}_n [\{\hat{Z}_{il} \hat{Z}_{jl}\}_{l=1}^q] \right)$$

- $q = 1$
  - $C_1 = 1$

$$q = \log_2(k)$$



$$\hat{H}_C = \frac{1}{2} (w_{12} + w_{13} + w_{23} + \\ - w_{12} \hat{Z}_1 \hat{Z}_2 - w_{13} \hat{Z}_1 \hat{Z}_3 + \\ - w_{23} \hat{Z}_2 \hat{Z}_3)$$

