

Ottocento anni di libertà e futuro

## QUANTUM COMPUTING FOR HIGH ENERGY PHYSICS

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"Let the computer itself be built of quantum mechanical elements which obey quantum mechanical laws."

#### **RICHARD FEYNMAN (1982)**

#### **QUANTUM COMPUTERS**









#### Superconductors



### QUANTUM COMPUTING



#### **QUANTUM COMPUTERS AND SIMULATORS**

#### **Universal Quantum Simulators**

Seth Lloyd

Feynman's 1982 conjecture, that quantum computers can be programmed to simulate any local quantum system, is shown to be correct.

**Table 1.** The asymptotic scaling of the number of quantum gates needed to simulate scattering in the strong-coupling regime in d = 1, 2 spatial dimensions is polynomial in p (the momentum of the incoming pair of particles),  $\lambda_c - \lambda_0$  (the distance from the phase transition), and  $n_{out}$  (the maximum kinematically allowed number of outgoing particles). The notation  $f(n) = \tilde{O}(g(n))$  means  $f(n) = O(g(n) \log^c(n))$  for some constant c.



S. Lloyd, Science (1996)

#### Quantum Algorithms for Quantum Field Theories

Stephen P. Jordan,<sup>1</sup>\* Keith S. M. Lee,<sup>2</sup> John Preskill<sup>3</sup>

Quantum field theory reconciles quantum mechanics and special relativity, and plays a central role in many areas of physics. We developed a quantum algorithm to compute relativistic scattering probabilities in a massive quantum field theory with quartic self-interactions ( $\phi^4$  theory) in spacetime of four and fewer dimensions. Its run time is polynomial in the number of particles, their energy, and the desired precision, and applies at both weak and strong coupling. In the strong-coupling and high-precision regimes, our quantum algorithm achieves exponential speedup over the fastest known classical algorithm.

S.P. Jordan et al., Science (2012)

#### **GAUGE THEORIES**

Theories with local symmetries (to be satisfied at every point)

CLASSICAL (electrodynamics)



#### **SIGN PROBLEM**



The current wisdom on the phase diagram of nuclear matter.

McLerran, L. Nucl.Phys.Proc.Suppl. 195 (2009) 275-280

## **LGT HAVE APPLICATIONS IN**



#### **QUANTUM SIMULATION OF HEP PROCESS**



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## CLOUD QUANTUM COMPUTING OF AN ATOMIC NUCLEI





#### **QUANTUM MACHINE LEARNING**

machine learning	quantum machine qua learning	antum information processing
simulated annealing	annealing quantum annealing	quantum gibbs sampling
markov chain monte-carlo	quantum BM quantum topological algorithms	
feed forward neural net neural nets	quantum perceptron quantum NN classification quantum clustering quantum data fitting quantum rejection sa	Quantum ODE solvers
reinforcement learning	control and metrology quantum control phase estimation hamiltonian learning	tomography

J. Biamonte et al. Nature (2017)

## **QUANTUM ALGORITHMS FOR BIG DATA**

#### Speed up

Grover like SQRT(N)

Shor algorithm (QFT) Exponential

#### Tasks

Solving sets of linear equations Ax = b

Find a concise function that approximates the data to be fitted and bound the approximation error

Support vector machine

Cluster assignment and finding

• • •

#### **QUANTUM PERCETRON**



F. Tacchino et al. npj Quantum Information (2019)

> Preparation of the system in an "easy" state



Slowly change the system Hamiltonian to reach another ground state which encodes the solution of the problem  $\downarrow\uparrow\downarrow\cdots\downarrow\downarrow\uparrow$ 

$$H_0 = -h_0 \sum_{i=1}^N s_i \qquad s_i = \{\uparrow, \downarrow\} \qquad H(t) = \left(1 - \frac{t}{T}\right) H_0 + \frac{t}{T} H_P$$



