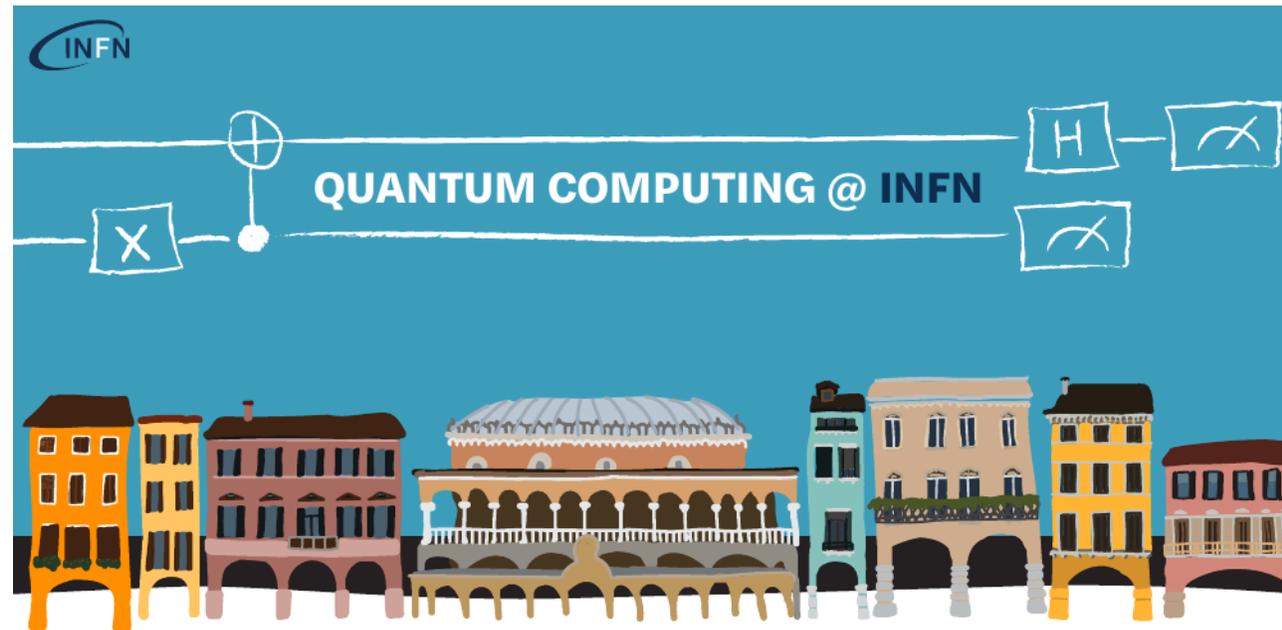


# Quantum Computing in experimental HEP

(quant-ph, hep-ph, hep-ex)



***Michele Grossi, PhD***

*Hybrid Quantum Computing Infrastructures and Algorithms Coordinator*



QUANTUM  
TECHNOLOGY  
INITIATIVE

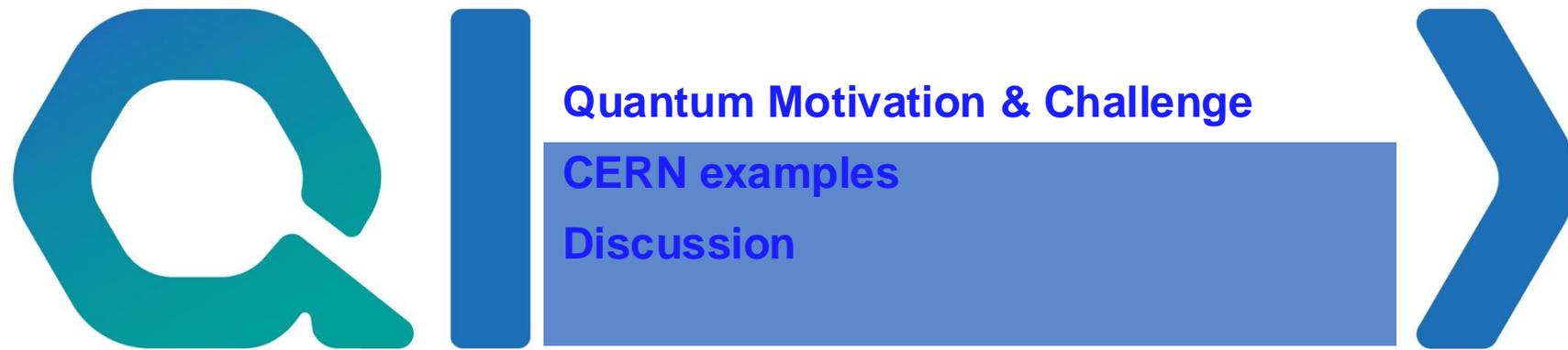


QUANTUM  
TECHNOLOGY  
INITIATIVE



**Quantum Motivation & Challenge**  
**CERN examples**  
**Discussion**





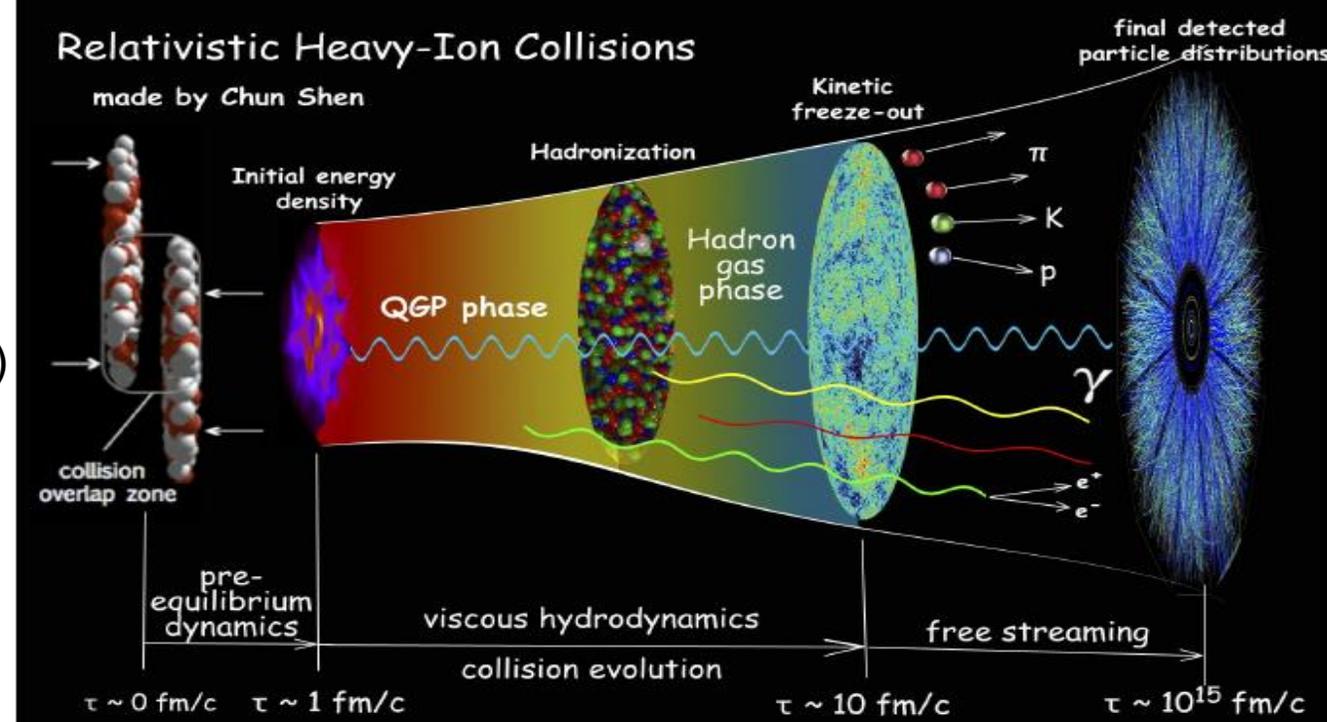
# QI

**Quantum Motivation & Challenge**  
**CERN examples**  
**Discussion**

# Motivation

## Theoretical challenge

- Non-zero chemical potential (QCD phase diagram)
- Real time dynamics
  - heavy ion collisions, scattering quenches



