

Exploration of Quantum Computing for Fusion Energy Science Applications

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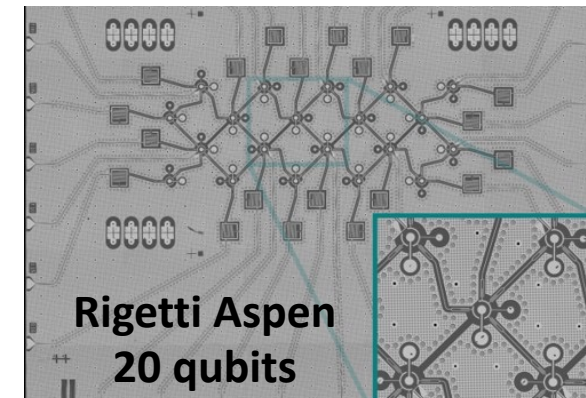
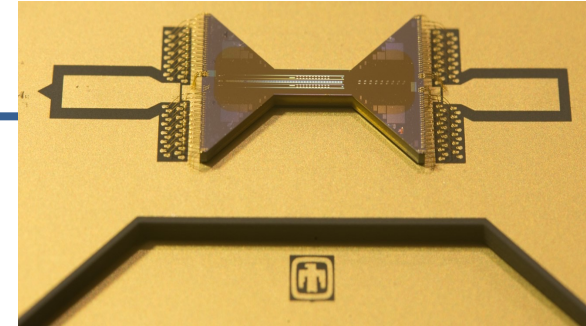
65th Annual Meeting of the
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Quantum computing may be a game-changer for fusion and science in general

- **Polynomial to exponential gains in memory and computational power**
 - Exponential speedup for the Fourier transform, linear solvers, factoring integers, ...
 - Quadratic speedup for unstructured search, optimization, sums & integrals, ...
- **Great progress has been made on quantum hardware & technology**
 - Multiple platforms: ion traps, neutral atom traps, superconducting circuits, ...
 - Google, IBM, & others now claim to have achieved **quantum supremacy** ...
- **But, we are still in the Noisy Intermediate-Scale Quantum (NISQ) era**
 - Many qubits, but no error correction
 - 1% error rate per gate → can only perform ~100 gate operations

Sandia Peregrine
6 qubit ion trap



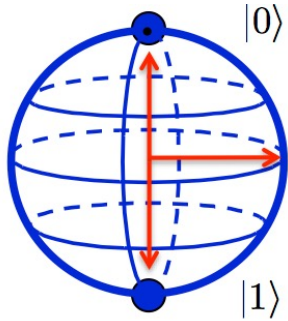
Rigetti Aspen
20 qubits



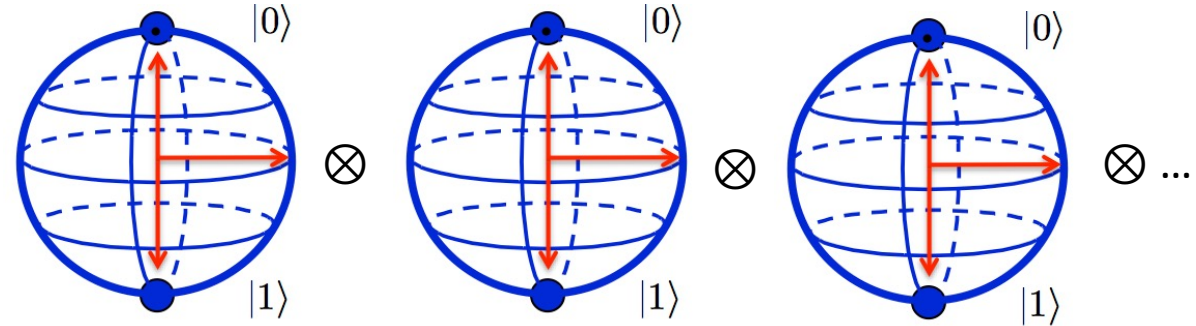
IBM Eagle
127 qubits

Quantum memory registers are “exponentially large”

Qubit: Dimension 2



n Qubits \rightarrow Hilbert Space Dimension: $N = 2^n$



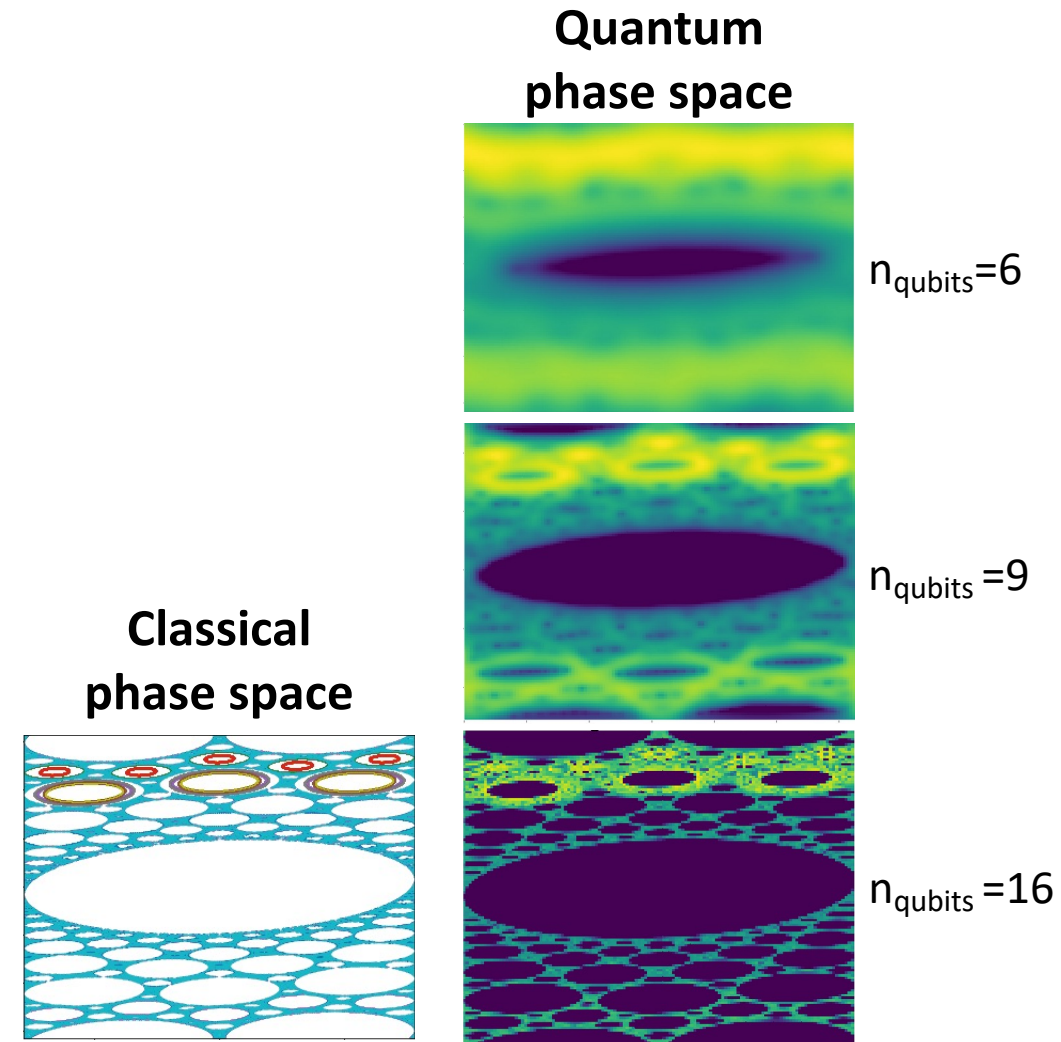
A few essential subroutines power the majority of quantum algorithms

- **Quantum Fourier Transform:** Cost of $(\log N)^2$ rather than classical $N \log N$
- **Amplitude Amplification:** Cost of \sqrt{N} rather than classical N
- **Quantum Walks:** Cost of N rather than classical N^2

Generic quadratic speedup \leq quantum density matrix DOFs / classical PDF DOFs

Outline: Quantum Computing for Fusion Science Applications

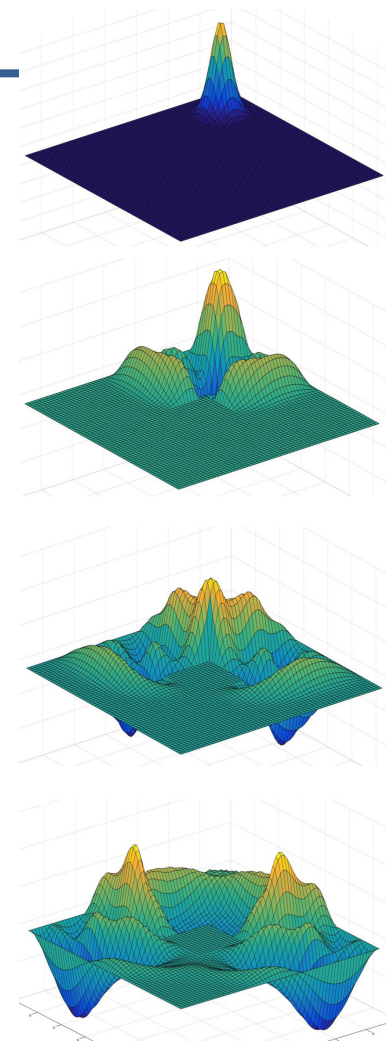
- **Quantum Algorithms for Diff Eqs**
 - Linear problems can be solved efficiently
 - Nonlinear problems can be embedded within linear problems
- **Testing Quantum Hardware Platforms**
 - Error mitigation
 - Error exploitation
- **Conclusions & Outlook**



Quantum algorithms for solving linear problems can achieve polynomial to exponential speedup

- Simple PDEs, e.g. Poisson or wave equation, have simple sparse Hamiltonians and can typically be solved with **exponential speedup**
 - The output is a wavefunction that encodes the solution
 - A few robust physical observables $\langle O_1 \rangle, \langle O_2 \rangle, \langle O_3 \rangle$
- However, outputting the data $\{\psi_x\}$ to a classical register, requires an **exponential** amount of work & reduces speedup to **quadratic** at best [1,2]
 - The same problem occurs for nontrivial initial condition and/or sources
- **“Hidden Spectral Problem”**: if you promise there is a basis in which the solution is exponentially sparse, then we can get exponential speedup
 - Like doing “X-ray crystallography”

$$|\psi\rangle = \sum_x \psi_x |x\rangle$$



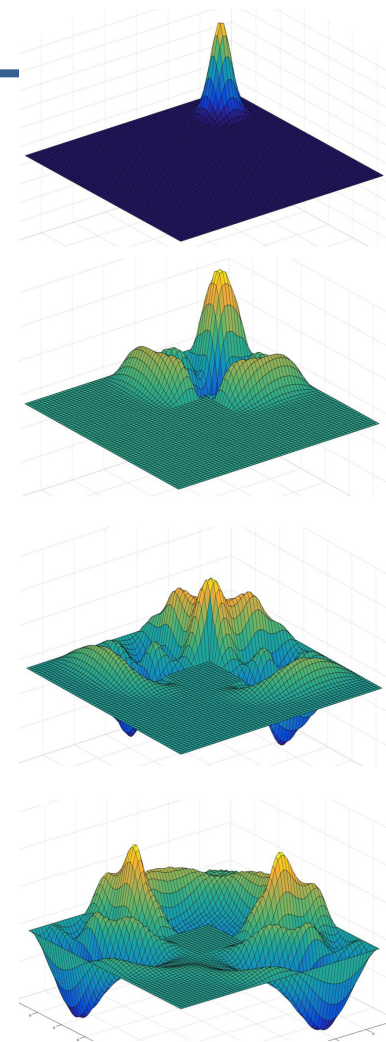
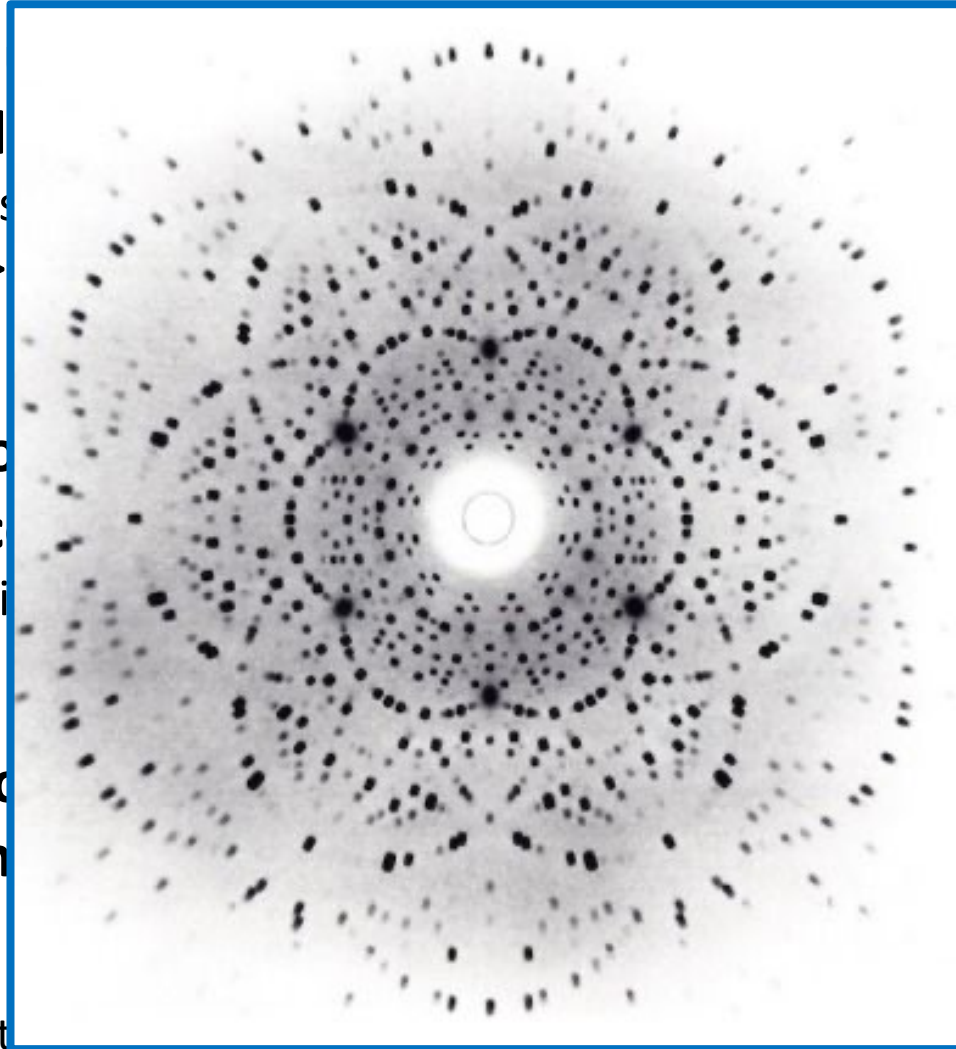
Costa, Jordan, Ostrander
Phys. Rev. A 2019

