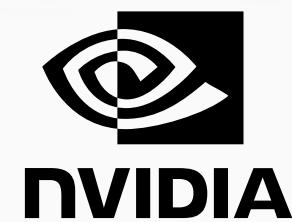


OPTIMIZING LABS

-

AN ANALYSIS OF HYBRID QUANTUM ALGORITHMS

Chengyi, Brayden, Jiani, Sofi, Zaara



MEET THE SQUACKXIONS

(in their excited state)



The Problem & Our Plan

Low Autocorrelation Binary Sequences (LABS)

Find the binary sequence that:

$$\min\{E(s)\} = \min\left\{\sum_{k=1}^{N-1} C_k^2\right\} \quad \text{with} \quad C_k = \sum_{i=1}^{N-k} s_i s_{i(i+k)}$$

CLASSICAL

“Golden standard”

- Memetic Tabu Search (MTS)
- Global search, greedy local descent
- $N \leq 64$
- $\approx O(1.25^N)$ CPU
- $\approx O(1.21^N)$ GPU

CLASSICAL + QUANTUM I

Phase 1

- QE: seeding initial population for MTS
- Measured improvement is shown only for $27 < N < 37$
- $\approx O(1.38^N)$

CLASSICAL + QUANTUM II

Phase 2

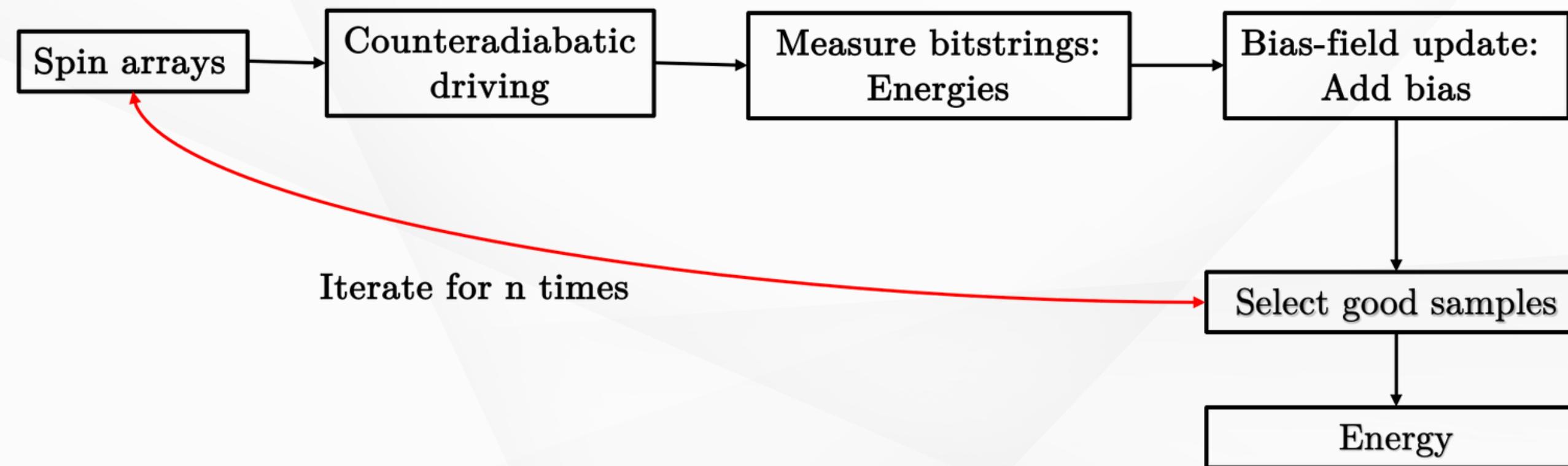
- QE improvement:
Bias in magnetization (spins) learned
- $\approx O(1.20^N)$

BF-DCQO Algorithm

Bias-Field Counterdiabatic Quantum Optimization

WHY?

