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SCALABLE BENCHMARKING OF QUANTUM CHEMISTRY ALGORITHMS USING CIRCUIT MIRRORING

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CURRENT STATE OF QUANTUM BENCHMARKS



- High-level Approaches
 - Scalable, measure general performance of a device
 - E.g., randomized benchmarking, IBM quantum volume benchmark
- Tomographic Approaches
 - Not scalable, but detailed
 - E.g., gate set tomography, process tomography
- What we want from an application-specific benchmark:
 - **Scalable** to many qubits
 - **Representative** of the performance of a specific algorithm
 - **Generalizable** to any algorithm

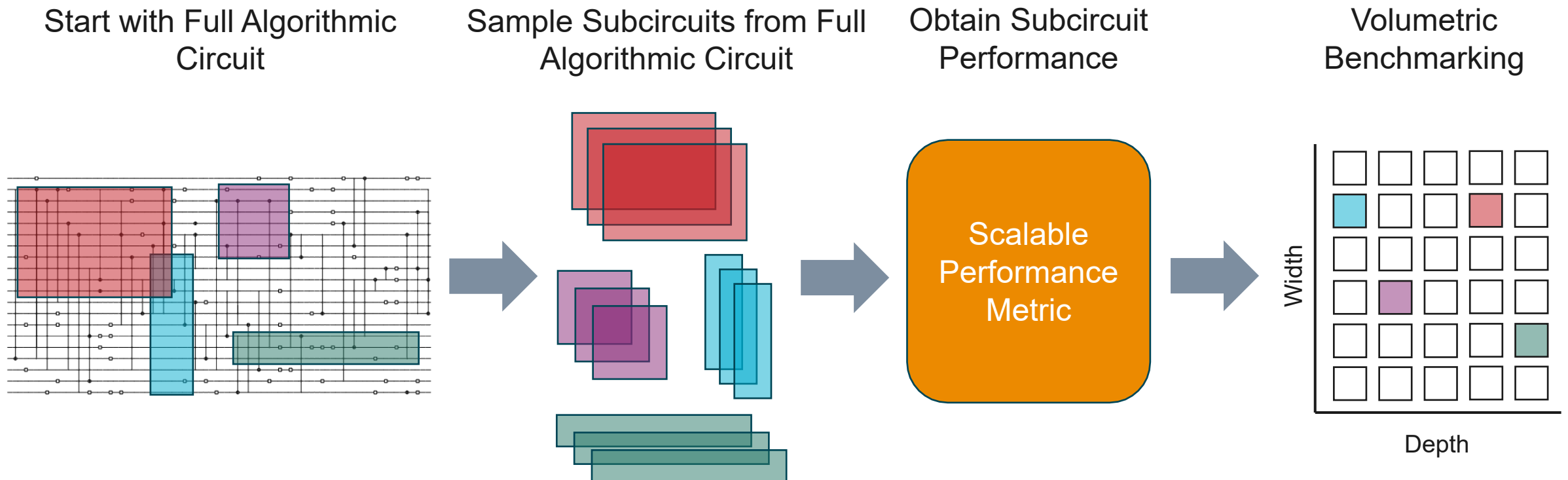
APPLICATION-SPECIFIC BENCHMARKS



- Current application-specific benchmarks typically...
 - Exploit properties of specific problems (i.e., efficiently verifiable circuit outcomes)
 - Lack scalability
- How can we **efficiently** construct benchmarks to measure the performance of **any algorithm**?
 - Use **circuit mirroring** on **subcircuits** of an application circuit

SUBCIRCUIT SELECTION

- Randomly sample subcircuits of different shapes from application circuit
- Obtain fidelities using a scalable performance metric
 - We use mirror circuit fidelity estimation (MCFE)



