Quantum Machine Learning Integration in the High Energy Physics Pipeline: CERN QTI perspective

BiOLEE CENTER, MILAN 2024

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How does CERN engage in Quantum Technologies?

Develop technologies required by the CERN scientific programme

Integrate CERN to future quantum infrastructures

Extend and share technologies available at CERN

Boost development and adoption of QT beyond CERN

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CERN QTI Phase 2

Launched January 2024

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- o Develop quantum sensors to provide new capabilities for particle physics research (dark matter search, axion search, gravitational wave detection…)
- o Focus areas: Superconducting RF cavities, hydrogen-like Rydberg ions, and Transition Edge Sensors

CERN QUANTUM **TECHNOLOGY** PLATFORMS

CERN's broad expertise and experimental facilities in many areas (superconducting materials, magnets, radiation effects, cryogenics, controls etc.) could be useful to support your developments.

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HYBRID QUANTUM COMPUTING AND ALGORITHMS

- Integration in the EU and US HPC+QCS infrastructures
- Development of hybrid classic+quantum algorithms for theoretical and experimental physics
- Lead the development of common libraries of quantum algorithms and tools for HEP and other sciences
- Simulation of high dimensional classical / quantum systems
- *Software stacks for quantum devices calibration and control systems*
- *Investigations of distributed quantum computing, resource optimisation, green computing*

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Why Quantum Computing for HEP?

Fundamental motivation

Utilise information and correlations inherent in HEP data.

Exploit "quantum remnants" in data.

Quantum Machine Learning (QML)

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Agliardi, Grossi, Pellen, Prati "**Quantum integration of elementary particle processes**." <https://doi.org/10.1016/j.physletb.2022.137228>

Form wat the namber of call to the algorithm, required to approximate
I can be reduced almost quadratically beyond the MC classical bound. With **QAE** the number of call to the algorithm, required to approximate

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Agliardi, Grossi, Pellen, Prati "**Quantum integration of elementary particle processes**." <https://doi.org/10.1016/j.physletb.2022.137228>

- IQAE: demonstrated speed up *(Grinko, Gacon, Zoufal, Woerner [npj QI 7, 52 \(2021\)\)](https://www.nature.com/articles/s41534-021-00379-1)*
- QGAN: potential bottleneck for data/function upload
- *Difficult to run on real HW*

 $\label{eq:reduced} \begin{split} \mathcal{I} &= -\frac{1}{4}F_{\mu\nu}F^{\mu\nu} \\ &+i\overline{\psi}B\Psi+h.c. \\ &+ \psi_i y_{ij}\psi_j\Phi+h.c. \\ &+ |\mathbb{D}_\mu\Phi|^2 - \mathbb{V}(\Phi) \end{split}$

- IQAE: demonstrated speed up *(Grinko, Gacon, Zoufal, Woerner [npj QI 7, 52 \(2021\)\)](https://www.nature.com/articles/s41534-021-00379-1)*
- Integrate trigonometric functions
- QNN encoding into Fourier series
- **QFIAE** applicable to n-D functions
- Good result (1% error) on HW

Agliardi, Grossi, Pellen, Prati "**Quantum integration of elementary particle processes**." <https://doi.org/10.1016/j.physletb.2022.137228>

Jorge J. Martinez de Lejarza, Michele Grossi, Leandro Cieri and German Rodrigo: arXiv: 2305.01686

- distinguish (C) and compute (R) the scattering • Build **a quantum supervised model** that can amplitude squared for related Feynman diagrams LO QED process
- Topology encoded in the adjacency matrix of the graph
- Particles (m, Q, S) encoded in the edges
- $\frac{1}{2}$ Time flow (initial state, interaction vertex, final state) encoded in the vertices

$$
f_{\rm{max}}
$$

 $\mathcal{L} = -\frac{1}{4}F_{\mu\nu}F^{\mu\nu}$

+ $(\overline{\psi} \mathbb{B} \psi + h.c.$
+ $\psi_i \mathbb{Y}_i \psi_j \phi + h.c.$
+ $|\mathbb{D}_\mu \Phi|^2$ - $\mathbb{V}(\Phi)$