[1] Costa et al PRA 2019

[2] Montanaro & Pallister PRA 2016

Quantum algorithms for solving linear problems can achieve polynomial to exponential speedup

- Simple PDEs, e.g. Poisson or wave eq Hamiltonians and can typically be solved
 - The output is a wavefunction that encodes
 - Or a few robust physical observables $\langle O_1 \rangle$
- However, outputting the data $\{\psi_x\}$ to **exponential** amount of work & reduces speedup
 - The same problem occurs for nontrivial initial conditions
- **“Hidden Spectral Problem”**: if you produce a sparse solution is exponentially sparse, then you can’t output it
 - Like doing “X-ray crystallography”



Costa, Jordan, Ostrander
Phys. Rev. A 2019

[1] Costa et al PRA 2019

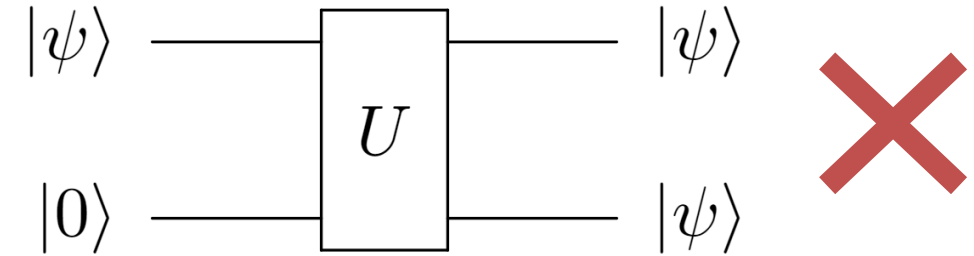
[2] Montanaro & Pallister

The No-Cloning Theorem fundamentally limits the ability of a quantum computer to efficiently compute nonlinear functions

- **No-Cloning Theorem:**

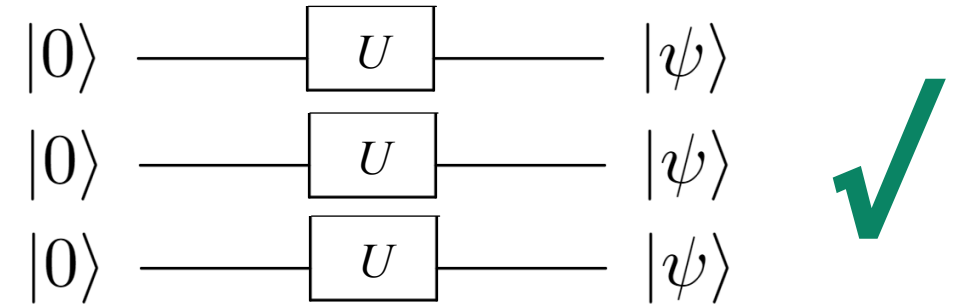
An **unknown quantum state cannot be copied**

- The only way to do this is to measure all components and prepare an identical state from scratch



- **If a state preparation process is reproducible, we can form multiple replicas of the state**

- A fault-tolerant quantum computer can run the same quantum program to create identical outputs



- **Iterative algorithms that require nonlinear operations are **exponentially** costly***

- If each iteration needs 2 replicas, then the next iteration needs 4 replicas, and T iterations needs **2^T replicas**

*S. K. Leyton, T. J. Osborne, [arXiv:0812.4423](https://arxiv.org/abs/0812.4423) (2008)

How about embedding nonlinear Diff Eqs within larger linear systems?

- **Quantization is an approximate embedding for Hamiltonian systems**
 - Dissipation can be included by embedding the system within a much larger ideal system [1]
- **Exact Koopman-von Neumann (KvN) [2] and Carleman [3] approaches**
 - The conservation law for the probability distribution function (PDF) is a perfect embedding of a nonlinear system ... within an infinite-dimensional system of equations for the PDF
 - Carleman embedding is a complex analytic form of KvN [4] that works well near fixed points
- **Special classes of PDEs may have more efficient types of embedding [5]**
 - PDEs that are reducible to ODEs can be embedded using the KvN approach for ODEs: Hamilton-Jacobi equation, advection equation
- **Integrable Diff Eqs also have special types of embedding**

[1] J. Yepez 2002, S. Lloyd 2020 [2] I. Joseph 2020 [3] Jin-Peng Liu 2021 [4] I. Joseph 2023 [5] S. Jin & N. Liu 2022

Approach # 1: Quantize the dynamics $i\hbar\partial_t\psi = \mathbf{H}\psi$

■ Point Example: Quantum Sawtooth Map*

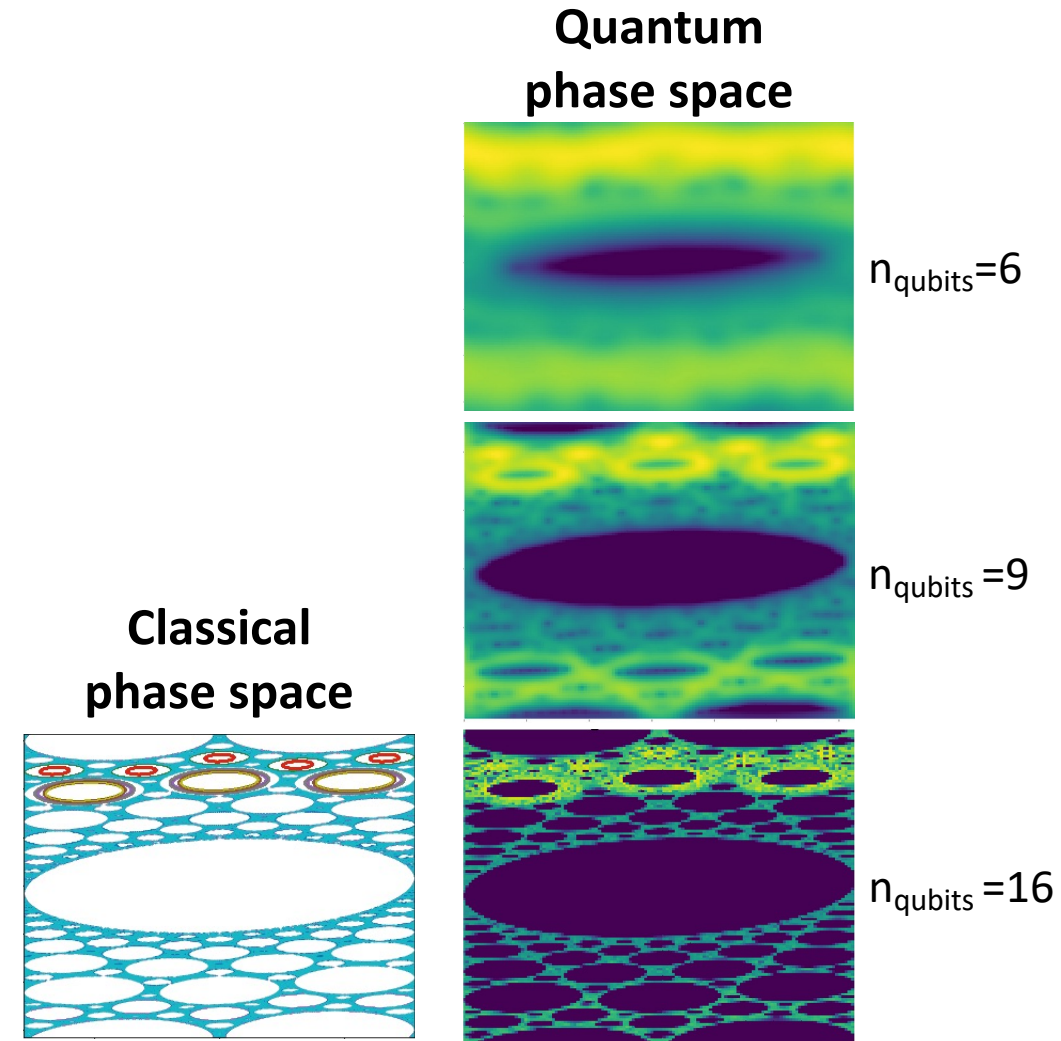
- Model for chaotic wave-particle interactions
- Converges to classical result as # of qubits increases

■ Advantages

- Quantum version may be the more accurate physical model
- Many quantum algorithms for quantum simulation
- Quantum algorithms can efficiently calculate classical quantities: Lyapunov exponent* & diffusion coefficient

■ Disadvantages

- **Quantum \neq Classical:** interference, diffraction, & tunneling
- Semiclassical limit requires very large quantum numbers
- Non-Hamiltonian systems, e.g. with dissipation, require embedding in a much larger ideal system



*M. D. Porter, I. Joseph, Quantum 6, 799 (2022)

Approach # 2: Nonlinear dynamics acts linearly on function spaces

- Consider a set of nonlinear Diff Eq's $dz/dt = V(z, t)$ with initial conditions $z_0 := z(t = 0)$

Lagrangian
picture

$$\partial_t z \Big|_{z_0} = +V \cdot \nabla z \quad \xleftrightarrow{\text{chain rule}} \quad \partial_t z_0 \Big|_z = -V \cdot \nabla z_0$$

- The advection equation expresses the evolution of a scalar function: $\theta(z, t)$

Koopman
evolution

$$\partial_t \theta \Big|_{z_0} = +V \cdot \nabla \theta \quad \xleftrightarrow{\text{chain rule}} \quad \partial_t \theta \Big|_z = -V \cdot \nabla \theta$$

Eulerian
picture

- The Liouville equation expresses conservation of probability: $f(z, t)$

$$\partial_t f \Big|_{z_0} = +\nabla \cdot (Vf) \quad \xleftrightarrow{\text{chain rule}} \quad \partial_t f \Big|_z = -\nabla \cdot (Vf)$$

Perron-Frobenius
evolution

Semiclassical wavefunction yields efficient unitary representation [1-2]

- Since quantum algorithms act naturally on wavefunctions, consider the “semiclassical” ansatz

$$\psi(z, t) = \sqrt{f(z, t)} e^{i\theta(z, t)}$$

- Where $f(z, t)$ evolves as a PDF and the phase $\theta(z, t)$ evolves as a scalar field with a source

$$\partial_t \theta \Big|_z = -V \cdot \nabla \theta + L(z, t)/\hbar$$

- Inserting the definitions leads to the “**Koopman-van Hove**” equation [1-2]

$$i\hbar \partial_t \psi \Big|_z = -i\hbar (V \cdot \nabla \psi + \nabla \cdot V \psi)/2 - L\psi$$

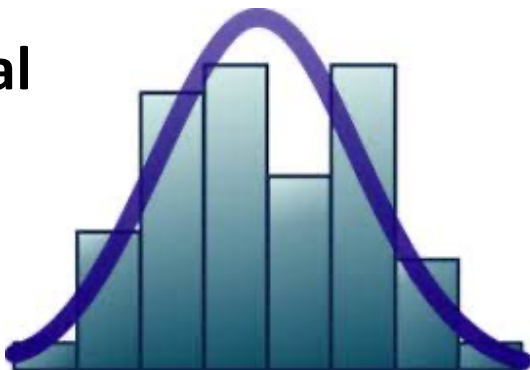
- The **classical Lagrangian** $L(z, t) = p \cdot \partial_p H - H(x, p)$ is the natural choice for semiclassical dynamics because it agrees with Feynman’s prescription for the path integral [2]

[1] I. Joseph, Phys. Rev. Research **2**, 043102 (2020)

[2] I. Joseph, [arXiv:2306.01865](https://arxiv.org/abs/2306.01865), J Phys A: Math Theor (2023)

Amplitude Estimation provides up to quadratic speedup for output

- Classical randomized Monte Carlo algorithms also provide an exponential speedup over Eulerian methods
 - Central limit theorem: direct sampling requires computational **cost $\sim 1/\text{accuracy}^2$**
- Amplitude estimation only requires computational **cost $\sim 1/\text{accuracy}$**
- Relative to deterministic algorithms, speedups increase for **high dimensions $d \rightarrow \infty$** and for **solutions that are not smooth $k, \alpha \rightarrow 0$**
 - Key Assumption: location of discontinuities are unknown \rightarrow *stochastic / randomized function*