METHODS



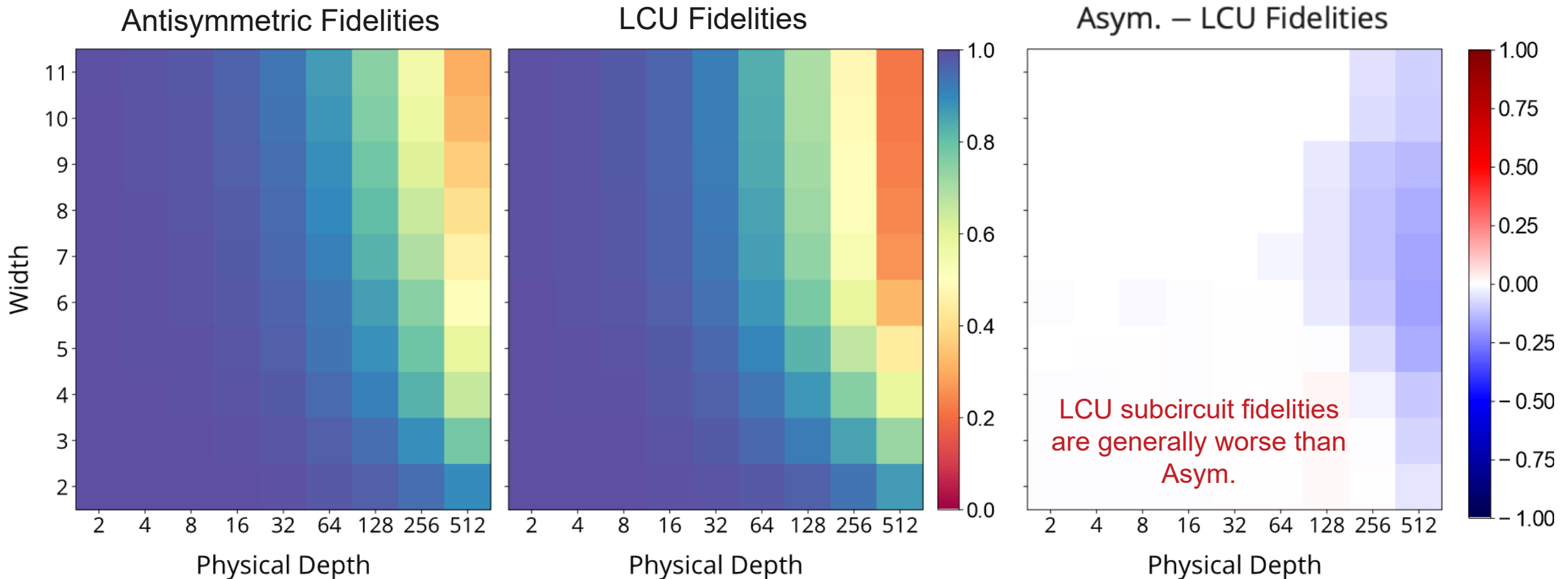
- Tested on two different applications
 - Antisymmetrization routine for first-quantized molecular state preparation
 - Uses a bitonic sorting network
 - Block-encoded Hamiltonian for second-quantized state preparation
 - Uses linear combination of unitaries (LCU)
- Used noisy numerical simulations to obtain subcircuit and full circuit fidelities with MCFE
 - 300 subcircuits per width-depth pair
 - MCFE Parameters: 100 mirroring samples per subcircuit, 100 reference experiments per width
 - Widths from 2 to 11 qubits
 - Depths from 2 to 512 (exponentially increasing)
 - Noise model with 1- and 2-qubit Hamiltonian and stochastic error generators
 - Hamiltonian generator strengths of 10^{-3} and 10^{-2} for 1- and 2-qubits
 - Stochastic generator strengths of 4×10^{-4} and 4×10^{-3} for 1- and 2-qubits

VOLUMETRIC BENCHMARKS



- Volumetric benchmarking (VB) plots our benchmark is **sensitive to differing performance** across algorithms

Geometric Mean of Subcircuit Fidelities by Width and Depth



PREDICTING FULL CIRCUIT FIDELITIES



- Our benchmark is sensitive to differences in performance
- How can we quantify these differences concisely?
 - Use an **error rate model**
 - Want to use benchmarking results to estimate error rate ε that allows us to predict full circuit fidelity

Compute average fidelity of (w, d)
subcircuits

$$F_{wd} = \text{GeomMean}\{\underbrace{F_{wd,1}, \dots, F_{wd,n}}_{\text{Individual subcircuit fidelities with width } w \text{ and depth } d}\}$$