## **QUANTUM SIMULATION OF MANY-BODY CORRELATED DYNAMICS**



M. Lukin's group Nature (2017)

M. Dalmonte et al (2019)

### **OPTIMAL CROSSING OF QPT**



$$i\frac{\partial}{\partial t}|\psi(t)\rangle = (H_0 + f(t)H_1)|\psi(t)\rangle$$

 $\min_{f(t)} J(|\psi(T)\rangle)$ 

Optimal control

Fast

P. Doria et al PRL (2010), T. Caneva et al. PRA (2014)

#### **GHZ STATE PREPARATION ON RYDBERG**



#### **QUANTUM GATES ON RYDBERG ATOMS**

#### **Optimization!**

RedCRAB

PHYSICAL REVIEW LETTERS 123, 170503 (2019)

**Editors' Suggestion** 

Featured in Physics

#### Parallel Implementation of High-Fidelity Multiqubit Gates with Neutral Atoms

Harry Levine<sup>(b)</sup>,<sup>1,\*</sup> Alexander Keesling<sup>(b)</sup>,<sup>1</sup> Giulia Semeghini<sup>(b)</sup>,<sup>1</sup> Ahmed Omran<sup>(b)</sup>,<sup>1</sup> Tout T. Wang<sup>(b)</sup>,<sup>1,2</sup> Sepehr Ebadi,<sup>1</sup> Hannes Bernien,<sup>3</sup> Markus Greiner,<sup>1</sup> Vladan Vuletić,<sup>4</sup> Hannes Pichler,<sup>1,5</sup> and Mikhail D. Lukin<sup>1</sup>
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#### Toffoli Gates



Bell state preparation



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# When do we really need a quantum simulation/computation?



## **TENSOR NETWORK ALGORITHMS**



- > State of the art in  $1 \mathbf{K}(\mathbf{paye})$
- ► No sign problem
- Extended to open quantum systems
- ► Machine learning
- ► Data compression (BIG DATA)
- Extended to lattice gauge theories
- Simulations of low-entangled systems of hundreds qubits!

U. Schollwock, RMP (2005)

A. Cichocki, ECM (2013) I. Glasser, et al. PRX (2018)



T. Pichler, E. Rico, M. Dalmonte, P. Zoller, and SM, PRX (2016)



Real time

**MESONS SCATTERING** 

T. Pichler, E. Rico, M. Dalmonte, P. Zoller, and SM, PRX (2016)

## TWO DIMENSIONAL SIMULATION OF A LGT AT FINITE DENSITY



## **COMPARISON WITH MACHINE LEARNING**



G. Carleo, MIV Trogera Sciencea (2013) (2018)

M. Collura et al, arXiv:1905.11351

## TN MACHINE LEARNING OF HEP DATA

**Hypothesis class:**  $f^{\ell}(\bar{x}) = \mathbf{W}^{\ell} \cdot \Phi(\bar{x})$ 

$$f^{\ell}(\bar{x}) = \sum_{\mathbf{s}} W^{\ell}_{s_1 s_2 \dots s_N} \phi(x_1)^{s_1} \phi(x_2)^{s_2} \dots \phi(x_N)^{s_N}$$

 $f^\ell$  map input data to the space of labels

**PROBLEM:** W is a N+1 order tensor that grows exponentially with the input data





#### **TENSOR NETWORK MACHINE LEARNING OF HEP EVENT**



In collaboration with A. Gianelle, D. Lucchesi, L. Sestini, D. Zuliani

## TAKE HOME MESSAGE

- Quantum technologies are fast developing, hybrid solutions will play a fundamental role
- Tensor network algorithms can be used to benchmark, verify, support and guide quantum simulations/computations
- Synergies between quantum technologies and high-energy physics can lead to unexpected developments:
  - Sign-problem-free solutions
  - ► Machine learning
  - > Quantum sensing
  - Optimized protocols
- Quest for quantum advantage is open

