



Quantum Algorithmic Breakeven: on scaling up with noisy qubits

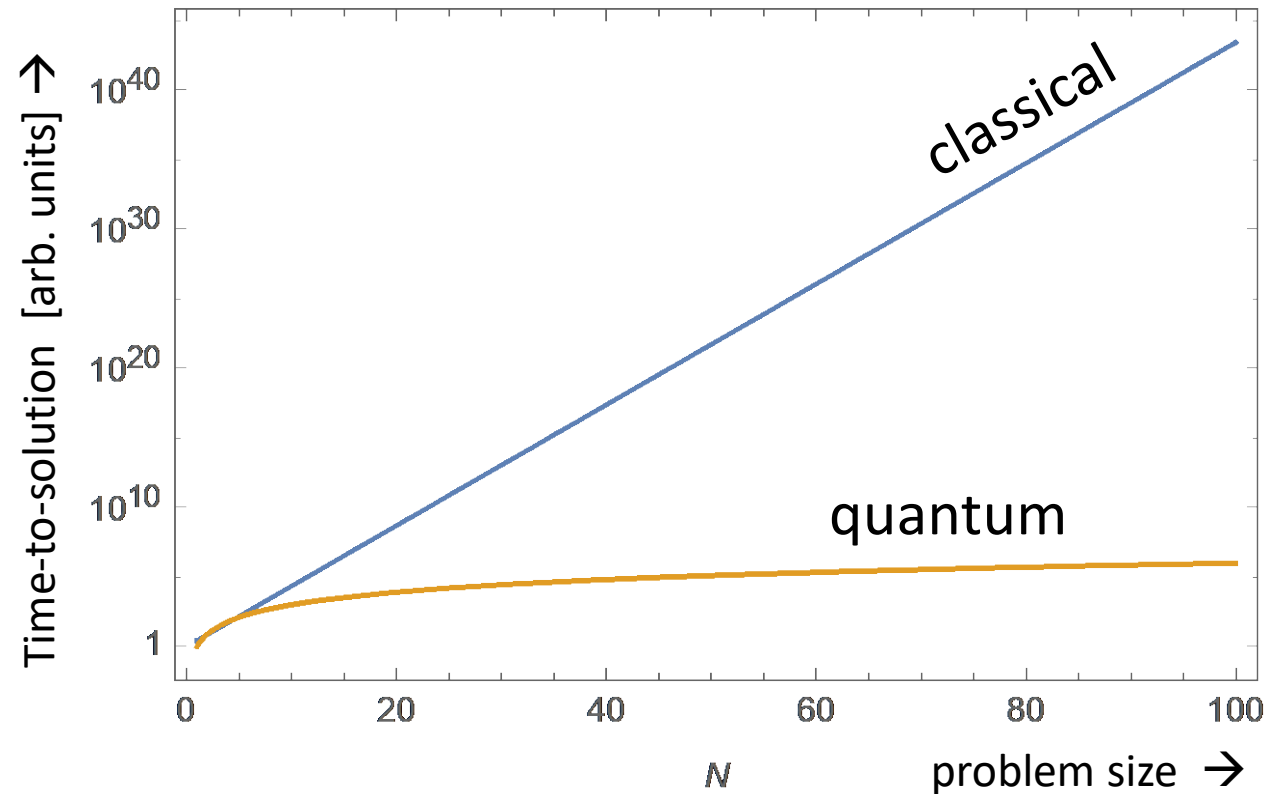
ISQST Quantum Symposium
Purdue
April 23, 2019

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USC

Algorithmic Scaling

Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware

ideal case: no decoherence

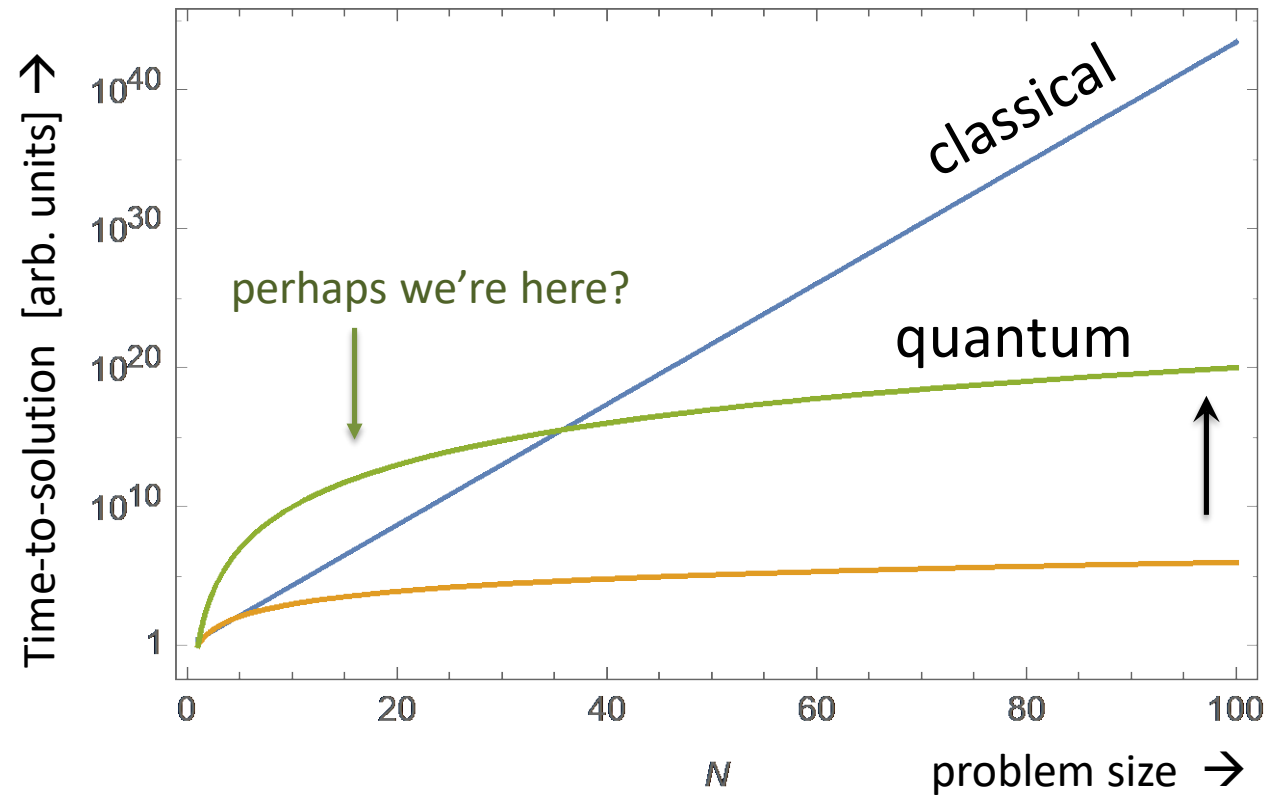


quantum scaling
advantage clear
for all N

So far never observed
on quantum hardware

Algorithmic Scaling

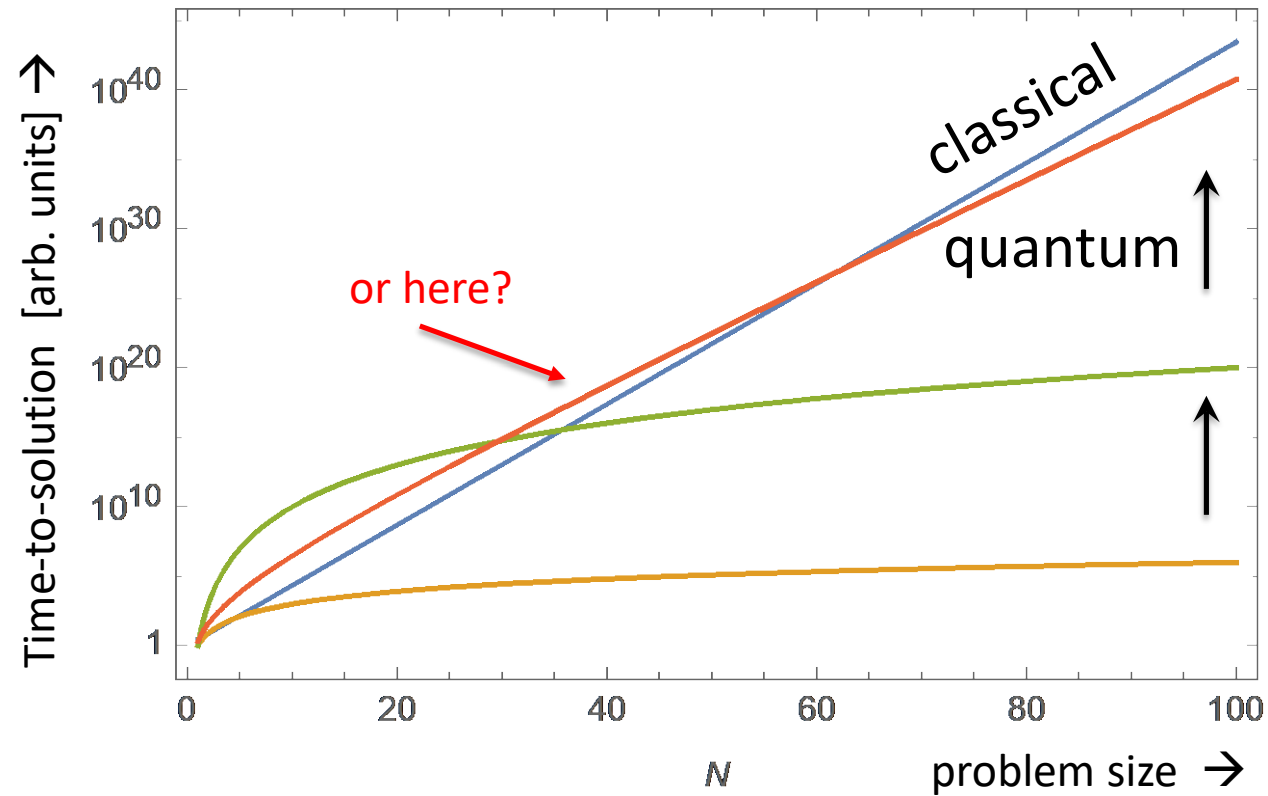
Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware
with decoherence



