				11.									•	•	•	
				•												•
		11 et (²⁴⁶ -1) 201-														
H R_2	\dots R_{n-1} n_n H \dots H	Rn-2 Rn-1	CALE R	- //												
	What											16				
	wnat		WIUN	an	ea		le		Q	U						
			CO	mn	117	ar										
				1000												



VECTOR INSTITUT INSTITUTE VECTEUR





the **matter lab**



Alán Aspuru-Guzik

Professor of Chemistry and Computer Science

University of Toronto

Vector Institute for Artificial Intelligence

Early classical mechanical simulators

Antykythera Mechanism circa 200 BC



Digital computer simulators



Without the computer-based simulation, the material culture of late-twentieth-century microphysics is not merely inconvenienced – It does not exist. [...] Machines [...] are inseparable from their virtual counterparts – all are bound to simulations.

-Peter Galison

From *Image and Logic: A material culture of microphysics* (1997) "[H]e produced a paper tape of his whole computer program and unrolled it along the length of the chemical lecture bench. There, in one roll, was something, of which one could ask a chemical question at one end and it would produce an answer at the other! . . . most of the audience probably thought the demonstration bizarre. But it was prescient"

Handy, Pople, Shavitt, JPCA (1996)



Richard P. Feynman, Simulating Physics with Computers Int. J. Theor. Phys. (1982) Think different.

267

T

ALC: NO

THE



The quantum simulation way



• • • • • • •

Disruption and quantum simulation



Quantum computers intro



quantum bit

Superposition

Entanglement

Collapse upon measurement

quantum computer

Collection of controllable qubits

Subject to decoherence

Ability for quantum error correction

Quantum computation



Quantum gates and circuits



Subroutines

Quantum Fourier Transform



Quantum computing with quantum circuits in one slide

Information is stored on qubit registers





Example of a universal gate set



Quantum compujting for Chemistry circa 2005



A. Aspuru-Guzik, A. D. Dutoi, P. J. Love, M. Head-Gordon, Science (2005) Full quantum circuit: J. D. Whitfield, et. al., Mol. Phys. (2011) Error correction: N. Cody Jones, J. D. Whitfield, et al. New. J. Phys.(2012)

Noisy intermediate scale quantum (NISQ) computers



"Quantum computing in the NISQ era and beyond" Preskill, 2018 https://aniv.org/abs/1801.00862

6

Noisy intermediate scale quantum (NISQ) algorithms

Noisy intermediate-scale quantum (NISQ) algorithms

Kishor Bharti,^{1, *} Alba Cervera-Lierta,^{2,3, *} Thi Ha Kyaw,^{2,3, *} Tobias Haug,⁴ Sumner Alperin-Lea,³ Abhinav Anand,³ Matthias Degroote,^{2,3,5} Hermanni Heimonen,¹ Jakob S. Kottmann,^{2,3} Tim Menke,^{6,7,8} Wai-Keong Mok,¹ Sukin Sim,⁹ Leong-Chuan Kwek,^{1,10,11,†} and Alán Aspuru-Guzik^{2,3,12,13,‡}

arXiv:2101.08448v1

Quantum Chemistry in the Age of Quantum Computing

Yudong Cao, Jonathan Romero, Jonathan P. Olson, Matthias Degroote, Peter D. Johnson, Mária Kieferová, Ian D. Kivlichan, Tim Menke, Borja Peropadre, Nicolas P. D. Sawaya, Sukin Sim, Libor Veis, and Alán Aspuru-Guzik*

Cite this: Chem. Rev. 2019, 119, 19, 10856–10915 Article Views Altmetric Citations

Quantum computational chemistry

Sam McArdle, Suguru Endo, Alán Aspuru-Guzik, Simon C. Benjamin, and Xiao Yuan Rev. Mod. Phys. **92**, 015003 – Published 30 March 2020



Sha

Variational quantum algorithms

Variational quantum eigensolver

Peruzzo et al.

