# Quantum Computing Applications: Opportunities at DESY and beyond

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- Introduction
- Variational quantum simulations
- Examples:
  - flight gate assignment
  - Heisenberg model
- Opportunities



# Problem description by Hamiltonian

- Consider two classes of problems
- Systems very hard for or not accessible to classical computers, e.g.
  - topological terms
  - non-zero quark density
  - Hubbard model away from half filling
  - real time evolutions
- Classical optimization problems  $\rightarrow$  quantum supremacy
  - flight to gate assignment
  - particle track reconstruction
  - air shower
  - traffic, etc.

#### Hamiltonian for the Quantum Computer

• Hamiltonian of a physical system is expressed in terms of Pauli matrices

$$\sigma_x = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad \sigma_y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \quad \sigma_z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

- Hamiltonian is direct product of Pauli matrices  $\Rightarrow$  obtain a  $2^N \otimes 2^N$  matrix
- general goal: find ground and excited states and corresponding wave functions
- problem scales exponentially: how can we do this?
- Variational Quantum Simulation (VQS) (alternative: imaginary time evolution)

#### Finding ground state: Variational Quantum Simulation

- start with some initial state  $|\Psi_{
  m init}
  angle$
- apply succesive gate operations  $\equiv$  unitary operations  $e^{iS\theta}$
- examples for S: Pauli matrices  $\sigma_x$ ,  $\sigma_y$ ,  $\sigma_z$ , parametric CNOT

$$|\Psi(\vec{\theta})\rangle = e^{iS_{(n)}\theta_n}\dots e^{iS_{(1)}\theta_1}|\psi_{\text{init}}\rangle$$

• with  $R_j := e^{iS_{(j)}\theta_j}$  cost function evaluated on quantum computer

$$C(\vec{\theta}) := \left\langle \psi_{\text{init}} \left| \left( \prod_{j=1}^{n} R_j \right)^{\dagger} H \prod_{j=1}^{n} R_j \right| \psi_{\text{init}} \right\rangle$$

- goal: minimize  $C(\vec{\theta})$  over the angles  $\vec{\theta}$  $\rightarrow$  obtain minimal energy, i.e. ground state
- minimization performed classically (hybrid classical-quantum approach)

   — also possible on quantum computer itself

# **Example: only qubit rotations**







unitary rotation with angle  $\theta$ 

### **Adding entanglement**

#### $\Psi^1_{\rm ini}:$ $\Psi_{\mathrm{fin}}^1(\theta_1^1,\theta_1^2)$ $e^{i\theta_1^2\sigma_x}$ $e^{i\theta_1^1\sigma_x}$ $e^{i heta_2^2\sigma_x}$ $\Psi_{\rm fin}^2(\theta_2^1,\theta_2^2)$ $\Psi_{\rm ini}^2:$ 2 $e^{i\theta_2^1\sigma_x}$ $\Psi_{\rm ini}^3$ : $\Psi_{\rm fin}^3(\theta_3^1,\theta_3^2)$ 3 $e^{i\theta_3^1\sigma_x}$ $e^{i\theta_3^2\sigma_x}$ $\Psi_{\rm ini}^4$ : $\Psi_{\rm fin}^4(\theta_4^1,\theta_4^2)$ $e^{i\theta_4^2\sigma_x}$ $e^{i\theta_4^1\sigma_x}$ 4 final w.f. initial w.f. layer

Adding entanglement

- (k) qubit k
- $e^{i\theta\sigma_x}$
- unitary rotation with angle  $\theta$
- entanglement gate

#### Variational quantum simulation



- evaluate cost function  $\langle \Psi(\vec{\theta)}|H|\Psi(\vec{\theta})\rangle$  on quantum device





• optimize over parameters  $\vec{\theta}$ on classical computer  $\rightarrow$  give back new set of  $\vec{\theta}$ 

# **Classical optimization problem: flight gate assignment**

• Find shortest path between two connecting flights

 $x_{i\alpha} = \left\{ \begin{array}{ll} 1, & \text{if flight } i \in F \text{ is assigned to gate } \alpha \in G \\ 0, & \text{otherwise} \end{array} \right.$ 