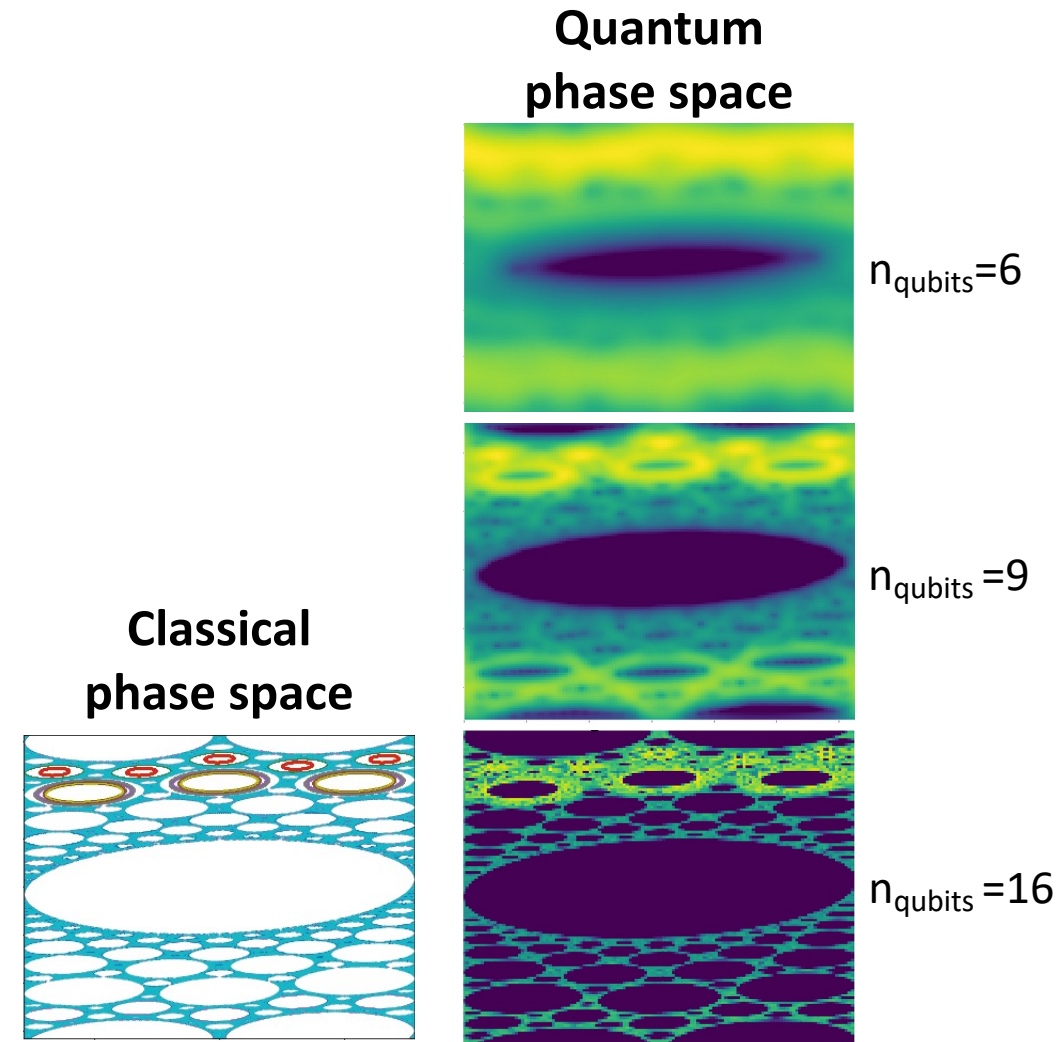
	function space	deterministic	randomized	quantum
Holder class	$L_p^N, 2 \leq p \leq \infty$	1	$n^{-1/2}$	n^{-1}
	$F_d^{k,\alpha}$	$n^{-(k+\alpha)/d}$	$n^{-(k+\alpha)/d-1/2}$	$n^{-(k+\alpha)/d-1}$
Sobolev class	$W_{p,d}^k, 2 \leq p \leq \infty$	$n^{-k/d}$	$n^{-k/d-1/2}$	$n^{-k/d-1}$

Convergence of error with
number of function calls n

Heinrich & Novak 2001
[arXiv:quant-ph/0105114](https://arxiv.org/abs/quant-ph/0105114)

Outline: Quantum Computing for Fusion Science Applications

- **Quantum Algorithms for Diff Eqs**
 - Linear problems
 - Nonlinear problems
- **Testing Quantum Hardware Platforms**
 - Error mitigation improves results (Yuan Shi's talk)
 - Errors can be exploited for certain calculations
- **Conclusions & Outlook**



The quantum sawtooth map (QSM) is the most efficient chaotic system to simulate on a quantum computer*

- Classical sawtooth map depends on kicking strength K

$$H_{saw} = \frac{1}{2}p^2 - \frac{1}{2}Kq^2 \sum_n \delta(t - n) \quad \text{for } q \bmod 2\pi$$

- Quantum sawtooth map also depends on \hbar

— Map eigenvalues of p to n qubits that represent $N = 2^n$ states

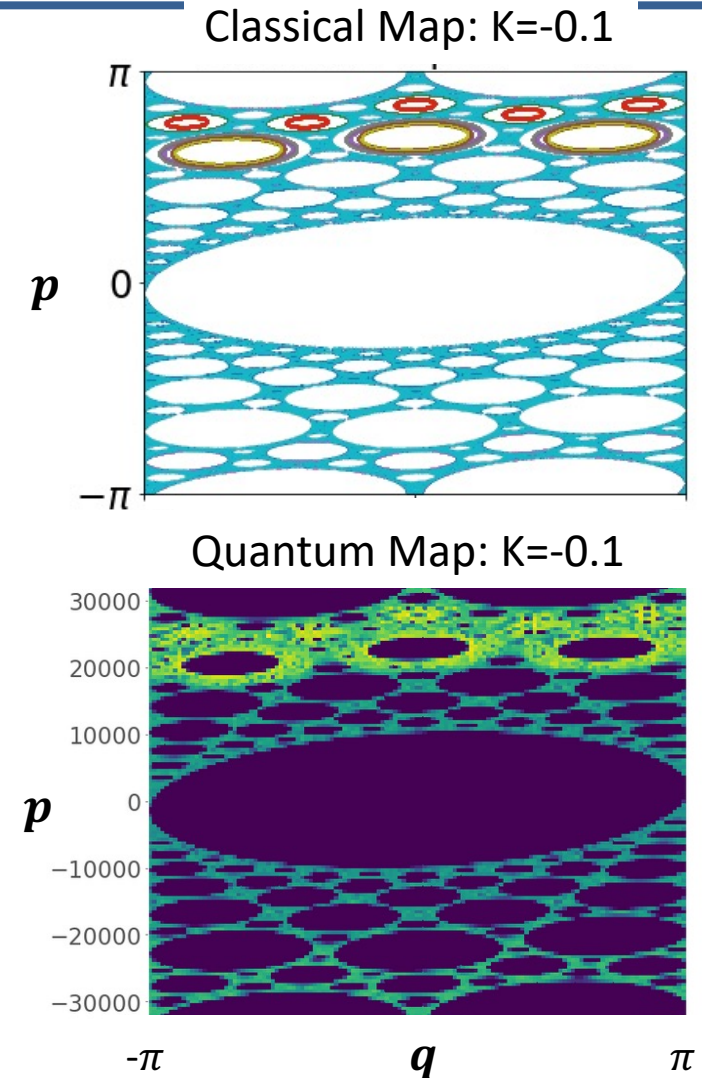
— \hbar is quantized in order to match periodicity in p

$$\Delta p = 2\pi = \hbar N \quad \hbar = 2\pi/N$$

— Quantum propagator has four stages:

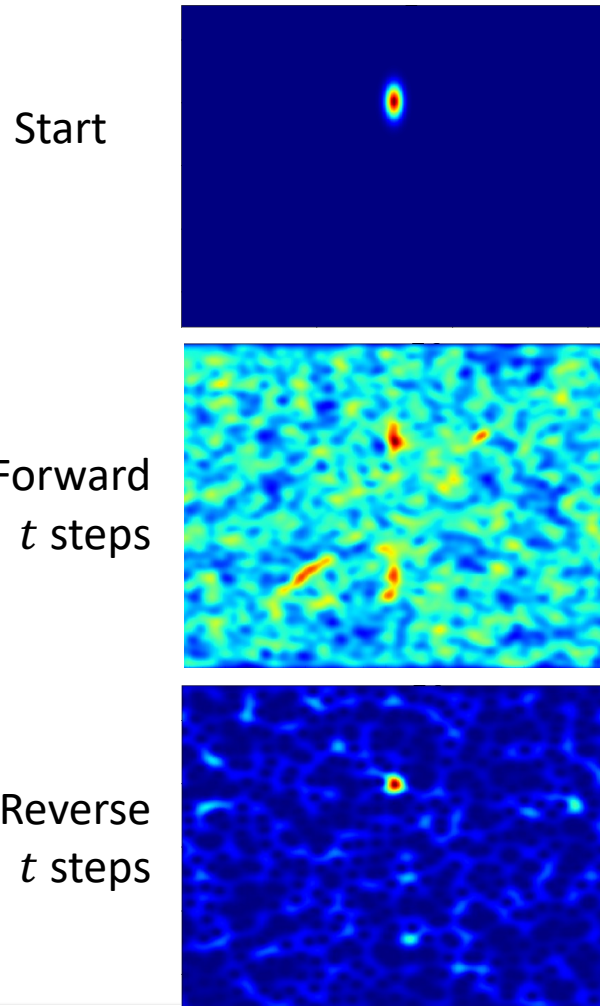
$$U_{QSM} = \hat{\mathcal{T}} e^{-i \int H_{saw} dt / \hbar} = U_{kin}(\hbar) U_{QFT}^{-1} U_{pot}(K/\hbar) U_{QFT}$$

*G. Benenti, et al, PRL **87** 227901 (2001)



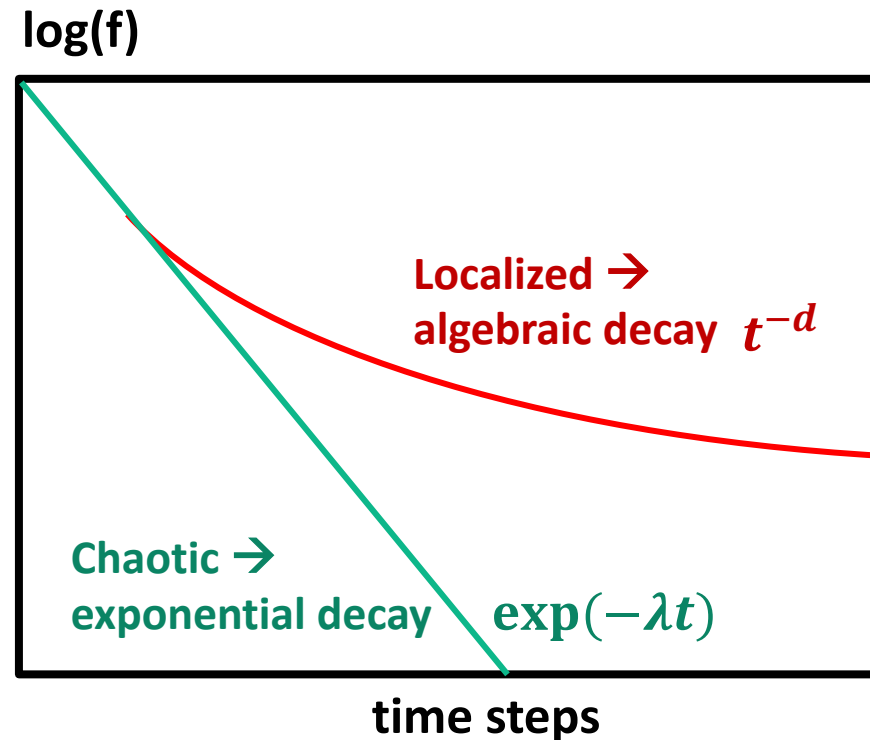
Noisy quantum computers can efficiently compute key signatures of chaos, such as the Lyapunov rate λ = the exponential separation of trajectories*

Chaotic $K = 0.5$

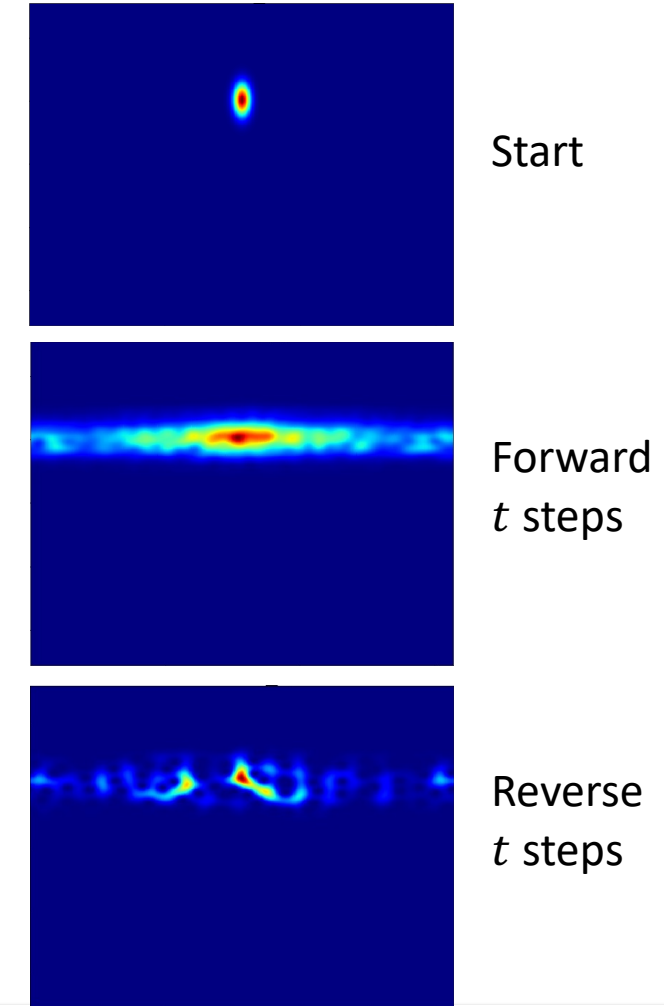


Simply measure the fidelity

$$f = |\langle \psi_{\text{actual}} | \psi_{\text{target}} \rangle|^2$$



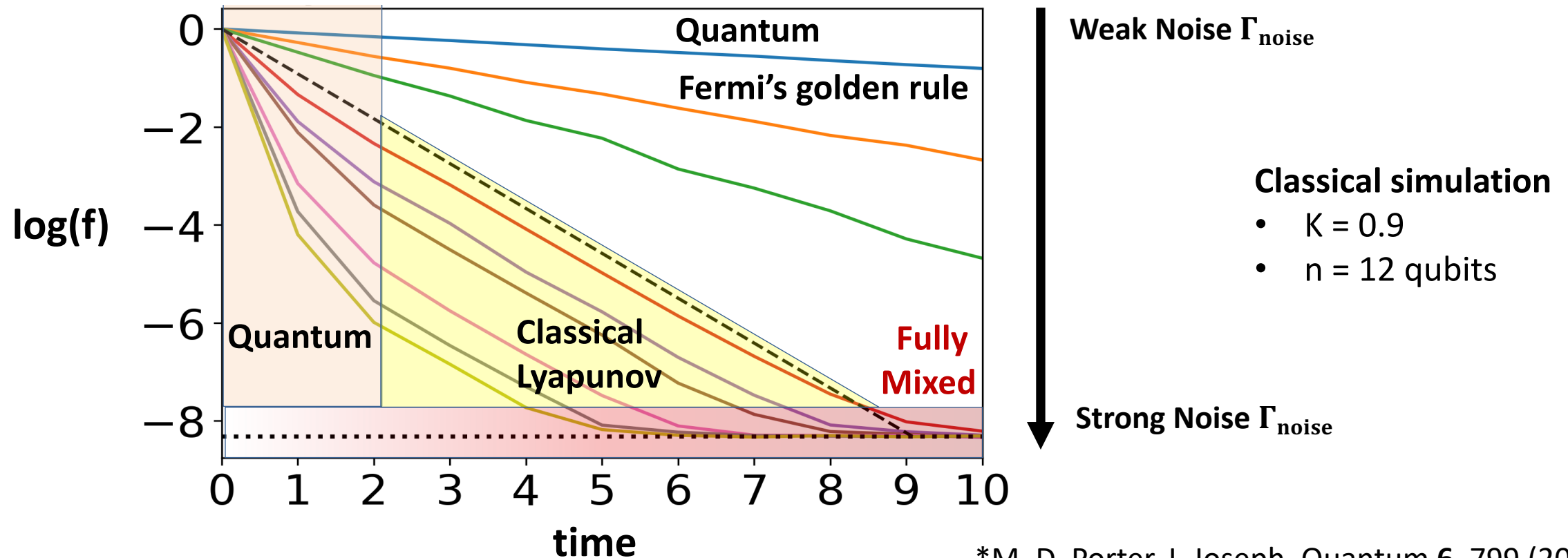
Localized $K = 0.1$



*G. Benenti, et al., Quantum Info. Proc. **3**, 273 (2004)