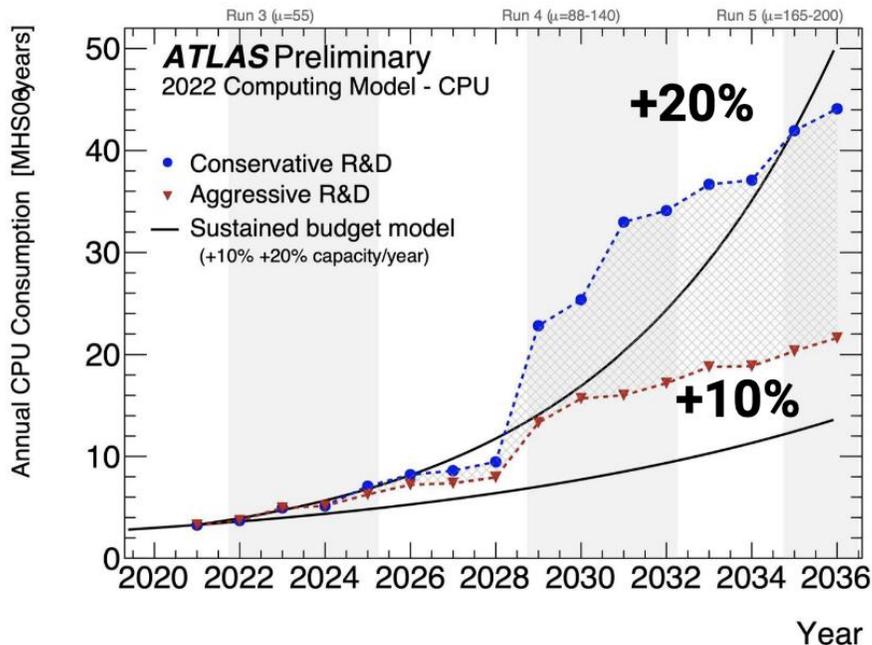
From: [10.1051/epjconf/20159700025](https://doi.org/10.1051/epjconf/20159700025)

Cartoon of the time evolution of an ultra-relativistic heavy-ion collision

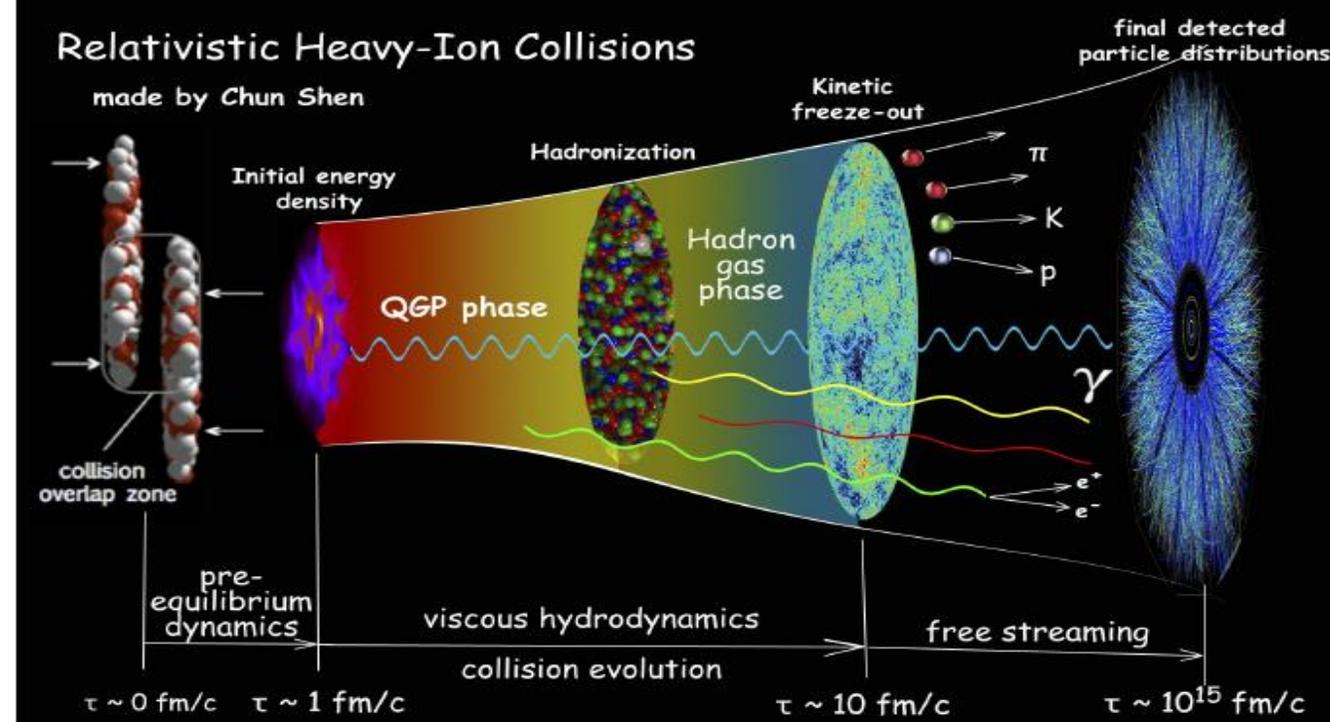
# Motivation

## Theoretical challenge

- Non-zero chemical potential
  - Real time dynamics
- heavy ion collisions, scattering quenches



From HL-LHC Projections - ATLAS Software and Computing HL-LHC Roadmap



From: [10.1051/epjconf/20159700025](https://doi.org/10.1051/epjconf/20159700025)