Nat Comm (2014)

Quantum adiabatic optimization algorithm Farhi et al. (2014)



Quantum autoencoders

Romero, et al. Quantum Sci. Technol. (2017)



Variational Quantum Generators

Romero, et al. arXiv:1901.00848 (2019)



NISQ algorithms: many body physics

Algorithm/Application	Proposed implementations
Variationa	l quantum eigensolver (VQE) and related solvers
VQE	(McClean et al., 2016; Peruzzo et al., 2014; Wecker et al., 2015)
Adaptive VQE	(Grimsley et al., 2019b; Ryabinkin et al., 2018b; Sim et al., 2020)
IQAE	(Bharti, 2020; Bharti and Haug, 2020a)
Krylov approaches	(Huggins et al., 2020; Jouzdani and Bringuier, 2020; Stair et al., 2020)
v **	(Bharti and Haug, 2020b; McArdle et al., 2019a; Motta et al., 2020; Sun
Imaginary time evolution	<i>et al.</i> , 2020c)
	VQE for excited states
Folded spectrum	(Peruzzo et al., 2014; Ryabinkin et al., 2018a)
Orthogonally constrained VQE	(Higgott et al., 2019; Kottmann et al., 2020b; Lee et al., 2018)
Subspace expansion and	
linear-response based	(McClean et al., 2017; Ollitrault et al., 2020; Takeshita et al., 2020)
Subspace-search VQE	(Nakanishi et al., 2019)
Multistate contracted VOE	(Parrish et al., 2019a)
Fourier transform of evolutions	(Aleiner et al., 2020; Roushan et al., 2017)
WAVES	(Santagati et al. 2018)
Adiabatically-Assisted	(Garcia-Saez and Latorre 2018: McClean <i>et al.</i> 2016)
Turabatically Tissisted	Hamiltonian simulation
	(Endo et al. 2020c: Kubo et al. 2020: Li and Benjamin. 2017: McArdle
Variational quantum simulator	et al. 2019a: Yuan $et al. 2019$
Subspace variational quantum	
simulator	(Here et al. 2019)
Variational fast forwarding	(Cirstoiu et al. 2020: Commean et al. 2020)
Quantum assisted simulator	(Christolii et al., 2020, Commeat et al., 2020) (Bharti and Haug, 2020b)
Quantum assisted simulator	(Diatri and Haug, 2020b)
Scrambling	(Holmos at al. 2020: Joshi at al. 2020: Londsmon at al. 2010)
Thermal state	(Vorden et al. 2010b)
Thermai state	(Verdoli et al., 2019b)
Concerne line of an existing of an external	Open quantum systems
cimulator	$(\mathbf{E}_{1} + \mathbf{e}_{1} + \mathbf{e}_{2}) = 0.200 \mathbf{e}_{1} \mathbf{E}_{1} \mathbf{e}_{2} \mathbf{e}_{2} \mathbf{e}_{1} \mathbf{e}_{2} \mathbf{e}_{2} \mathbf{e}_{1} \mathbf{e}_{2} \mathbf{e}_$
Conceptional amontum assisted	(Endo et al., 2020c, End et al., 2020b, Tuan et al., 2019)
Generalized quantum assisted	(II
simulator Thattan inclution	(Haug and Bharti, 2020)
Trotter simulation	(Hu et al., 2020; Koppenhoier et al., 2020)
	State preparation
NT 111 - 1 - 1	(Yoshioka et al., 2020)(Endo et al., 2020b; Jaderberg et al., 2020; Kreula
Non-equilibrium steady state	$\frac{et al., 2016}{2016}$
Gibbs-state	(Chowdhury et al., 2020; Endo et al., 2020c; Haug and Bharti, 2020)
Many-body ground state	(Ho and Hsieh, 2019; Ho <i>et al.</i> , 2019; Wauters <i>et al.</i> , 2020a)
	Quantum autoencoder
	(Bondarenko and Feldmann, 2020; Bravo-Prieto, 2020; Huang <i>et al.</i> , 2020a,b;
Quantum autoencoder	Pepper et al., 2019; Romero et al., 2017)
	Quantum computer-aided design
Optical setups	(Kottmann et al., 2020c)
Superconducting circuits	(Kyaw <i>et al.</i> , 2020b)

Table I NISQ algorithms for Many-body physics and chemistry applications from Sec. V.A.