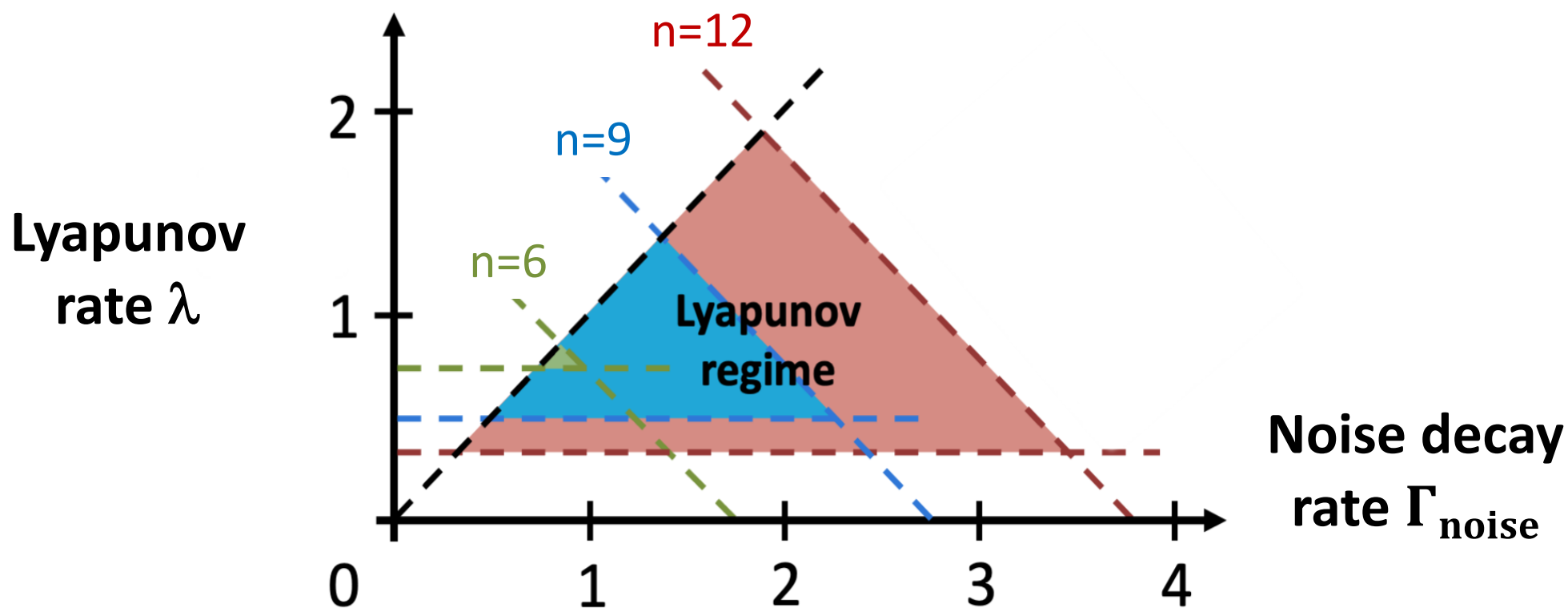
Semiclassical theory predicts that fidelity has two components that decay at different rates*

$$f(t) \approx f_{\text{Quantum}} e^{-\Gamma_{\text{noise}} t} + f_{\text{Classical}} e^{-\lambda_{\text{Lyap}} t} + 1/N$$



*M. D. Porter, I. Joseph, Quantum 6, 799 (2022)

Fidelity phase diagram determines whether the Lyapunov rate can be observed



Key Limitations

- Dynamics must be chaotic
- Lyapunov rate < noise decay rate
- Overall decay rate cannot be too fast
- Noise cannot be too strong or too weak

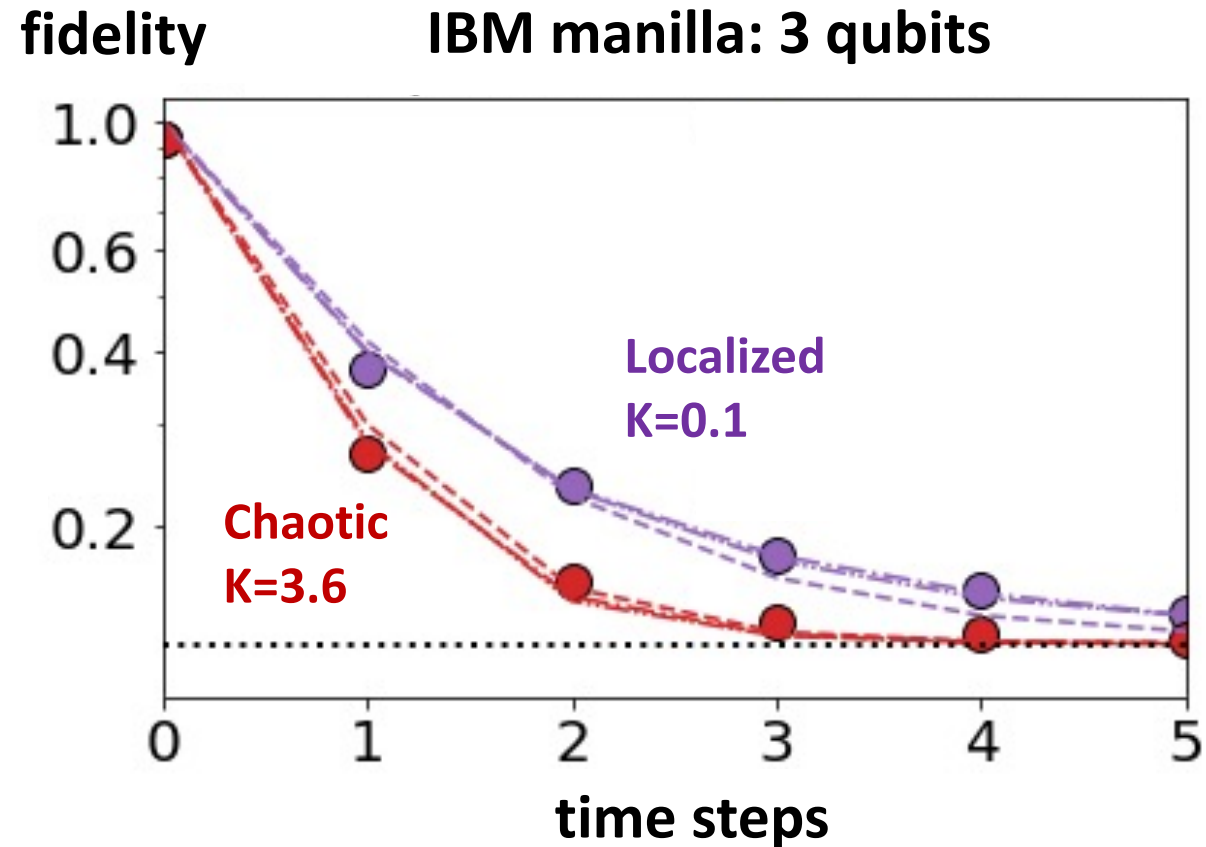
Key Requirements

- At least **6 qubits**
- Noise must be reduced by **10-100x**
- Depends on architecture
 - Parallelization, layout, etc.

*M. D. Porter, I. Joseph, Quantum **6**, 799 (2022)

We performed the first gate-based quantum simulation showing that fidelity decay depends on dynamics in addition to gate-count*

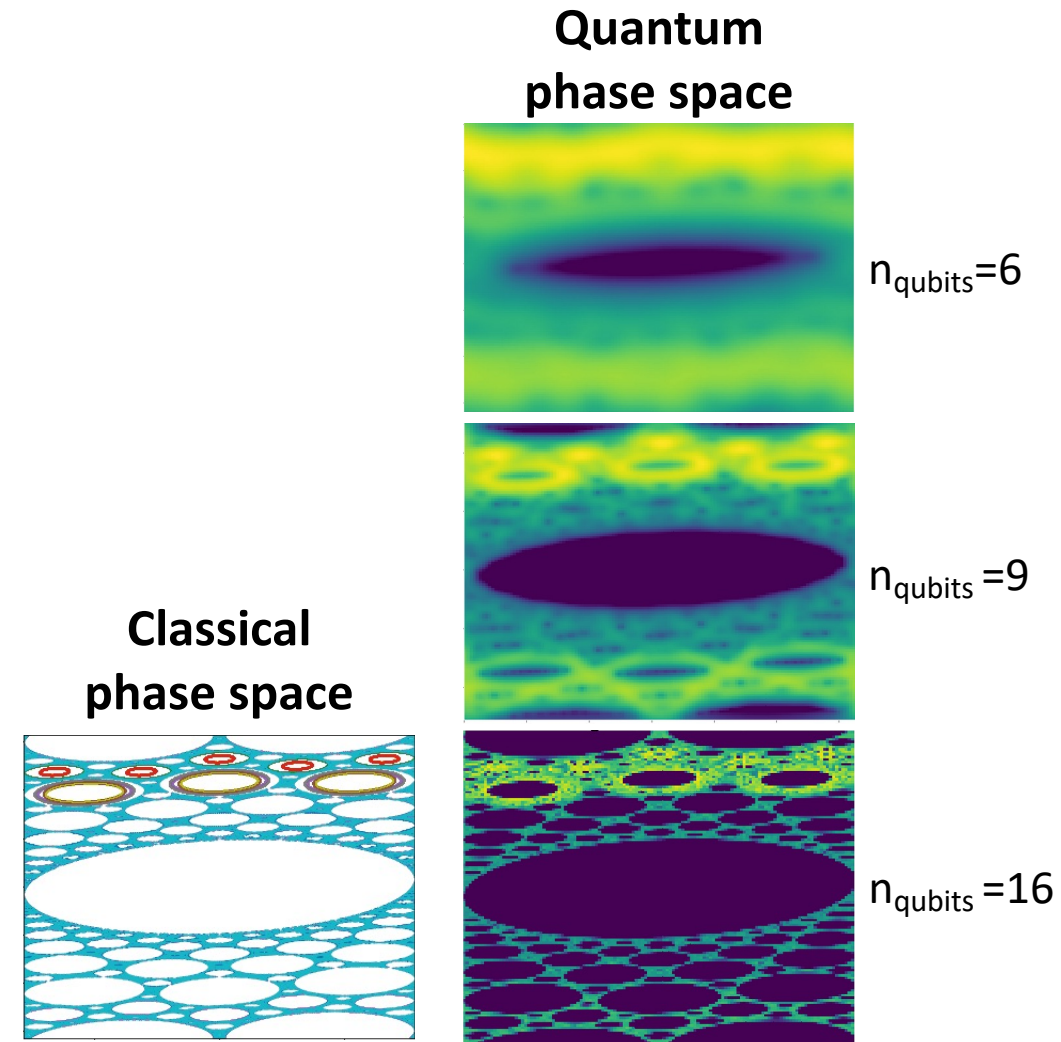
- **Decay rate is faster for chaotic dynamics with same # of gates**
 - Only single-qubit rotation angles change
- **Saturates at low and high values of K**
 - Increases during the transition to chaos, but does not keep increasing with Lyapunov rate
- **Chaos generates delocalized entangled states that are more sensitive to noise**
 - Actual error rates are 3 – 5x larger than reported
 - Lindblad decoherence model infers 3x larger dephasing rate $1/T_2^*$



*M. D. Porter, I. Joseph, [arXiv:2206.04829](https://arxiv.org/abs/2206.04829)

Outline: Quantum Computing for Fusion Science Applications

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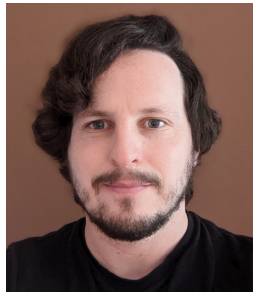


- **Quantum computing holds great promise for accelerating scientific discovery**
 - Efficient Fourier transforms, sparse linear solvers, sparse Hamiltonian simulation, variational eigensolvers, ...
 - Chemistry, materials science, high-energy physics, nuclear physics, ..., **fusion energy science!**
- **Quantum simulation of the PDF of nonlinear dynamical systems can achieve exponential speedup over Eulerian methods and up to quadratic speedup over Monte Carlo methods**
 - Simulations of fluids, plasmas, molecular dynamics, finance, ecology, epidemiology, ...
 - Quadratic speedup attained for high dimension and lack of smoothness
 - Exponential speedup for end-to-end app's likely requires problems with special structure
- **Algorithms that utilize noise have potential for near-term quantum advantage**
 - Simulate open system dynamics with an open quantum system
 - We've also explored passive and active error mitigation, e.g. quantum optimal control
 - Decoherence controls the **"information confinement time"**

Our team consists of experts in MFE, HEDS and QIS

■ Core Team

- **MFE:** Ilon Joseph (PI), Vasily Geyko, *Jeff Parker (FY20)*
- **HEDS:** Frank Graziani, Stephen Libby, Yuan Shi (U Colorado Boulder)
- **QIS:** Jonathan DuBois (co-PI), Al Castelli, Max Porter (Sandia), Gabriel Woolls (Caltech/Berkeley)



