

# Quantum Computing for Data Analysis in High-Energy Physics



Andrea Delgado (she/her/ella)

On behalf of the whitepaper team

*arXiv:2203.08805*

Visit [Workshop Website!](#)

Recordings of talks available

# 2021 SNOWMASS WORKSHOP ON QUANTUM COMPUTING FOR HIGH-ENERGY PHYSICS

DECEMBER 1-3, 2021

Virtual Event

## QUANTUM MACHINE LEARNING - DATA ANALYSIS - SIMULATION

This workshop aims to bring HEP scientists currently working on or interested in quantum computing applications to HEP and showcase state-of-the-art algorithms for data analysis, and simulation applications.

**Call for abstracts open until  
November 17th at the Indico website.**  
<https://indico.phy.ornl.gov/event/144/>

### Questions? Contact the Workshop Chairs:

Andrea Delgado (delgadoa@ornl.gov)

Jean-Roch Vlimant (jvlimant@caltech.edu)

### Local Organizing Committee

Marcel Demarteau

Raphael Pooser

Kathleen Hamilton

Alex McCaskey





# Quantum Computing for High-Energy Physics

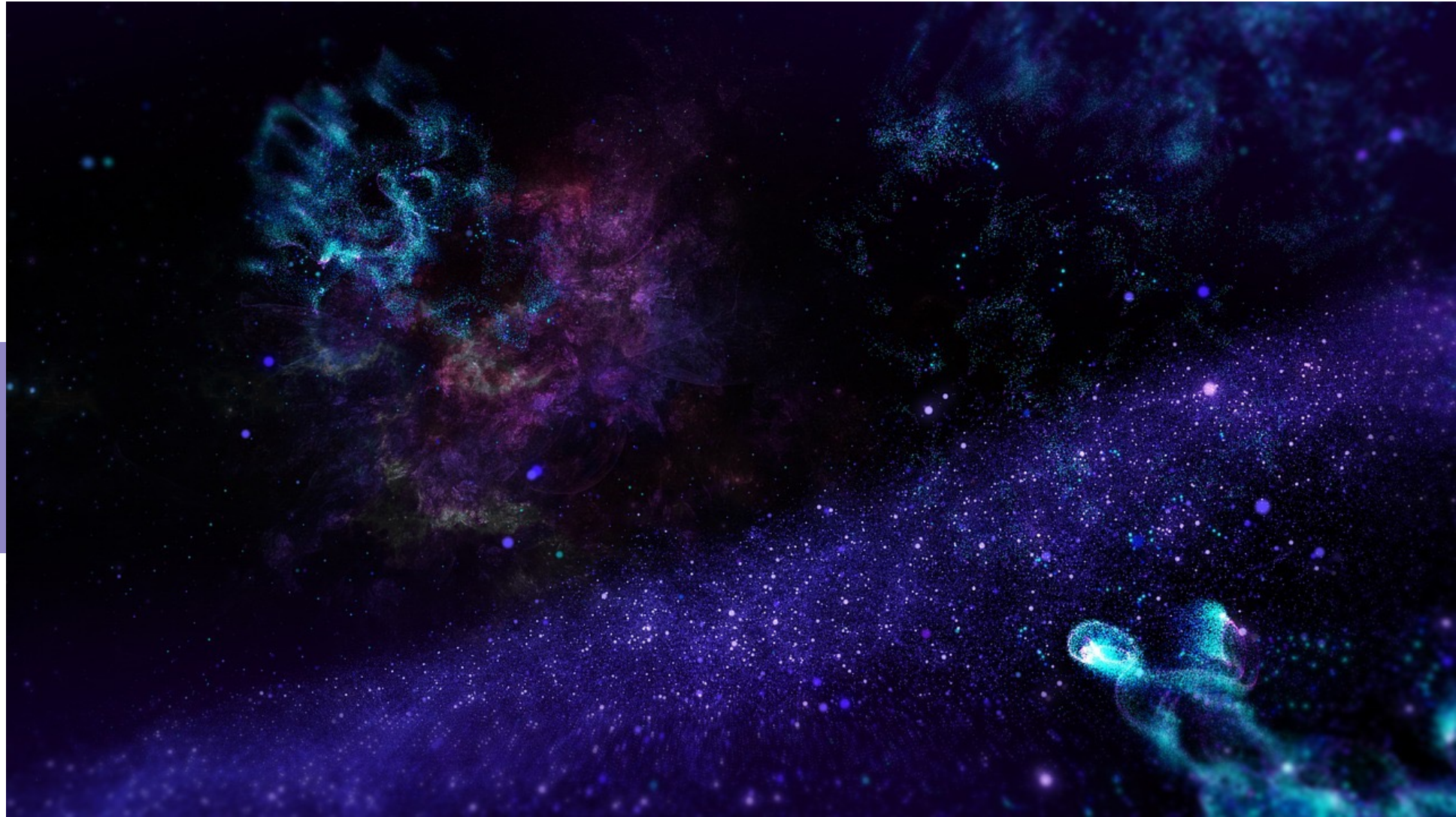
## Hilbert space is a big place!