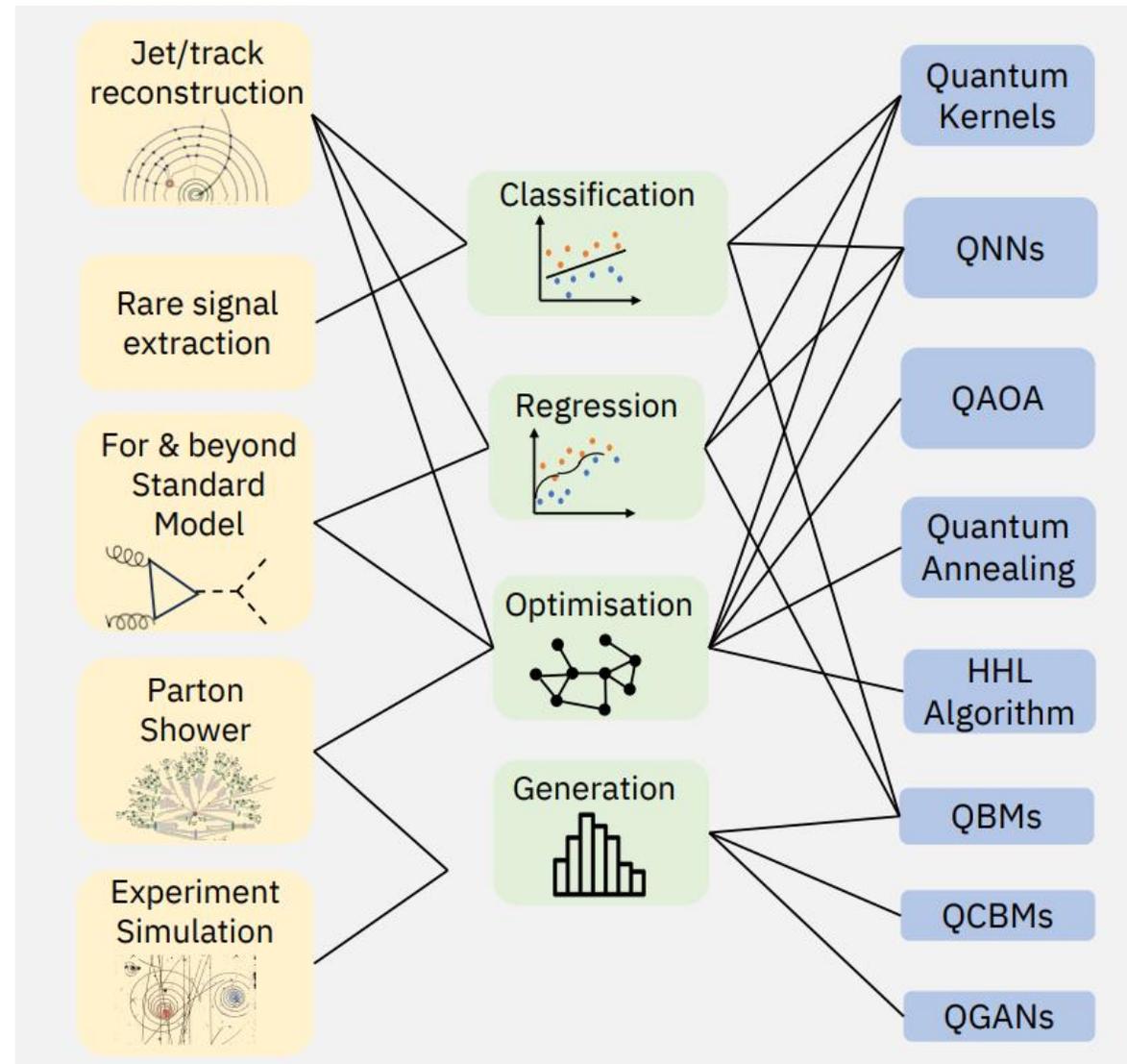
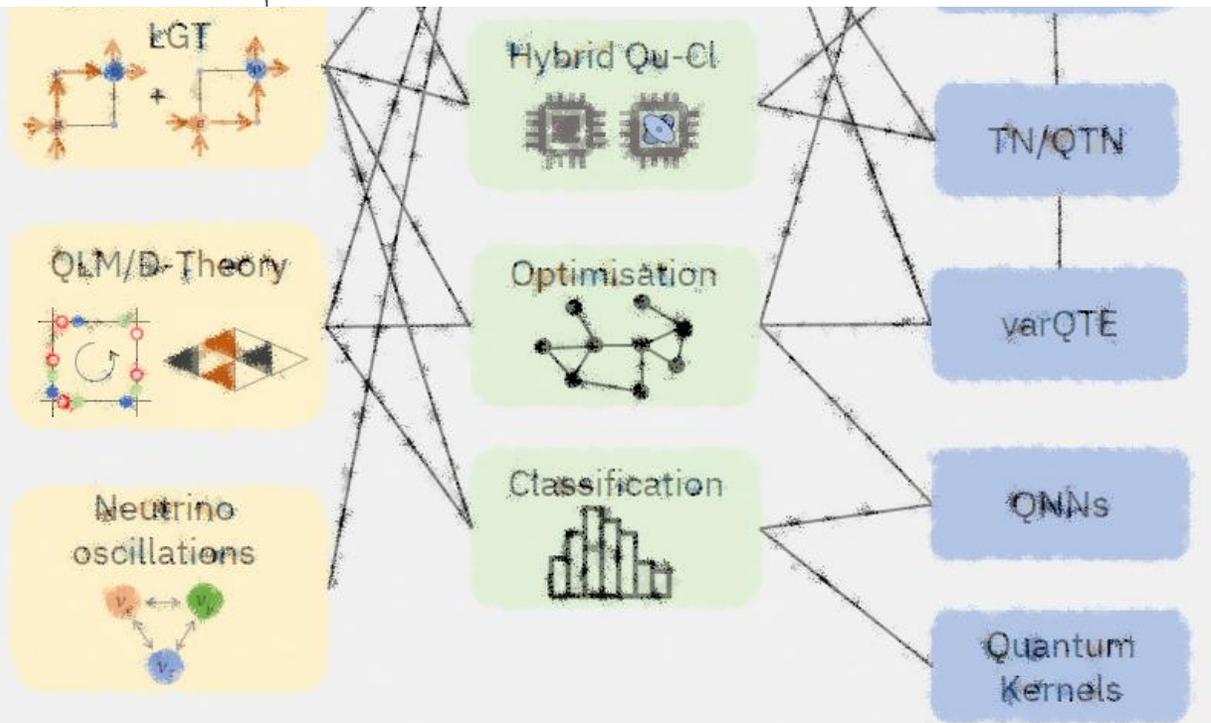
Cartoon of the time evolution of an ultra-relativistic heavy-ion collision

## Computing challenge for High-Lumi LHC

- Simulation and analysis
- need: new technology, algorithms and methods

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

Alberto Di Meglio<sup>1,\*</sup>, Karl Jansen<sup>2,3,†</sup>, Ivano Tavernelli<sup>4,‡</sup>, Constantia Alexandrou<sup>3,5</sup>, Srinivasan Arunachalam<sup>6</sup>, Christian W. Bauer<sup>7</sup>, Kerstin Borras<sup>8,9</sup>, Stefano Carrazza<sup>1,10</sup>, Arianna Crippa<sup>2,11</sup>, Vincent Croft<sup>12</sup>, Roland de Putter<sup>6</sup>, Andrea Delgado<sup>13</sup>, Vedran Dunjko<sup>12</sup>, Daniel J. Egger<sup>4</sup>, Elias Fernández-Combarro<sup>14</sup>, Elina Fuchs<sup>1,15,16</sup>, Lena Funcke<sup>17</sup>, Daniel González-Cuadra<sup>18,19</sup>, Michele Grossi<sup>1</sup>, Jad C. Halimeh<sup>20,21</sup>, Zoë Holmes<sup>22</sup>, Stefan Kühn<sup>2</sup>, Denis Lacroix<sup>23</sup>, Randy Lewis<sup>24</sup>, Donatella Lucchesi<sup>1,25</sup>, Miriam Lucio Martinez<sup>26,27</sup>, Federico Meloni<sup>8</sup>, Antonio Mezzacapo<sup>6</sup>, Simone Montangero<sup>1,25</sup>, Lento Nagano<sup>28</sup>, Vincent R. Pascuzzi<sup>6</sup>, Voica Radescu<sup>29</sup>, Enrique Rico Ortega<sup>30,31,32,33</sup>, Alessandro Roggero<sup>34,35</sup>, Julian Schuhmacher<sup>4</sup>, Joao Seixas<sup>36,37,38</sup>, Pietro Silvi<sup>1,25</sup>, Panagiotis Spentzouris<sup>39</sup>, Francesco Tacchino<sup>4</sup>, Kristan Temme<sup>6</sup>, Koji Terashi<sup>28</sup>, Jordi Tura<sup>12,40</sup>, Cenk Tüysüz<sup>2,11</sup>, Sofia Vallecorsa<sup>1</sup>, Uwe-Jens Wiese<sup>41</sup>, Shinjae Yoo<sup>42</sup> and Jinglei Zhang<sup>43,44</sup>



A group of business professionals in a meeting, looking at a tablet. The text 'Are there other indications towards a quantum approach?' is overlaid in white. The background shows people in suits, some holding coffee cups, and a tablet displaying a document with charts and text.