quantum scaling
advantage only
clear for large N

Algorithmic Scaling

Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware
with more decoherence

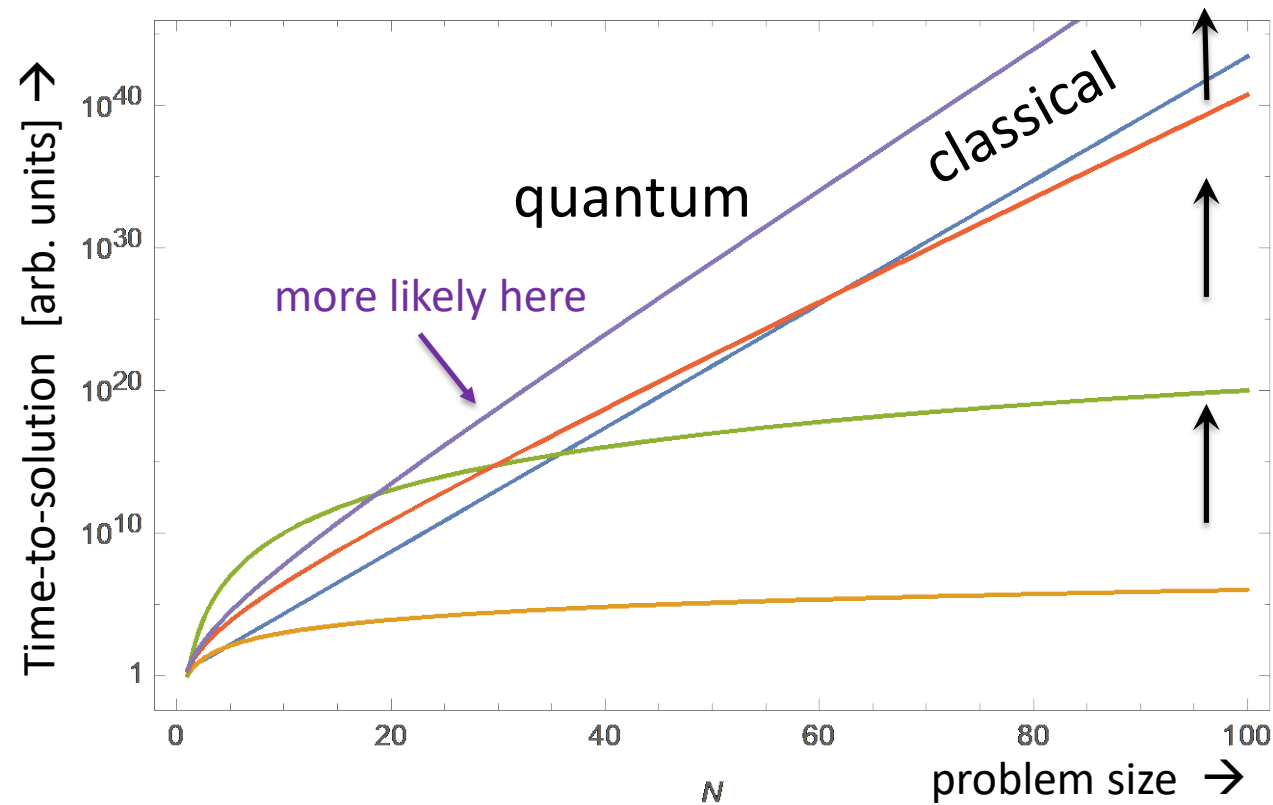


quantum scaling
advantage only
clear for even
larger N

Algorithmic Scaling

Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware

with too much decoherence

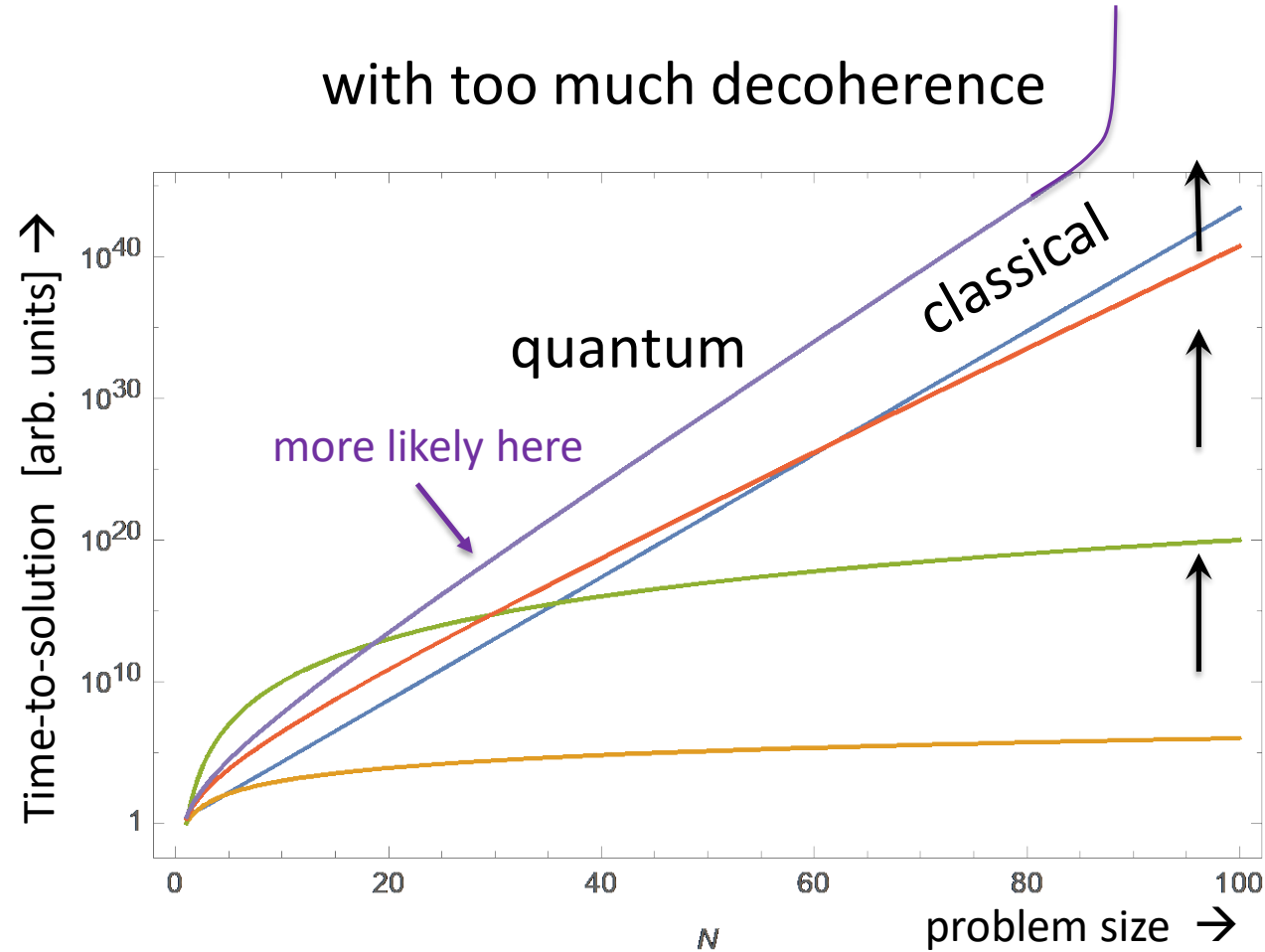


quantum scaling
disadvantage

Algorithmic Scaling

Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware

with too much decoherence



quantum blowup

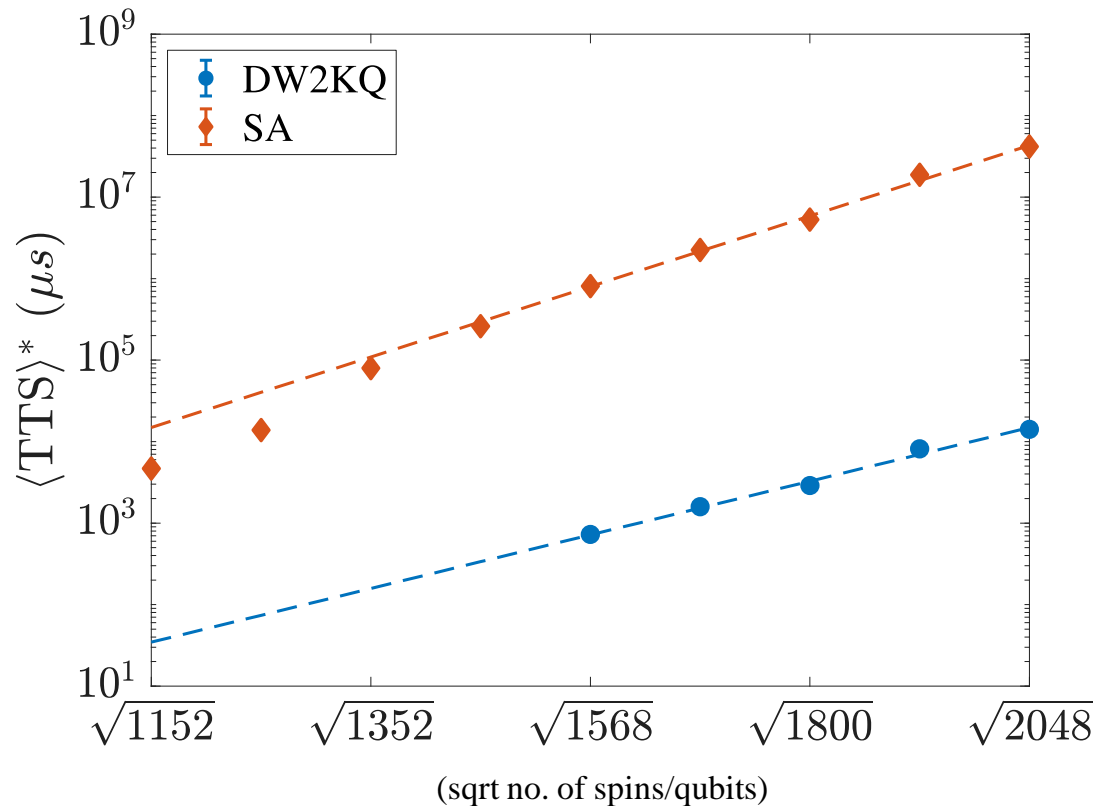
Trying for speedup: D-Wave 2000Q vs simulated annealing