*Carlton Caves*

With 275 qubits, we can represent more basis/computational states than the number of atoms in the observable universe.

$$2^{275}$$

The potential to speed up certain calculations, simulate quantum mechanical systems through the manipulation of quantum mechanical properties such as entanglement.





# Quantum Computing for High-Energy Physics

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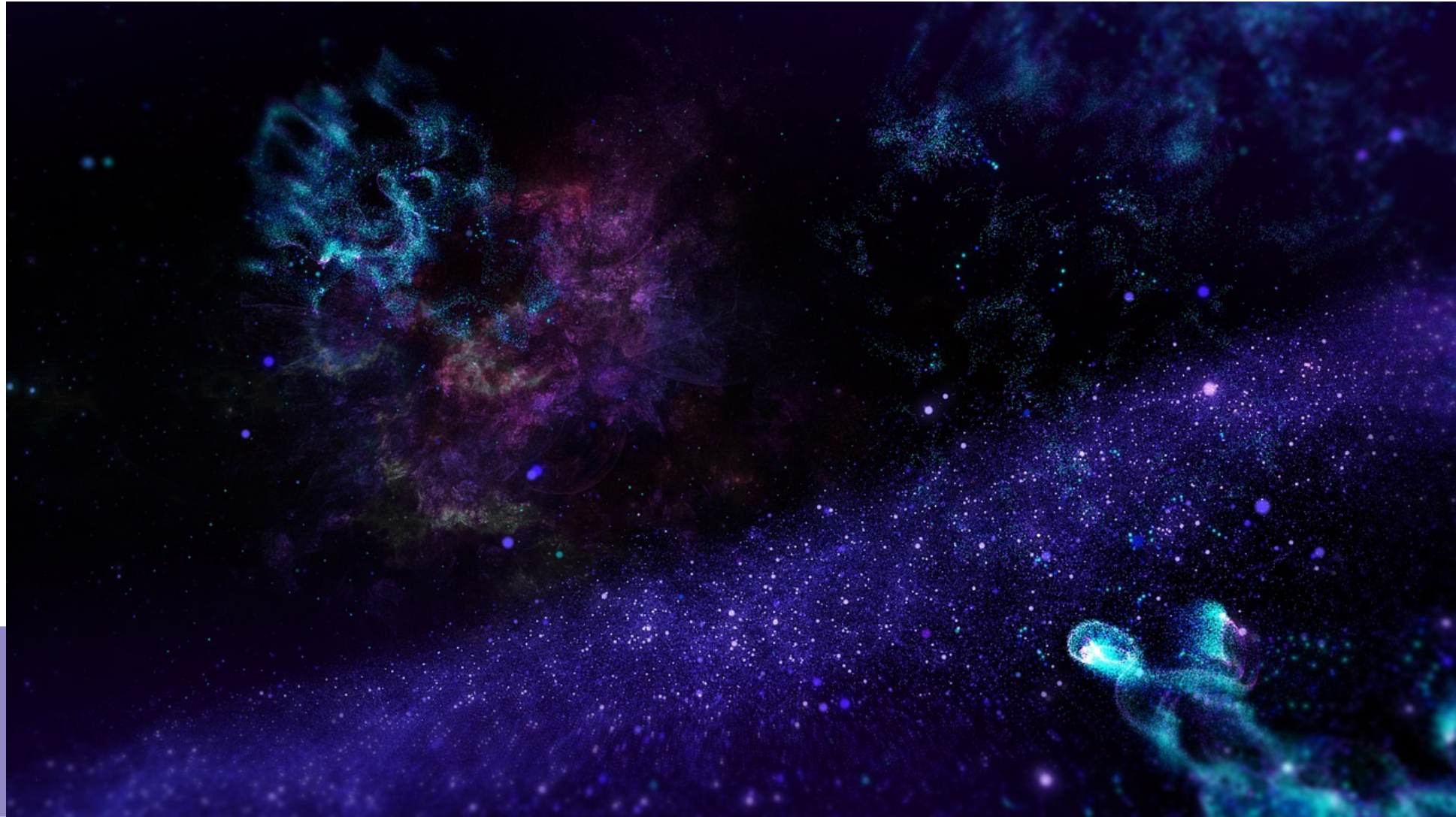
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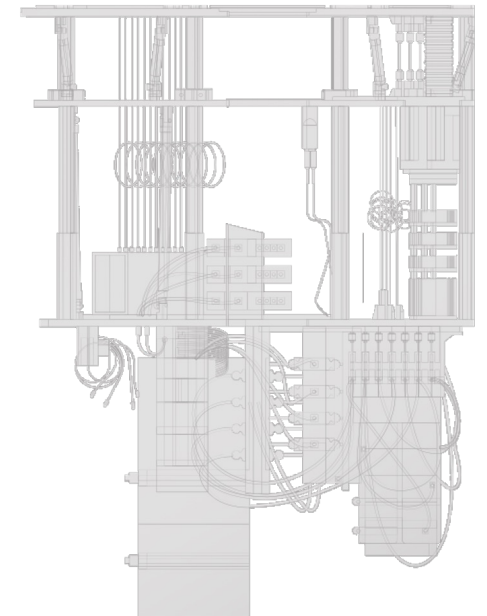
The potential to speed up certain calculations, simulate quantum mechanical systems through the manipulation of quantum mechanical properties such as entanglement.