Individual subcircuit
fidelities with width
 w and depth d

Calculate error rate from
average fidelity

$$\longrightarrow \varepsilon_{wd} = 1 - (F_{wd})^{1/wd}$$

Assume fidelity loss occurs
equally at each circuit location

Estimate full circuit fidelity
from error rate

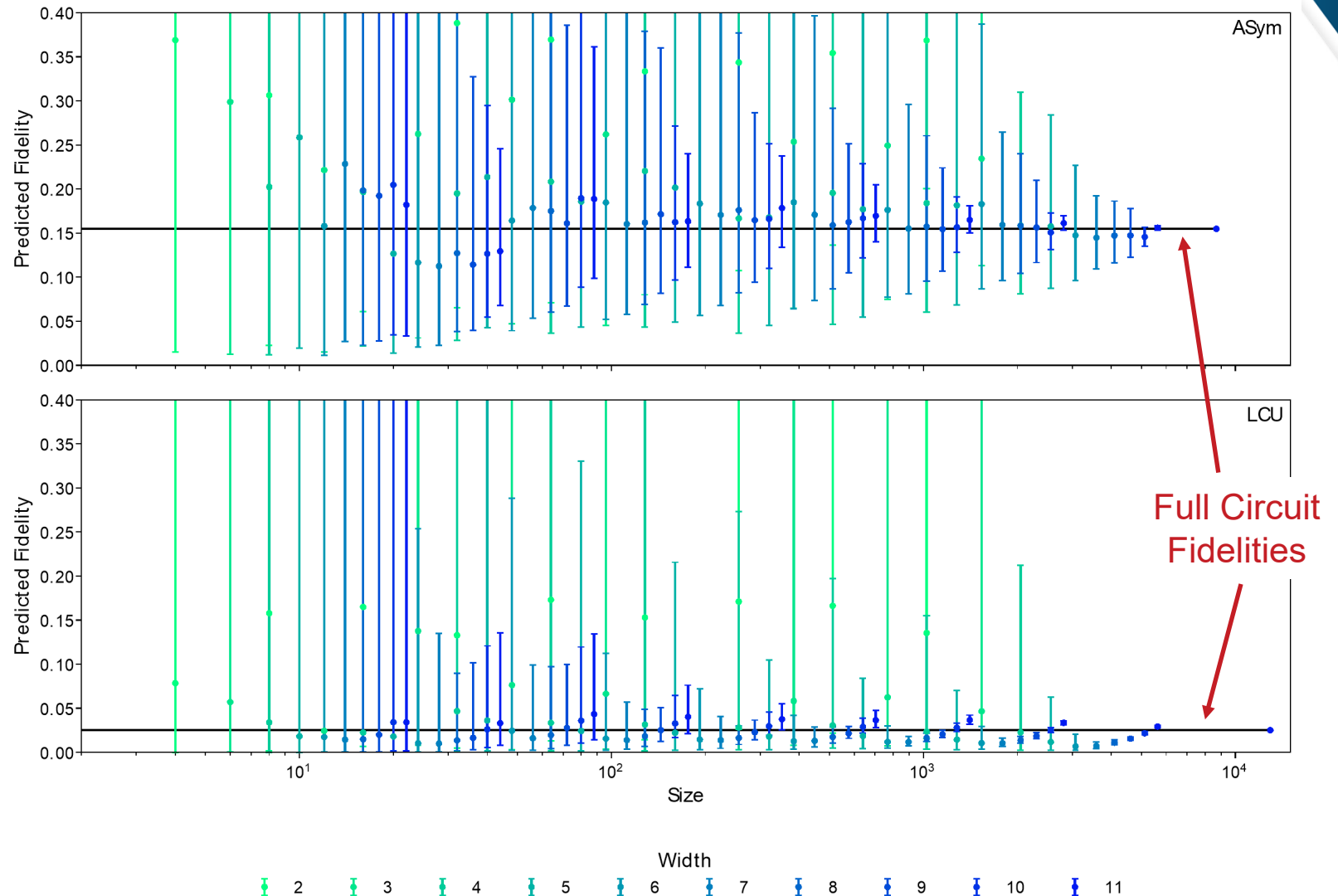
$$\longrightarrow F_{\text{est},wd} = (1 - \varepsilon_{wd})^{WD}$$