Al Castelli



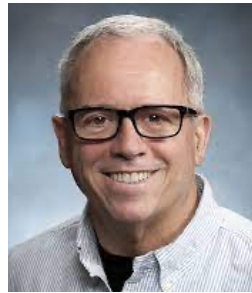
Jonathan DuBois



Vasily Geyko



Ilon Joseph



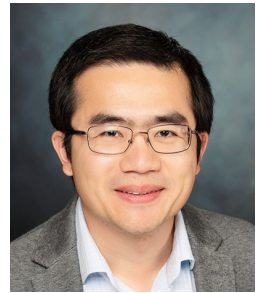
Frank Graziani



Steve Libby



Max Porter



Yuan Shi



Gabriel Woolls

■ LLNL Collaborators

- **Quantum Coherent Device Physics Group:** Kristi Beck (formerly IonQ), Yaniv Rosen, Xian Wu (now at Rigetti)
- **NACS:** Roger Minich, Kyle Wendt; **CASC:** Anders Petersson, Stephanie Gunther
- **Students:** Jessica Tucker (SJSU), Chris Yang (Caltech, SSGF Fellowship)

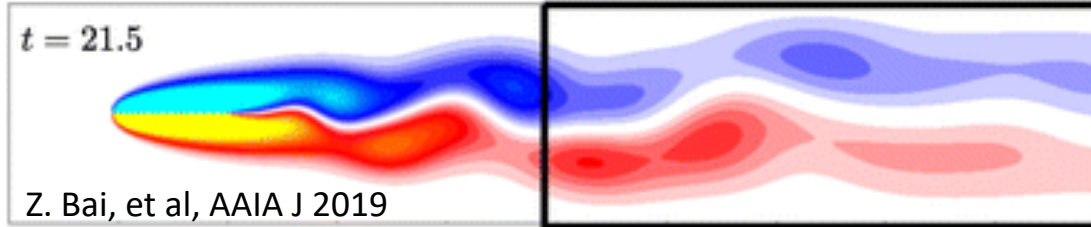
■ External Collaborators

- Robin Blume-Kohut (Sandia), Denys Bondar (Tulane), Frank Gaitan (LPS), Cesare Tronci (U. Surrey, Tulane), R. Tyler Sutherland (Quintinuum)

Killer App? Use quantum machine learning to develop exponentially reduced-order models of quantum simulation/data

- Quantum machine learning [1] and principle component analysis [2] are potentially powerful techniques

- But, they have the difficulty of getting the database of information in and out



$$|\psi\rangle = \sum_x \psi_x |x\rangle$$



- Quantum algorithms for **reduced-order modeling** of native quantum simulation or experimental data [3] could be the **killer app!**

- Quantum data assimilation [4] and closure of dynamical systems [5]
- Imagine a quantum dynamic mode decomposition: qDMD or quantum sparse identification of nonlinear dynamics qSINDy

[1] P. Rebentrost PRL 2014, M. Schuld PRA 2016, J. Biamonte, Nature Phys 2017 [2] S. Lloyd, Nature Phys., 2014

[3] B. Kiani PRA 2022 [4] D. Giannakis PRE 2019, D. Freeman PNAS 2023 [5] D. Freeman [arXiv:2208.03390](https://arxiv.org/abs/2208.03390)

The key insight ...



Nature isn't classical, dammit, and if you want to make a simulation of nature, you'd better make it quantum mechanical, and by golly it's a wonderful problem, because it doesn't look so easy.

— *Richard P. Feynman* —

AZ QUOTES

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Quantum information science (QIS) may soon lead to game-changing capabilities for fusion & science in general

- **Quantum Sensing:** improves measurement sensitivity
 - Heisenberg limit for noise/signal ratio scales as $1/N$ instead of $1/\sqrt{N}$
- **Quantum Communications:** secure information transfer
 - Intrinsically parallel data transfer / data compression
- **Quantum Computing:** polynomial or exponential gains in effective memory and computational power
 - Fourier transform, linear solvers, Hamiltonian simulation, ...
 - **Today = Noisy Intermediate-Scale Quantum (NISQ) era**

Quantum-Enhanced Advanced LIGO Detectors in the Era of Gravitational-Wave Astronomy

M. Tse *et al.* Phys. Rev. Lett. **123**, 231107 (2019)

China Demonstrates Quantum Encryption By Hosting a Video Call

A. Nordrum, IEEE Spectrum (2017-10-03)

Quantum supremacy using a programmable superconducting processor

F. Arute, *et al.* Nature. **127**, 180502 (2019)

Strong Quantum Computational Advantage Using a Superconducting Quantum Processor

Yulin Wu, *et al.* Phys. Rev. Lett. **127**, 180501 (2021)

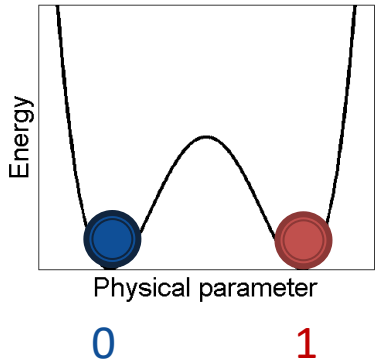
Han-Sen Zhong, *et al.* Phys. Rev. Lett. **127**, 180502 (2021)

Phase-Programmable Gaussian Boson Sampling Using Stimulated Squeezed Light



The qubit is the simplest complex Hilbert space

Classical Information



$|0\rangle\langle 0|$

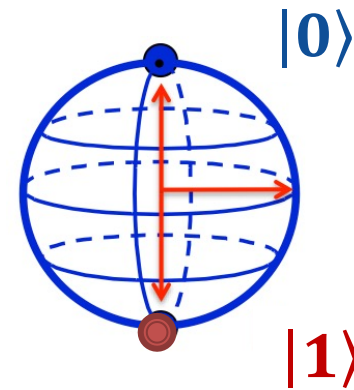
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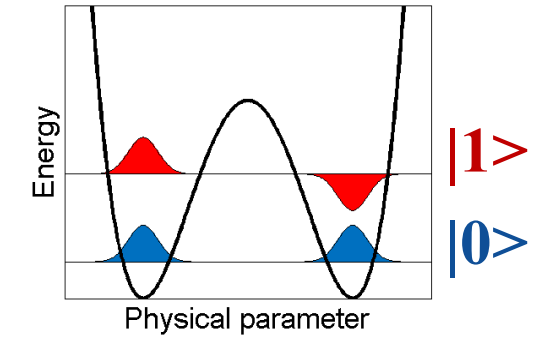
$|1\rangle\langle 1|$



Qubit



Quantum Information



- Qubit wavefunction $\psi \in \mathbb{C}^2$ is a normalized superposition of the basis states $|0\rangle$ and $|1\rangle$
- Probabilistic mixture of pure states described by the density matrix $\rho = \rho^\dagger \in \mathbf{H}_4 \sim \mathbb{R}^4$
- Probability distribution function (PDF)
 $f \in \mathbb{R}^2$

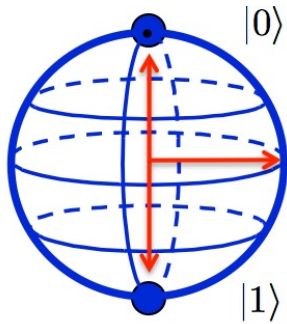
$$\psi = \cos \theta e^{-i\phi/2} |0\rangle + \sin \theta e^{+i\phi/2} |1\rangle$$

$$\rho = \rho_{00} |0\rangle\langle 0| + \rho_{01} |0\rangle\langle 1| + \rho_{10} |1\rangle\langle 0| + \rho_{11} |1\rangle\langle 1|$$

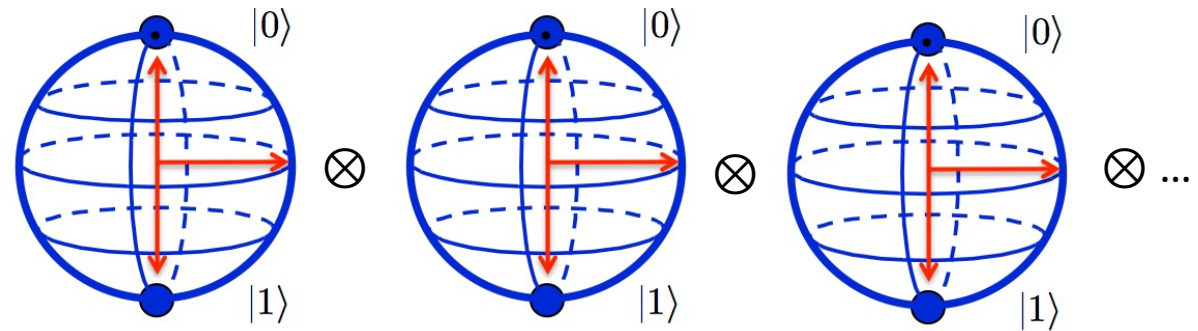
$$f = \text{Diag}(\rho) = \rho_{00} |0\rangle\langle 0| + \rho_{11} |1\rangle\langle 1|$$

Quantum memory registers are “exponentially large”

Qubit: Dimension 2



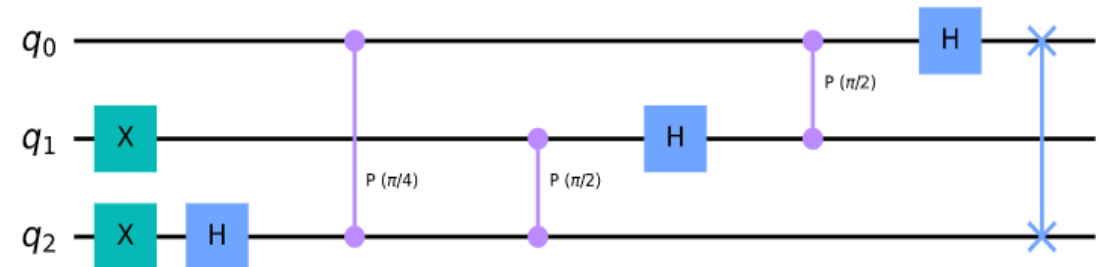
n Qubits \rightarrow Hilbert Space Dimension: $N = 2^n$



- Direct quantum simulation is extremely difficult due to exponentially large Hilbert space
- Why not use a “quantum machine” to simulate quantum physics?
 - **Early-1980's:** Think of the challenge into an opportunity — Feynman, Manin, Bennett & Brassard
 - **Mid-1990's:** Factoring integers, unstructured search, quantum counting — Shor, Grover, Brassard, Hoyer, Tapp
 - **Late-1990's:** Efficient simulation algorithms based on Trotter-Suzuki decompositions — Lloyd & Abrams
 - **Early 2000's:** Linear solver algorithms — Harrow Hassidim & Lloyd, Ambianis, Childs Kothari & Somma, ...
 - **2015-present:** Accelerated linear solver, linear diff eq & simulation algorithms — Berry, Childs, Low & Chuang

Digital quantum computing model has power and simplicity

- Quantum states can be transformed efficiently via linear unitary operations
 - $\psi = U\psi_0$ where $U = e^{-iHt}$ is a unitary $UU^\dagger = I$ evolution operator and $H = H^\dagger$ a Hermitian Hamiltonian
- While there are a huge number, $N^2 = 2^{2n}$, of unitary operations, they are generated by a small number $\sim O(n)$ of basic operations called a “gate set”
 - Single qubit operations can be achieved efficiently with a few standard gates, e.g. RX and RZ
 - Adding one nontrivial 2-qubit gate, e.g. CNOT or CZ, between nearest neighbors generates the rest
- Many useful computations can be performed in $O(\text{poly}(n))$ basic gate operations!
 - Key resources are **superposition and interference**, often called **quantum parallelism** 😊
 - Any reversible classical computation can also be performed, but typically without a speedup ☹️
- **Key Limitation:** Approximating an arbitrary unitary is exponentially hard
 - Only certain unitaries can be performed efficiently
 - Initializing all quantum information is exponentially hard
 - Measuring all quantum information is exponentially hard



A few essential subroutines power the majority of quantum algorithms

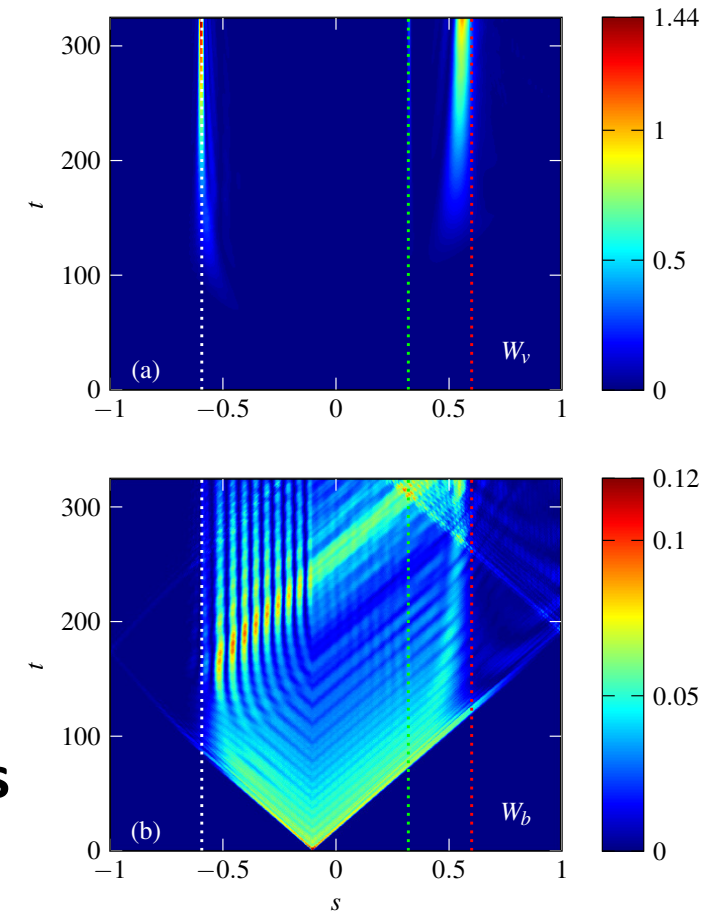
- **Quantum Fourier Transform: Cost of $(\log N)^2$ rather than classical $N \log N$**
 - Phase estimation, factoring integers, and taking discrete logarithms – Peter Shor 1994
 - Powers many Hamiltonian simulation algorithms
 - Hamiltonian simulation powers linear solvers, linear diff. eq. solvers, and variational eigensolvers, etc.
- **Amplitude Amplification: Cost of \sqrt{N} rather than classical N**
 - Amplitude amplification first used in Grover's search algorithm – Lov Grover 1996
 - Amplitude estimation & Quantum counting – Brassard, Hoyer, Mosca, Tapp 2000
 - Powers many Monte Carlo and integration algorithms – Heinrich & Novaks 2000, Montanaro 2015
- **Quantum Walks: Cost of N rather than classical N^2**
 - Early models turned into a computational framework – Aharonov, Ambianis, Kempe, Vazirani 2001
 - Graph search, element uniqueness, ... – Ambianis, Childs, Kempe
 - Hamiltonian simulation, state preparation – Szegedy 2004, Childs 2010
 - Qubitization, Quantum Signal Processing, Quantum Singular Value Transformation – Low & Chuang 2017

Hamiltonian simulation can achieve super-polynomial to exponential speedup by exploiting special structure and/or sparsity

- Trotter-Suzuki & Lie group decompositions work well for specific Hamiltonians [1]

$$e^{n(A+B)} \approx \prod_{j=1}^n e^{B/2} e^A e^{B/2} = e^{-B/2} \left(\prod_{j=1}^n e^B e^A \right) e^{B/2}$$

- **Black box methods work well for sparse Hamiltonians**
 - “Efficiently row-computable sparse”
 - Linear Combination of Unitaries (LCU) [2]; spectral methods [3]
- **Quantum signal processing (QSP) & qubitization [4], eigenvalue & singular value transformation [5] use **block-encoded** Hamiltonians**
 - Block encoding allows **non-unitary operations** to be performed!