Not as easy as it sounds!



# Quantum Computing in the NISQ Era

- ★ Motivated by access to **cloud-based** processors and commercial applications.
- ★ Applications in **Quantum Machine Learning (QML)** spurred by the release of Xanadu's PennyLane / Google's Tensorflow.
- ★ **Co-design**:
  - *Algorithmic development/research is adapting to match the pace of hardware development.*
- ★ Applications in quantum annealing, superconducting qubits, and continuous-variable explored.
- ★ Developed for deployment on **NISQ** devices.
  - Few qubits,
  - Noisy,
  - Low gate fidelity.
- ★ Hybrid frameworks to leverage benefits of both classical and quantum computing - **variational quantum circuits**.



# Data Analysis in High-Energy Physics

Standard Model  
(and beyond)

$$\begin{aligned}\mathcal{L} = & -\frac{1}{4}F^{\mu\nu}F_{\mu\nu} \\ & + i\bar{\psi}\not{D}\psi \\ & + \bar{\psi}_i Y_{ij} \psi_j \phi + \text{h.c.} \\ & + |D_\mu \phi|^2 - V(\phi)\end{aligned}$$



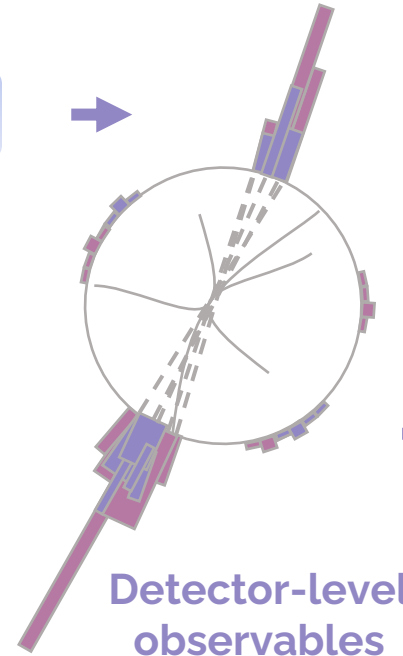
Simulation



Nature



Experiment



Detector-level  
observables

Analysis





# Data Analysis in High-Energy Physics

See Christian Bauer's talk on "Quantum Simulation"

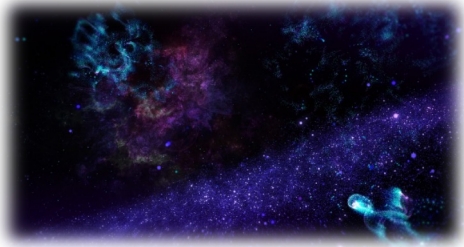
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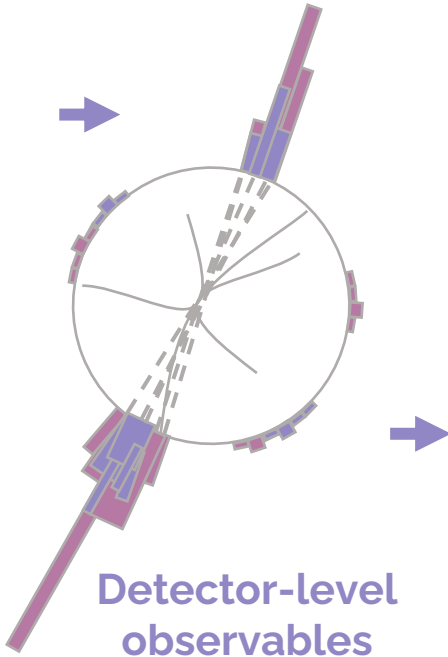
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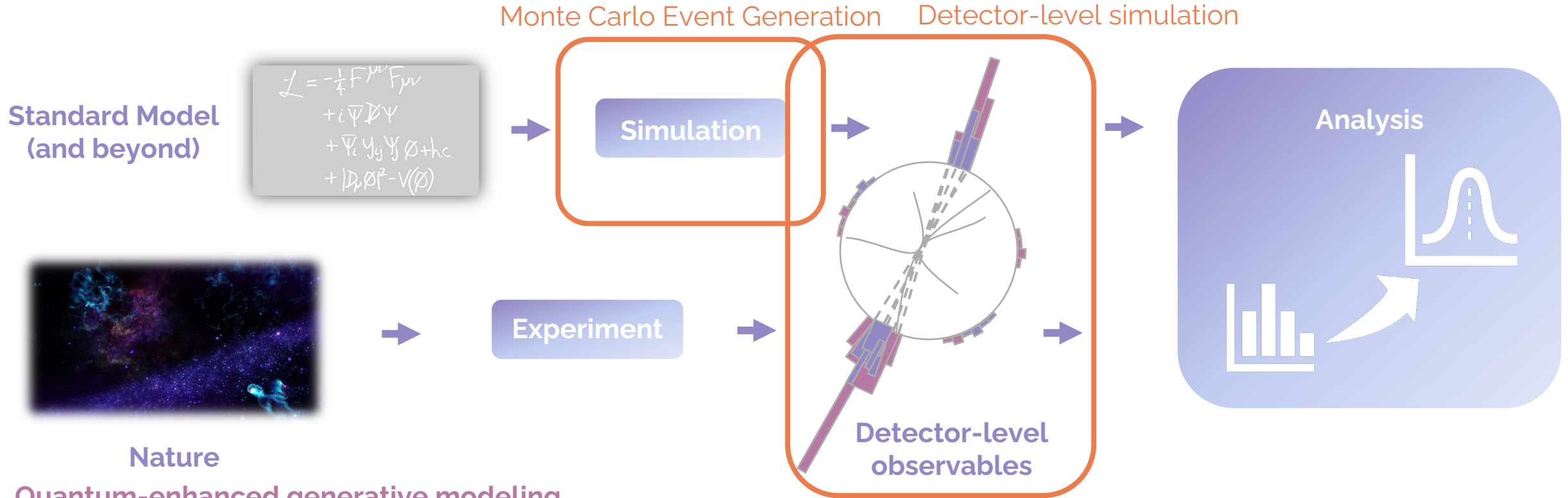
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Nature