**Are there other indications  
towards a quantum approach?**

**Observation of quantum entanglement in top-quark pairs using the ATLAS detector**  
 ATLAS Collaboration

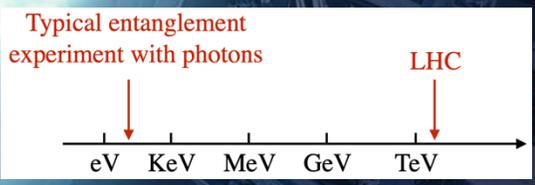
**Abstract:** Entanglement is an intrinsic property of quantum mechanics and is predicted to be exhibited in the particles produced at the Large Hadron Collider. A measurement of the extent of entanglement in top quark-antiquark ( $t\bar{t}$ ) events produced in proton-proton collisions at a center-of-mass energy of 13 TeV is performed with the data recorded by the CMS experiment at the CERN LHC in 2016, and corresponding to an integrated luminosity of  $36.3 \text{ fb}^{-1}$ . The events are selected based on the presence of two leptons with opposite charges and high transverse momentum. An entanglement-sensitive observable  $D$  is derived from the top quark spin-dependent parts of the  $t\bar{t}$  production density matrix and measured in the region of the  $t\bar{t}$  production threshold. Values of  $D < -1/3$  are evidence of entanglement and  $D$  is observed (expected) to be  $-0.480^{+0.026}_{-0.029}$  ( $-0.467^{+0.026}_{-0.029}$ ) at the parton level. With an observed significance of 5.1 standard deviations with respect to the non-entangled hypothesis, this provides observation of quantum mechanical entanglement within  $t\bar{t}$  pairs in this phase space. This measurement provides a new probe of quantum mechanics at the highest energies ever produced.

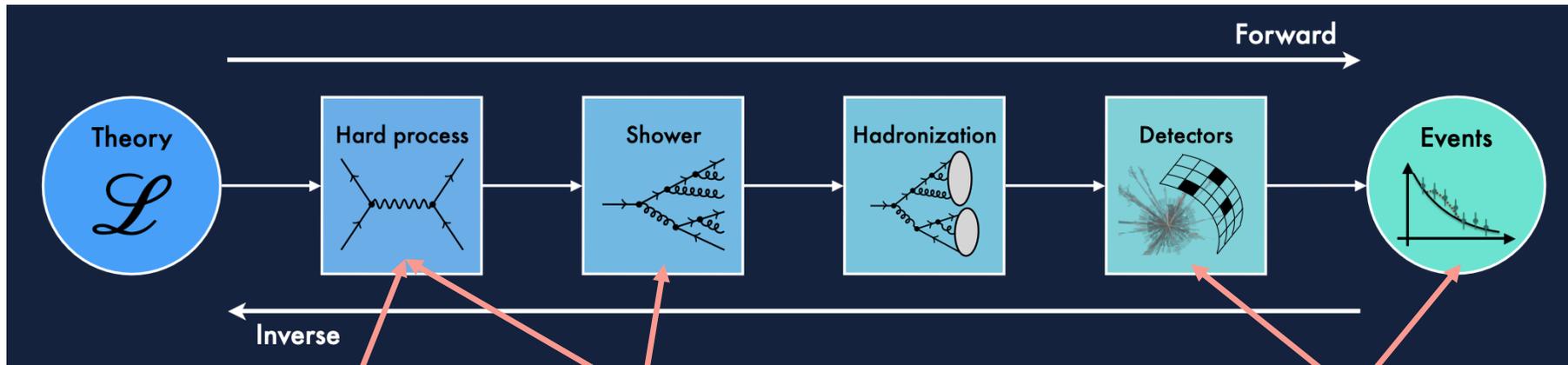
CMS Collaboration  
 Submitted to Reports on Progress in Physics  
 6 June 2024

**From angular coefficients to quantum observables:  
 a phenomenological appraisal in di-boson systems**

arXiv:2409.16731

Michele Grossi,<sup>a</sup> Giovanni Pelliccioli,<sup>b</sup> Alessandro Vicini<sup>c</sup>



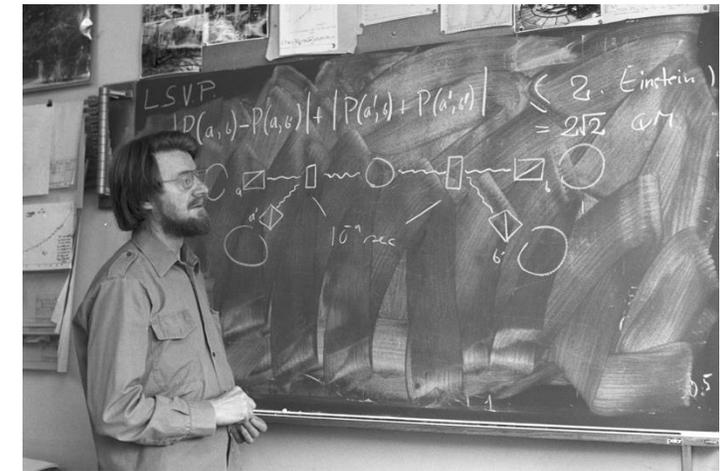


entanglement [1703.02989]      interference [2110.10112]      spin correlations [1907.03729]      Bell inequalities [2102.11883, 2203.05582]

## Fundamental motivation

Utilise information and correlations inherent in HEP data.

Exploit “quantum remnants” in data.



# CERN QTI Phase 2

*Launched January 2024*

HYBRID QUANTUM  
COMPUTING AND  
ALGORITHMS

QUANTUM  
NETWORKS AND  
COMMUNICATIONS

CERN QUANTUM  
TECHNOLOGY  
PLATFORMS

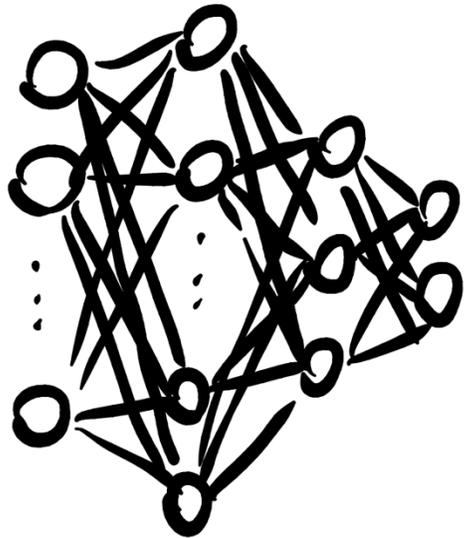
COLLABORATION  
FOR IMPACT



*A 5 years research plan*



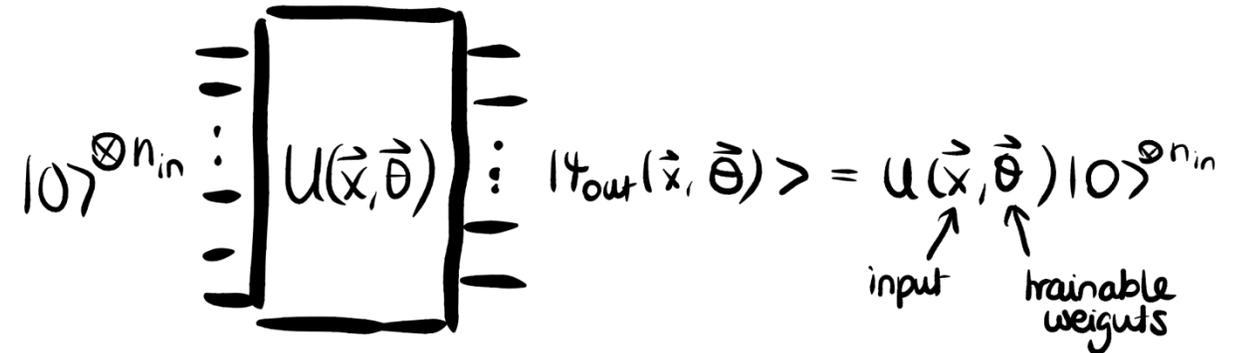
# From ML to QML: easy?



$$\Psi(\vec{x}, \vec{\theta}) : \mathbb{R}^{n_{in}} \rightarrow \mathbb{R}^{n_{out}}$$