NISQ algorithms: machine learning and optimization

Algorithm/Application	Proposed implementations
	Supervised learning
	(Havlíček et al., 2019; Kusumoto et al., 2019; Schuld et al., 2020b; Schuld
Quantum kernel methods	and Killoran, 2019)
	(Farhi and Neven, 2018; Lloyd et al., 2020; Mitarai et al., 2018; Pérez-Salinas
Variational quantum classifiers (VQC)	et al., 2020a; Schuld et al., 2020a,c; Vidal and Theis, 2019)
Encoding strategies in VQA	(Cervera-Lierta et al., 2020; Mitarai et al., 2019)
	(Chien and Whitfield, 2020; Fujii and Nakajima, 2017; Ghosh et al., 2019;
	Mitarai et al., 2018; Nakajima et al., 2019; Negoro et al., 2018; Nokkala
Quantum reservoir computing	et al., 2020)
Supervised QUBO classifier	(Li et al., 2018)
	Unsupervised learning
Quantum Boltzmann machines	
(QBM)	(Amin et al., 2018; Kieferová and Wiebe, 2017; Zoufal et al., 2020)
	(Alcazar et al., 2020; Benedetti et al., 2019a; Coyle et al., 2020a; Hamilton
Quantum circuit Born machines	et al., 2019; Leyton-Ortega et al., 2019; Liu and Wang, 2018; Rudolph et al.,
(QCBM)	2020)
Quantum generative adversarial	(Dallaire-Demers and Killoran, 2018; Hu et al., 2019; Lloyd and Weedbrook,
networks (QGAN)	2018; Romero and Aspuru-Guzik, 2019; Situ <i>et al.</i> , 2020; Zeng <i>et al.</i> , 2019)
	Reinforcement learning
	(Albarrán-Arriagada et al., 2020; Cárdenas-López et al., 2018; Chen et al.,
	2020; Crawford et al., 2016; Jerbi et al., 2019; Lamata, 2017; Lockwood and
Reinforcement learning	Si, 2020a,b; Yu <i>et al.</i> , 2019)

Table II NISQ algorithms for machine learning applications from Sec. V.B.

Algorithm/Application	Proposed implementations
	(Bravyi et al., 2019; Farhi et al., 2014; Hastings, 2019; Headley et al., 2020;
Max cut	Otterbach et al., 2017)
Max clique	(Arrazola and Bromley, 2018; Banchi et al., 2020a)
Maximum independent set	(Choi et al., 2020; Saleem, 2020; Utkarsh et al., 2020)
Max hafnian	(Arrazola et al., 2018)
Vertex cover	(Cook et al., 2019)
Exact cover	(Bengtsson et al., 2020; Garcia-Saez and Latorre, 2018; Vikstål et al., 2020)
Knapsack	(de la Grand'rive and Hullo, 2019)
Graph multi-coloring	(Oh <i>et al.</i> , 2019)

Table III NISQ algorithms for combinatorial optimization from Sec. V.C.

NISQ algorithms: numerical solvers, finance

Algorithm/Application	Proposed implementations
Factoring	(Anschuetz et al., 2019; Karamlou et al., 2020)
SVD	(Bravo-Prieto et al., 2020; Wang et al., 2020c)
Linear systems	(Bravo-Prieto et al., 2019; Huang et al., 2019; Xu et al., 2019b)
	(Gaitan, 2020; Haug and Bharti, 2020; Kyriienko et al., 2020; Lubasch et al.,
Non-linear differential equations	2020)

Table IV NISQ algorithms for numerical solvers applications from Sec. V.D.

Algorithm/Application	Proposed implementations
	(Bouland et al., 2020; Cohen et al., 2020; Egger et al., 2020a; Marzec, 2016;
Portfolio optimization	Rosenberg et al., 2016; Venturelli and Kondratyev, 2019)
Fraud detection	(Egger et al., 2020a,b; Zoufal et al., 2020)
Option pricing	(Kubo <i>et al.</i> , 2020)

