

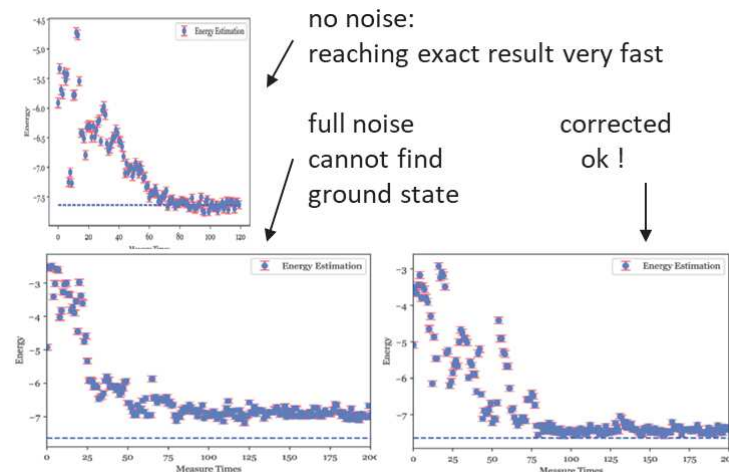
Methods for reliable Quantum Computing Calculations

Increase the Reliability for Quantum Computing Calculations

$$H = \sum_{i=1}^N \beta [\sigma_x(i)\sigma_x(i+1) + \sigma_y(i)\sigma_y(i+1) + \sigma_z(i)\sigma_z(i+1)] + J\sigma_z(i)$$

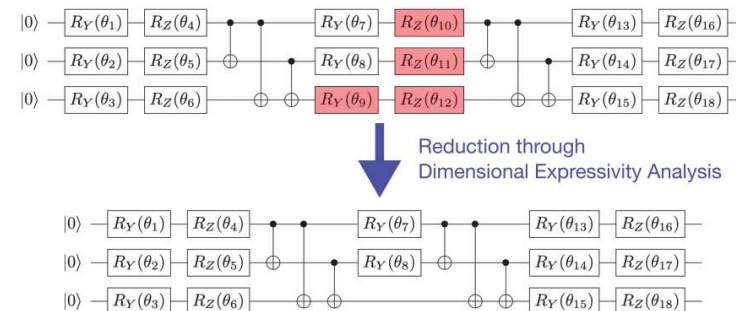
➤ Example for Error Mitigation in VQS

- Model in Condensed Matter Physics:
1-Dimensional Heisenberg model
very sensitive to QC noise
cured by own developed error mitigation methods



➤ Optimize Dimensional Expressivity of a Quantum Gate Circuit

- Gate Operations are noisy
- Develop methods for Dimensional Expressivity Analysis
- **Generate as many/complicated states as possible with fewest number of gates**



Software engineering for Noise Model Benchmarks



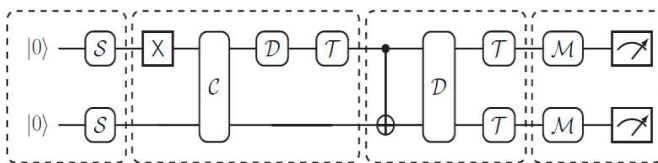
Increase the Reliability for Quantum Computing Calculations



➤ Software Engineering Models for Error Mitigation

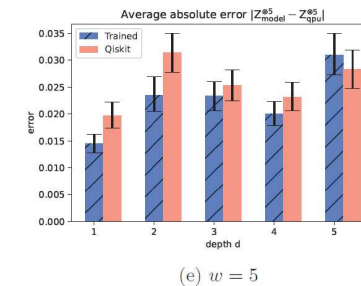
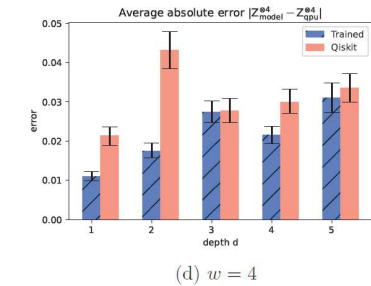
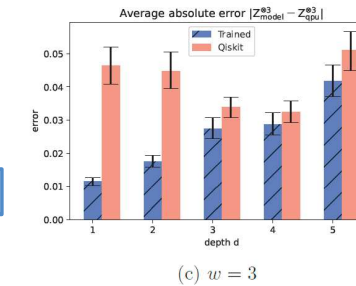
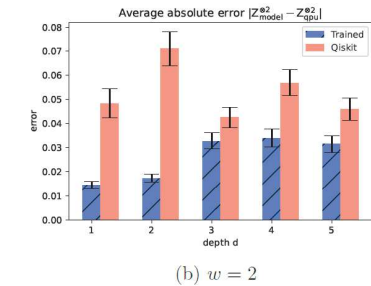
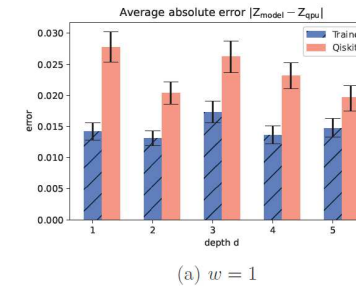
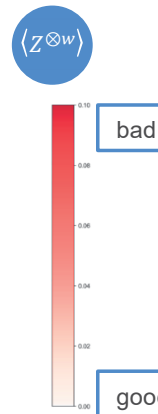
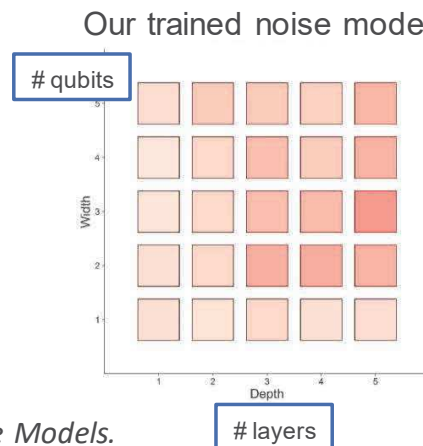
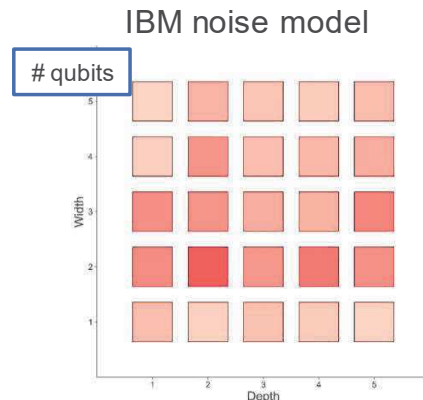
- Systematic approach to train error models with Machine Learning and perform benchmarks for quantum computing applications