↑input    ↑trainable weights

Classical Neural Network

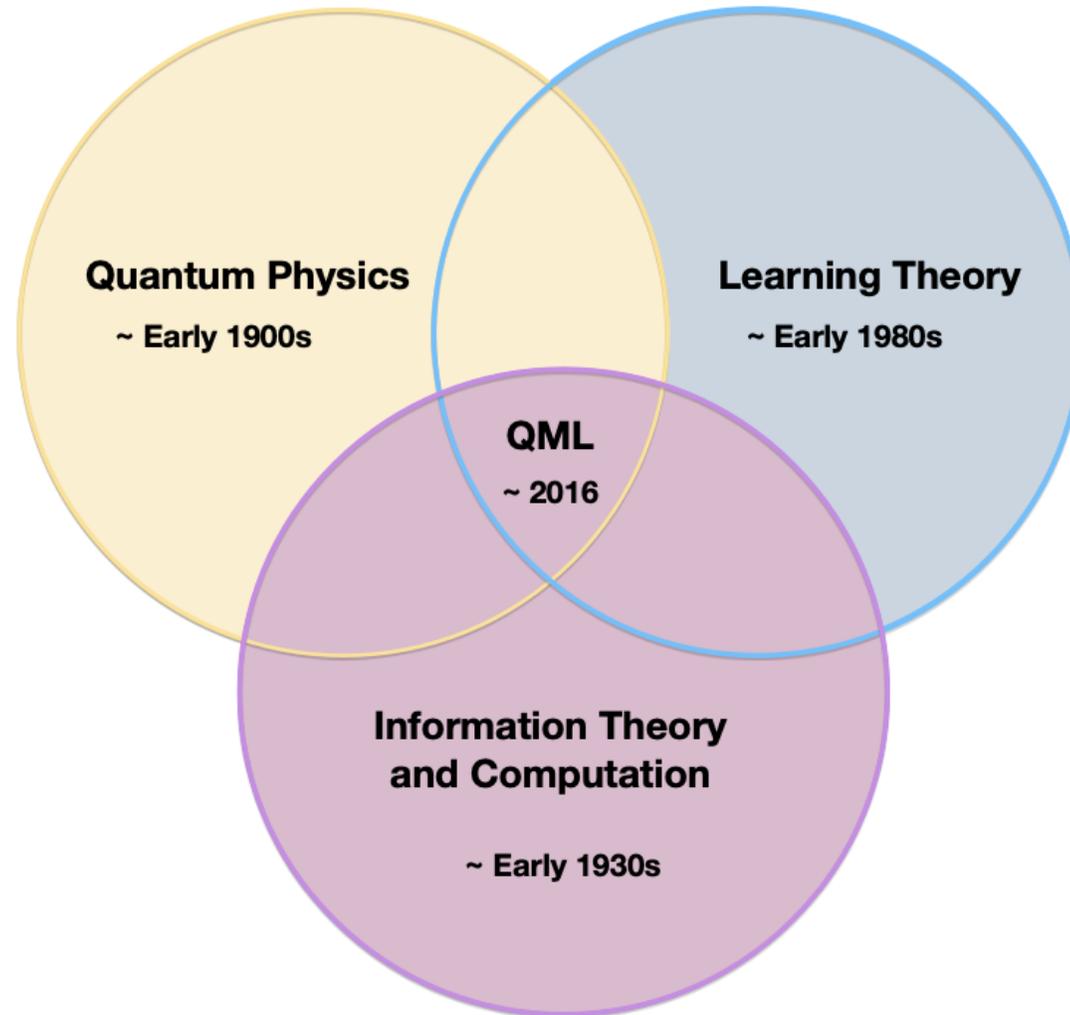


$$y(\vec{x}, \vec{\theta}) = \langle \Psi_{out}(\vec{x}, \vec{\theta}) | \hat{O} | \Psi_{out}(\vec{x}, \vec{\theta}) \rangle$$

Parametrized Quantum Circuit

Input at different stage of computation  
Unitary operations

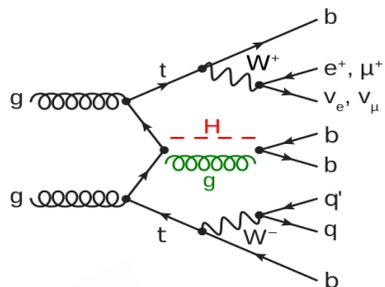
# Quantum Machine Learning (QML)



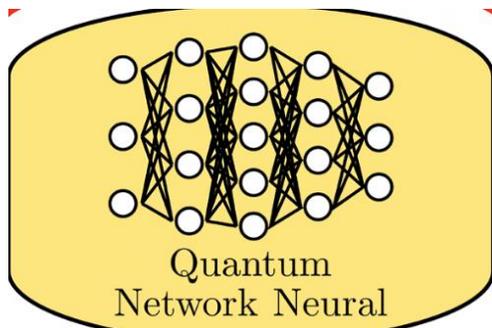
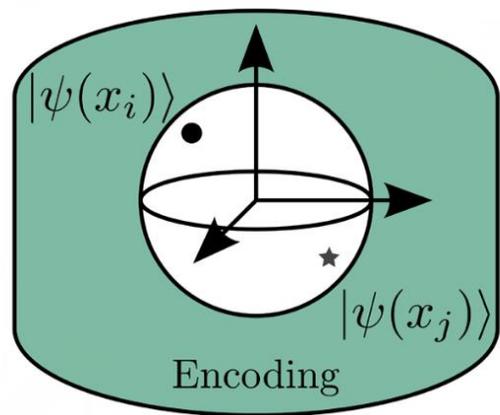
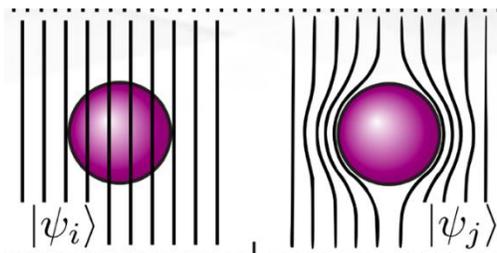
# Quantum Circuits for QML

Theoretically:  $P(0) = \text{Tr}[|0\rangle\langle 0|\rho]$ .  
 In practice:  $P(0) = \frac{N_0}{N}$  with a statistical precision of order  $\frac{1}{\sqrt{N}}$

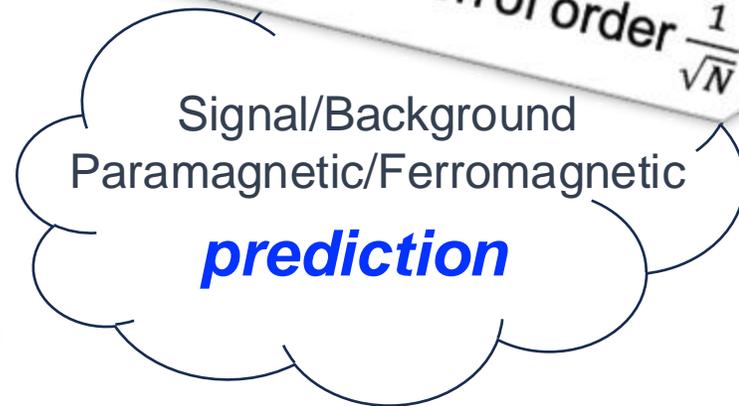
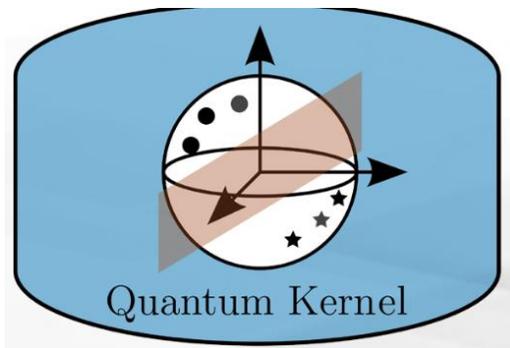
## Classical Data



## Quantum Data



*training*

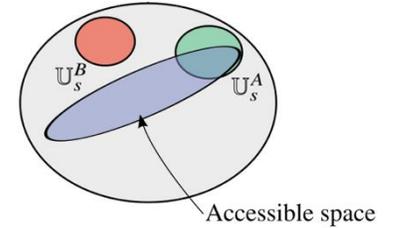


# Variational Quantum Algorithms – the Challenge

## 1. Efficient data handling and data embedding

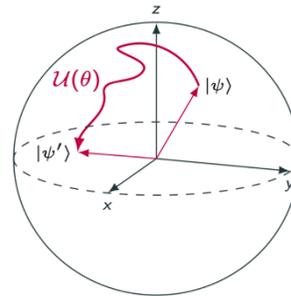
## 2. Ansatz choice

Can we find the most suitable ansatz for the given problem?  
How well can we survey the Hilbert space (SYMMETRY?!)?