# Data Analysis in High-Energy Physics



## Quantum-enhanced generative modeling

- GANS and QCBMs.
- Studied in the context of data augmentation.
- Challenges associated with the encoding of target distributions and scalable quantum error correction techniques.
- Important for building up the “quantum pipeline” – in anomaly detection settings, intermediary encoders, etc.

“Style-based quantum generative adversarial networks for Monte Carlo events”, Bravo-Prieto, Baglio, Cè, Francis, Grabowska, Carrazza, [arXiv:2110.06933](#)

“Quantum integration of elementary particle processes”, Agliardi, Grossi, Pellen, Prati, [DOI:10.1016/j.physletb.2022.137228](#)

“Unsupervised Quantum Circuit Learning in High-Energy Physics”, Delgado, Hamilton [arXiv:2203.03578](#)

“Conditional Born machine for Monte Carlo events generation”, Kiss, Grossi, Kajomovitz, Vallecorsa [arXiv:2205.07674](#)



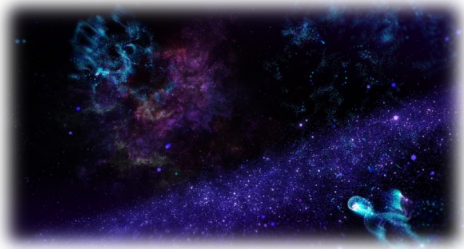
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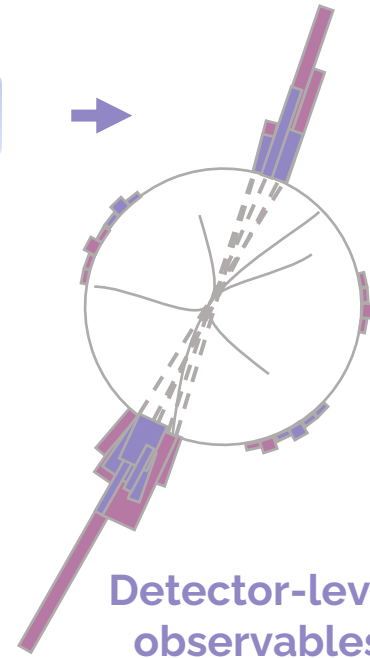
Simulation



Nature



Experiment



Detector-level  
observables

Track reconstruction, jet  
clustering, classification.

Analysis



## Casting Data Analysis Tasks as Optimization Problems

- Reformulating clustering tasks as binary unconstrained satisfaction problems (QUBO).
- Motivated by access to D-Wave QPUs.
- Large number of available qubits, relatively easy to program.

Zlokapa, A., Anand, A., Vlimant, J.R. *et al.* **Charged particle tracking with quantum annealing optimization.** *Quantum Mach. Intell.* **3**, 27 (2021).

Quiroz, G., Ice, L., Delgado, A., Humble, T.S., **Particle track classification using quantum associative memory**, Nuclear Instruments and Methods in Physics Research Section A, Volume 1010 (2021).

