EXTREME SIMULATIONS FOR QUANTUM TECHNOLOGIES

F

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Verona 25/10/18

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25

20

815

10

5

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100

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(QUANTUM) TECHNOLOGIES

1800		1950)	2000
Thermodynamics	Electromagnetism	Quantun	n mechanics	Quantum science
Engines	Circuits	Lasers, semi	conductors	2nd quantum revolution
Hz		KHz	Canal Cana	GHz ???
N COM	Zuse's Z3		Intel 4004	
Babbage difference engine	Numeric of	cal solution PDEs	ا Den ٦	attice calculations sity Functional theory ensor networks

2ND QUANTUM REVOLUTION

Quantum Technologies Timeline







Demo: IBM Quantum Experience



(±)

IBM Quantum Computing Lab Tour

QUANTUM SIMULATORS

System



Model

 $H = \sum \sigma_i^z \sigma_{i+1}^z + \sum_i \sigma_x$

Energy function

Simulator



classical

- Original Feynman's motivation for QC
- Cold atoms, trapped ions, superconductors...
- Adiabatic quantum computation



quantum

J. I. Cirac and P. Zoller, Nat. Phys. (2012).

CLASSICAL PROBLEMS AND AQC

Partitioning problem For every configuration H>0 $S = \{n_1, \dots, n_N\}$ partitioning Complete subgraph finding (clique)Is it possible to divide it in two sets of equality integer programming $E_{qs} = 0$ NP-comeletevering and packaging problems exists! Define the energy function. Minimal maximal matching... $H = A\left(\sum_{i=1}^{N} n_i s_i\right)$ the mismatch is minimized (NP-hard) $s_i = \{\uparrow, \downarrow\}$

ALL-TO-ALL TO LGT MAPPING

$$H_{I} = \sum \sigma_{x}^{[k]}$$
$$H_{F} = \sum_{i < j} V^{[i,j]} \sigma_{z}^{[i]} \sigma_{z}^{[j]}$$

$$H_F = \sum_{k=1}^{K} f^{[k]} \sigma_z^{[k]} + H_C = -\sum_{p=1}^{P} c^{[p]} \sigma_z^{[k_1]} \sigma_z^{[k_2]} \sigma_z^{[k_3]} \sigma_z^{[k_4]}$$



W. Lechner, P. Hauke, and P. Zoller, Sci. Adv.. (2015).

ADIABATIC QUANTUM COMPUTATION

Preparation of the system in an "easy" state

- $\downarrow \downarrow \downarrow \cdots \downarrow \downarrow \downarrow$
- Change of a system parameter to reach another ground state which encodes the problem solution $\downarrow \uparrow \downarrow \dots \downarrow \downarrow \uparrow$

$$H_{0} = -h_{0} \sum_{i=1}^{N} s_{i} \qquad s_{i} = \{\uparrow, \downarrow\} \qquad H(t) = \left(1 - \frac{t}{T}\right) H_{0} + \frac{t}{T} H_{P}$$
Slow
$$\int \frac{\mathsf{E}}{\mathsf{Phase II}} \int \frac{\mathsf{Adiabatic}}{\mathsf{strategy}}$$

WHAT IS NEEDED?

Quantum hardware



Validation and certification

Algorithms/Protocols



Calls for:

- Classical numerical simulations
- Optimisation
- Software engineering

CHALLENGES?

System description









NUMERICAL SIMULATIONS

The art of high-performance computing

MANY-BODY WAVE FUNCTION

 $|\psi\rangle = \psi_0|0\rangle + \psi_1|1\rangle$

 $|\psi\rangle = \psi_{00}|00\rangle + \psi_{10}|10\rangle + \psi_{01}|01\rangle + \psi_{11}|11\rangle$

 $|\psi\rangle = \psi_{000}|000\rangle + \psi_{010}|010\rangle + \psi_{001}|001\rangle + \psi_{011}|011\rangle + \psi_{100}|100\rangle + \psi_{110}|110\rangle + \psi_{101}|101\rangle + \psi_{111}|111\rangle$

N=2

N=3

N=I

$$|\psi\rangle = \sum \psi_{\alpha_1,\alpha_2,\dots,\alpha_N} |\alpha_1,\alpha_2,\dots,\alpha_N\rangle$$

MANY-BODY WAVE FUNCTION



TENSOR NETWORK STATES

"Simple" variational Ansatz to describe faithfully interesting quantum dynamics!