Full circuit has width W
and depth D

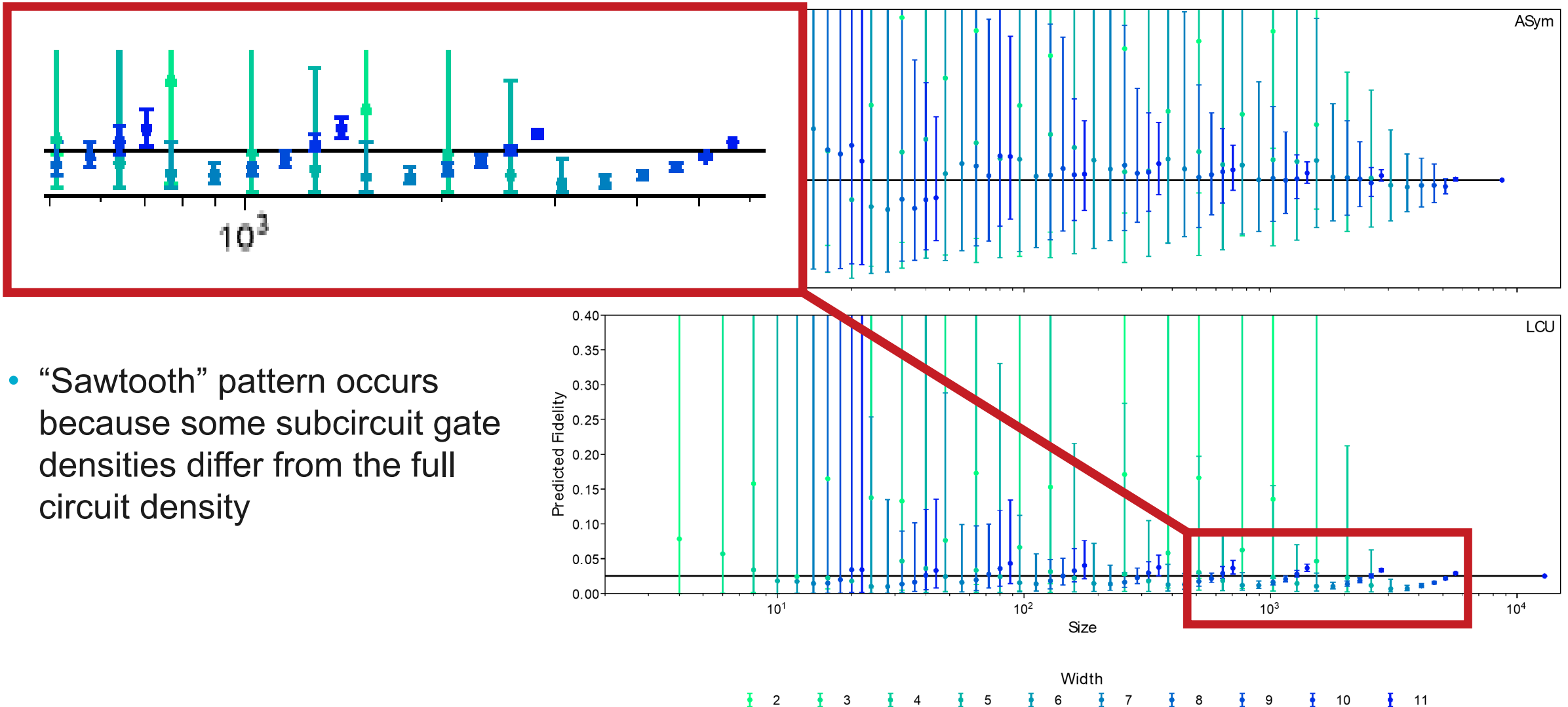
PREDICTING FULL CIRCUIT FIDELITIES



- Low width subcircuits are poor predictors of performance
 - So are 1Q/2Q error rates
- High width subcircuits are generally good predictors of full circuit fidelities, even at lower depths



PREDICTING FULL CIRCUIT FIDELITIES



PREDICTING FULL CIRCUIT FIDELITIES (REVISITED)



- To account for the difference in density between subcircuits and the full circuit, adjust the subcircuit fidelities using:

$$\tilde{F}_{wd,i} = (F_{wd,i})^{\xi / \xi_{wd,i}}$$

“density-adjusted fidelity”

ξ = full circuit density

$\xi_{wd,i}$ = density of (w, d) subcircuit i

Compute average fidelity of (w, d) subcircuits

$$\tilde{F}_{wd} = \text{GeomMean}\{\underbrace{\tilde{F}_{wd,1}, \dots, \tilde{F}_{wd,n}}\}$$