- Performs DCQO iteratively to use the solution from each to bias the next iteration
- Circumvent local minima & lead to faster convergence



The Plan: Acceleration

CLASSICAL

- From [Zhang et al. paper](#) : 26x speedup over 16-core CPU implementations
- How:
 - All in GPU
 - Two Level Parallelism
 - Shared Memory Data Structures

QUANTUM

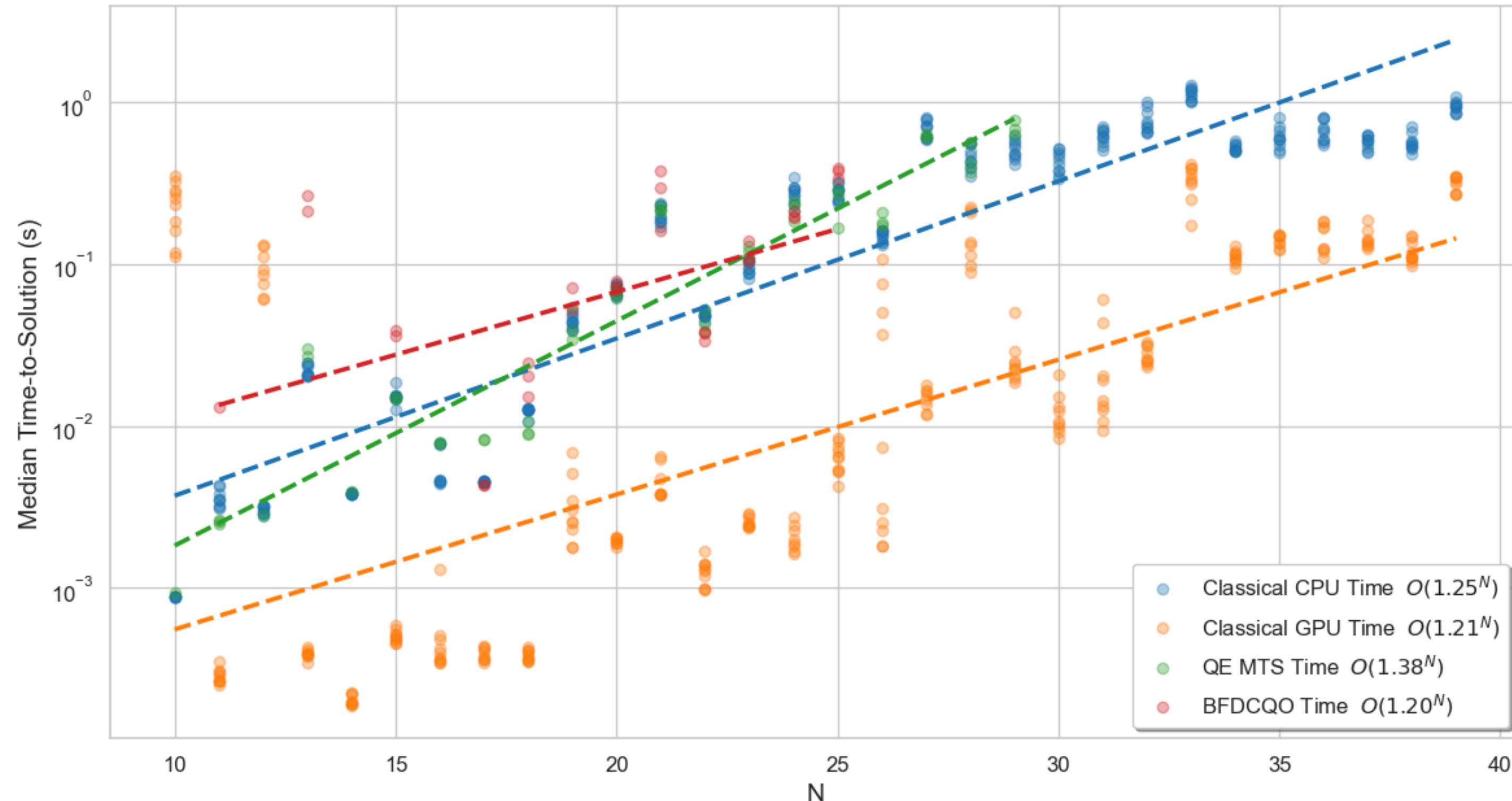
- Switch to NVIDIA GPUs as backends (A100 for dev/H100 for benchmark)
- MGPU/MQPU Backends
- Experimentation on superconducting quantum computers (QPU's) from IQM Resonance