## 3. Trainability

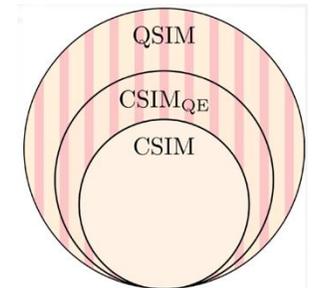
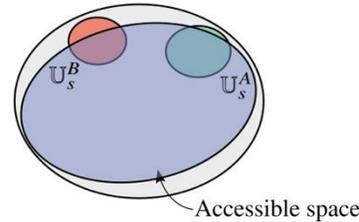
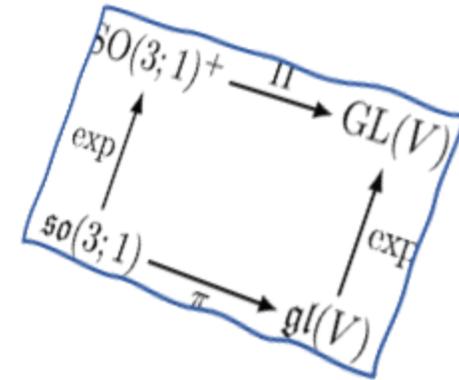
Can the parameters be updated?



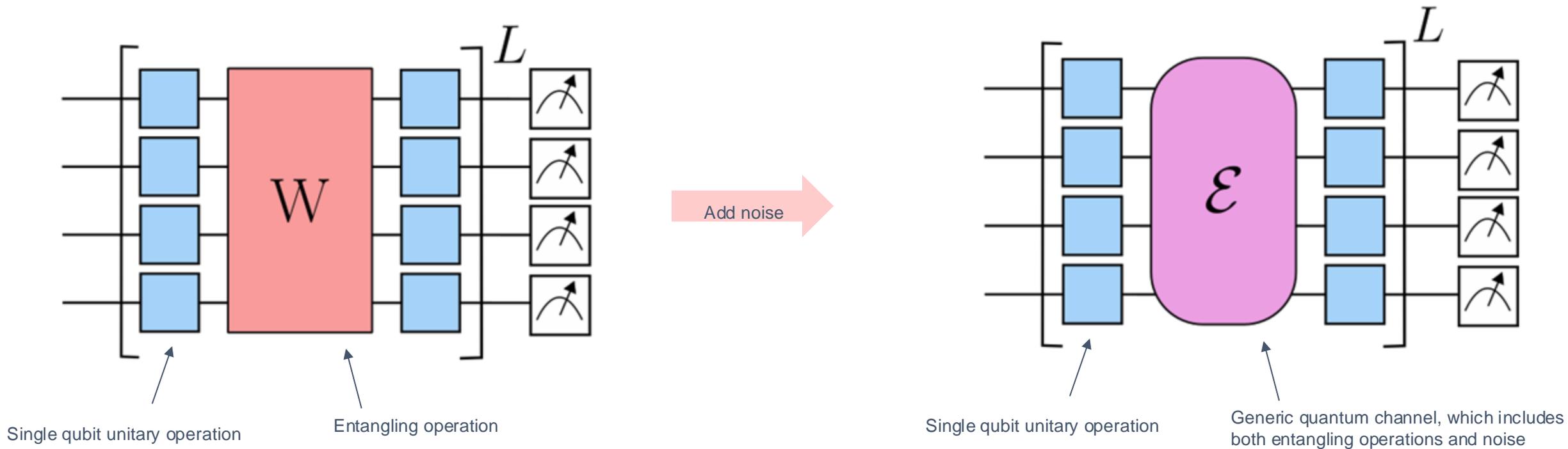
## 4. Classical Simulability

Are the quantum simulations classically simulable?  
No need for a quantum computer!?

*Just because we can simulate a loss, does not mean it is practical to do so!*



# What about noise? Non-unitary QML



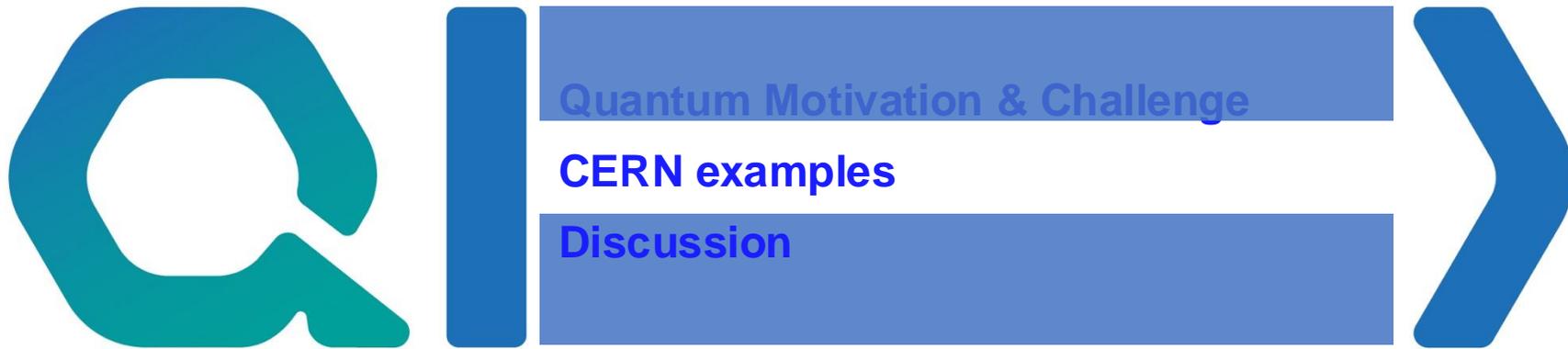
The presence of noise is often overlooked in such analyses

→ Symmetry breaking in geometric quantum machine learning in the presence of noise

[\[MG et al. PRX Quantum 5, 030314\]](#)

→ Estimates of loss function concentration in noisy parametrized quantum circuits

[\[G. Crognalotti., GM, et al – arXiv:2410.01893\]](#)