Table V NISQ algorithms for finance applications from Sec. V.E

NISQ algorithms: other applications

Algorithm/Application	Proposed implementations
	Quantum foundations
Bell inequalities	(Alsina and Latorre, 2016)
Contextuality	(Kirby and Love, 2019; Kirby <i>et al.</i> , 2020)
Variational consistent history (VCH)	(Arrasmith <i>et al.</i> , 2019)
	Quantum optimal control
	(Dive et al., 2018; Li et al., 2017a; Lu et al., 2017; Magann et al., 2021;
Quantum optimal control	Policharla and Vinjanampathy, 2020)
	Quantum metrology
	(Beckey et al., 2020; Kaubruegger et al., 2019; Koczor et al., 2020; Ma et al.,
Quantum metrology	2020; Meyer <i>et al.</i> , 2020)
	Fidelity estimation
Fidelity estimation	(Cerezo $et al., 2020a$)
	Quantum error correction (QEC)
Quantum variational error corrector	
(QVECTOR)	(Johnson et al., 2017)
Variational circuit compiler for QEC	(Xu et al., 2019a)
	Nuclear physics
	(Avkhadiev et al., 2020; Dumitrescu et al., 2018; Hauke et al., 2013; Klco
	et al., 2018; Kokail et al., 2019; Liu and Xin, 2020; Martinez et al., 2016;
Nuclear physics	Roggero et al., 2020)
	Entanglement properties
Schmidt decomposition	(Bravo-Prieto et al., 2019; Wang et al., 2020a)
Multipartite entanglement	(Pérez-Salinas <i>et al.</i> , 2020b)
Entanglement spectrum	(Cerezo <i>et al.</i> , 2020b; LaRose <i>et al.</i> , 2019)

Table VI NISQ algorithm for other applications listed in Sec. V.F.

Variational Quantum Algorithms (VQAs)



Peruzzo, McClean, et al. Nature Communications 5 4213 (2014)

												•
												•
												•
	10											
The Variational (Juantum E	· (n	S	\cap	JF	٦r					
	- \	<u> </u>		$\mathbf{\bigcirc}$	$\mathbf{\nabla}$	V N						

.

Variational Quantum Eigensolver



Peruzzo, McClean, et al. Nature Communications 5 4213 (2014)

How are things done under the hood in a VQE? send instructions Generators: Hermitian Operators $L = \langle H \rangle^{2}$ $\frac{dL}{d\Theta} = 2 \langle H \rangle \frac{d \langle H \rangle}{d\Theta}$ Quantum Computer $G_{abkl} = i(a_a^{\dagger}a_i a_b^{\dagger}a_j - h.c.)$ sample Circuits from unitaries expectation values $U(\theta) = e^{-i\frac{\theta}{2}G}$ Fermionic operators are mapped to Pauli strings $a_k^{\dagger} = 1^{\otimes k-1} \sigma_k^- \sigma_Z^{\otimes n-k}$ Variational optimization

 $\min_{\theta} \left(\langle H \rangle_{U_{\theta}} \right) \equiv \min_{\theta} \left(\langle 0 | U^{\dagger} \left(\theta \right) H U \left(\theta \right) | 0 \rangle \right)$





• • • • •

Example: The Measurement reduction race





Total: Pauli strings in a Hamiltonian Qubit-wise commuting: Izmaylov, et al. JCP (2020). Fully commuting: Izmaylov, et al. JCTC (2020). Google+Berkeley: Huggins, et al. arXiv (2020). Cartan subalgebra: T.C. Yen, A.F. Izmaylov, arXiv (2020).

Plot : TC Yen and Artrur Izmaylov (UofT)

The Tequila Package



https://github.com/aspuru-guzik-group/tequila

Quantum Science and Technology

ACCEPTED MANUSCRIPT

TEQUILA: A platform for rapid development of quantum algorithms.

Jakob Kottmann¹, Sumner Alperin-Lea², Teresa Tamayo-Mendoza³, Alba Cervera-Lierta⁴, Cyrille Lavigne⁵, Tzu-Ching Yen⁶, Vladyslav Verteletskyi⁷, Philipp Schleich⁸, Abhinav Anand⁹, Matthias Degroote⁴ + Show full author list Accepted Manuscript online 11 February 2021 • © 2021 IOP Publishing Ltd

What is an Accepted Manuscript?

The Tequila Package: Hello world



active = {"b1u": [0, 1]}
mol = tq.chemistry.Molecule("beh2.xyz", "6-31g", active)
H = mol.make_hamiltonian()
U = tq.gates.Ry("a", 0)
U += tq.gates.CNOT(0, 1) + tq.gates.CNOT(0, 2)
U += tq.gates.CNOT(1, 3) + tq.gates.X([2, 3])
expv = tq.ExpectationValue(U, H)
result = tq.optimizer_scipy.minimize(expv, "bfgs")
wfn = tq.simulate(U, variables=result.angles)