Individual subcircuit fidelities with width w and depth d

Calculate error rate from average fidelity

$$\tilde{\epsilon}_{wd} = 1 - \left(\tilde{F}_{wd}\right)^{1/wd}$$

Assume fidelity loss occurs equally at each circuit location

Estimate full circuit fidelity from error rate

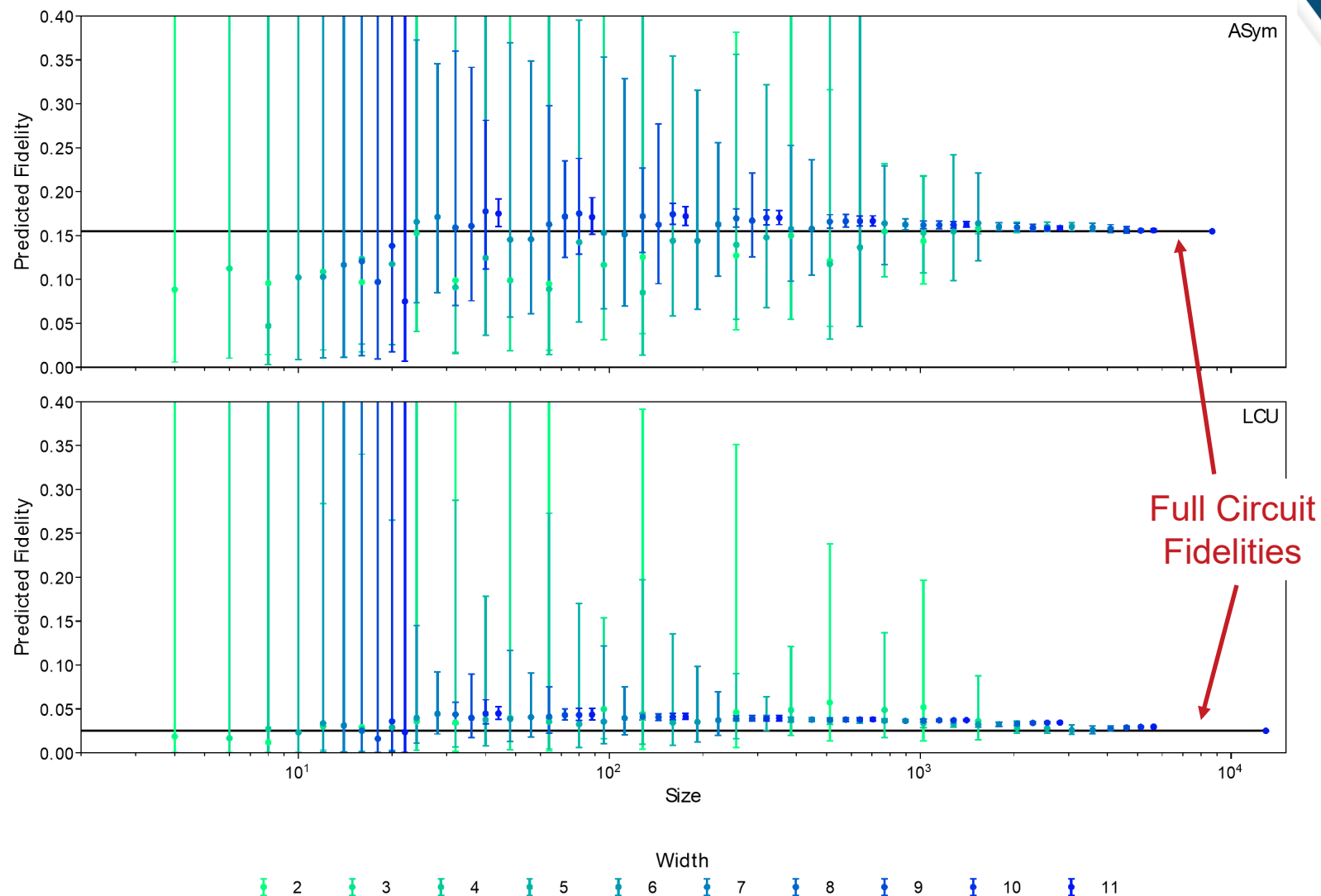
$$\tilde{F}_{\text{est},wd} = (1 - \tilde{\epsilon}_{wd})^{WD}$$

Full circuit has width W and depth D

PREDICTING FULL CIRCUIT FIDELITIES (DENSITY-ADJUSTED)



