# Southampton

### State Preparation using Quantum Algorithms within Hamiltonian Formulations of Quantum Field Theories

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### Cambridge-Southampton QA for QFT

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### Outline

#### • Motivations and Project Structure

 $\circ$   $\,$  Big-picture reasons to consider QAs for QFTs, and our collaboration  $\,$ 

#### Quantum State Preparation

- Why is quantum state preparation a hard problem?
- How are variational algorithms effective solutions?
  - The choice of an ansatz

#### • Specialising towards Lattice Gauge Theories

- Adiabatic Preparation for the Schwinger Model
- $\circ$   $\,$  VQE and QAOA for the Schwinger Model  $\,$
- Accelerating VQE with equivariant ansatze

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# Quantum Scale-up

- All low-qubit, low-depth QAs can be simulated classically.
- Noisy intermediate-scale quantum (NISQ) technology may show quantum advantage.
  - **50-100s of qubits**
  - 1000s of gate operations before noise dominates.
- Stepping stone towards fault-tolerant QC (2030s)
  - Why wait to do physics?



[Figure by Funcke (Lattice2022)]

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# **Physics drives**

- Collider phenomenology
  - Parton showers, S-matrices
- Strongly-interacting systems

   Lattice QCD, hadronisation
- Neutrino (astro)physics
  - Oscillations, mean-field
- Early-universe cosmology
  - Inflation, CP-violation
- Gauge-gravity duality
  - **QEC, SYK / matrix models**



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### **Quantum Simulation Protocols**



Circuit depth O(2<sup>n</sup> / n) needed to prepare generic n-qubit state
 Can be reduced to O(n) with O(2<sup>n</sup>) ancilla qubits

• We want eigenstates of given Hamiltonians, physicality/symmetry advantages

#### Quantum Phase Estimation (QPE) vs Variational Quantum Algorithms

**QPE**:



Purely quantum algorithm, O(1) shots needed but O(1/ $\epsilon$ ) depth to achieve precision  $\epsilon$ .  $\circ$  M = 1/log2( $\epsilon$ ) ancilla qubits needed

#### Quantum Phase Estimation (QPE) vs Variational Quantum Algorithms

**OPE**:



- Purely quantum algorithm, O(1) shots needed but O(1/ε) depth to achieve precision ε.
   M = 1/log2(ε) ancilla qubits needed
- Iterative quantum state preparation with a classical optimiser
  - $\circ$  O(1/ $\epsilon^2$ ) shots needed but only O(1) depth, ideal for NISQ.

#### Quantum Phase Estimation (QPE) vs Variational Quantum Eigensolver



- Iterative quantum state preparation with a classical optimiser a "hybrid" algorithm.
  - $\circ$  O(1/ $\epsilon^2$ ) shots needed but only O(1) depth,
  - Although QPE has better asymptotics, VQE ideal for NISQ

### **Ansatz Selection**

- Desirable qualities:
  - Low circuit depth
  - Minimally contain physical spectrum
  - Avoid barren plateaus for larger # qubits
- Heuristic approaches: k-local gatesets
- Adaptive approaches: ADAPT-VQE, Generative VQE
- Model-driven approaches:
  - Coupled-cluster
  - Quantum Approximate Optimisation (QAOA)
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[Gard et al., 2020]



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### (1+1)d Kogut-Susskind Hamiltonian

• Lattice gauge theory with group G, staggered fermion:

$$H = w \sum_{n=1}^{N} \left( \phi_n^{\dagger} U_n^{j} \phi_{n+1} + h.c. \right) + m \sum_{n+1}^{N} (-1)^n \phi_n^{\dagger} \phi_n + J \sum_{n=1}^{N} \vec{L}_n^2$$

- In (1+1)d, gauge redundancy given boundary conditions
  - $\circ$   $\,$  Solve Gauss's Law to "fermionise" the Hamiltonian  $\,$
  - For d > (1+1), we will have to be clever.
- Map fermions to qubits, e.g. Jordan-Wigner, Bravyi-Kitaev  $\circ$  For G = SU(n) or U(n), one fermion  $\Rightarrow$  n qubits.

$$\begin{split} \hat{Q}_n^a &= \sum_{r,s} \hat{\phi}_n^{r,\dagger} \left[ \hat{T}_j^a \right]_{rs} \hat{\phi}_n^s, \\ \hat{R}_n^a &= \sum_b 2 \text{Tr} \left[ \hat{U}_n \hat{T}_j^a \hat{U}_n^\dagger \hat{T}_j^b \right] \hat{L}_n^b, \\ \hat{G}_n^a &= \hat{L}_n^a - \hat{Q}_n^a - \hat{R}_n^a = 0 \\ \hline \phi^\dagger U \phi &\mapsto X \otimes X + Y \otimes Y, \\ \phi^\dagger \phi &\mapsto Z, \\ L^2 &\mapsto \sum_{1 \le l \le n} Z_l \\ \end{split}$$

### Adiabatic State Prep, Schwinger Model

- Take G = U(1) with theta term. [Chakraborty et al., PRD 105.9 (2022)]
- Evolve from H[m=0,w=0,th=0] vacuum to vacuum of interest.



### **ASP for the Schwinger Model**

- Example for 4 qubits, lattice spacing a = 1.43 , masses m = 0,1
  - Deterministic evolution to true vacuum



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### **VQE for the Schwinger Model**

- VQE guarantees exact convergence for good ansatz
  - Six-parameter ESP ansatz which spans charge-subspace (adjustable)



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### **QAOA** for the Schwinger Model

- QAOA sacrifices exactness for ease of construction
  - Ansatz determined by Hamiltonian, plus flexible choice of "mixer"



### **Efficiency Comparison**

#### • Total "runtime" = evaluation count \* circuit depth



# **Equivariance and Expressivity**

- Geometric ML:
  - Predictions invariant to symmetries of data.
- Geometric VQE:
  - Observables invariant to symmetries of Hamiltonian.
- Equivariant embeddings restrict to a symmetry sector
  - Also provides desirable parameter-dependence
  - $\circ$  Must sacrifice some expressivity of the ansatz
- In simple (1+1)d cases, equivariance reproduces ESP, better explanatory power
   In general, will provide a (very) competitive toolbox for ansatz design



### Summary

- State preparation is a crucial bottleneck for near-term quantum simulations.
- Variational algorithms are a leading contender for low-depth QAs for physics.
- Continued developments of ansatze required if we are to:
  - Take maximal advantage of the hardware frontier as it moves.
  - Extend QAs to otherwise inaccessible LGTs, etc.
- <u>Next step: (2+1)d</u>

### **THANK YOU FOR LISTENING!**

### **Works Cited**

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#### Thank you to UKLFT, and our hosts here in Plymouth!

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