Adaptive system description tunable between mean field and exact

TENSOR NETWORKS

$$\psi_{lpha_1,lpha_2,...lpha_N} \quad \mathcal{O}(d^N)$$

$$A_{lpha_1}^{eta_1} A_{lpha_2}^{eta_1 eta_2} \dots A_{lpha_N}^{eta_{N-1}} \mathcal{O}(Ndm^2)$$



PEPS



Tree Tensor Network (Hierarchical Tucker decomposition)

TENSOR NETWORK ALGORITHMS



- Polynomial effort (state of the art in ID)
- No sign problem
- Works also with open system (Lindblad master equation)
- Increasingly applied in quantum chemistry
- Recently extended to Lattice Gauge Theories

GAUGE THEORIES

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

CLASSICAL (electrodynamics)



$$\rho = \vec{\nabla} \cdot \vec{E}$$

Gauss' law

$$\psi_x^{\dagger}\psi_x|\Psi\rangle = \Delta E_{x,x+a}|\Psi\rangle$$

Cirac, Lewenstein, Zoller, Verstraete, Reznik, QTFLAG...

LGT HAVE APPLICATIONS IN



simulations, ...

SU(2) LATTICE GAUGETHEORY



P. Silvi, E. Rico2 F. Tschirsich I, M. Dalmonte, and SM, Quantum I (2017)

ADDITIONAL APPLICATIONS

Tensor decomposition for Big Data and Optimization

TABLE II: Similarities and links between tensor networks (TNs) and graphical models used in Machine Learning (ML) and Statistics. The categories are not exactly the same, but they closely correspond.

Tensor Networks	Graphical Models in ML/Statistics		
TT/MPS	Hidden Markov Models (HMM)		
HT/TTNS	Gaussian Mixture Model (GMM)		
TNS/PEPS	Markov Random Field (MRF) and Conditional Random Field (CRF)		
MERA	Deep Belief Networks (DBN)		
DMRG and MALS Algs.	Forward-Backward Algs., Block Nonlinear Gauss-Seidel Methods		

Tensor networks for image processing



FIG. 3: Comparison of the original 512x512 8-bit grayscale image (upper-left) with images compressed with the MPS algorithm (upper-right, DCR=17.97, SSIM=0.8311, and lowerleft, DCR=7.64, SSIM=0.9014) and JPEG (lower-right, DCR=33.14, SSIM=0.8311). Note how the compression artifacts are characteristic of each algorithm, even though the quality measure is the same.

A. Cichocki, ECM (2013).

A. Bloque and J.I Latorre

QT, TN & MACHINE LEARNING

	machine learning		qua	quantum information processing	
	simulated annealing		qu	quantum gibbs sampling	
	narkov chain monte-carlo		quantum BM	quantum topological algorithms	
	feed forward neural net neural nets		quantum perceptron	quantum PCA quantum SVM quantum NN classification quantum clustering quantum data fitting	Quantum ODE solvers
		quantum rejection sampling / IIIIL			
reinforcement learning control and metrology number of tomography hamiltonian learning					

J. Biamonte et al. Nature (2017).

I. Glasser, et al. Phys. Rev. X (2018).

GAME OF LIFE (CONWAY'S)

Cellular automata

Two states: "dead" or "alive"

Set of simple rules that generate complexity, self-organization

Universal Turing machine

Non unitary



QUANTUM GAME OF LIFE

Unitary One dimensional

$$H = \sum_{i=3}^{L-2} (b_i + b_i^{\dagger}) \cdot (\mathcal{N}_i^3 + \mathcal{N}_i^2)$$
$$\mathcal{N}_i^2 = \sum_P n_{\alpha} n_{\beta} \bar{n}_{\gamma} \bar{n}_{\delta}$$
$$\mathcal{N}_i^3 = \sum_{P'} n_{\alpha} n_{\beta} n_{\gamma} \bar{n}_{\delta}$$

GI GI G2 G3

Blinker

Two possible states: "active only if surrounded by 2 or 3 alive sites"

D. Bleh, T. Calarco, SM EPL, 97 (2012) 20012



D. Bleh, T. Calarco, SM EPL, 97 (2012) 20012



OPTIMAL CONTROL

Steering quantum systems at your will

Classical OCT problem



Quantum OCT problem

QUANTUM OPTIMAL CONTROL



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

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

- Few-body: standard optimal control (high-accuracy, many iterations, complete knowledge...)
- Many-body: dCRAB (high-efficiency, few iterations, minimal knowledge...)