Problem: find the ground state of a certain family of hard spin-glass instances of $N \leq 2048$ spins

Trying for speedup: D-Wave 2000Q scaling **advantage** against simulated annealing

Problem: find the ground state of a certain family of hard spin-glass instances of $N \leq 2048$ spins

Simulated Annealing (SA) with single-spin updates
D-Wave 2000Q unequivocally beats SA



Fit curves to
 $a \exp(bL)$

Solver	b [95% CI]
DW2KQ	0.760 ± 0.017
SA	1.002 ± 0.066

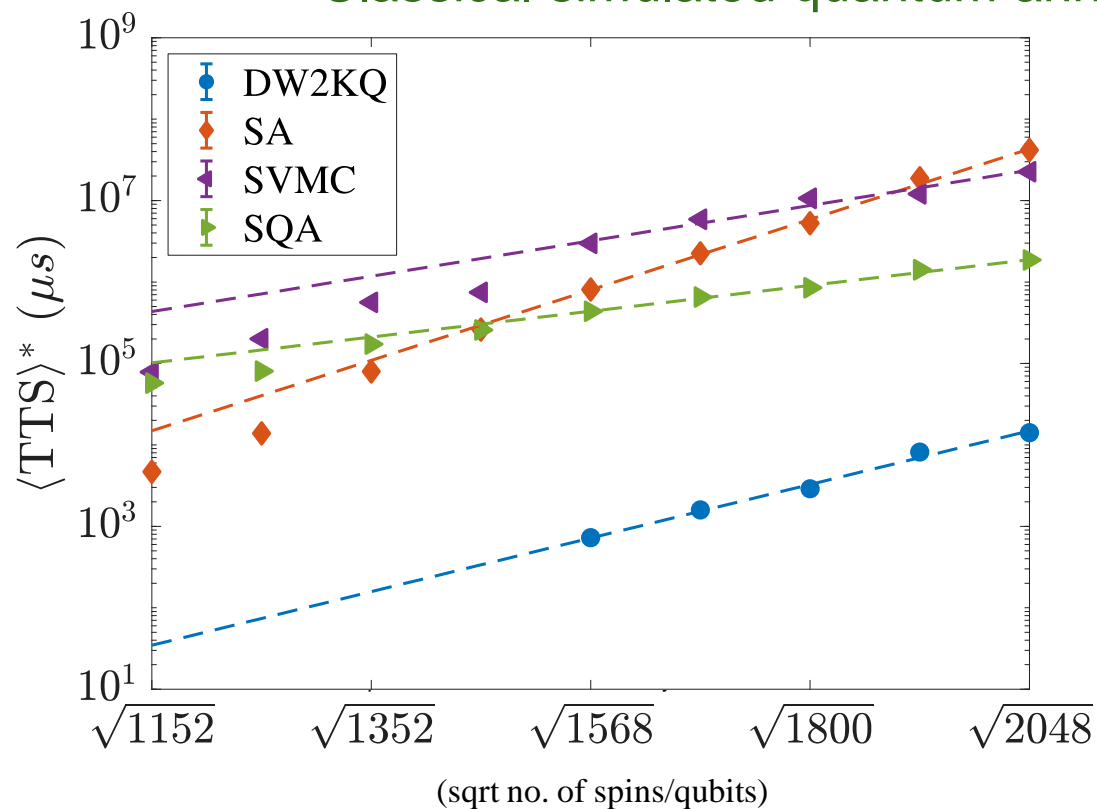
scaling of the median of the instance distribution

First example of an experimental quantum scaling advantage on a non-trivial problem relative to a generic classical algorithm

D-Wave 2000Q scaling **disadvantage** against better classical heuristic algorithms

Problem: find the ground state of a certain family of hard spin-glass instances of $N \leq 2048$ spins

Simulated Annealing (SA) with single-spin updates
 D-Wave 2000Q unequivocally beats SA
 Classical Spin Vector Monte Carlo (SVMC) beats both
 Classical simulated quantum annealing (SQA) beats all



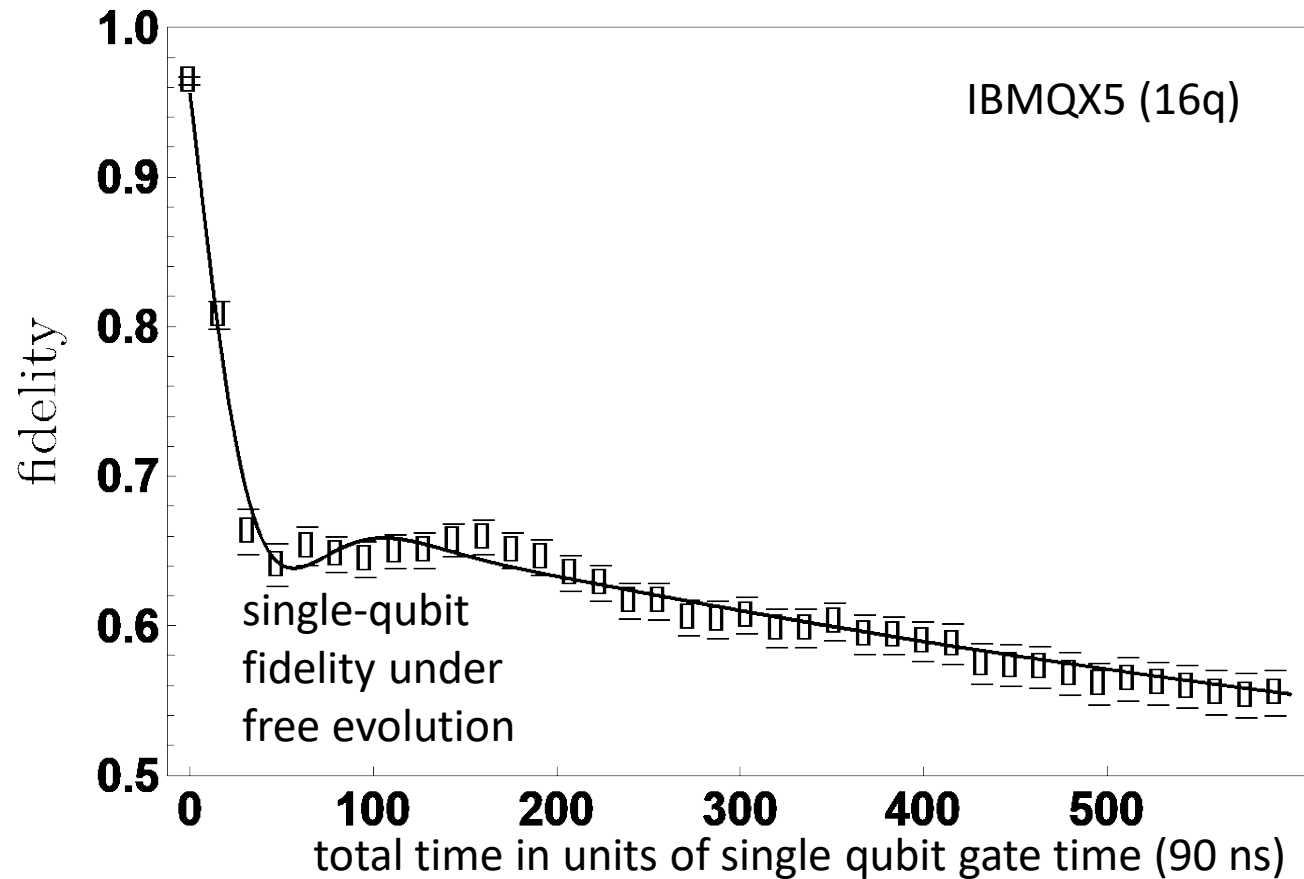
Fit curves to
 $a \exp(bL)$

Solver	b [95% CI]
DW2KQ	0.760 ± 0.017
SA	1.002 ± 0.066
SVMC	0.500 ± 0.029
SQA	0.370 ± 0.052

scaling of the median of the instance distribution

Why no speedup?