• • • • •

The Tequila Package: Benzene example

https://github.com/aspuru-guzik-group/tequila

• • •





The Tequila Package: How it works







Multi-resolution analysis Variational Quantum Eigensolver



• • Kottman, Schleich, Tamayo-Mendoza, Aspuru-Guzik J Phys. Chem. Lett. 2021, 12, 1, 663-673

Multi-resolution analysis Variational Quantum Eigensolver



Hydrogen molecule VQE/MRA

System	Metric	$\operatorname{Qubits}/\operatorname{MRA}$	$\operatorname{Qubits}/\operatorname{GBS}$	More
He	MAX	4	4-10	Fig. 3
Be	MAX	10	10-18	Fig. 3
H_2	NPE	4	20-56	Figs. 5, 4
H_2	NPE	8	20-56	Figs. 5, 4
H_2	NPE	20	56 - 120	Figs. 5, 4
H_2	MAX	4	8	Figs. 5, 4
H_2	MAX	8	20-56	Figs. 5, 4
H_2	MAX	20	56	Figs. 5, 4
LiH	NPE	12	20-38	Figs. 5, 4
LiH	NPE	20	38	Figs. 5, 4
LiH	MAX	12	20-38	Figs. 5, 4
LiH	MAX	20	170-288	Figs. 5, 4
BH	NPE	12-20	38-88	Figs. 5, 4
BH	MAX	12-20	38-88	Figs. 5, 4
NH ₃	ΔE	12-18	58-100	Fig. 2

Mutual-Information Adaptive Unitary Coupled Cluster



Strategy:

Leverage classical-computing DMRG mutual information to iteratively construct compact entangler circuits.



Also: Qubit-Coupled Cluster by Izmaylov, and co-workers, Adapt-VQE by Mayhall and co-workers

• • • Zhang, Kyaw, Kottmann, Degroote, Alan Aspuru-Guzik Quant. Sci. Tech. (2021) Accepted manuscript online.



Strategy: Add encoding layer to learn parameterized Hamiltonians: **QML + VQE** hybrid. See also: Mitarai, et al. PR Applied 11 044087 (2019)

Cervera-Lierta, Kottman, Aspuru-Guzik arXiV:2009.13545



$$|0\rangle^{\otimes n} \not\xrightarrow{n} \mathcal{S}\left(\vec{\lambda}, \vec{\Phi}_{opt}\right) - \mathcal{U}\left(\vec{\Theta}_{opt}\right) - \overline{\langle H(\vec{\lambda}) \rangle}$$

Strategy: Add encoding layer to learn parameterized Hamiltonians: **QML + VQE** hybrid. See also: Mitarai, et al. PR Applied 11 044087 (2019)



Strategy: Add encoding layer to learn parameterized Hamiltonians: **QML + VQE** hybrid. See also: Mitarai, et al. PR Applied 11 044087 (2019)

Cervera-Lierta, Kottman, Aspuru-Guzik arXiV:2009.13545





Example: H₄ molecule as the atomic square is deformed into a rectangle

A feasible approach for derivatives for UCC



enerator form	Gradient cost	Strategy
$p_{\mathbf{pq}} = \sum c_i \boldsymbol{\sigma}_i$	$\mathscr{O}(2^{2n})$	Shift-rule (6)
$T_{ m pq} = rac{1}{2}(G_+ + G)$	4	Fermionic-shift (16)
eal wavefunctions $a_{m} = \frac{1}{2}(G_{1} + G_{2})$	2	Fermionic-shift (19)
enerator approximation $pq \approx G_{\pm}$	2	Shift-rule (6)



H₂ minimal basis 4 orbital example

Kottman, Anand, Aspuru-Guzik Chemical Science (2021) In Press. Hot Article https://doi.org/10.1039/D0SC06627C

							•	•		
									•	
									•	
									٠	

Quantum Computer-Aided Design (QCAD)

Kyaw, Menke, Sim, Sawaya, Oliver, Guerreschi, Aspuru-Guzik, arXiV:2006.03070 (2020)

Kottman, Krenn, Kyaw, Alperin-Lea, Aspuru-Guzik arXiv:2006.03075 (2020)

Kottman, Krenn, Tischler, Aspuru-Guzik, arXiV:2005.06443 (2020)

Sawaya, Menke, Kyaw, Johri, Aspuru-Guzik, Guerresci, npJ Quantum Information 6, 1 (2020)

Enter Sycamore



Courtesy: Google Quantum

Google researchers claim to have attained "quantum supremacy" for the first time. Their 53-bit quantum computer, named Sycamore, took 200 seconds to perform a calculation that, according to Google, would have taken the world's fastest supercomputer 10,000 years. Technology Review

77

... and unconfirmed Gaussian Boson Sampling machine!