 $x \in \{0,1\}^{F \otimes G} \to x$  binary variable  $\to x \in \{-1,1\}^{F \otimes G}$ 

eigenstate of third Pauli matrix  $\sigma_z$ 

$$H = \sum_{j=1}^{n} Q_{jj} \sigma_j^z + \sum_{\substack{j,k=1\\j < k}}^{n} Q_{jk} \sigma_j^z \otimes \sigma_k^z$$

- $Q_{ij}$  coeffecients specific for a real given airport
- Goal: find ground state (shortest path)
- contraints:
  - every flight can only be assigned to a single gate
  - no aircraft can be at the same gate at the same time





# VQS for FGA

(L. Funcke, T. Hartung, S. Kühn, T. Stollenwerk, P. Stornati, K.J.)

- use variational quantum simulation to find ground state
  - use 6 qubits on simulator
  - overlap:  $\langle \Psi_{VQS} | \Psi_{\mathrm{exact}} 
    angle$





- Remarks:
  - Hamiltonian is diagonal  $\rightarrow$  classical optimization
  - QC helpful through principles of superposition and entanglement?
  - the same Hamiltonian can be used in particle track reconstruction

#### A condensed matter physical model

• 1-dimensional Heisenberg model

 $H = \sum_{i=1}^{N} \beta \left[ \sigma_x(i) \sigma_x(i+1) + \sigma_y(i) \sigma_y(i+1) + \sigma_z(i) \sigma_z(i+1) \right] + J \sigma_z(i)$ 

• Pauli matrices

$$\sigma^{x} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \quad , \quad \sigma^{y} = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix} \quad , \quad \sigma^{z} = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$$

- nearest neighbour interaction, tensor products
- Hamiltonian expressed in Pauli matrices  $\rightarrow$  suitable for quantum computer
- shows phase transitions, critical behaviour, non-trivial spectrum

 $H = \sum_{i=1}^{N} \beta \left[ \sigma_x(i) \sigma_x(i+1) + \sigma_y(i) \sigma_y(i+1) + \sigma_z(i) \sigma_z(i+1) \right] + J \sigma_z(i)$ 

```
from pyquil.quil import Program

import pyquil.api as api

from pyquil.gates import *

qvm = api.QVMConnection() [hardware \rightarrow qvm = api.QPUConnection()]

import numpy as np

from pyquil.api import QVMConnection

from scipy.optimize import minimize

from grove.pyvqe.vqe import VQE

from pyquil.paulis import ID, sX, sY, sZ
```

```
def smallansatz(params):
return Program(RX(params[0], 0))
```

```
\begin{split} & \text{beta}{=}0.12578, J{=}1.87 \\ & \text{hamiltonian}{=}0 \\ & \text{for k in range(3):} \\ & \text{I}{=}(k{+}1)\%3 \\ & \text{hamiltonian} +{=}\text{beta}*(sX(k)*sX(l){+}sY(k)*sY(l) + sZ(k)*sZ(l)) + J*sZ(k) \\ & \text{print(hamiltonian)} \end{split}
```

initialangle = [0.0]

vqeinst = VQE(minimizer=minimize,minimizerkwargs='method': 'nelder-mead')

```
angle = 2.0
vqeinst.expectation(smallansatz([angle]), hamiltonian, None, qvm)
result = vqeinst.vqerun(smallansatz, hamiltonian, initialangle, None, qvm=qvm)
print(result)
```

# General measurement error mitigation in NISQ area

(L. Funcke, T. Hartung, S. Kühn, P. Stornati, X. Wang, K.J., arXiv:2007.03663)