D-Wave quantum annealers are NISQ-era devices
Current NISQ-era devices are indeed Noisy

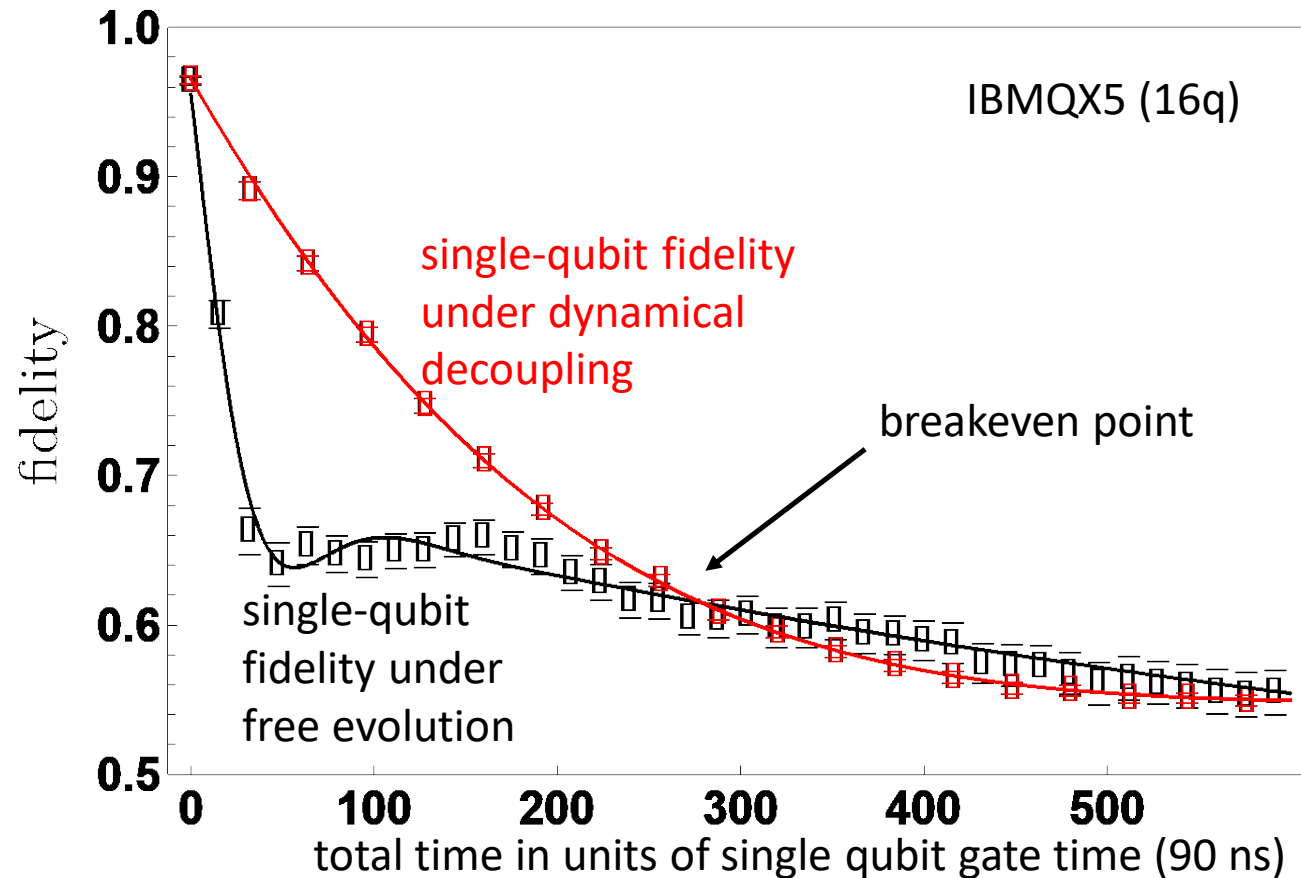


Why no speedup?

D-Wave quantum annealers are NISQ-era devices

Current NISQ-era devices are indeed noisy

But improvements are possible via error suppression methods



how well can we do with error correction?

Enter Quantum Error Correction (QEC)

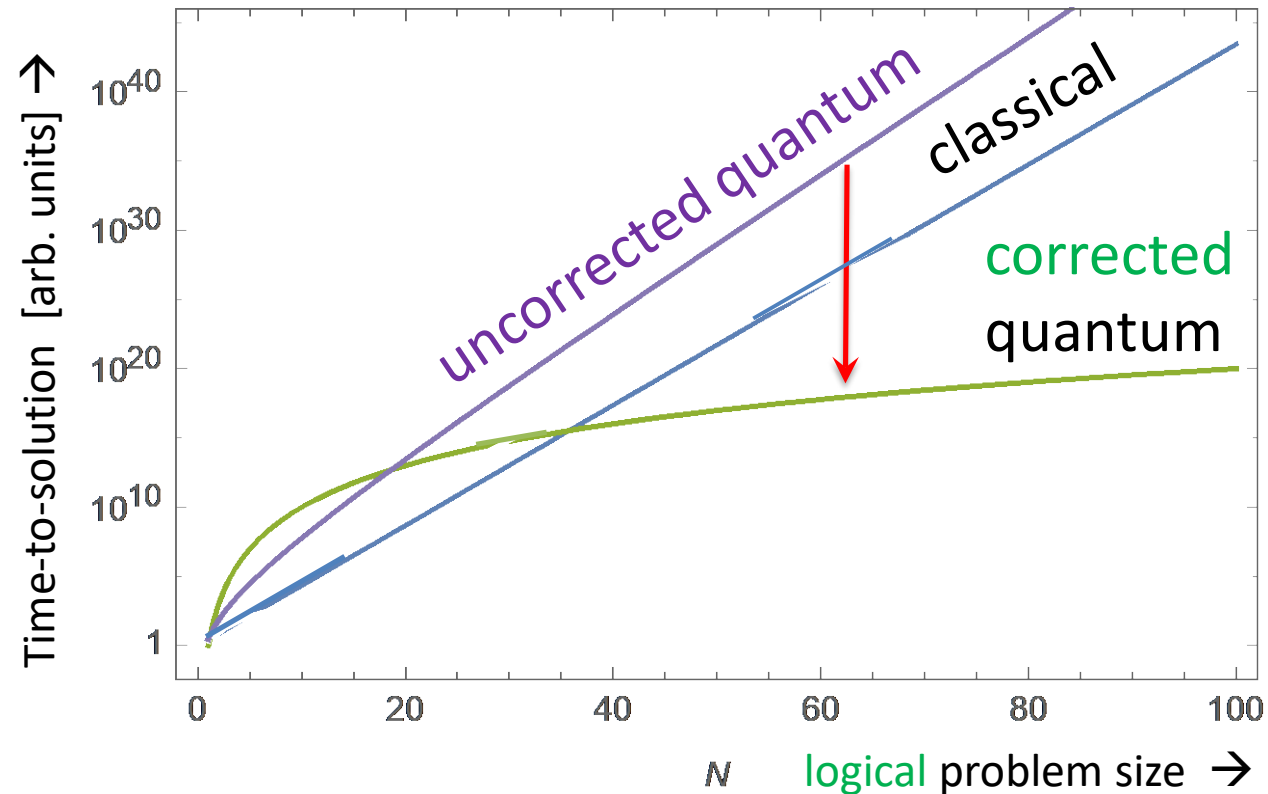
Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware

with decoherence + QEC

Algorithmic Success with QEC

Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware

with decoherence + QEC



quantum scaling advantage returns

achieving this with quantum hardware =

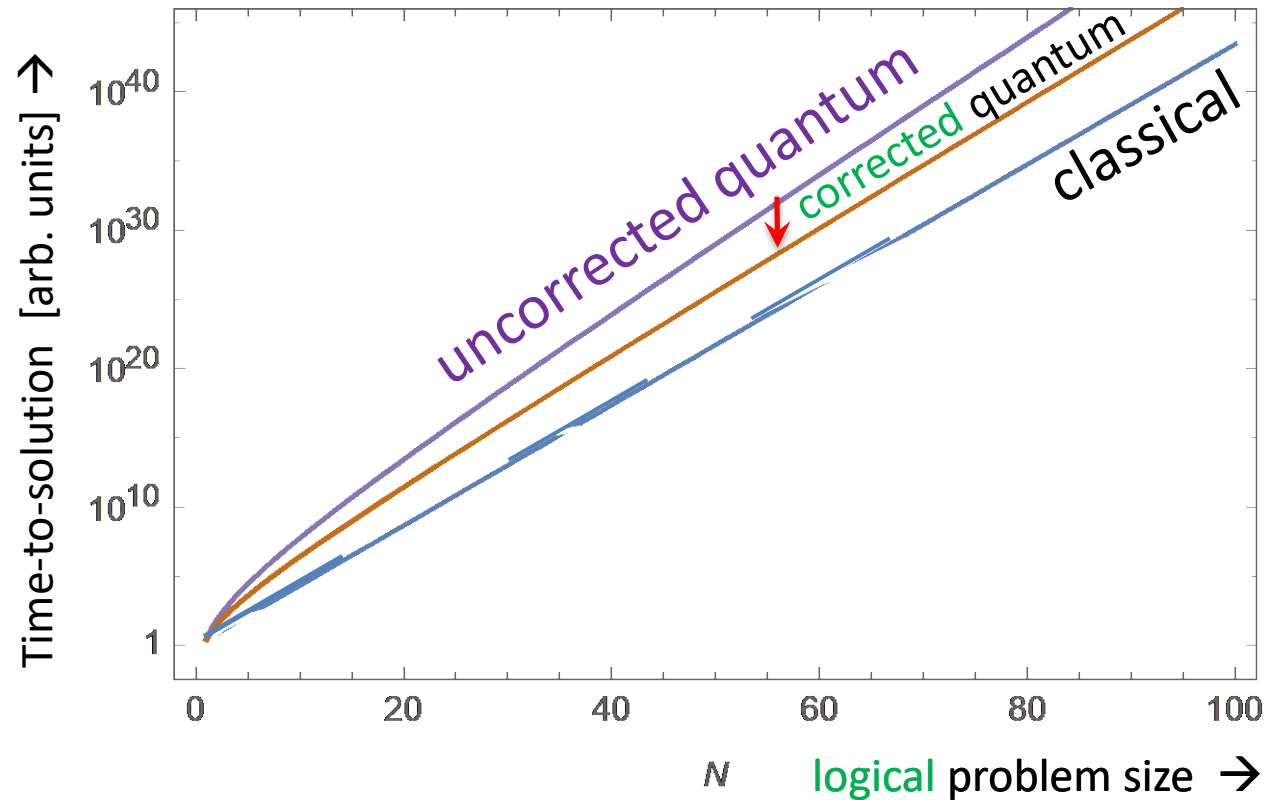


Algorithmic success with QEC: corrected quantum scaling is better than both uncorrected quantum & classical

Algorithmic Breakeven with QEC

Run an algorithm with exponential quantum speedup (e.g., quantum simulation) on quantum hardware

with decoherence + QEC



More modest:

Can **this** be achieved with *existing* quantum hardware?