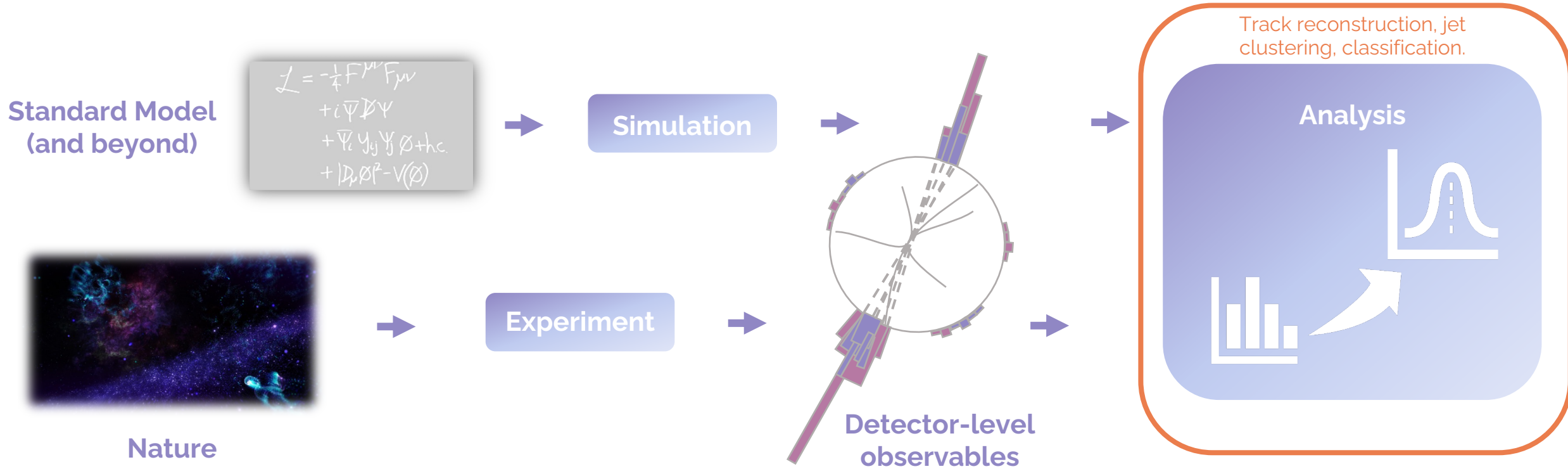
Wei, A. Y., Naik, P., Harrow, A. W., Thaler, J., **Quantum Algorithms for Jet Clustering** Phys. Rev. D101:094015 (2020)

Delgado, A., Thaler, J., **Quantum Annealing for Jet Clustering with Thrust** arXiv: 2205.02814 (2022).

Pires, D., Omar, Y., Seixas, J. **Adiabatic Quantum Algorithm for Multijet Clustering in High Energy Physics** arXiv:2012.14514 (2020).

Anschuetz, E., Funcke, L., et al., **Degeneracy Engineering for Classical and Quantum Annealing: A Case Study of Sparse Linear Regression in Collider Physics**, arXiv:2205.10375 (2022).

# Data Analysis in High-Energy Physics



## Variational Quantum Circuits as Machine Learning Models

- Parameterized circuit training for classification tasks such as event classification, track reconstruction.
- Advanced techniques such as data-reuploading, layer wise training.
- Deployed on IBM hardware, IonQ, CV platforms.

Terashi, K., Kaneda, M., Kishimoto, T., et al., **Event classification with quantum machine learning in high-energy physics**, Computing and Software for Big Science, 5(1) (2021).

Belis, V., Gonzalez-Castillo, S., Reisel, C., Vallecorsa, S., et al., **Higgs analysis with quantum classifiers**, EPJ Web of Conferences, 251:03070 (2021).