[1] S. Lloyd 1996 [2] D. Berry 2017 [3] A. Childs 2021

[4] G. Low & I. Chuang 2017 [5] A. Gilyen 2019

Novikau, Startsev, Dodin
Phys. Rev. A 2022

Quantum algorithms for ODEs and PDEs come in many flavors

- **Linear vs. Nonlinear**

- For sparse Hamiltonians, quantum computers can exponentially speed up linear operations
- Koopman & Carleman: Nonlinear systems can be embedded within an infinite-dimensional linear system

- **Deterministic vs. Stochastic**

- Amplitude estimation can quadratically speed up Monte Carlo sums and integrals = observable estimation
- Quantum walks can quadratically speed up the mixing time of Markov chains = time to solution

- **Variational Algorithms and Quantum Machine Learning**

- Classical computer can efficiently perform nonlinear operations that drive a quantum computer
- Quantum machine learning can potentially avoid the use of classical computers altogether except for I/O

- **Continuous Variable Model for classical and quantum field theory (CFT / QFT)**

- Uses a quantum field theory as the computational basis
- Classical limit is a classical field theory, i.e. a set of PDEs such as Maxwell's equations

Solving for wavefunction is efficient for sparse linear evolution

- Choose a basis for finite-dimensional numerical discretization

- Possible choices of basis functions $\phi_n(z)$

- Spectral [1,2] e^{inz} , orthogonal polynomials $H_n(z)$, etc., ...
- Finite difference & finite element: local orthogonal polynomials
- Carleman linearization [3,4]: polynomials z^k
- Reproducing Kernel Hilbert Spaces [2]: allow pointwise evaluation

$$\psi(z, t) = \sum_n \psi^n(t) \phi_n(z)$$

$$i\hbar \frac{d}{dt} \psi^n = \sum_m \psi^m H_m^n$$

- If the evolution is unitary, use the quantum Hamiltonian simulation algorithm (QHSA) [1,2]

$$\psi(t) = \mathbf{U}_{approx} \psi(0) \approx \mathcal{T} e^{-i \int \mathbf{H} dt / \hbar} \psi(0)$$

- Otherwise, use the quantum linear differential equation solver algorithm (QLDA) [3,4]

- Uses quantum linear solver algorithm (QLSA) to propagate forward for small timesteps Δt

$$1 = \alpha + \beta \quad (1 + i\alpha \mathbf{H} \Delta t / \hbar) \psi(t + \Delta t) \approx (1 - i\beta \mathbf{H} \Delta t / \hbar) \psi(t)$$

semi-implicit time
splitting

[1] I. Joseph, Phys. Rev. Research 2, 043102 (2020)

[2] D. Giannakis, A. Ourmazzi, P. Pfeffer, et al., arXiv:2012.06097 (2022)

[3] Jin-Peng Liu, H.Ø. Kolden, H.K. Krovi, et al., PNAS 118, e2026805118 (2021)

[4] A. Engel, G. Smith, S. P. Parker, Phys. Plasmas 28, 062305 (2021)

Amplitude estimation of physical observables¹⁻³ is up to quadratically more efficient than best classical methods

- **Expectation value $\langle \mathcal{O} \rangle = \sum_x \mathcal{O}(x) f(x)$ can be found by simulating a reversible classical computation of**

$$\phi(x) := \mathcal{O}^{1/2}(x)\psi(x) \quad \phi'(x) := \sqrt{1 - |\phi(x)|^2} \quad |\phi\rangle := \sum_x \phi(x) |x\rangle / \|\phi\|$$

- **Add an ancillary qubit to $|\psi\rangle$ and compute a state proportional to $|\phi\rangle$**

$$\mathcal{R}_\phi |\psi\rangle |0\rangle := \sum_{x=1}^N |x\rangle (\phi'(x) |0\rangle + \phi(x) |1\rangle) / N^{1/2} = \cos(\theta) |\phi'\rangle |0\rangle + \sin(\theta) |\phi\rangle |1\rangle$$

- **Amplitude estimation of the ancillary $|1\rangle$ state probability yields $\sin^2(\theta) = \langle \mathcal{O} \rangle / N$ with complexity $\sim Q_H / \epsilon$**

¹D. S. Abrams and C. P. Williams, arXiv:quant-ph/9908083 (1999)

²S. Heinrich and H. Novak, *Proc. 4th Int. Conf. on Monte Carlo and Quasi-Monte Carlo Methods, Hong Kong 2000*, Springer-Verlag (2002)

³A. Montanaro, *Proc. R. Soc. London, Ser. A* 471, 20150301 (2015)

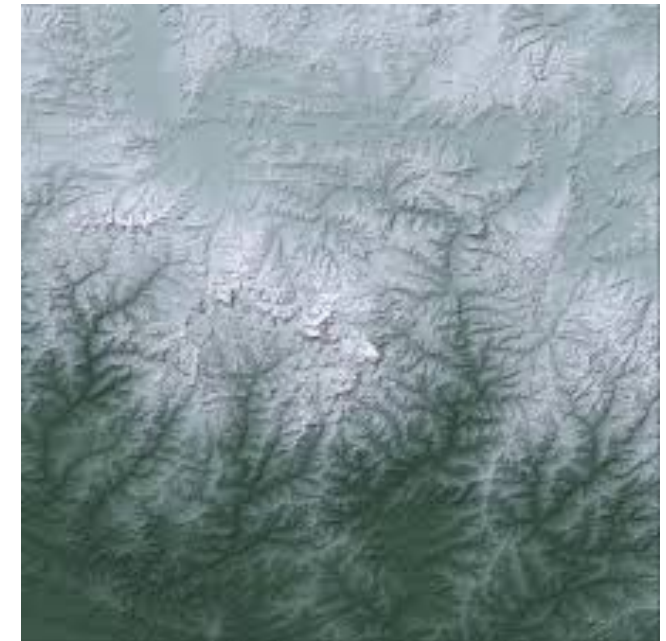
Variational algorithms have a few key steps ... and some key **limitations**

For each time step, iterate until convergence [1]:

- Prepare initial ansatz
- Solve equations using Hamiltonian simulation
- Measure cost function and, potentially, gradients of the cost function
- Execute step of classical optimization algorithm

Optimization landscape may have **intrinsic difficulties such as ...**

- Many local maxima and minima
- Barren plateaus with little information on the gradient of the cost function



NASA Earth Observatory:
Himalayas

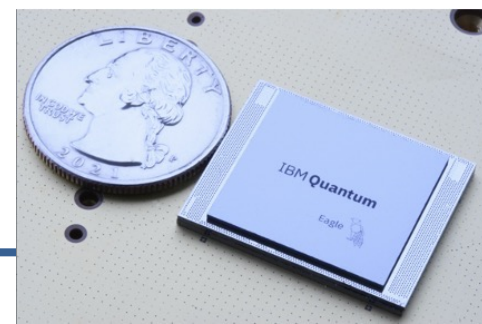
NP-complete optimization problems may not have any quantum advantage at all [2]

[1] M. Lubasch, et al, PRA 2015

[2] L. Bittel & M. Kliesch, PRL 2021

To date, we've used superconducting hardware platforms & have begun to use ion traps

IBM-Q
Eagle
127 qubits



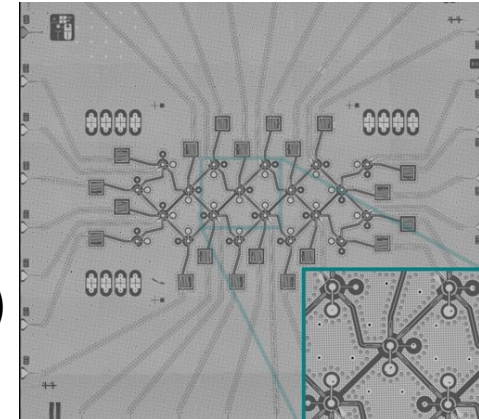
■ IBM-Quantum Experience

- Open but limited access to 5 qubit devices with relatively good fidelity

■ Rigetti Quantum Cloud Services

- Rigetti-LLNL-USC Collaboration

Rigetti
20 qubit device
(Aspen-M3 has 79 qubits)

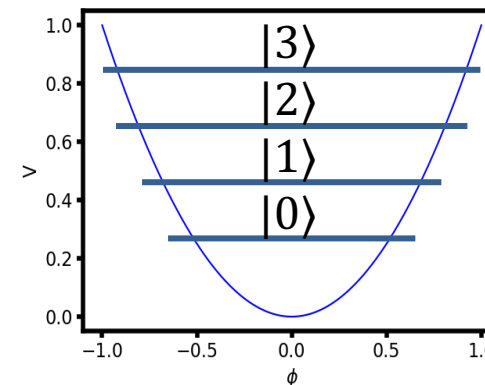
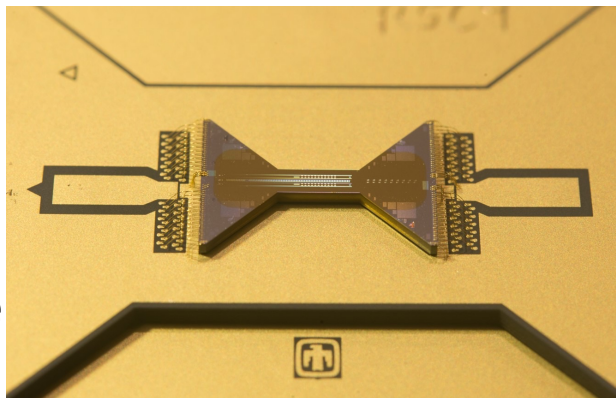


■ LLNL Quantum Design and Integration Testbed (QuDIT)

- Open access to 3-level and 4-level qudits rather than 2-level qubits

■ Sandia QSCOUT ion trap

Sandia
Peregrine
6 qubits

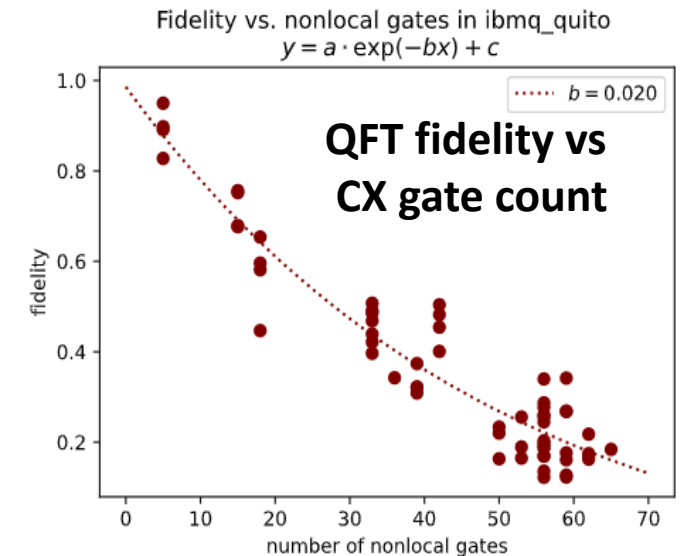
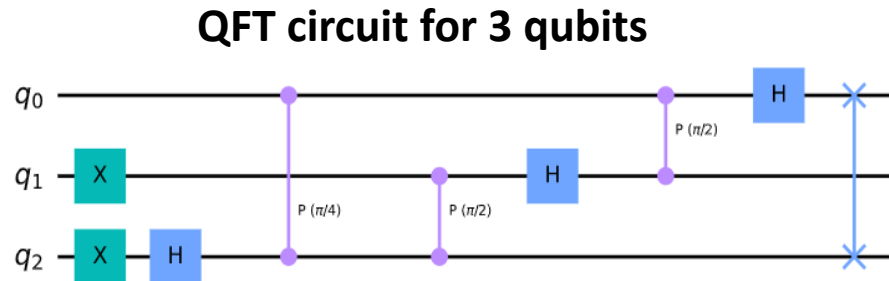


LLNL
Quantum
Design &
Integration
Testbed
(QuDIT)
~ 6 qubits



Lack of error-correction limits fidelity of present-day calculations

- **“Fidelity” is the figure of merit:** $F = |\langle \psi_{\text{expected}} | \psi_{\text{actual}} \rangle|^2$
 - Single qubit gate fidelity (IBM-Q): 99.9% --> 700 useful operations before 50%
 - Two qubit gate fidelity (IBM-Q): 99.5% --> 140 useful operations before 50%
 - State preparation & measurement (SPAM): 95% --> error at beginning and end



- **Infidelity for realistic calculations is ~2-5x worse than expected**
 - Fidelity does not always decay at an exponential rate
 - Coherent gate errors are important and need to be corrected for
 - Coherent errors can be much more damaging to intended calculations

First plasma application: simulating three-wave interactions [1]

■ Cubic couplings are ubiquitous in plasmas, fluids, & nonlinear media

- Examples: nonlinear optics, laser-plasma interactions, weak turbulence, gauge theory, lattice QED ...