Algorithmic breakeven with QEC: corrected quantum scaling is better than uncorrected quantum, but not necessarily better than classical

Algorithmic breakeven with quantum annealing

Brief intro to D-Wave processors

They are programmable quantum annealers;
Designed to solve optimization problems
formulated as Ising spin-glass Hamiltonians:

Given the Hamiltonian

$$H_{\text{Ising}} = \sum_{i=1}^N h_i \sigma_i^z + \sum_{i<j} J_{ij} \sigma_i^z \sigma_j^z$$

Find the minimizing spin configuration $\{\sigma_i^z = \pm 1\}$

Solve by adiabatically evolving the transverse field Ising
Hamiltonian

$$H(t) = A(t) \sum_{i=1}^N \sigma_i^x + B(t) H_{\text{Ising}}$$

from ground state of $\sum_{i=1}^N \sigma_i^x$ to ground state of H_{Ising}

4 generations so far:

D-Wave 1:

$N=128$ (USC)

K_{17} for ideal

K_{14} for actual

D-Wave 2:

$N=512$ (USC, NASA)

K_{33} for ideal

K_{32} for actual

D-Wave 2X:

$N=1152$ (USC, NASA, LANL)

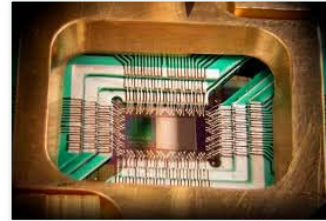
K_{49} for ideal

K_{44} for actual

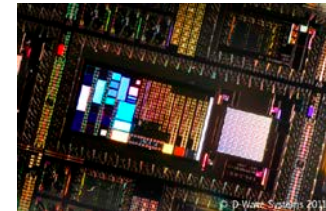
D-Wave 2000Q

$N=2048$ (NASA, LANL)

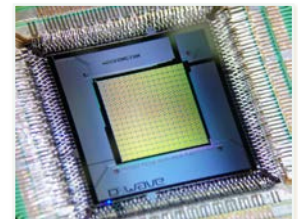
K_{65} for ideal



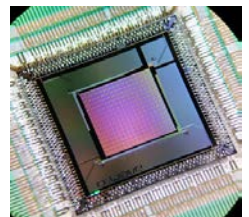
Oct 2011



March 2013



March 2016



Sep 2017

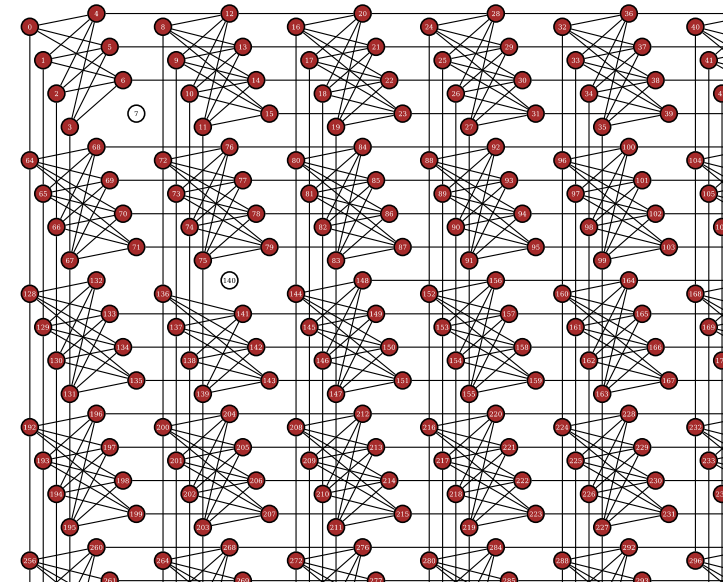
Brief intro to D-Wave processors

- Use superconducting Nb flux qubits, each coupled to up to 6 other qubits (“Chimera graph”)
- $T_1, T_2 \sim 10 - 100ns$, annealing time $t_f \geq 1\mu s$
- minimum gap(H) can be $\ll T \sim 10mK$
- “Stoquastic”: efficient classical simulation possible in many cases using Quantum Monte Carlo

How
quantum?
“somewhat”

A testbed for
algorithmic scaling with noisy qubits
and error correction

Chimera graph:
square grid of $L \times L$ unit cells
number of qubits $N = 8L^2$



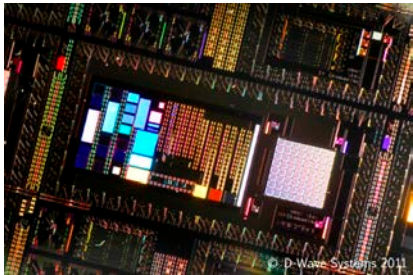
Algorithmic breakeven with quantum annealing

Problem:
find ground state energy of

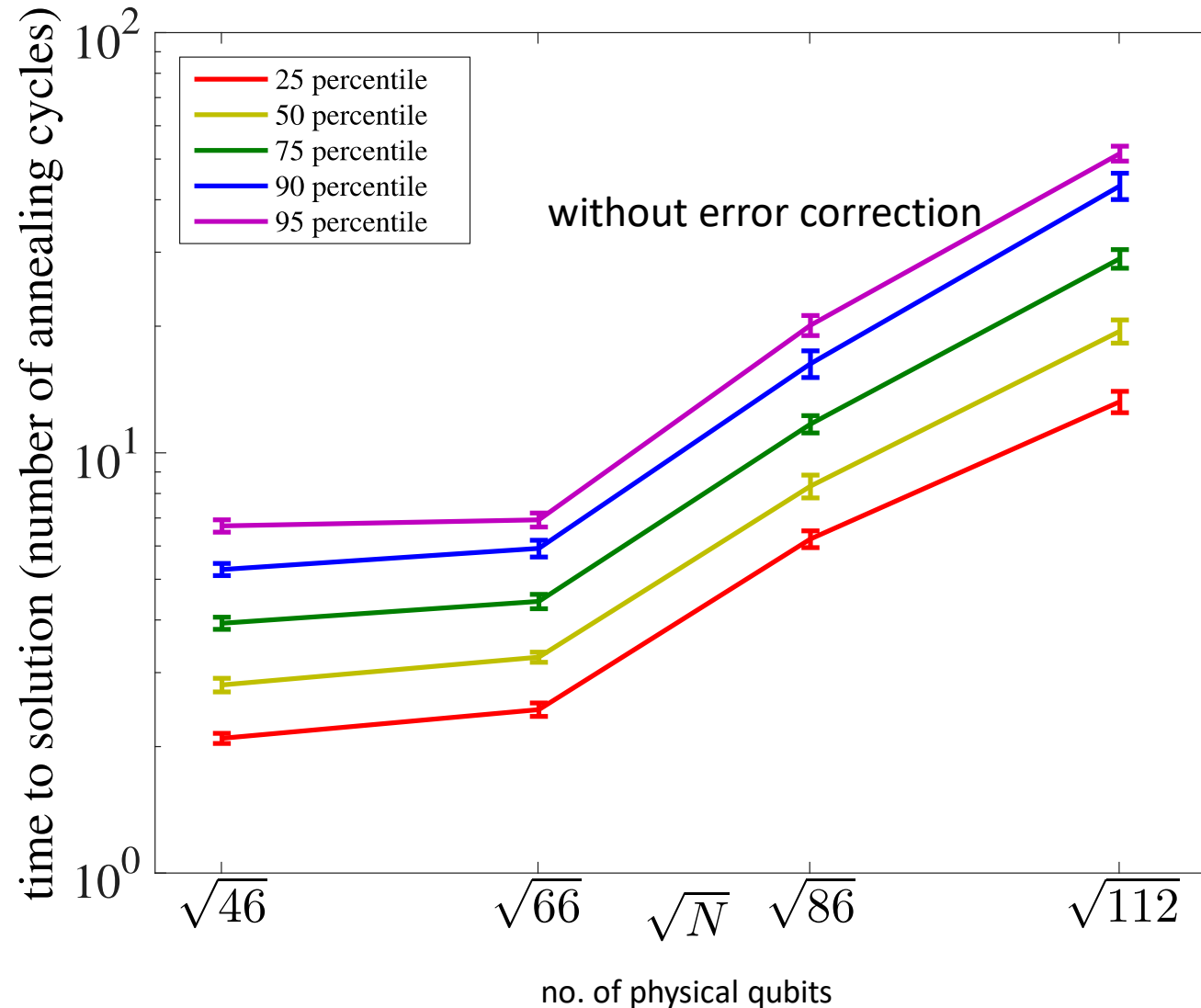
$$H_{\text{Ising}} = \sum J_{ij} \sigma_i^z \sigma_j^z$$

with random

$$J_{ij} \in \left\{ \pm \frac{1}{6}, \dots, \pm \frac{5}{6}, \pm 1 \right\}$$



Run on a D-Wave 2
(503 qubits)



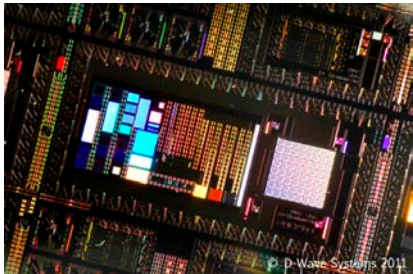
Algorithmic breakeven with quantum annealing ✓