0.0000

 0.5

Successful training:

- Is able to learn several diagrams at the same time
- Can learn diagrams with same topology but different particles
- Task difficult with classic approaches

 1.0

 1.5

 2.0

Scattering Angle (rad)

loop-channel predictions

 3.0

 0.5

 1.0

 1.5

 2.0

Scattering Angle (rad)

ground truth

 2.5

 2.5

 3.0

F.Rehm et al., **Precise image generation on current noisy quantum computing devices,** *Quantum Sci. Technol.* **9** 015009

 (b) (c) Geant4 QAG 1.00 1.00 0.75 0.75 -0.50 0.50 $\frac{1}{2}$ $\frac{1}{4}$ pixel
4 -0.25 č 0.25 $0.00 -$ Ŧ 0.00 $-0.25\,$ ලි $-0.25\,\overline{5}$ -0.50 **150** $3 \quad 4 \quad 5 \quad 6 \quad 7$ $0 \quad 1 \quad 2$ \mathbf{O} $1 \quad 2$ 3 4 5 6 7 pixel pixel

a value of 1 indicates a perfect positive correlation.

Data **Generation**

- Quantum angle generator (QAG): a full quantum machine learning model designed to generate accurate images on current quantum devices
- Reproduces average values, AND, complex pixel-wise correlations

Theory **Are we using the right data?Where is NEW PHYSICS?**

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Theory **Are we using the right data? Where is NEW PHYSICS?**

Data Analysis

Subleading Jet

 p_{T2} Jet 1, pt: 70.0 Ge Leading Jet p_{T1}

let 0, pt: 205.1 GeV

acquisition *starts gathering information about nature* Re-embracing the scientific method:

… *our baseline is the SM (from 1970!)* → *let's change the approach*

Rather than specifying a signal hypothesis upfront, we could start looking at our data

signal hypothesis Based on what we see (e.g., clustering alike objects) we could formulate a

extraction → *QCD dijet events*

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T_{rel} **Quantum Anomaly Detection**

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Quantum Anomaly Detection

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QC research directions in HEP

PRX QUANTUM 5, 037001 (2024)

Ouantum Computing for High-Energy Physics: State of the Art and Challenges

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- To go beyond the hype we need **concrete challenges**
	- What are the most promising applications?
	- How to **define performance metrics** and validate results?
- **Experimental data has high dimensionality**
	- Can we train **Quantum Machine Learning** algorithms effectively?
	- Can we reduce **the impact of data reduction techniques**?
- Experimental data is shaped by physics laws
	- Can we leverage them to build better algorithms?

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Methods and applications

Phase Detection with Anomaly Detection

 2.0

 1.5

 -1.0

 $0.5¹$

 $0.0_{0.0}$

 0.25

- Quantum equivalent of an Autoencoder to learn an effective unitary operation capable of compressing all the information in the Pauli-Z expectation values of a subset of the qubits

- Minimization of the loss function
- *All anomaly detection models were trained to compress the point (κ, h) = (0, 0) of the Hamiltonian*
- *Training: single state selected to achieve compression*
- *Cost is assigned to compressed state allowing the outline of all phases*

Compression:

 $|\psi\rangle \rightarrow |\phi\rangle \otimes |0\rangle^k$

FIG. 13: Compression Scores C of the AD circuits trained on the $(\kappa, h) = (0, 0)$ point of the ANNNI model phase diagram at different system sizes $N: 6$ (left), 12 (middle), and 18 (right). The scores are showcased as a function of the interaction strength ratio $(\kappa = -J_2/J_1)$ and the external magnetic field $(h = B/J_1)$. Lower compression scores indicate better disentanglement of trash qubits from others, as defined by eq. $\sqrt{2}$.

Exploring the Phase Diagram of the quantum one-dimensional ANNNI model <https://arxiv.org/abs/2402.11022>

QT4HEP 2025 - save the date

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