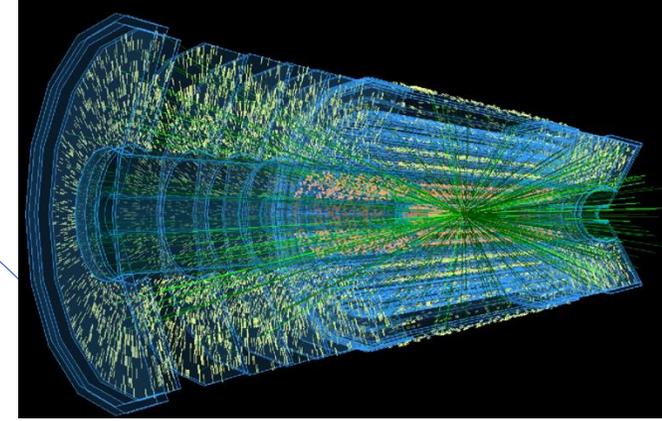


# HEP Pipeline

Theory

Calculate (differential) cross sections

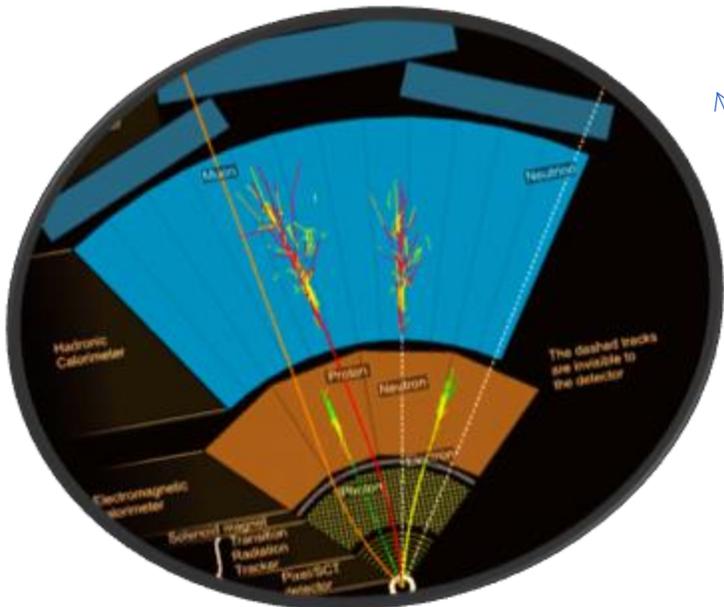
$$d\sigma = \frac{1}{\text{flux}} dx_a dx_b f(x_a) f(x_b) d\Phi_n \langle |M_{\lambda,c,\dots}(p_a, p_b | p_1, \dots, p_n)|^2 \rangle$$



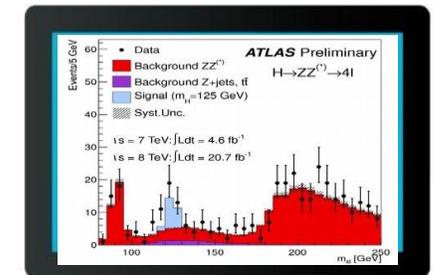
Data Analysis

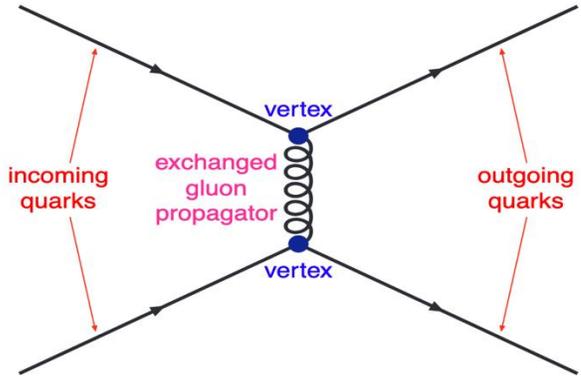


Data Generation



Feature Extraction





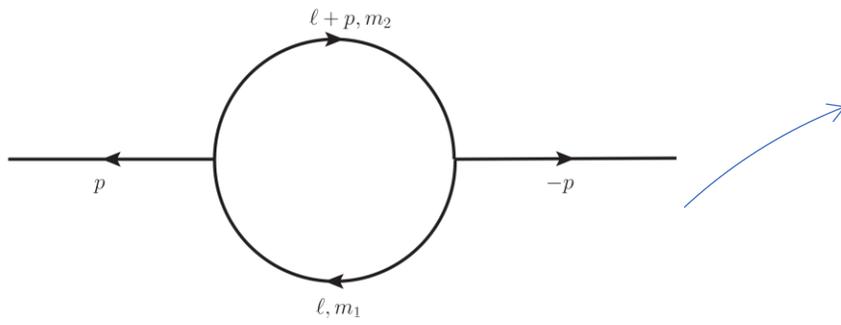
Theory

**3 billions CPU hours/year  
15% is MC integration**

$$\sigma = \frac{1}{F} \int d\Phi |M|^2 \Theta(\Phi - \Phi_c)$$

*phase-space factor* (points to  $\frac{1}{F}$ )  
*integrand* (points to  $|M|^2 \Theta(\Phi - \Phi_c)$ )  
*probability distributions/ matrix element* (points to  $|M|^2$ )  
*phase-space cuts* (points to  $\Theta(\Phi - \Phi_c)$ )

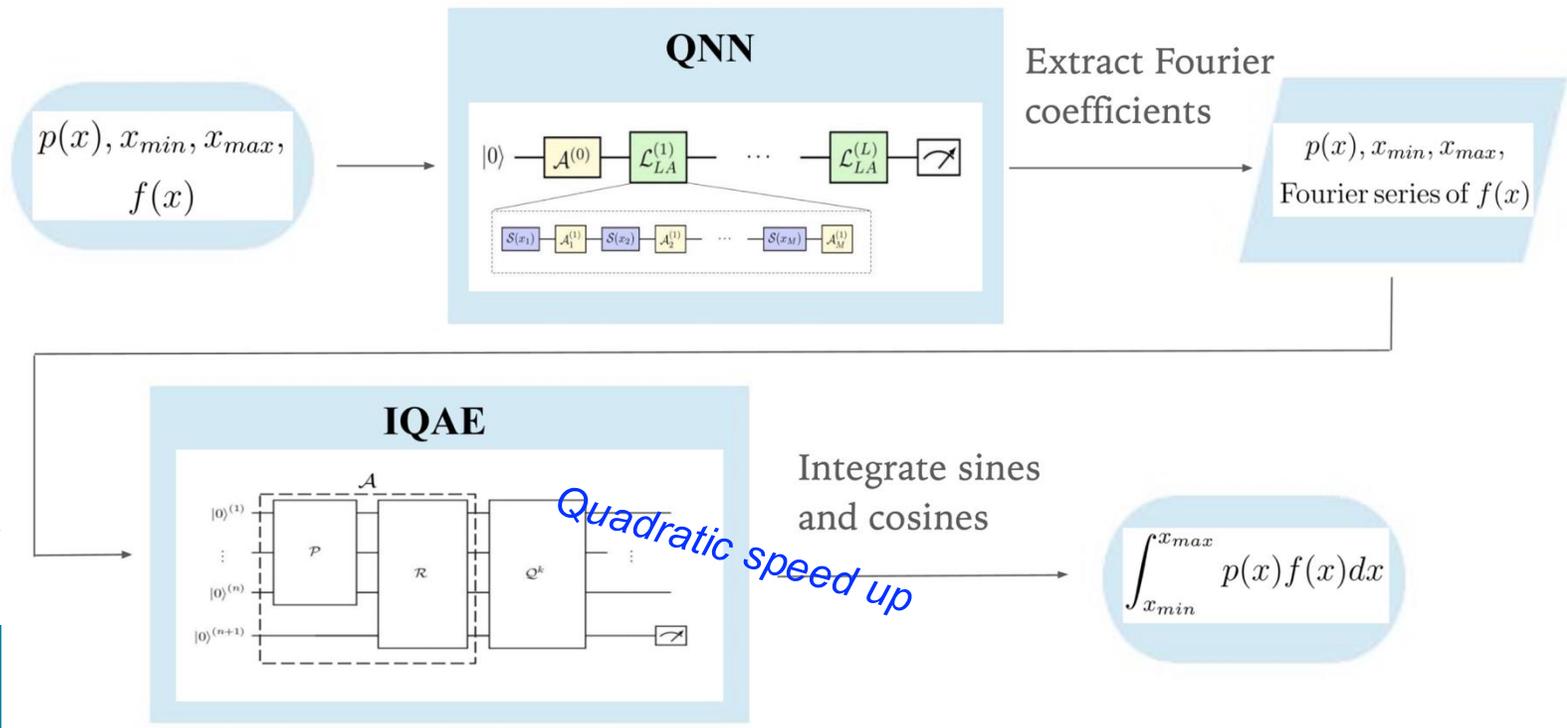
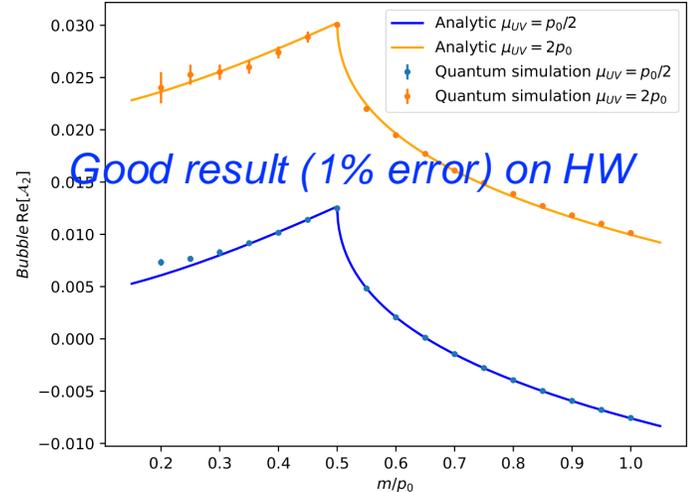
Agliardi, Grossi, Pellen, Prati "Quantum integration of elementary particle processes." <https://doi.org/10.1016/j.physletb.2022.137228>