Problem:
find ground state energy of

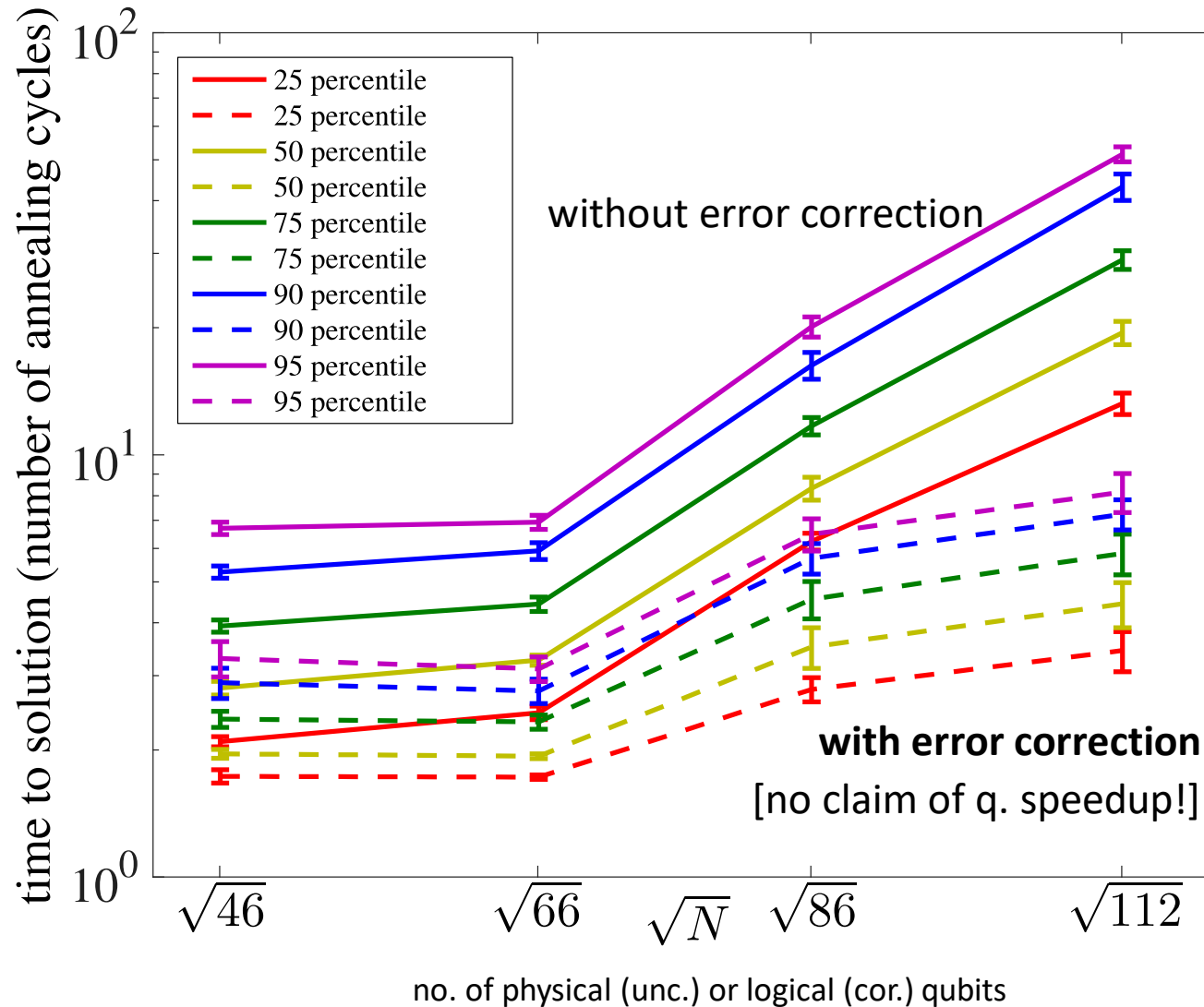
$$H_{\text{Ising}} = \sum J_{ij} \sigma_i^z \sigma_j^z$$

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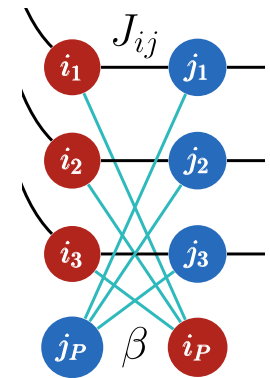
$$J_{ij} \in \left\{ \pm \frac{1}{6}, \dots, \pm \frac{5}{6}, \pm 1 \right\}$$



Run on a D-Wave 2
(503 qubits)



Quantum Annealing Correction



QAC works by:

- introducing energy penalty against excitations
- majority-vote decoding of logical qubits

Up the ante: analog control errors

Ideally, a quantum annealer evolves adiabatically according to $H(t) = A(t)H_X + B(t)H_P$

$$\begin{aligned}
 \downarrow H_X &= H_{\text{transverse}} = \sum_{j \in V} \sigma_j^x \\
 \uparrow H_P &= H_{\text{Ising}} = \sum_{j \in V} h_j \sigma_j^z + \sum_{(i,j) \in E} J_{ij} \sigma_i^z \sigma_j^z
 \end{aligned}$$

programmable

In reality: analog errors \rightarrow J-chaos
 H implemented $\neq H$ intended

$$+ \sum_j \delta h_j \sigma_i^z + \sum_{ij} \delta J_{ij} \sigma_i^z \sigma_j^z$$

$$\delta h_i, \delta J_{ij} \sim \mathcal{N}(0, \eta)$$

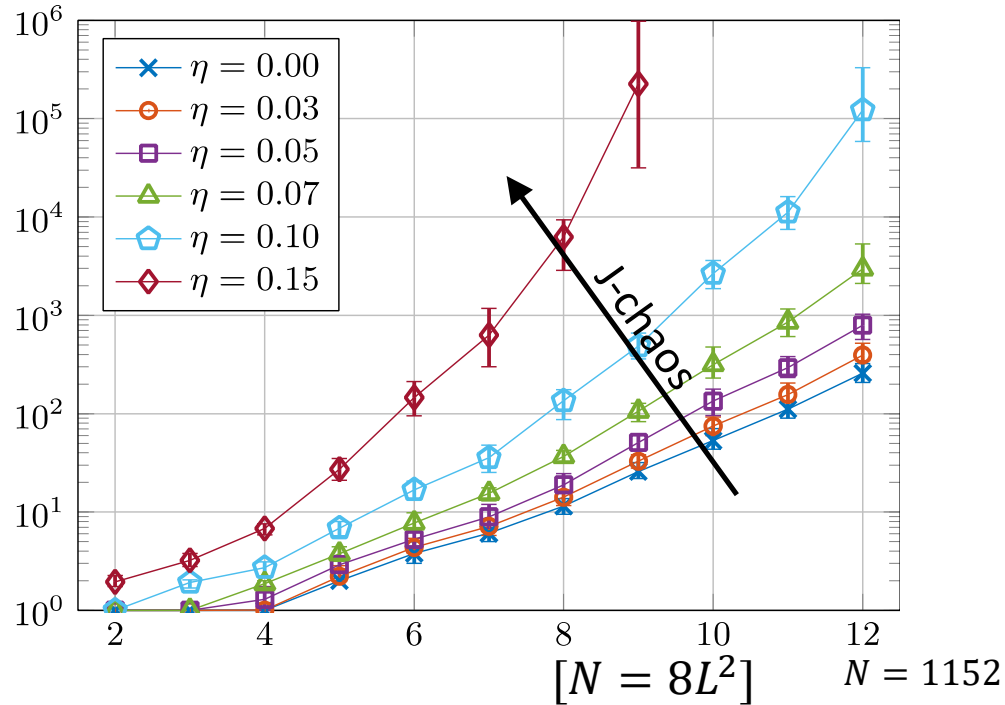
D-Wave 2X chip intrinsic analog errors: $\frac{\delta h}{h_{\max}}, \frac{\delta J}{J_{\max}} \sim N(0, 0.03)$

Let's add artificial Gaussian noise $\delta J_{ij} \sim N(0, \eta_{\text{int}} + \eta)$ on top of intrinsic analog errors: $\eta_{\text{int}}=0.03$
 $\eta \in \{0.03, 0.05, 0.07, 0.10, 0.15\}$

Time-to-solution as a function of problem size and noise

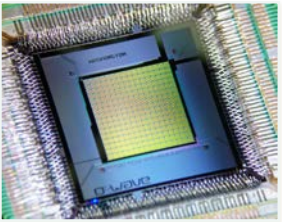
Time-to-solution (# of repetitions R at $t_f = 5\mu\text{s}$) for median instances