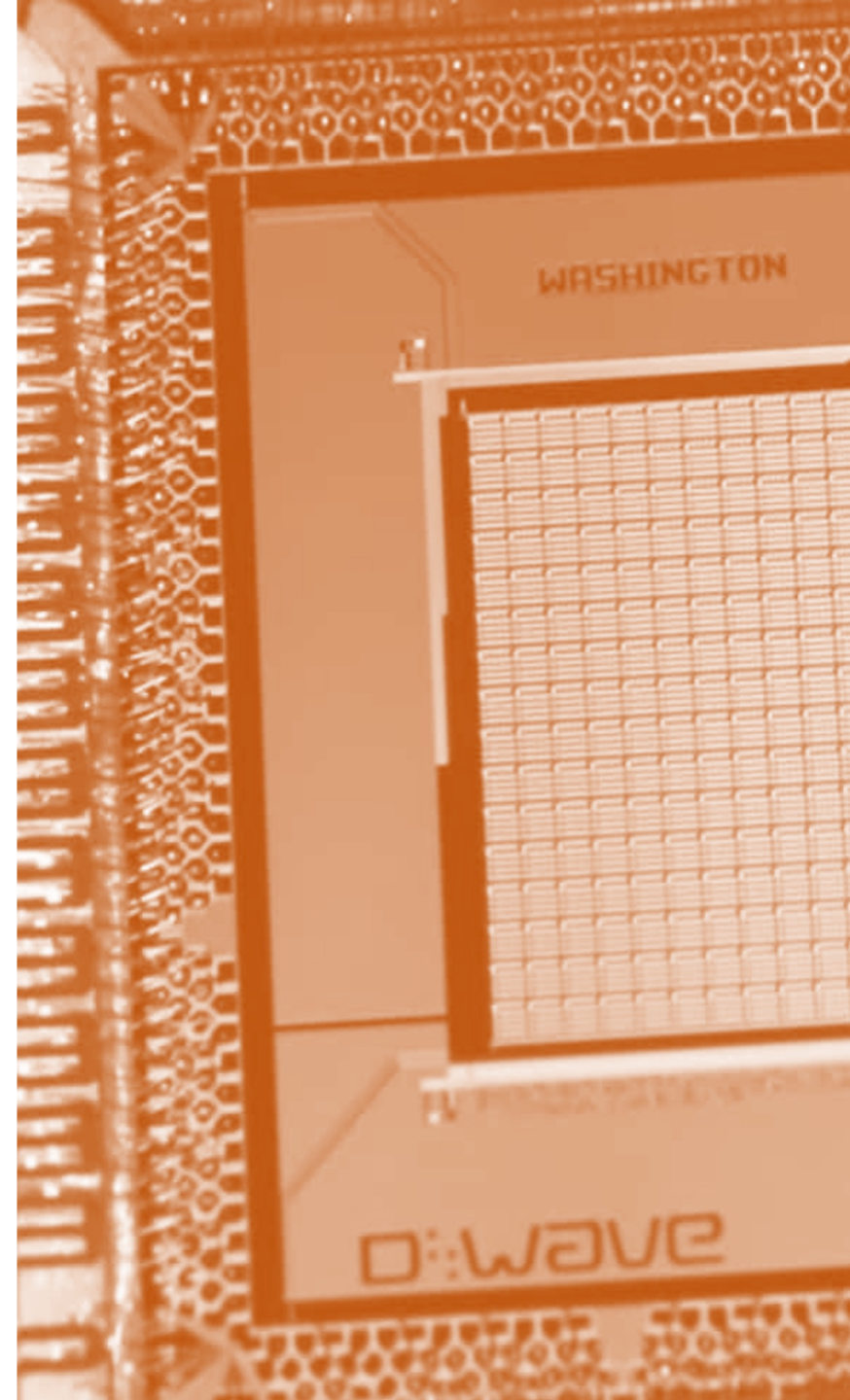
Peters, E., Caldeira, J., Ho, A., et al., **Machine learning of high-dimensional data on a noisy quantum processor**, npj Quantum Information, 7(1) (2021).

Gianelle, A., Koppenburg, P., Lucchesi, D., et al., **Quantum machine learning for b-jet identification**, arXiv:2202.13943 (2022)

# Challenges

## Quantum Annealing

- Large overhead associated with calculation of QUBO coefficients.
- Low connectivity among physical variables.
- Significant post-processing required, hyperparameter optimization.
- Problem decomposition tools lacking precision needed for HEP analysis.