South China Morning Post



Chinese quantum computer declared a million times greater than Sycamore

- Physicist Pan Jianwei says his team achieved quantum supremacy but 'further verification' is necessary
- Pan's team has received generous and consistent financial support from the Chinese government



How will we simulate large-scale quantum computers?



Number of transmons in a processor is growing exponentially

Case study: Transmons.

How will we simulate large-scale quantum computers?



- Number of transmons in a processor is growing exponentially
- Classical hardware simulation capacity hits roadblock just above ten transmons

Case study: Transmons.

How will we simulate large-scale quantum computers?



- Number of transmons in a processor is growing exponentially
- Classical hardware simulation capacity hits roadblock just above ten transmons
- Quantum computers can simulate one *physical* transmon per $\log_2(16) = 4$ *data* qubits in the computer
- Quantum simulation capabilities will soon surpass classical capacity

Future quantum computers need to be designed with the aid of existing quantum computers

Case study: Transmons.

Methods for simulating transmon hardware on digital quantum computers

- Encoding of the transmon Hamiltonian into Pauli strings
- Energy spectrum from variational simulations
- Gate operation from Suzuki-Trotter simulations
- Quantum simulation of large quantum computer modules

Kyaw, Menke, Sim, Sawaya, Oliver, Guerreschi, Aspuru-Guzik, arXiV:2006.03070 (2020)

Efficient encoding of d-level Hamiltonians into Pauli Strings





Gray encoding is simpler and more efficient.

Hamming distance = 1

N. Sawaya et al., npj Quantum Information 6, 1 (2020)

Efficient encoding of d-level Hamiltonians into Pauli Strings

$$\hat{H}_{\text{transmon}} = 4E_{\text{C}}\hat{N}^2 - 2E_{\text{J}}|\cos(2\pi\Phi_{\text{ext}}/\Phi_0)| \frac{\cos\left(2\pi\hat{\phi}/\Phi_0\right)}{2\pi\hat{\phi}/\Phi_0}$$



Mapped Hamiltonian can be simulated on any type of digital quantum computer (superconducting, ion traps, quantum optics, cold atoms, etc.)

Kyaw, Menke, Sim, Sawaya, Oliver, Guerreschi, Aspuru-Guzik, arXiV:2006.03070 (2020)

QCAD Example: Qubit Energy Levels



physical qubit



data qubits used for simulation of physical qubit Ansatz gates: Fully-connected ion-trap qubits

Find ground-state by VQE, Determine Excited States by Variational Quantum Deflation (VQD)

Y. Cao et al., Chem. Rev. 119, 10856 (2019) O. Higgot et al., Quantum 3, 156 (2019)

QCAD Example: Qubit Energy Levels



- VQD algorithm finds transmon energy levels to experimentally relevant accuracy
- Spectrum informs frequency range, noise sensitivity estimates, and gate operation

Kyaw, Menke, Sim, Sawaya, Oliver, Guerreschi, Aspuru-Guzik, arXiV:2006.03070 (2020)

QCAD for quantum optical experiments





Kottman, Krenn, Kyaw, Alperin-Lea, Aspuru-Guzik arXiv:2006.03075 (2020)

QCAD for quantum optical experiments





FIG. 2. Digital simulation of a Boson sampling experiment [31]. (a) In the abstract representation of the setup each path is represented by two qubits (allowing to represent 0-3 photons in each path). The setup consists of beam-splitters (B) and phase shifters (P) and is initialized with three photons in paths a, c and $d(|1_a0_b1_c0_d1_e\rangle)$. (b) Percentage of physically valid states (obeying photon number conservation) as an indicator of the error introduced by the Trotter expansion. (c) Simulated distribution of three photon states with each photon in a separate path. At 10 Trotter steps the error with respect to the exact quantum optical setup is about 2 percent, and consistent with the experimental results presented in [31].