without error correction



number of runs
required to find
the ground state

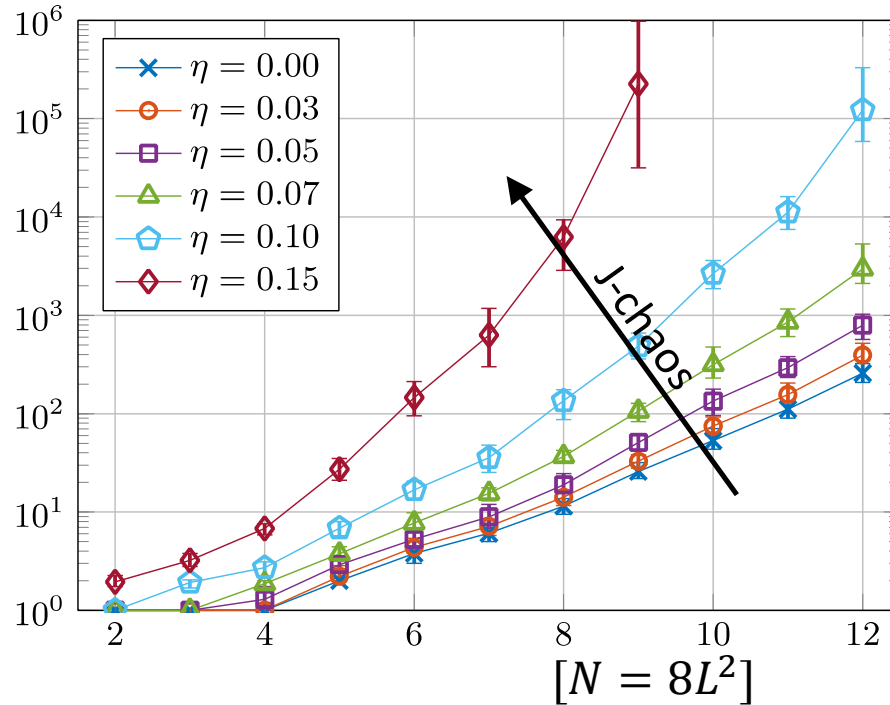
run on USC's DW2X



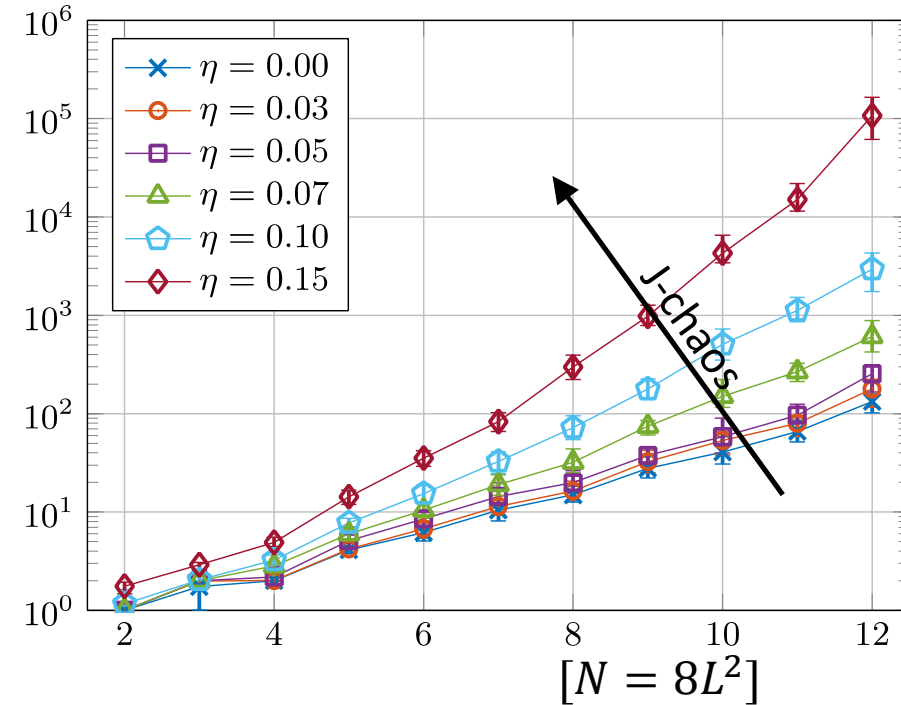
Time-to-solution as a function of problem size and noise

Time-to-solution (# of repetitions R at $t_f = 5\mu s$) for median instances

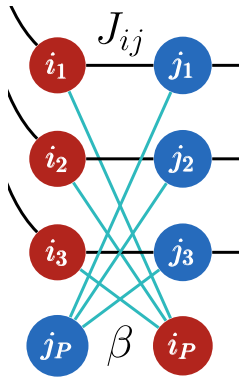
without error correction



with error correction

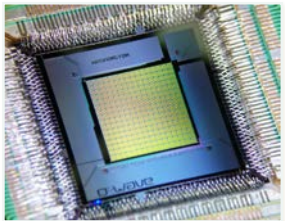


Quantum Annealing Correction



number of runs required to find the ground state

run on USC's DW2X



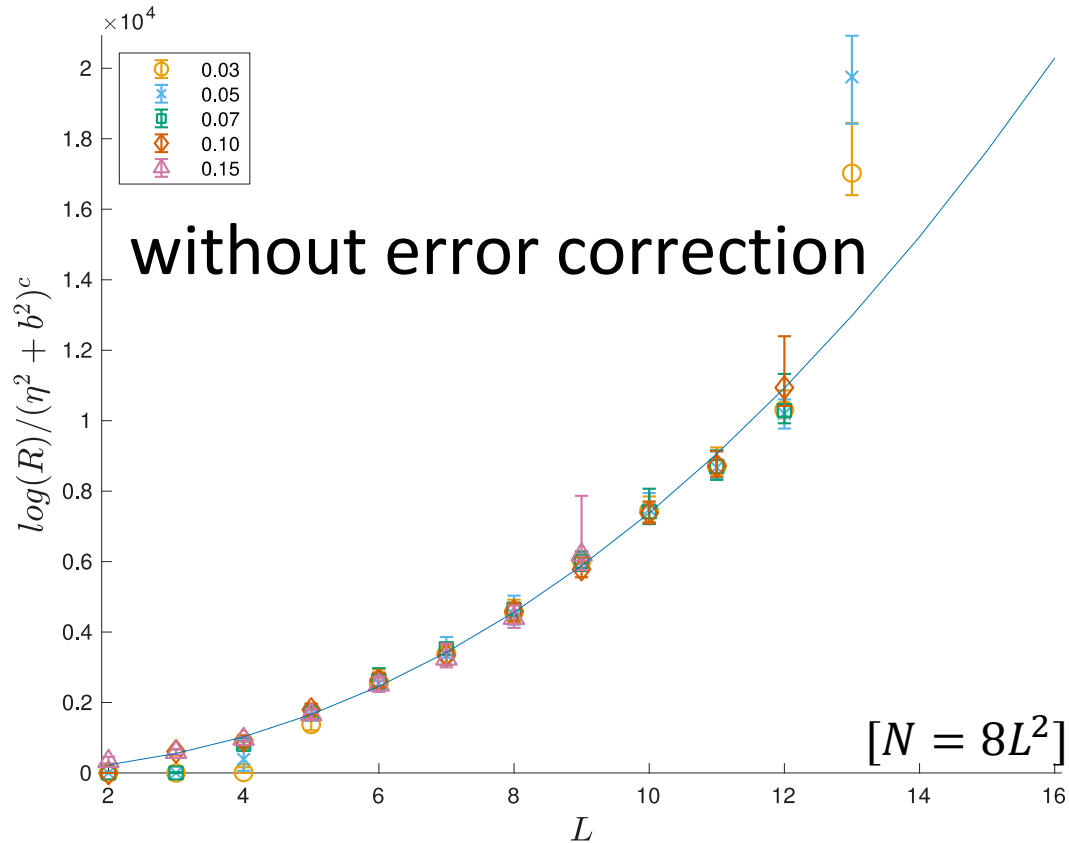
scaling with Quantum Annealing Correction is better than without but how good is this scaling?

perform data collapse, extract finite-size scaling

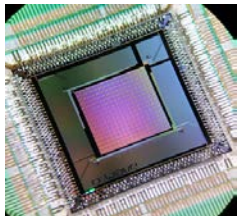
algorithmic breakeven
error-corrected scaling that is better than uncorrected on a non-trivial computational problem

Time-to-solution data collapse & finite-size scaling:

$$\text{fit to } R = e^{a(\eta^2 + b^2)^c} L^d$$



run on
NASA's
DW2000Q



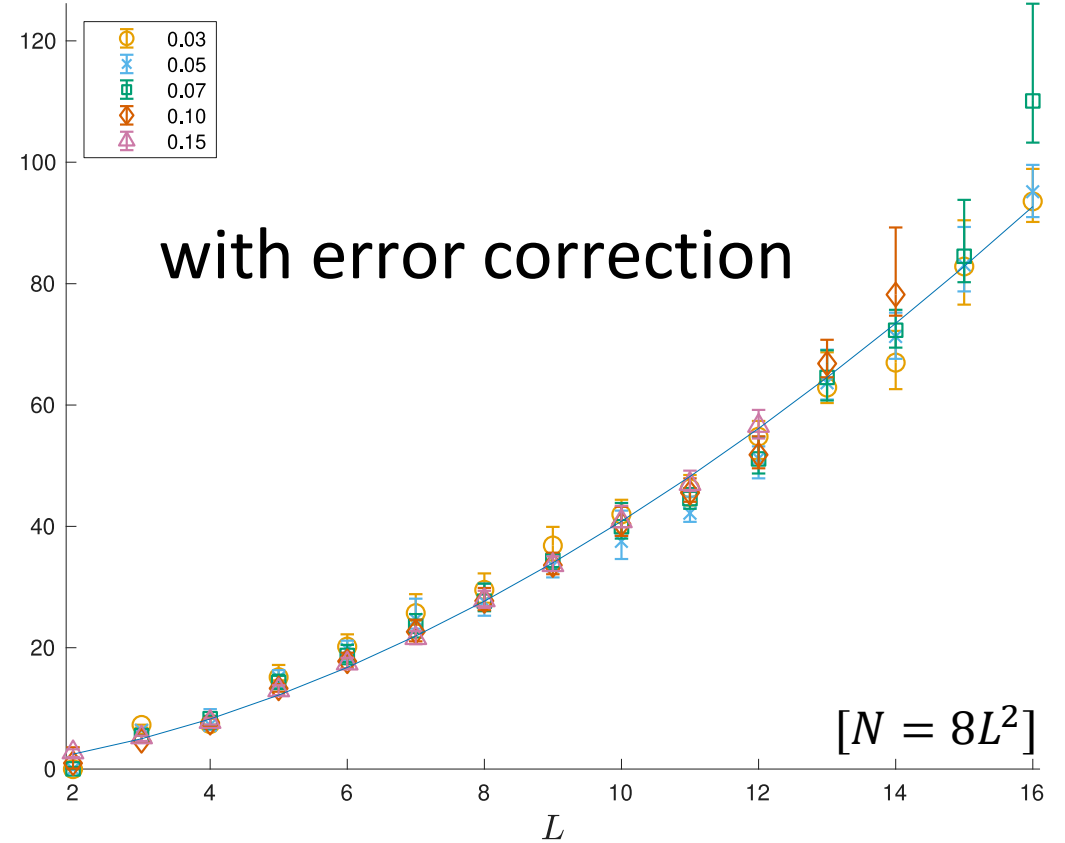
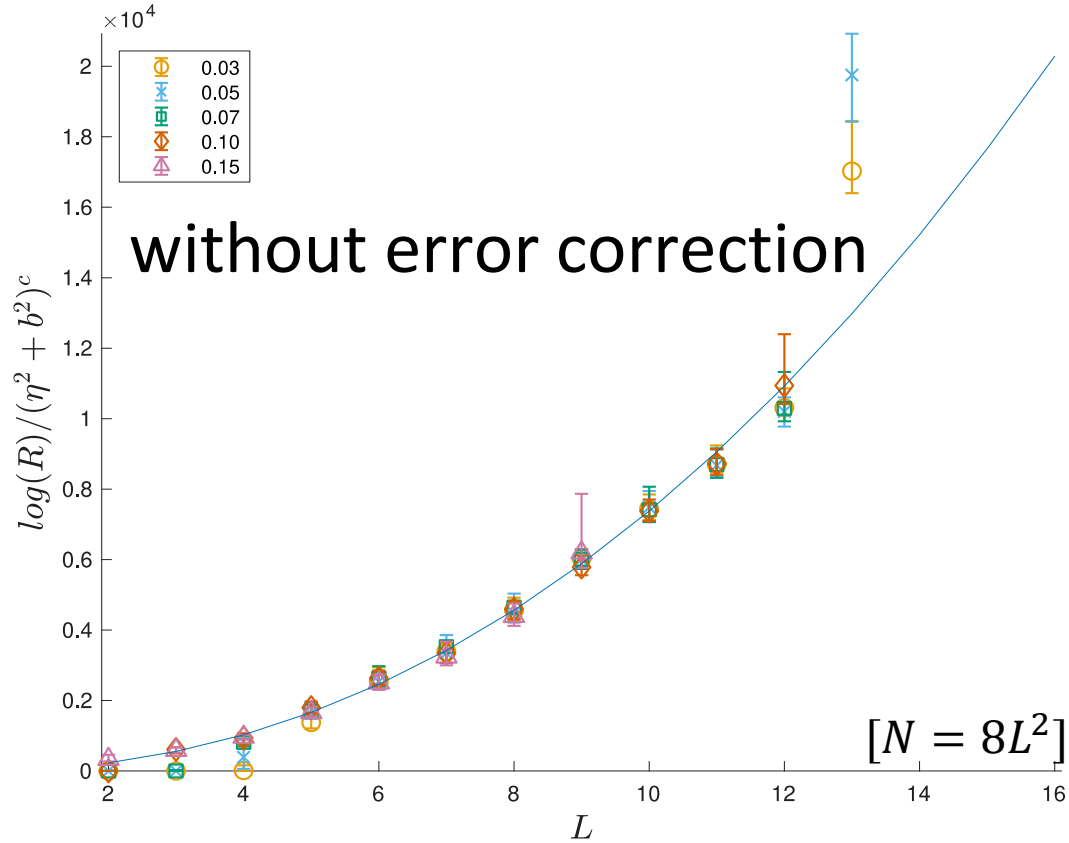
$$\begin{aligned} a &= 51.97 \\ b &= 0.16 \\ c &= 2.04 \\ d &= \mathbf{2.15} \end{aligned}$$