- Interaction Hamiltonian $H_I = igA_1^\dagger A_2 A_3 - ig^* A_1 A_2^\dagger A_3^\dagger$

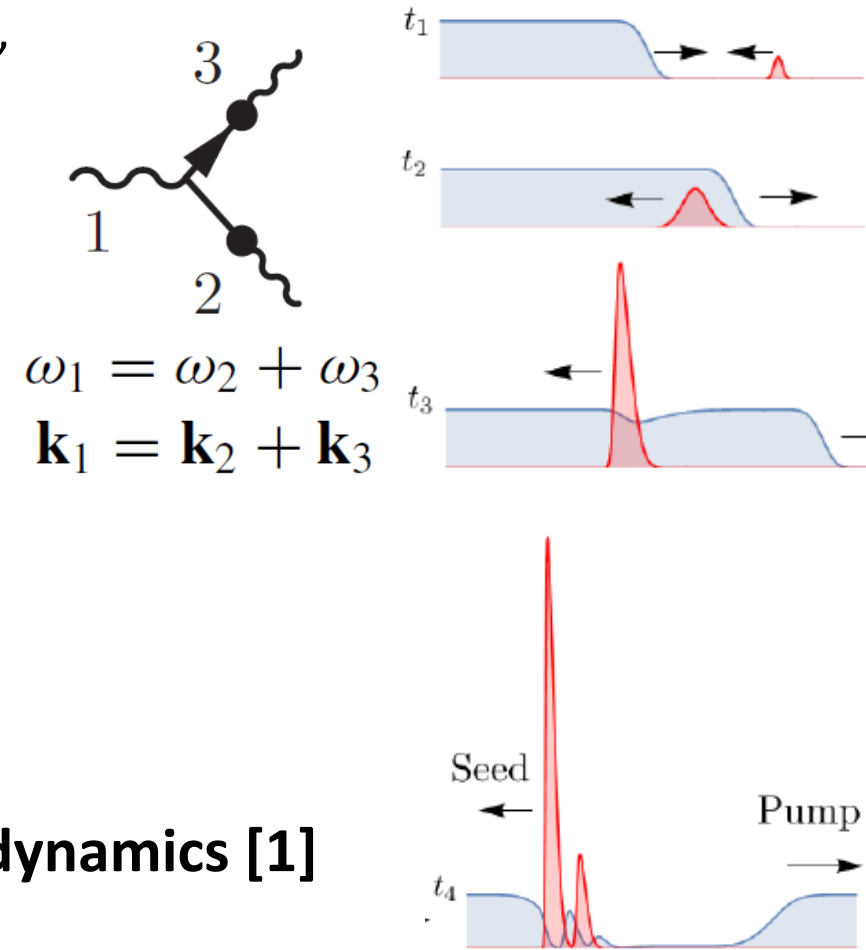
- Envelope equations for resonant interactions

$$d_t A_1 = g A_2 A_3, \quad d_t A_2 = -g^* A_1 A_3^\dagger, \quad d_t A_3 = -\textcircled{g}^* A_1 A_2^\dagger$$

- Quantized version $[A_j, A_l^\dagger] = \delta_{jl}$

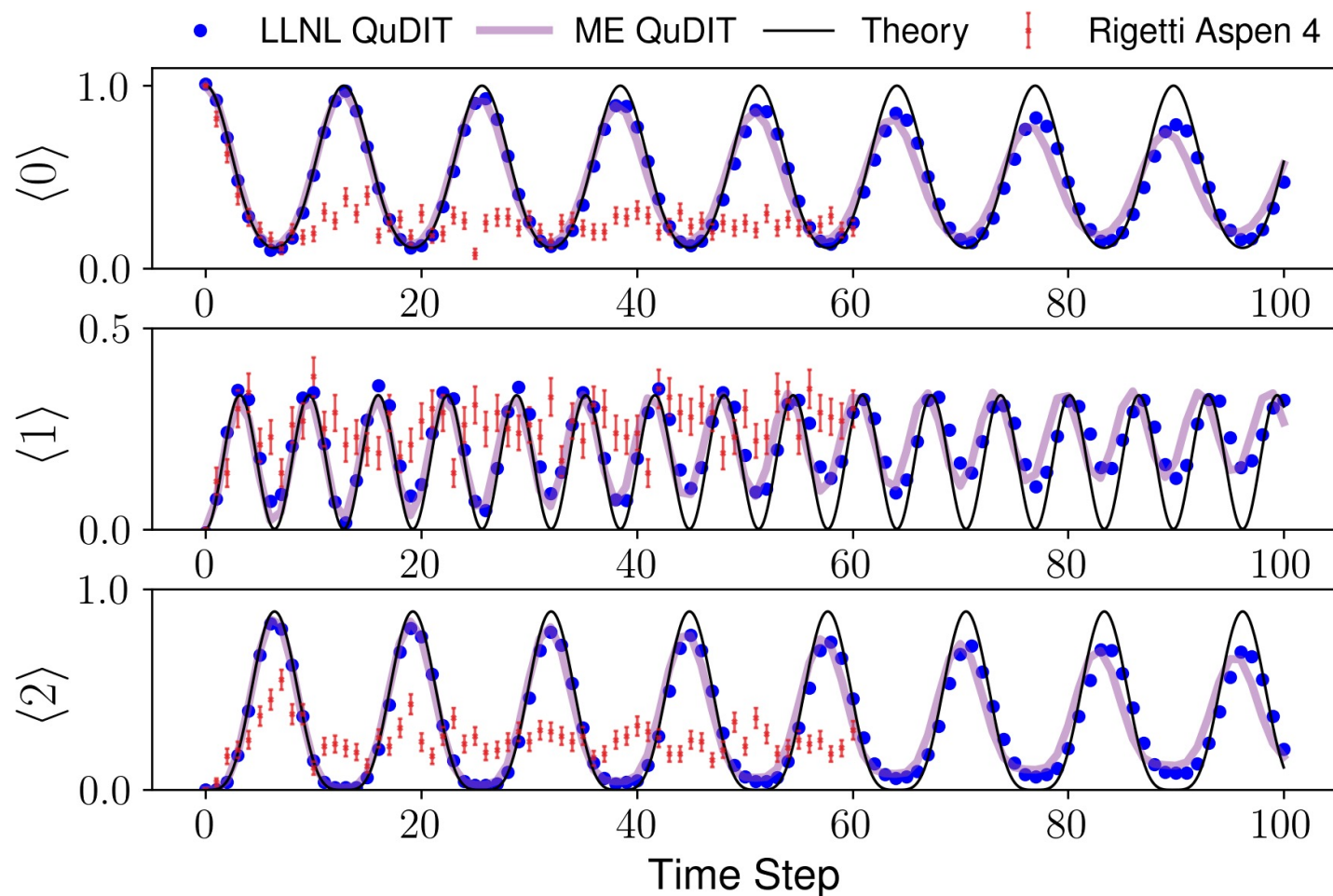
■ Developed a new quantum algorithm for simulating 3-wave dynamics [1]

- Transform to action-angle variables
- Evolve a sparse tridiagonal Hamiltonian system



[1] Y. Shi, A. R. Castelli, X. Wu, et al, Phys. Rev. A 103, 062608 (2021)

Optimal control approach to 3-wave yields ~10x improvement on LLNL QuDIT [1]



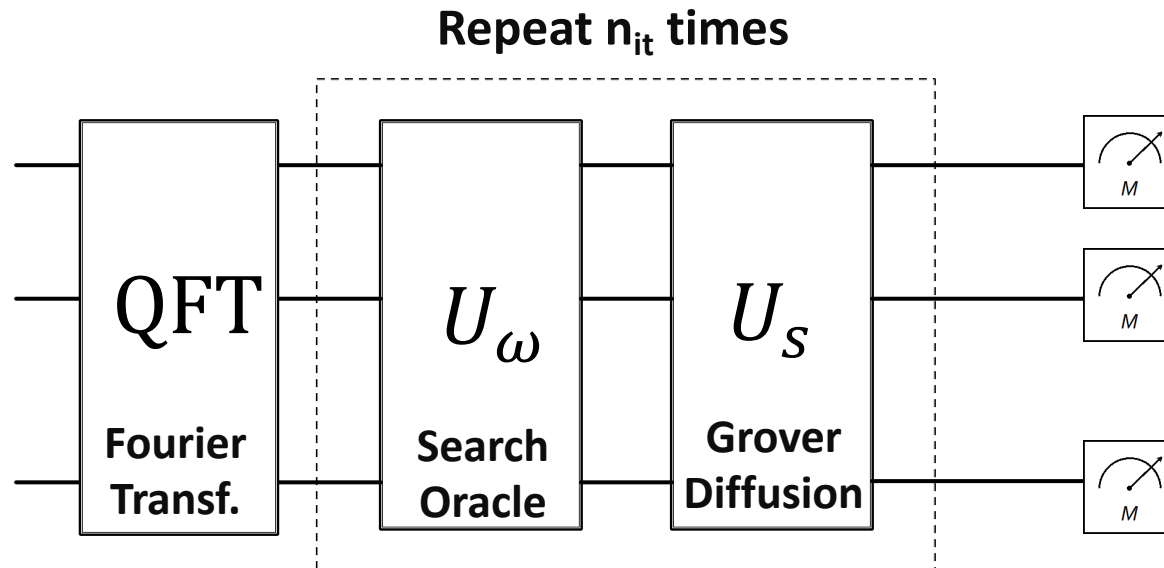
- Results of **LLNL QuDIT (blue)** are close to **analytic solution (black)** & match **Lindblad Master Equation (ME) simulation (purple)**
- Results of **Rigetti Aspen-4 platform (red)** perform well for first ~9 time-steps, but use **17x as many gates per step**
- On both platforms, decay and dephasing noise limit the fidelity after ~100 gate repetitions
- Combining gates into single control pulse improves long-term fidelity



[1] Y. Shi, A. R. Castelli, X. Wu, et al, Phys. Rev. A 103, 062608 (2021)

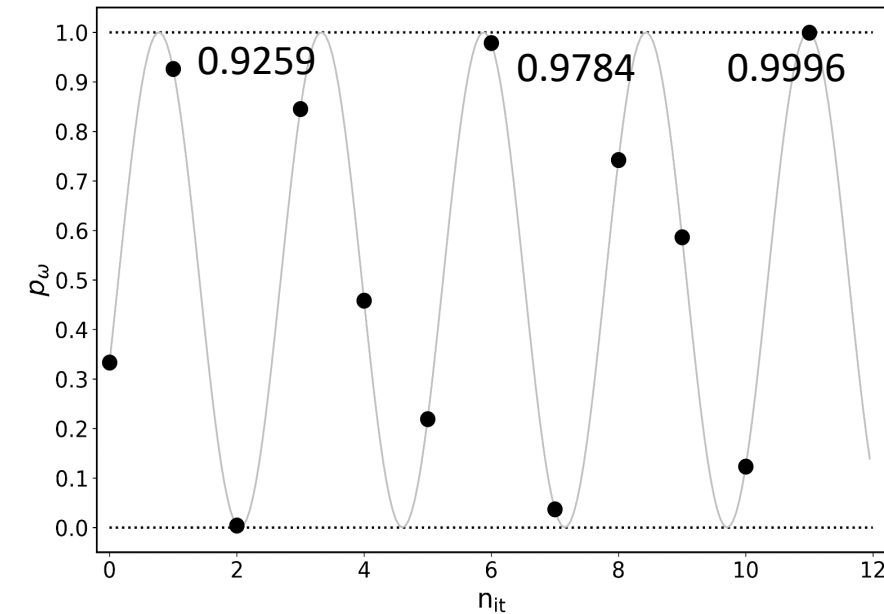
Grover's search for 3-level qudits developed & improved using optimal control [1]

- We developed a modified Grover's search algorithm for LLNL QuDIT
 - Search on 3 items has a 92.6% success probability on the first iteration
 - Compare to 2-qubit case: search on 4 items has 100% success on 1st iteration

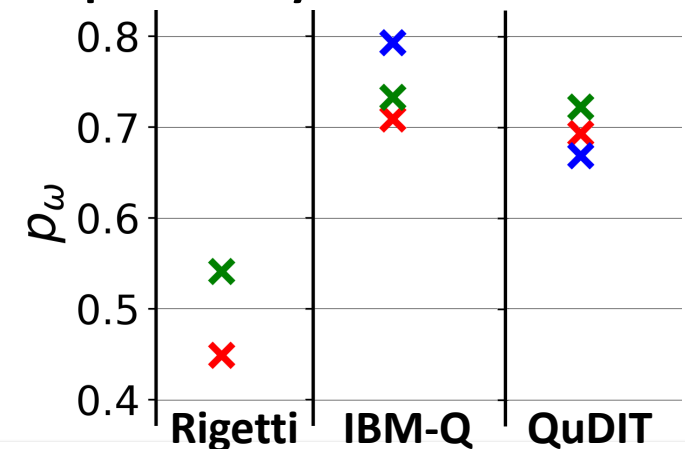


- Tests on IBM-Q, Rigetti, and LLNL QuDIT demonstrate reasonably good performance for 1-11 iterations
 - Optimal control effectively improves hardware performance

[1] V. I. Geyko, et al., "Using Grover's search algorithm to test present state-of-the-art quantum computing platforms," in preparation



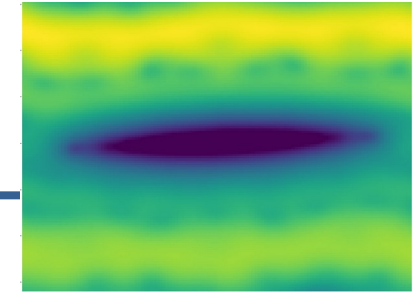
Expt. fidelity after 11 iterations



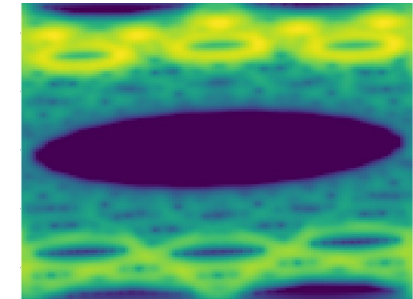
Motivation: Can we simulate chaotic dynamics on near-term universal quantum devices?