symbol	error	parameters	number of parameters
\mathcal{S}	state preparation	$p_{sp}(q)$	N
\mathcal{D}	depolarization	$\lambda_g(q)$	$4N - 1$
\mathcal{C}	crosstalk	$\phi_g(q)$	$2N$
\mathcal{T}	thermal relaxation	$T_{1,2}(q)$	$2N$
\mathcal{M}	measurement	$p_{0 \rightarrow 1}(q), p_{1 \rightarrow 0}(q)$	$2N$
total			$11N - 1$



Weber, T. et al.

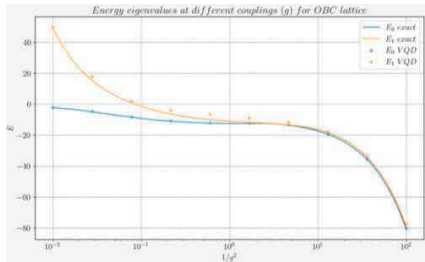
Volumetric Benchmarking of Quantum Computing Noise Models. Submitted to IOP Quantum Science and Technology.



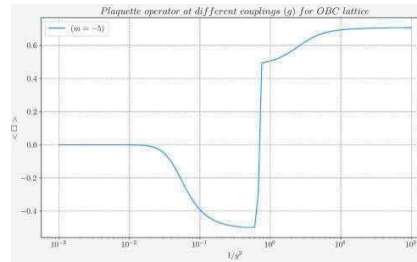
Quantum Computing: From Theory towards Applications

From QED in 2+1 dimensions to Flight Gate Assignments

➤ Variational Quantum Simulations (VQS) for QED



Particle Mass
 $\Delta = E_1 - E_0$
 → physical quantity

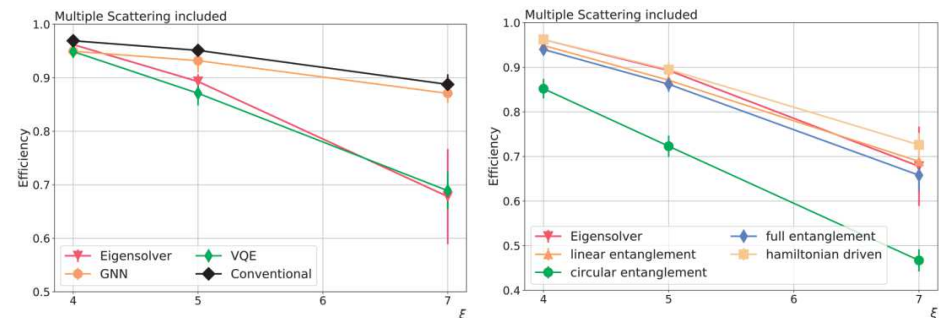
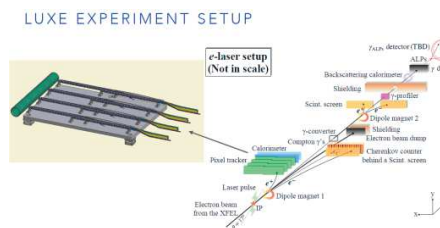


Detecting a phase transition at negative mass
 → not possible with MC methods

Clemente G. et al, Strategies for the Determination of the Running Coupling of (2+1)-dimensional QED with Quantum Computing, <https://arxiv.org/abs/2206.12454>