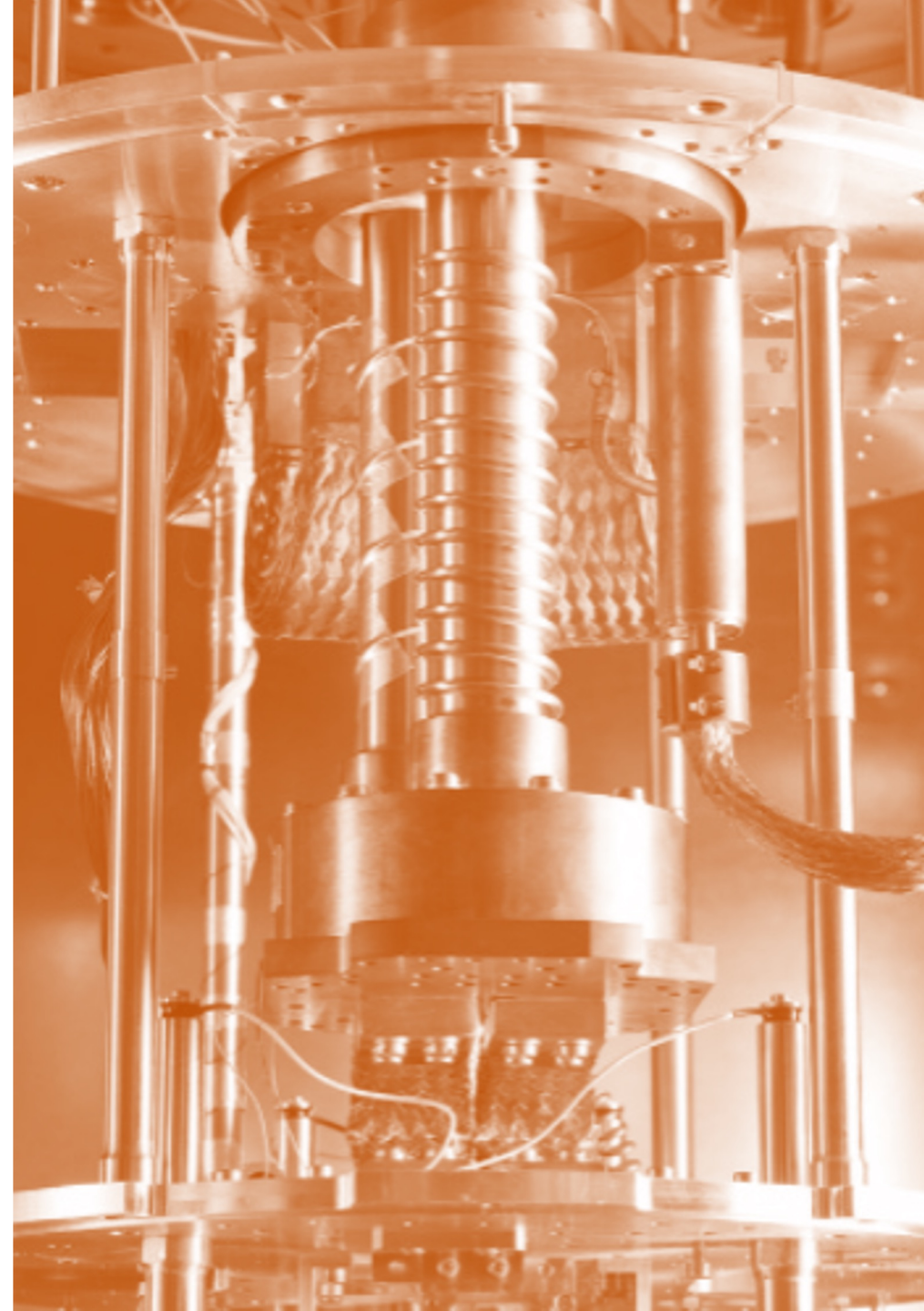
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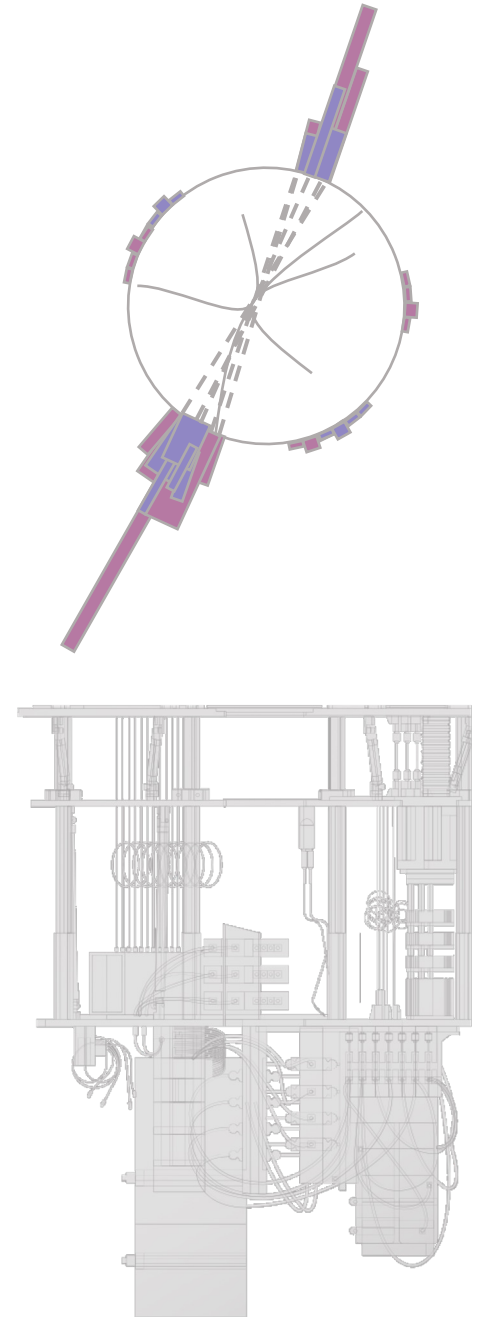
## Gate-based Quantum Computing

- Limited number of qubits, low connectivity.
- High overhead in embedding classical data into quantum states.
- Training limitations associated with cloud-based access.
- Scalable error correction techniques.



# Outlook

- ★ A large number of quantum computing applications to data analysis in HEP have been developed over the last decade.
  - First applications on **D-Wave**, followed by circuit-based applications targeted for **IBM** hardware, mostly.
  - CV continues largely unexplored.
  - Huge development in the area of QML: largest kernel-based classifier (27 qubits) and generative model (12 qubits) to-date developed for HEP applications.