Quasiprobability (Husimi Q)

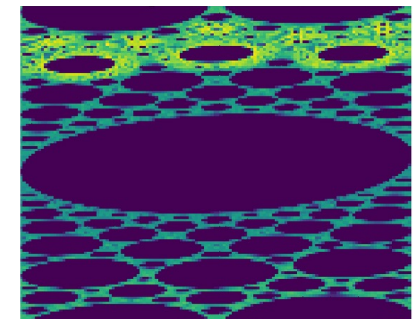
$n_{\text{qubits}}=6$



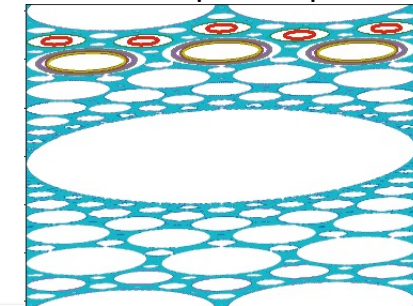
$n_{\text{qubits}}=9$



$n_{\text{qubits}}=16$



Classical phase space



- **Interesting quantum simulations usually contain chaotic regions**
 - Quantum chaotic simulations are important for many body localization, black hole information scrambling, classical chaos, ...
- **Efficient detection of quantum chaos can come from quantum-classical correspondence**
 - Quantized classical systems recover classical limit at small \hbar , but requires many qubits
 - Quantum fidelity decay of perturbed Hamiltonian evolution can reveal classically chaotic or regular dynamics^{1,2}
 - For chaotic dynamics, quantum fidelity can decay at the rate of the Lyapunov exponent λ , which measures the exponential divergence of classical trajectories³
- **Quantum maps allow efficient simulation of chaotic dynamics⁴**
 - A quantum map decaying at the Lyapunov rate may be the most resource-efficient signature of quantum chaos

¹A. Peres, *PRA* 30.4 (1984) 1610

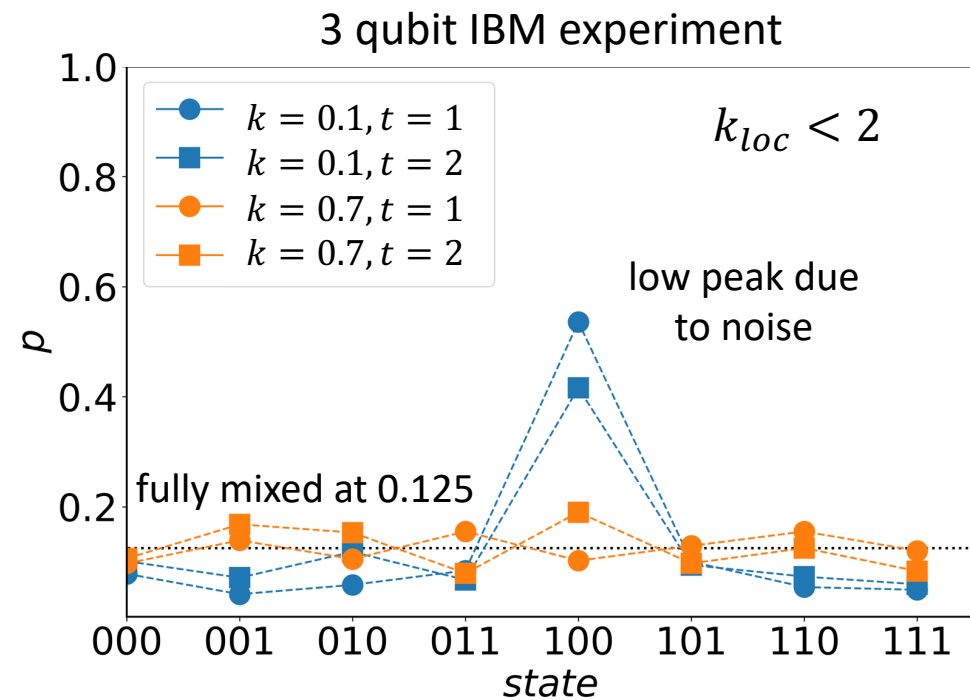
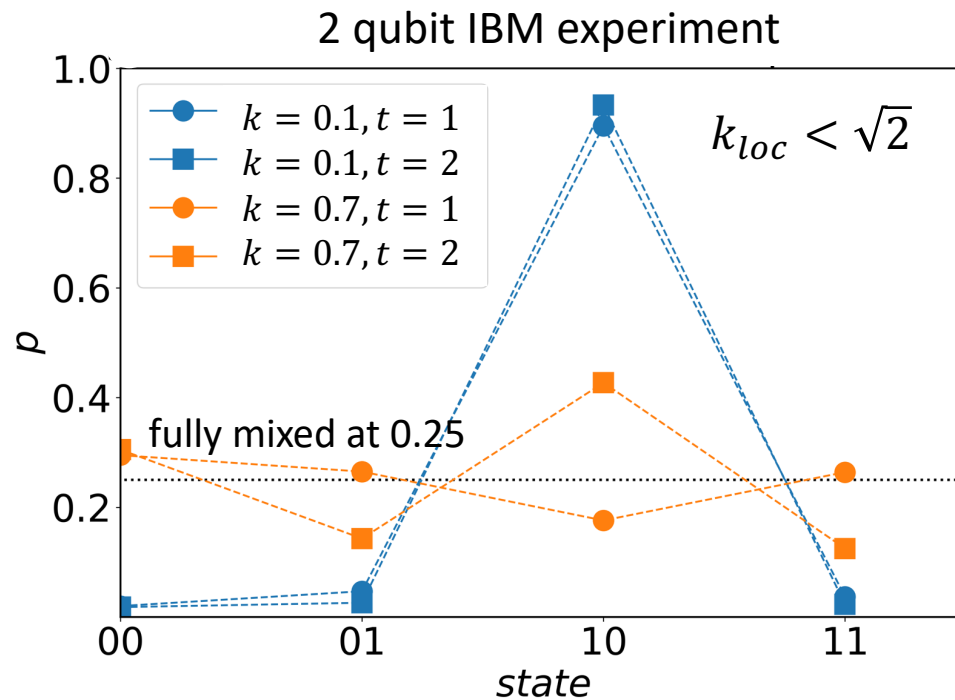
²Ph. Jacquod et al., *Advances in Physics* 58.2, 67-196 (2009)

³R. A. Jalabert, H. M. Pastawski, *PRL* 86.12, 2490 (2001)

⁴G. Benenti et al., *PRL* 87.22, 227901 (2001)

The phase transition between diffusive and localized dynamics is clearly observable on IBM-Q

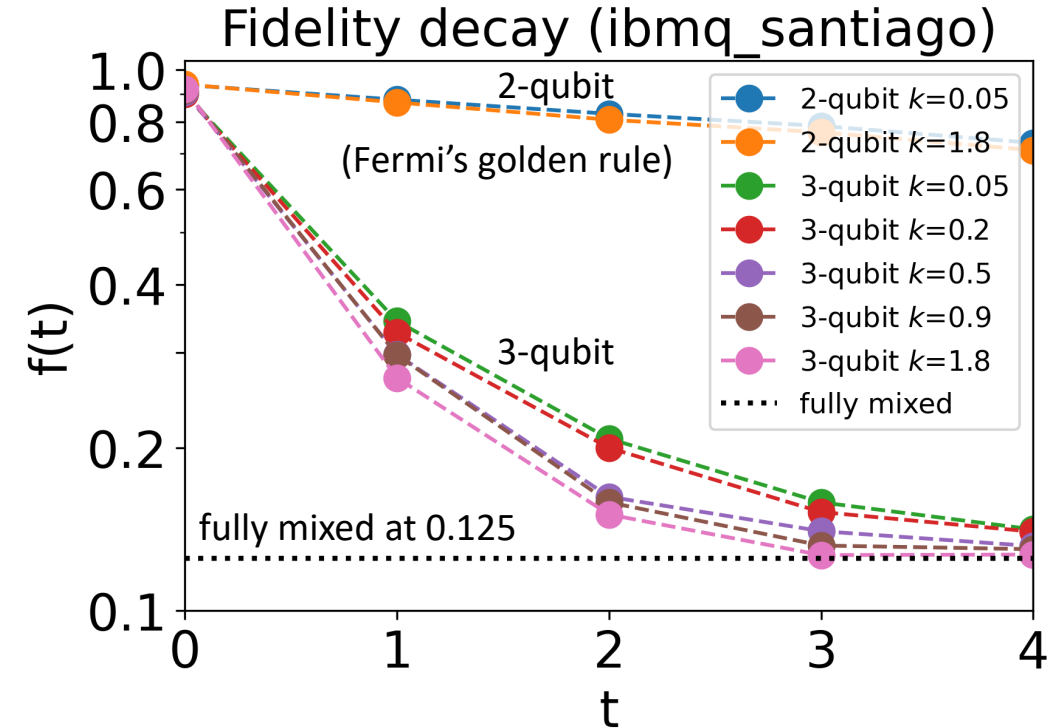
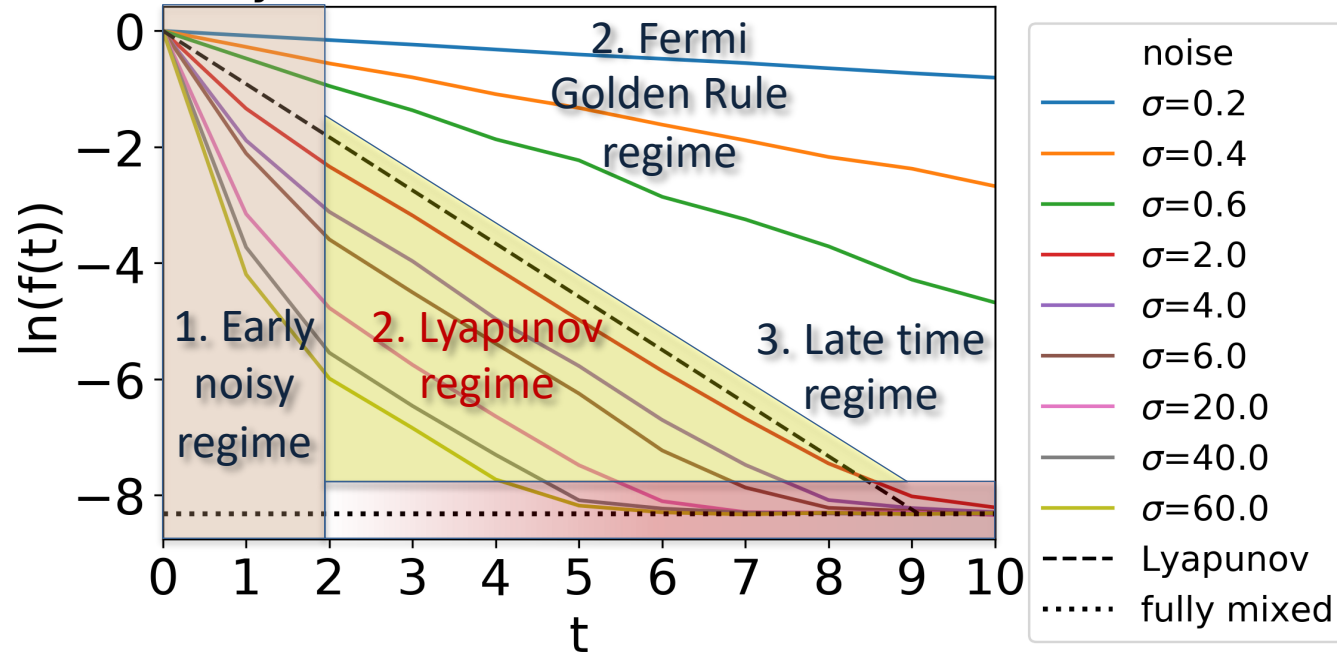
- We see localization and diffusion for both 2 and 3 qubits
 - Noise causes localization length decay



Statistical error
 $= 1/\sqrt{8,192}$
 ~ 0.01

Theoretical and experimental studies have shown that the fidelity does in fact depend on the chaotic/diffusive nature of the dynamics

Fidelity: $n=12$, $k \approx 587$, $K=0.9$, $\lambda=0.92$



- **Lyapunov regime can be observed with enough qubits if the noise processes are small enough [1]**
 - Semiclassical regime requires at least 6 qubits
 - Experimental noise must be reduced by 2.5-13x
- **Experiments with 3 qubits on IBM-Q observe an effect that matches the expected trend [2]**
 - Diffusive decay is more rapid than localized
 - Effective noise is larger by 2x

[1] M. D. Porter, I. Joseph, [arXiv:2110.07767](https://arxiv.org/abs/2110.07767), Quantum **6**, 799 (2022) [2] M. D. Porter, I. Joseph, [arXiv:2206.04829](https://arxiv.org/abs/2206.04829)

What about Generalized Eigenvalue Problems?

- **Generalized Eigenvalue Problem:** $\mathbf{A}v = \lambda \mathbf{B}v$
 - Assume \mathbf{A} is Hermitian and sparse
 - Assume \mathbf{B} is symmetric positive definite (SPD)
- Any SPD matrix, \mathbf{B} , has a unique **SPD square root** $\mathbf{B}^{1/2}$
- The problem can be reduced to standard Hermitian form using the transformation

$$u = \mathbf{B}^{1/2}v \qquad \mathbf{H} = \mathbf{B}^{-1/2}\mathbf{A}\mathbf{B}^{-1/2} \qquad \longrightarrow \qquad \mathbf{H}u = \lambda u$$

- In general, \mathbf{H} will not be sparse and, hence, QPE will not be efficient, unless ...
- **For special \mathbf{B}** , e.g. diagonal or block diagonal, then both $\mathbf{B}^{-1/2}$ and \mathbf{H} are also sparse

J. B. Parker and I. Joseph arXiv:2002.08497 (2020)

FES Application: MHD plasma stability

- Linear Ideal MHD is routinely used for plasma stability calculations of magnetic confinement fusion experiments and reactor designs

$$\mathbf{F} \cdot \boldsymbol{\xi} = -\omega^2 \rho \boldsymbol{\xi}$$

- Fundamental Theorem of MHD:** Force operator, $\mathbf{F}(\mathbf{x})$, is a self-adjoint 2nd order differential operator [I. B. Bernstein et al. (1958)]
- Numerical approximations such as finite differences, finite volume, and finite elements in the position, \mathbf{x} , basis typically lead to a sparse banded matrix for \mathbf{F} and block-diagonal ρ

Hermitian form: $\mathbf{u} = \rho^{1/2} \boldsymbol{\xi} \quad \mathbf{H} = \rho^{-1/2} \mathbf{F} \rho^{-1/2} \quad \longrightarrow \quad \mathbf{H} \cdot \mathbf{u} = -\omega^2 \mathbf{u}$

Quantum phase estimation can be applied to ideal MHD stability

Is this a route to fast stability calculations for design optimization or feedback control?

J. B. Parker and I. Joseph arXiv:2002.08497 (2020)



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