- generated state  $|\Psi(ec{ heta})
  angle$  is a bit string |00110011100101
  angle
  - false measurement  $|0\rangle \rightarrow |1\rangle$  with probability  $p_1$  $|1\rangle \rightarrow |0\rangle$  with probability  $p_2$
- setting (for simplicity)  $p_1 = p_2 = p$
- measuring *s*-times: get k correct and s - kincorrect results distributed as

$$f(k,s,1-p) = \begin{pmatrix} s \\ k \end{pmatrix} (1-p)^k p^{s-k}$$

- recompute <u>exact</u> energy from noisy measurments
- can be generalized to arbitrary number of qubits
- will be devloped further in DASHH



example: transverse ising model

# Opportunities

- theoretical particle physics
  - explore matter anti-matter asymmetry
  - CP-violation
  - early universe
  - heavy ion collisions
  - event and parton shower generation
  - raise your hand! 🖖
- experimental particle physics
  - particle track reconstruction (B. Heinemann, N. Styles)
  - jet cclustering (K. Borras, D. Krücker)
  - Higgs Physics (C. Issever)
  - Particle shower generation (D. Krücker, K. Borras)
  - raise your hand! 🖖





# **Opportunities in photon science**

- group of N. Rohringer:
  - Hamiltonian for electrons in the Born-Oppenheimer approximation

$$\widehat{H} = -\frac{\hbar^2}{2m_e^2} \sum_i \Delta_i - e^2 \sum_{i,J} \frac{Z_J}{r_{iJ}} + e^2 \sum_{i>j} \frac{1}{r_{ij}}$$

- corresponding Schrödinger equation

$$i\hbar\frac{d}{dt}|\Psi(t)
angle = \widehat{H}|\Psi(t)
angle$$

- cast in <u>fermionic</u> stochastic differential equations
  - $\rightarrow$  solve with quantum algorithms
- group of F. Kärtner
  - use wave guides for quantum simulations?
- group of R. Röhlsberger
  - explore potential of one-way-computing
- raise your hand

#### Some publications

- Zeta-regularized vacuum expectation values from quantum computing simulations
   T. Hartung and K.J., J.Math.Phys. 60 (2019) 9, 093504
- Measurement Error Mitigation in Quantum Computers Through Classical Bit-Flip Correction
   L. Funcke, T. Hartung, S.Kühn, P. Stornati, K.J., arxiv:2007.03663
- A resource efficient approach for quantum and classical simulations of gauge theories in particle physics
   J.F. Haase, L. Dellantonio, A.Celi, D.Paulson, A. Kan, K.J., C.A. Muschik, arxiv:2006.14160
- Towards simulating 2D effects in lattice gauge theories on a quantum computer
   D. Paulson, L. Dellantonio, J.F. Haase, A. Celi, A. Kan, A. Jena,
   C. Kokail, R. van Bijnen, K.J., P. Zoller, C. A. Muschik, arxiv:2008.09252
- Simulating Lattice Gauge Theories within Quantum Technologies M.C. Baüls et.al., Eur.Phys.J.D 74 (2020) 8, 165
- General quantum circuit analysis
   L. Funcke, T. Hartung, S.Kühn, P. Stornati, K.J., in preparation
- Flight gate assignment with variational quantum simulations
   L. Funcke, T. Hartung, S.Kühn, P. Stornati, T. Stollenwerk, K.J., in preparation



# **Succesful Innovation Pool Projects**

- Laser Und XFEL Experiment (LUXE), Coordinator: B. Heinemann
  - use quantum algorithms for particle track reconstruction
- Accelerating Science with Artificial Intelligence and Machine Learning (ACCLAIM), Coordinator: F. Gaede
  - explore potential of QC for AI/ML



- develop hybrid quantum/classical variational algorithms

#### About to submit:

Helmholtz Innovationsschub Projekt für Quanten Computing und Sensing



# Conclusion

- Quantum computing very active field
  - simulators run on local machines
  - hardware with small number of noisy qubits available
  - algorithms and methods are being developed:
    - $\rightarrow$  variational quantum simulations
    - $\rightarrow$  error mitigation and error correction
  - first benchmark models have been simulated
- New opportunities
  - explore potential of quantum computing
  - prepare for next generation of quantum computers
- Spin-off for chemistry, biology, material science, ...
- Quantum computing: a chance that we should take now