Orquestra

							•	•	
									•
									•
									۰



Orquestra The Unified Quantum Operating Environment



WE CAN SOLVE THESE CHALLENGES WITH WORKFLOW MANAGEMENT TOOLS



COMPOSE

Build guantum-enabled workflows[™] from Zapata tasks, your own tasks, and open-source libraries.



OPEN SOURCE TOOLS



Local Infrastructure

CONDUCT

Smart backend deploys workflows, orchestrating across quantum and classical hardware.



HARDWARE

Supercondi Photonic qu Ion traps	ucting qubits bits*	
CL CL	QUANTUM ANNEALERS* DEDICATED ASSICAL HARDWARE* QUANTUM IRCUIT SIMULATORS	20

Zapata Cloud/Customer Environment Infrastructure (eg. Azure, AWS)

RECORD

Analyze intermediate and final data, iterate and reproduce at scale.



DATA ANALYSIS & VIZ









Using Tableau + Orquestra Data Correlation Service (ODCS) to visualize results.

ata Analytics +	Pages	iii Columns AGG(Number of qubit.					
io.zapOS.v1alpha1.qubit		⊞ Rows	ATTR(Noiseles	s) ATTR(N	oisy)		
imensions III ₽ ▼ ■ evaluate-qubit-operator ■ io.zapOS.v1alpha1.intera	Filters	Correlatio	Noisy				
io.zapOS.v1alpha1.intera					0	Noiseless	
io.ZapOS.v1alpha1.opti	Marks	2000					
Abc _ld	∼ All 员						
Abc Expression	∨ A 00						
Abc ID	^ A 0⊡	Hu)					
Abc Schema	Aut 👻	ergy					
easures +		5 1000					
io.zapOS.v1alpha1.intera	Co Size La	latio					
io.ZapOS.v1alpha1.opti		Corre		0			
io.ZapOS.v1alpha1.value	De To Sh	500					
# Noiseless			0				
NOISY Number of qubits	# M. E		0				
# Number of Records		0	0 0	0	0		
# Measure Values			4		10 12		
Parameters							
# Hartree-Fock energy (Ha)			1	vaniser of quoits			

Superconducting VQE for H₂



Used Xmon qubits to compute energy surface of molecular hydrogen Started in Hartree-Fock state, used unitary coupled cluster, got chemical accuracy

Superconducting VQE vs Phase Estimation



Predicted dissociation energy without exponentially costly compilation for first time Substantial robustness to systematic errors seen

P. O'Malley, et al. Physical Review X 6 031007 2016

Ion trap implementation



Collaboration with Rainer Blatt (Innsbruck)



Cornelius Hempel, Christine Maier, Jonathan Romero, Jarrod McClean, Thomas Monz, Heng Shen, Petar Jurcevic, Ben P. Lanyon, Peter Love, Ryan Babbush, Alán Aspuru-Guzik, Rainer Blatt, and Christian F. Roos Phys. Rev. X 8, 031022

Ion trap implementation (LiH)



Cornelius Hempel, Christine Maier, Jonathan Romero, Jarrod McClean, Thomas Monz, Heng Shen, Petar Jurcevic, Ben P. Lanyon, Peter Love, Ryan Babbush, Alán Aspuru-Guzik, Rainer Blatt, and Christian F. Roos Phys. Rev. X 8, 031022

Variational Eigensolver by IBM team!



Hardware-efficient Quantum Optimizer for Small Molecules and Quantum Magnets

Abhinav Kandala,* Antonio Mezzacapo,* Kristan Temme, Maika Takita, Jerry M. Chow, and Jay M. Gambetta IBM T.J. Watson Research Center, Yorktown Heights, NY 10598, USA (Dated: April 18, 2017)

Kandala, et al Nature 549 242 (2017)

Ground-state energy estimation of the water molecule on a trapped ion quantum computer

Yunseong Nam,^{1,*} Jwo-Sy Chen,¹ Neal C. Pisenti,¹ Kenneth Wright,¹ Conor Delaney,¹ Dmitri Maslov,² Kenneth R. Brown,^{1,3} Stewart Allen,¹ Jason M. Amini,¹ Joel Apisdorf,¹ Kristin M. Beck,¹ Aleksey Blinov,¹ Vandiver Chaplin,¹ Mika Chmielewski,^{1,4} Coleman Collins,¹ Shantanu Debnath,¹ Andrew M. Ducore,¹ Kai M. Hudek,¹ Matthew Keesan,¹ Sarah M. Kreikemeier,¹ Jonathan Mizrahi,¹ Phil Solomon,¹ Mike Williams,¹ Jaime David Wong-Campos,¹ Christopher Monroe,^{1,4} and Jungsang Kim^{1,3,†}