The Pivot

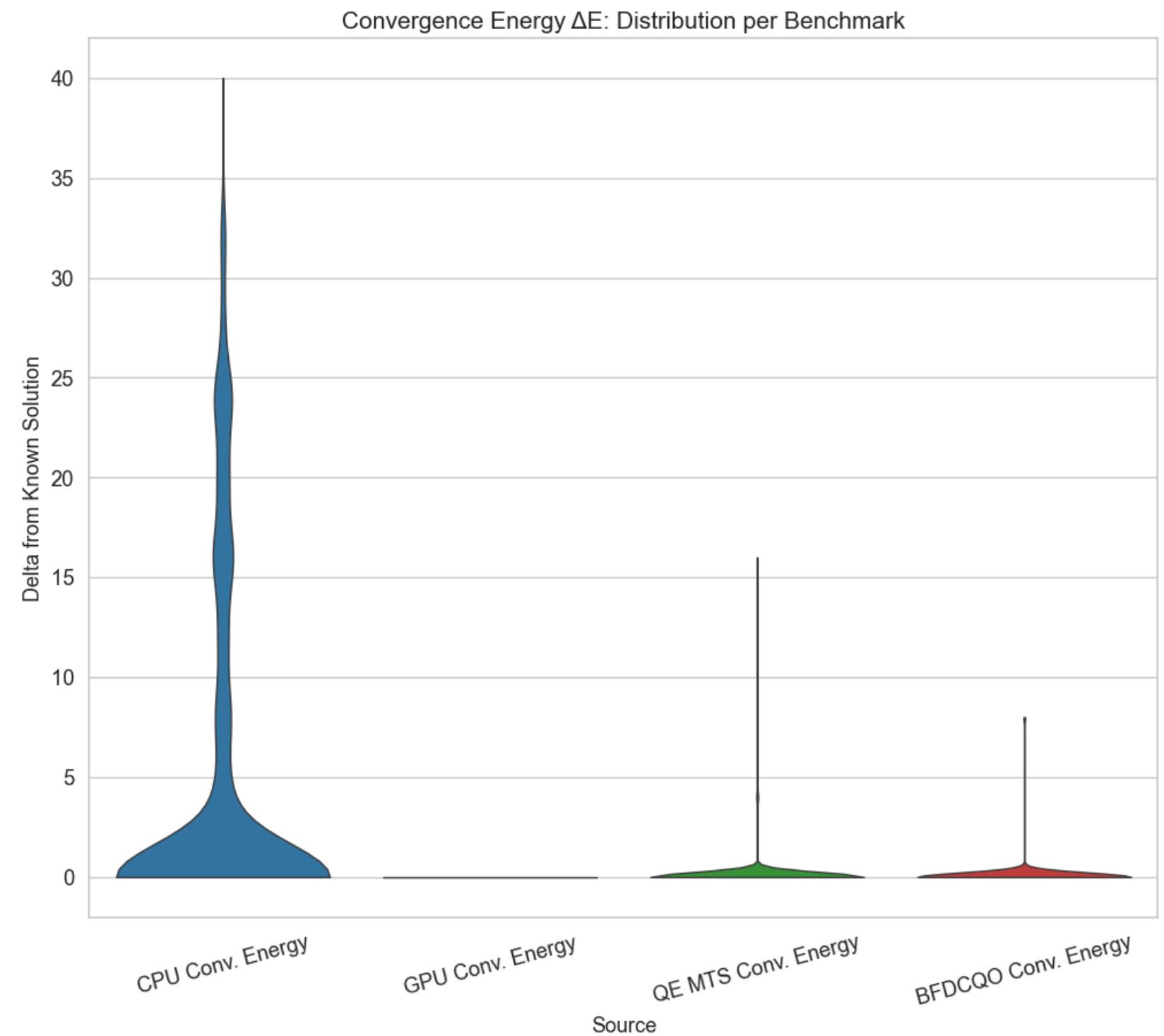
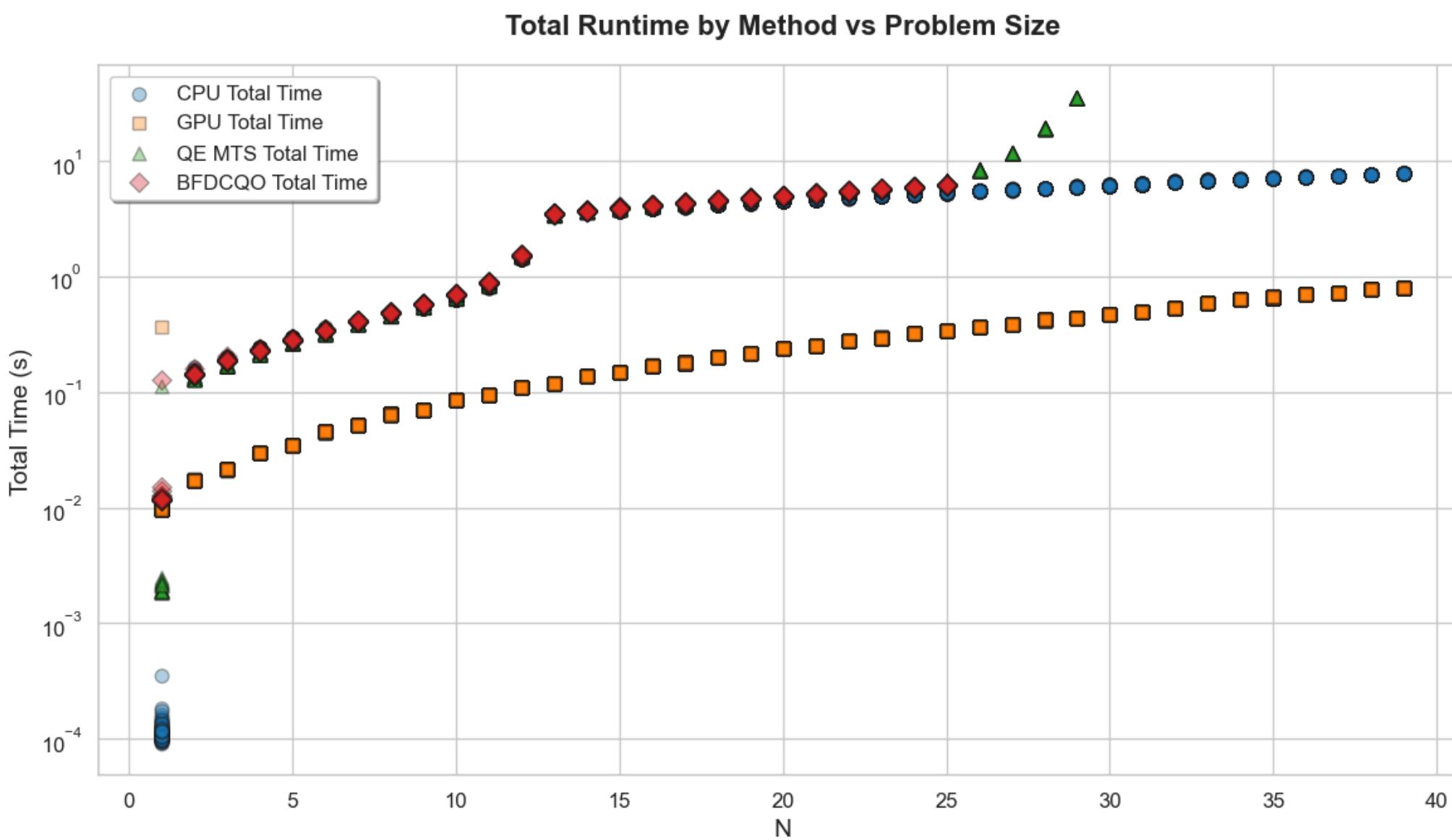
- Diversified Computing Platforms
 - qBraid for benchmarking, brev for dev
 - More compute power!
- Accelerating quantum algorithm by running on IQM's Quantum Computers → future work