# Outlook

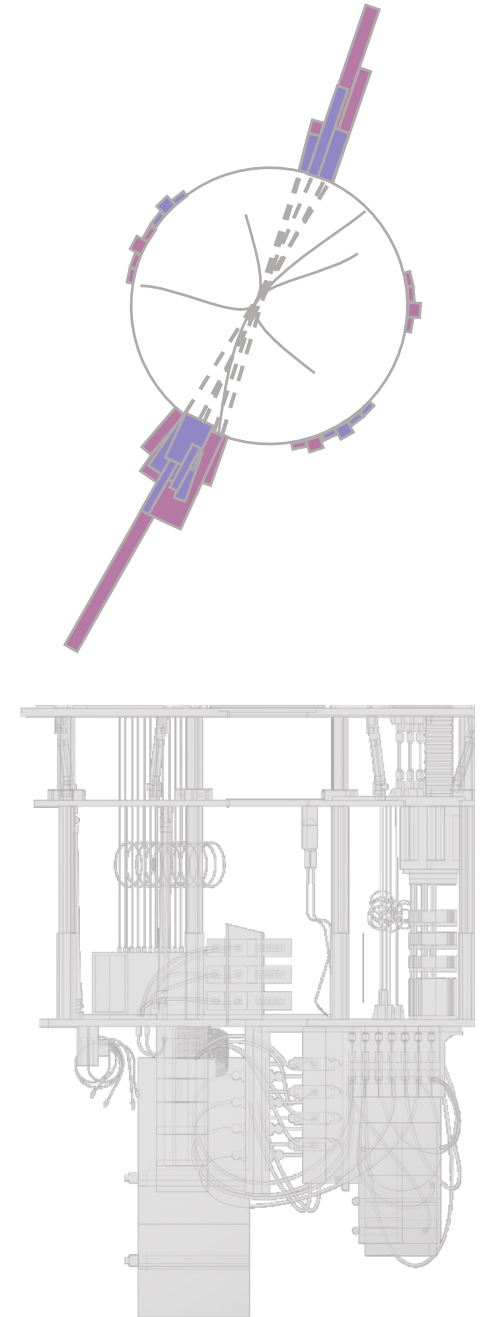
- ★ A large number of quantum computing applications to data analysis in HEP have been developed over the last decade.
- ★ **Quantum Machine Learning will have an important role in quantum computing for HEP in the next decade.**
  - Most likely not revolutionary in the analysis of classical data.
  - But impactful in the analysis of quantum data, complementing quantum-enhanced searches for BSM physics and simulation of quantum systems.





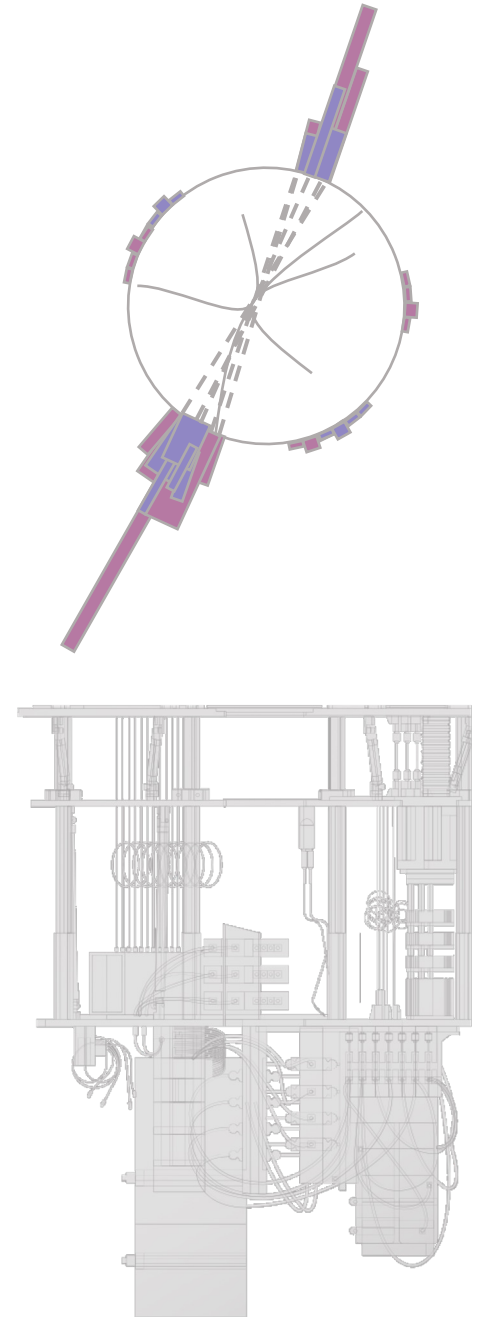
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- ★ A large number of quantum computing applications to data analysis in HEP have been developed over the last decade.
- ★ Quantum Machine Learning will have an important role in quantum computing for HEP in the next decade.
- ★ **There is still a lot of work to do.**
  - Understand the role of entanglement in model trainability,
  - Scalable error correction/mitigation techniques.
  - Understanding choice of Ansatz in model expressibility.
  - How to make better use of available Hilbert space, in terms of efficient measurements, metrics.



# Outlook

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- ★ Quantum Machine Learning will have an important role in quantum computing for HEP in the next decade.
- ★ There is still a lot of work to do.
- ★ **Not covered on this talk:**
  - ★ Quantum-inspired models: Tensor networks as ML models.
  - ★ Storing an IceCube event into quantum memory - See [Jeffrey Lazar's talk](#) at the workshop or read the [whitepaper](#) 😊



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# Thank you!