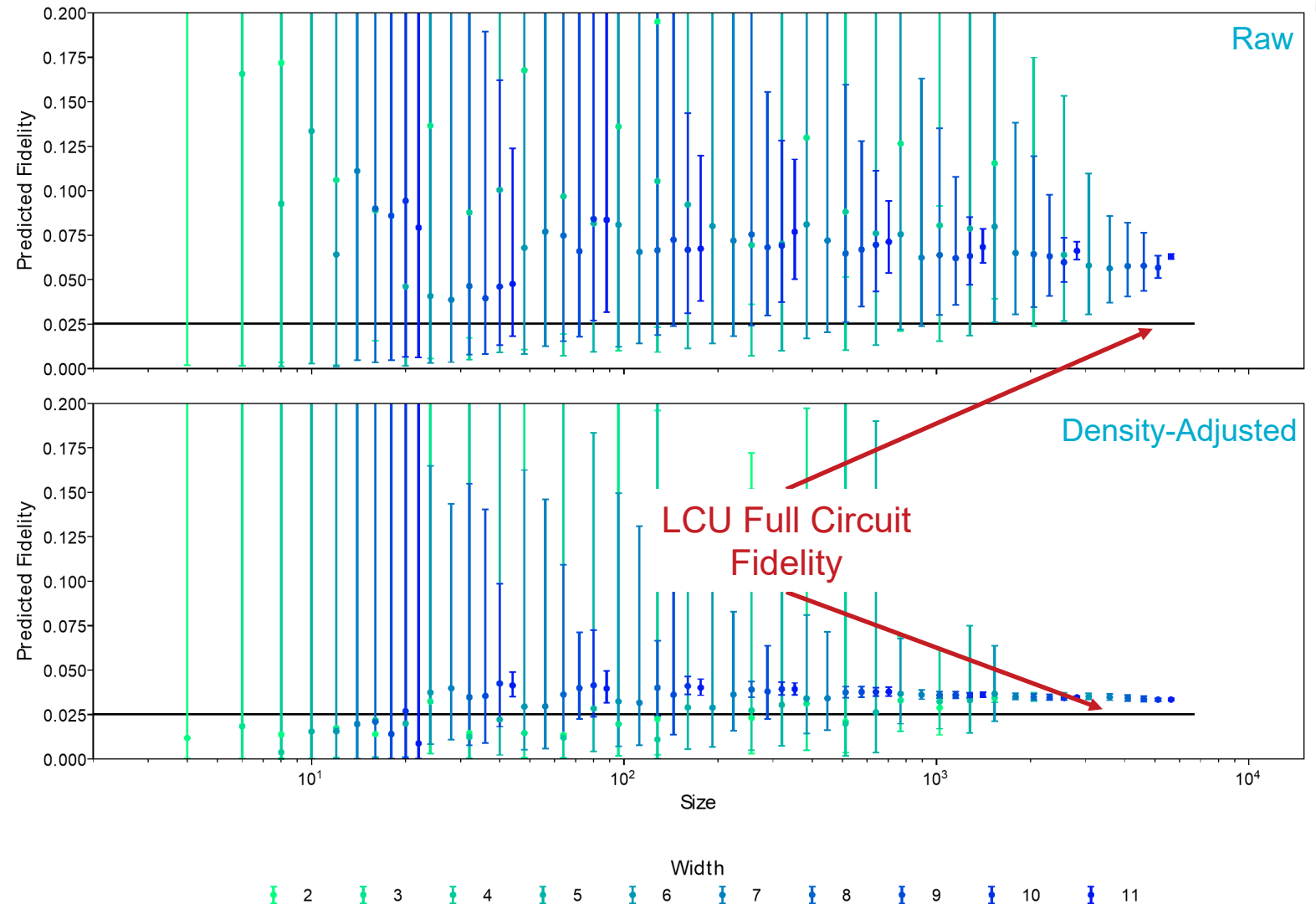
- Accounting for subcircuit density yields more consistent convergence to full circuit fidelity than before



CROSS-PREDICTING CIRCUIT FIDELITIES FAILS



- If we account for circuit width, depth, and density, can we predict the LCU full circuit fidelity from the ASym benchmarking data?
 - No; subcircuit benchmarking data reveals **structural** differences in performance



CONCLUSIONS



- Application-specific benchmarks using subcircuits can detect differences in performance across applications due to **structural** differences in the circuits
 - Volumetric benchmarks show performance differences in subcircuits
- Using error rate models, can succinctly quantify algorithmic circuit performance using subcircuit benchmarking data
 - Performance differs even after accounting for width, depth, and density
 - Predicting performance of one algorithmic circuit from benchmarking data for another does not work