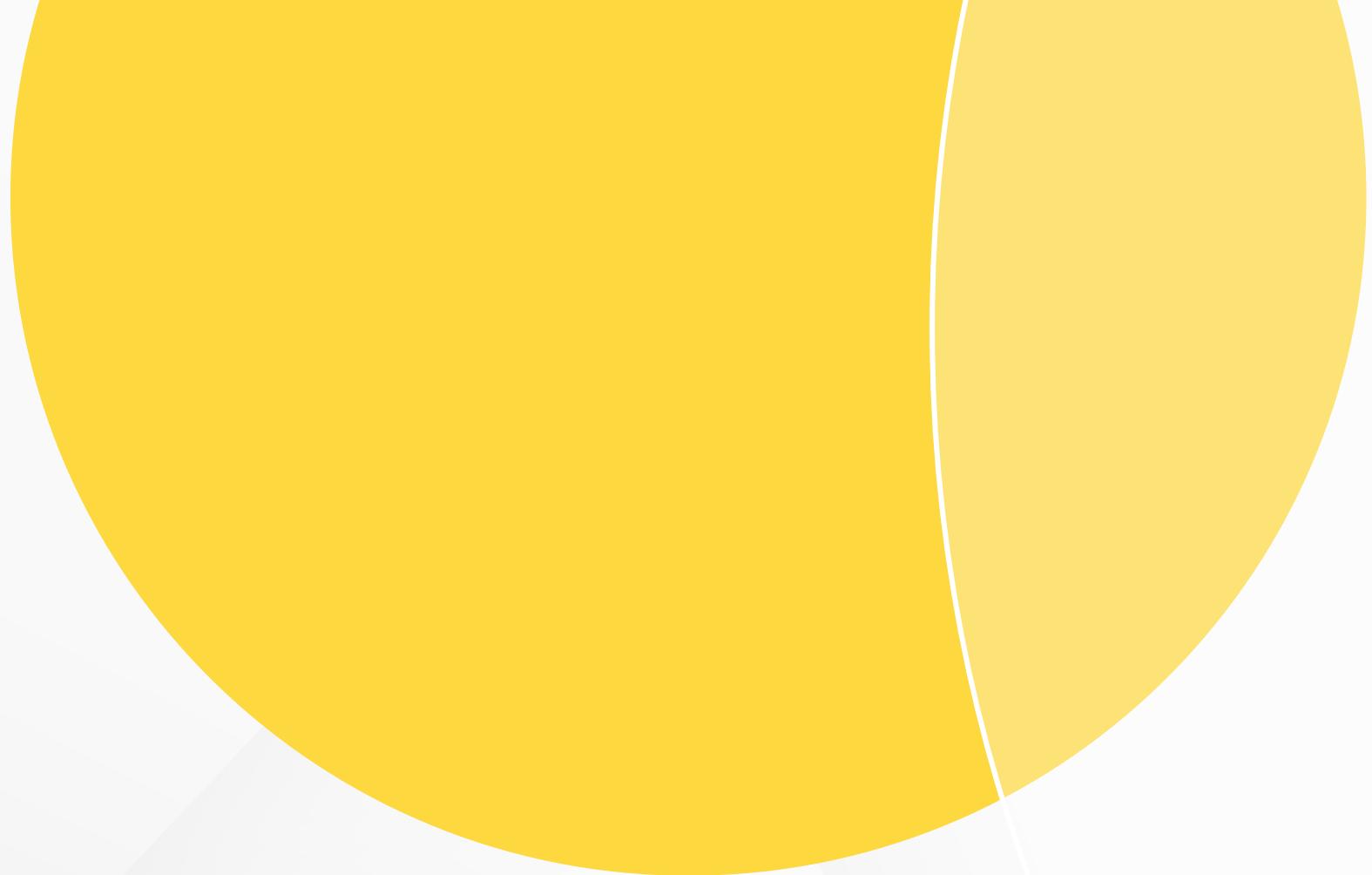
Results

Median Time-to-Solution (s) vs Problem Size

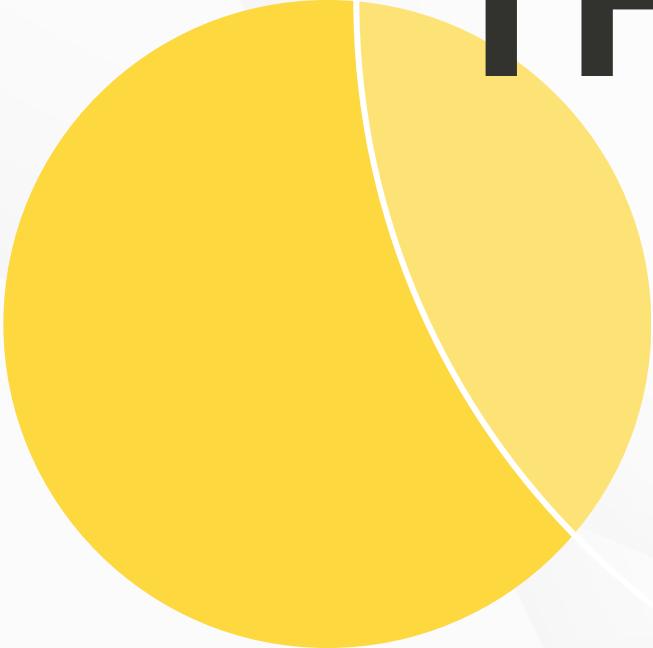


Results





Team reflections



THANK YOU!