¹IonQ, Inc., College Park, MD 20740, USA ²National Science Foundation, Alexandria, VA 22314, USA ³Department of Electrical and Computer Engineering, Duke University, Durham, NC 27708, USA ⁴Joint Quantum Institute and Department of Physics, University of Maryland, College Park, MD 20742, USA



Y. Nam, et al. Arxiv:1902.10171 (2019)

Largest simulation to date: Hartree Fock, 12 qubits. Hydrogen chains.



Google AI Quantum and Collaborators*†

VQE Realistic estimations and work needed

Zapata





Ethanol

inol





Ethane

Methane

bb

Ethyne



Ethene

Propane

Propene Propyne

To conclude, our estimation for the number of qubits $N_{\rm q}$ necessary to obtain accurate dynamical correlation energies is at least

$$N_{\rm q} \approx 13 N_{\rm el}$$
 (16)

where $N_{\rm el}$ is the number of active electrons in the system.

Identifying challenges towards practical quantum advantage through resource estimation: the measurement roadblock in the variational quantum eigensolver

Jérôme F. Gonthier,¹ Maxwell D. Radin,¹ Corneliu Buda,² Eric J. Doskocil,² Clena M. Abuan,³ and Jhonathan Romero¹ ¹Zapata Computing, Inc., 100 Federal St., Boston, MA 02110, USA ²Applied Chemistry and Physics Centre of Expertise, BP Group Research, 150 West Warrenville Road, Naperville, IL 60563, USA ³Digital Science and Engineering, BP Innovation and Engineering, 501 Westlake Park Blvd, Houston, TX 77079, USA (Dated: December 9, 2020)

arXiV:2012.04001 (2020)

Molecule	H ₂ O	CO ₂	CH ₄	CH ₄ O	C_2H_6	C_2H_4	C ₂ H ₂	C ₂ H ₆ O	C_3H_8	C_3H_6	C ₃ H ₄
N _{el}	8	16	8	14	14	12	10	20	20	18	16
Nq	104	208	104	182	182	156	130	260	260	234	208
$K \cdot 10^{-3}$	1.9	16	1.6	8.4	8.5	6.6	3.1	24	16	23	18
$M \cdot 10^{-9}$	3.9	32	3.2	17	17	13	6.2	48	31	46	36
t (days)	2.3	39	1.9	18	18	12	4.6	71	47	62	44

TABLE IV. Estimated runtimes *t* in days for a single energy evaluation using the number of measurements *M* from extrapolated values of *K* (Equation 17 and Table III), with $\epsilon = 0.5$ mHa and the effect of RDM constraints included by a factor of 1/2 (see Equation 18). The number of qubits N_q is computed from the number of active electrons N_{el} and our empirical estimations of active space size (Equation 16).

Gonthier et al arXiV:2012.04001 (2020)

Timeline of key events: Quantum chemistry on quantum computers

chemica

uantum



IBM Roadmap

.

Scaling IBM Quantum technology



IBM Q System One (Released)		(In development)		Next family of IBM Quantum systems		
2019	2020	2021	2022	2023	and beyond	
27 qubits Falcon	65 qubits Hammingbird	127 qubits Eagle	433 qubits Osprey	1,121 qubits Condor	Path to 1 million qubit and beyond Lorge scale systems	
			\sim			
Key advancement		Key advancement	Key advancement		Ney advancement	
Optimized lattice	Scalable readout	Novel packaging and controls	Miniaturization of components	Integration	Build new infrastructure, quantum error correction	

https://www.ibm.com/blogs/research/2020/09/ibm-quantum-roadmap/

Sponsors: Canada 150 Research Chairs, Google Focused Award, DOE BES, Intel DARPA AMD ARPA-E, Samsung, NSF, Tata Steel ARO, ONR, AFOSR, Samsung, Sloan Foundation, Camille and Henry Dreyfus Foundation, DTRA, DARPA, Anders Froseth

http://matter.toronoto.edu Twitter: A_Aspuru_Guzik aspuru@utoronto.ca







Aspuru-Guzik Group