H. Rabitz et.al., NJP (2009)

P. Doria, T. Calarco, SM PRL (2011)

APPLICATIONS



Quantum annealing



Light-harvesting dynamics



Entanglement/Squeezing manipulation



RED CRAB

Server is online!



PNAS (2018)



OPTIMAL EXPERIMENTAL PROTOCOLS

Cold atoms in optical lattices, Bose-Einstein condensates on atom chip, NV-centers in diamonds, Rydberg atoms, circuit QED, Trapped ions, Light-harvesting complexes...

OPTIMAL QPT CROSSING

SUPERFLUID-MOTT INS. QPT

Optimal loading of cold atoms in optical lattices

Bose-Hubbard model with external trapping potential

P. Doria, T. Calarco, SM Phys. Rev. Lett. (2011)

I. Bloch's group (MPQ-Munich)

One dimensional tubes

EXPERIMENTAL RESULTS

Speed up of one order of magnitude Compatible with the quantum speed limit

Ulm-Munich collaboration, Scientific Reports (2016)

FUNDAMENTAL LIMITS AND COMPLEXITY

Where no man has gone before

PHYSICAL LIMITS

Reachability

Directions available

- Energy
 - Power available
- Information

Channel capacity

• Dynamical Lie algebra $V^{(0)} \equiv [[iH'_0, iH_1], iH'_0]$ $V^{(\ell-1)} \equiv \hat{[}[iH'_0, V^{(\ell-2)}], iH'_0] - \omega_1^2 V^{(\ell-2)}$

- Energy-time uncertainty
 - $\Delta E \Delta t \ge \hbar/2$
- Shannon-Hartley theorem $k_s = \log(1 + S)$

OPTIMAL CONTROL COMPLEXITY

- The (smoothed) complexity of an optimal control problem scales polynomially with the size of (time-polynomial) reachable states D_{W^+}
- Efficiently simulatable dynamics (Integrable, TN, DMFT, HF-like, etc...) can be efficiently optimally controlled
- Final error scales exponentially with the optimal control bandwidth

$$\varepsilon \ge 2^{-\frac{T \,\Delta\Omega \,\kappa_S}{D}} \mathcal{W}^+$$

The optimal control bandwidth, total time and the size of the set of reachable states are such that:

TAKE HOME MESSAGES

- Tensor network algorithms field of application is rapidly expanding
- What can be simulated can be controlled
- A large class of "complex quantum systems" can be efficiently characterized, simulated and controlled.
- We aim to support the development of quantum technologies and to study novel interesting phenomena

OUTLOOK

Every journey begins with a single step

OPEN QUESTIONS

- New tools and algorithms to simulate quantum systems
- Software engineering and optimization
- New protocols for quantum technologies
- Fundamental limits

...

- Classical versus quantum annealing
- Quantum advantage definition
- Universal quantum computers?

VISION

Extreme simulations of quantum technologies

> Software development

> > System modelling

Algorithm

Computer science Quantum mechanics

Physics

NECESSARY KNOW-HOW

- Algorithms
- Software engineering
- Complexity theory
- Graph theory
- Optimization

....

High performance computing

- Quantum information theory
- Many-body quantum systems
- Quantum hardware
- Tensor networks
- Optimal control

....

Parallel computations

OPTIMAL ENGINEERING OF COMPLEX QUANTUM PHENOMENA

- Dynamics of quantum phase transition
- Optimal driving of correlated matter
- Quantum chemistry
- Quantum transport
- Quantum thermodynamics
- Quantum technologies: sensing (NVcenters, graphene nano-ribbons), Quantum simulations (of LGT)...

HOME

NEWS

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New Trends in Complex Quantum Systems Dynamics

9-12 April 2019

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Submit a talk or poster

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