worse than classical *deterministic* upper bound

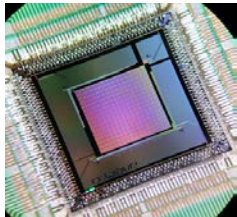
$$R \sim \exp(\xi L^2)!$$

Time-to-solution data collapse & finite-size scaling:

$$\text{fit to } R = e^{a(\eta^2 + b^2)^c} L^d$$



run on
NASA's
DW2000Q



$a = 51.97$
 $b = 0.16$
 $c = 2.04$
 $d = 2.15$

better than
classical
upper bound
(lower bound
unknown)

$a = 0.74$
 $b = 0.06$
 $c = 0.44$
 $d = 1.74$

} all smaller

error correction
restores hope :-)

Beyond D-Wave: The IARPA Quantum Annealing Consortium

IARPA Quantum Enhanced Optimization (QEO) Program

Goal: find out the ultimate capabilities of quantum annealing. Is there a quantum speedup?

Building a 100-qubit quantum annealer using *high-coherence* (Al) superconducting flux qubits, for quantum optimization and sampling applications. *Built-in error suppression and correction.*

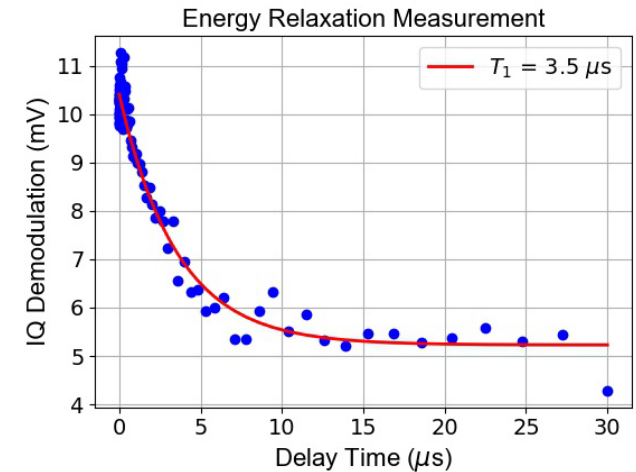
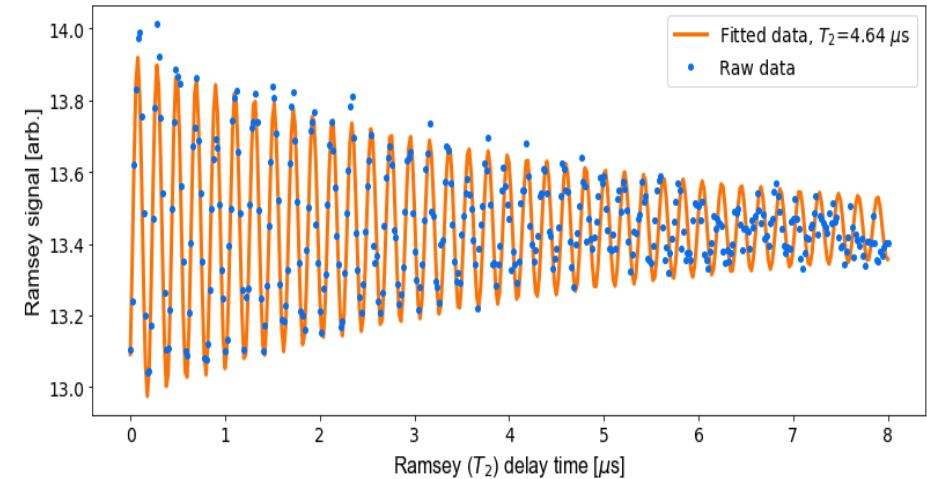


I A R P A



LOCKHEED MARTIN

NORTHROP GRUMMAN

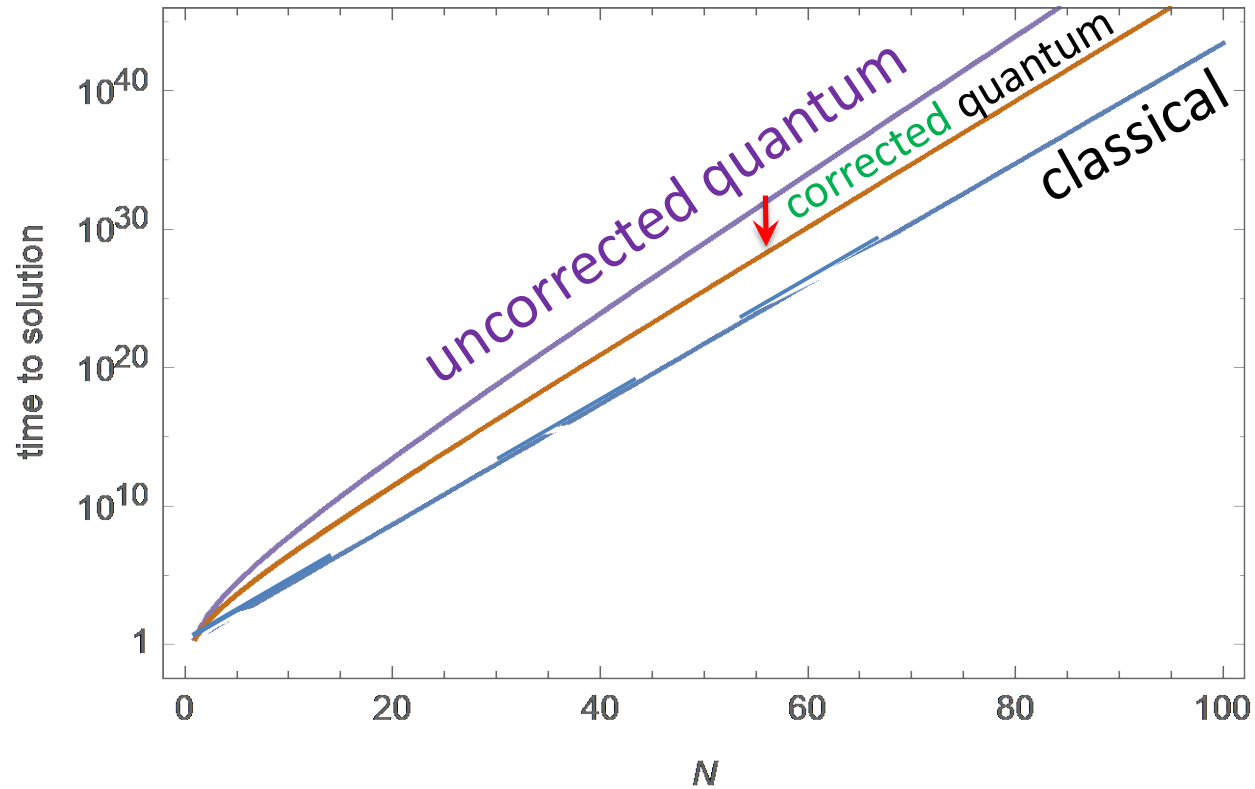


$T_1, T_2 \sim 4 \mu\text{s}$ for annealing-capable qubits
coupling $|J_{ij}| \sim 2 \text{ GHz}$

Quantum Algorithmic Breakeven

Intermediate goal for QC, similar in spirit to “quantum supremacy”:

Demonstrate error-corrected scaling that is better than uncorrected on a non-trivial computational problem



Already achievable in quantum annealing

An invitation for gate-model quantum computation

Thanks!