➤ Very similar methods for tracking at the LUXE Experiment

Q-GNN and VQE for particle tracking in the LUXE Experiment (Laser Und XFEL Experiment) study of the influence of entanglement



Efficiency as a function of the field intensity ξ

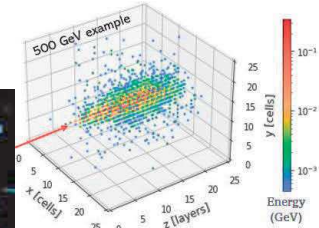
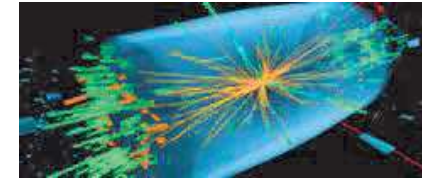
Crippa A. et al, Quantum algorithms for charged particle track reconstruction in the LUXE experiment, <https://arxiv.org/abs/2304.01690>

Methods and Quantum Machine Learning for Experiment

Early examples in Experimental Particle Physics

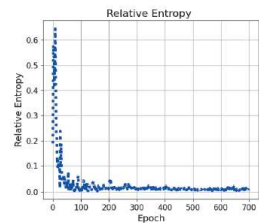
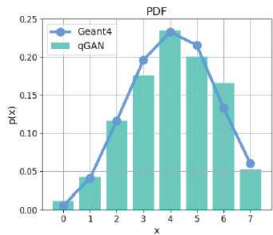
➤ Quantum Machine Learning proved to be robust against noise

- High Luminosity LHC needs vast amount of simulations with 200 pile-up events → a formidable computing challenge



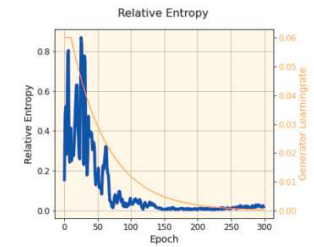
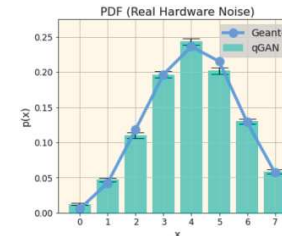
➤ Develop methods and machine learning applications for QC

- Q-GAN simulations for detectors (Quantum Generative Adversarial Network)

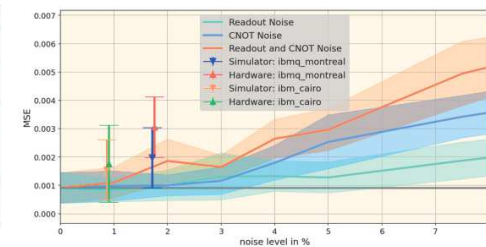
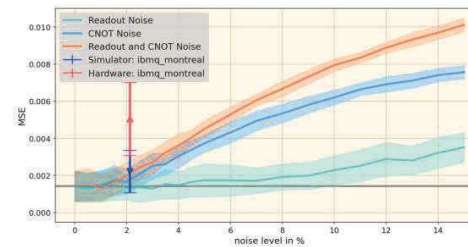
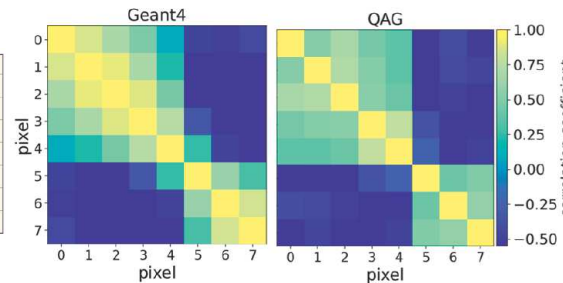
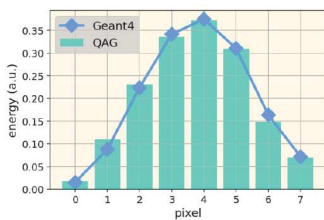
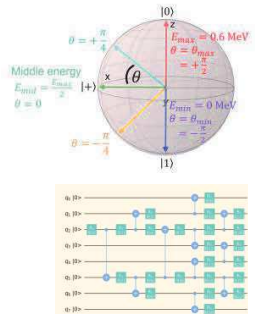


← readout noise

full noise
real hardware →



- Full Quantum Angle Generator → modelling correlations and robust against noise



Rehm, F. et al. Precise Image Generation on Current Noisy Quantum Computing Devices. Subm. to IOP Quantum Science and Technology / PhD Thesis RWTH Aachen

DESY. QUANTUM and CQTA in Zeuthen

Special focus on Quantum Computing Applications

- Innovation-Funding from Brandenburg:
15 M€ over 5 years for CQTA
 - access to quantum computer → **new IBM – Hub**
 - develop applications and use cases
 - enable quantum simulations
 - benchmark and verify emerging hardware platforms
 - provide training in quantum computing
 - make new generation "quantum ready"

Already very vibrant and intense connections to industry

Brandenburg Quantentechnology-Network meets in Zeuthen
Research-State-Secretary Dünow visits researchers