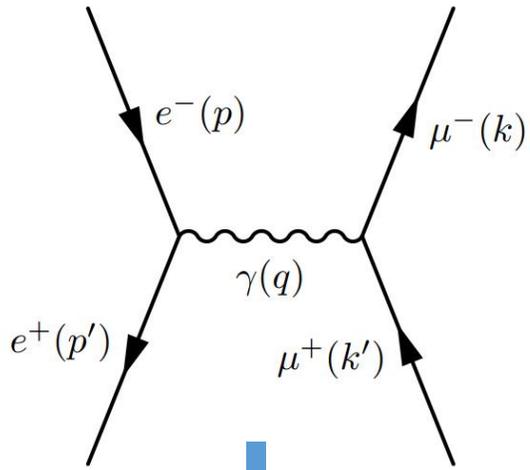
# Theory

Loop Feynman integral (Bubble)

$$\mathcal{A}_2^{(1)}(p, m_1, m_2) = \int_{\ell} \prod_{i=1}^2 G_F(q_i) = \int_{\ell} \frac{1}{(\ell^2 - m_1^2 + i0) ((\ell + p)^2 - m_2^2 + i0)}$$



*PhysRevD.110.074031 - Martinez de Lejarza, GM., et al.*  
 Loop Feynman integration on a quantum computer

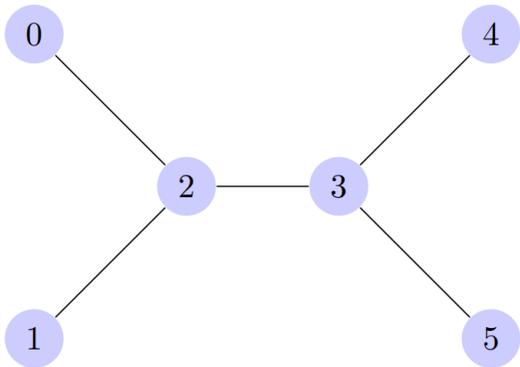


# Theory

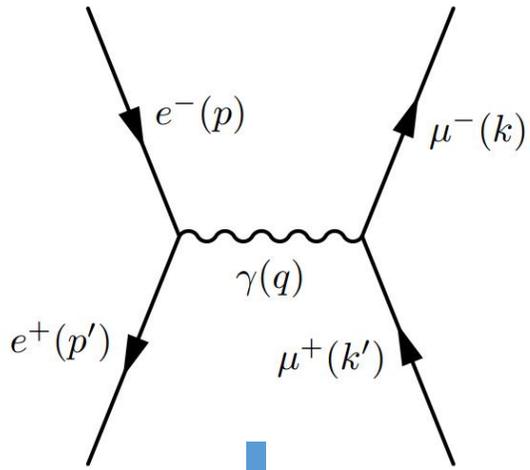
$$\mathcal{L} = -\frac{1}{4}F_{\mu\nu}F^{\mu\nu} + i\bar{\Psi}\not{D}\Psi + h.c. + \bar{\Psi}_i\gamma_{ij}\Psi_j\Phi + h.c. + |\mathcal{D}_\mu\Phi|^2 - V(\Phi)$$

$$\sigma = \frac{1}{F} \int d\Phi |M|^2 \Theta(\Phi - \Phi_c)$$

*phase-space factor* (pointing to  $d\Phi$ )  
*matrix element* (pointing to  $|M|^2$ )  
*phase-space cuts* (pointing to  $\Theta(\Phi - \Phi_c)$ )



s-channel

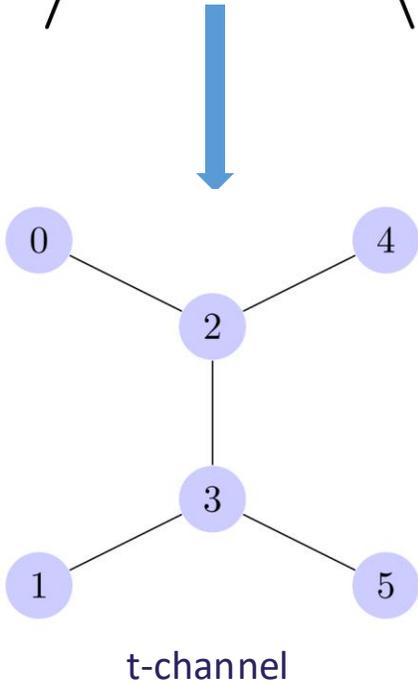


# Theory

$$\mathcal{L} = -\frac{1}{4}F_{\mu\nu}F^{\mu\nu} + i\bar{\Psi}\not{D}\Psi + h.c. + \bar{\Psi}_i\gamma_{ij}\Psi_j\Phi + h.c. + |D_\mu\Phi|^2 - V(\Phi)$$

$$\sigma = \frac{1}{F} \int d\Phi |M|^2 \Theta(\Phi - \Phi_c)$$

*phase-space factor*
*matrix element*
*phase-space cuts*



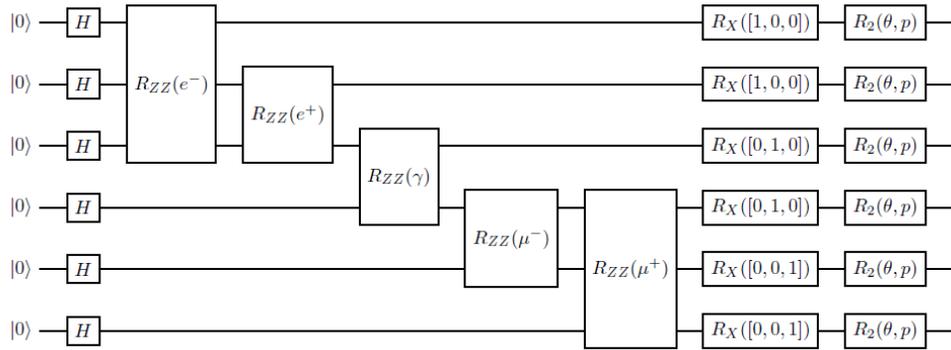
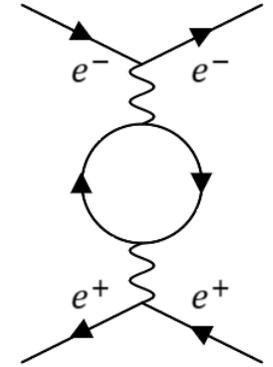
- Build a **quantum supervised model** that can distinguish (C) and compute (R) the scattering amplitude squared for related Feynman diagrams LO QED process
- Topology encoded in the adjacency matrix of the graph
- Particles (m,Q,S) encoded in the edges
- Time flow (initial state, interaction vertex,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

# Theory

$$\sigma = \frac{1}{F} \int d\Phi |M|^2 \Theta(\Phi - \Phi_c)$$

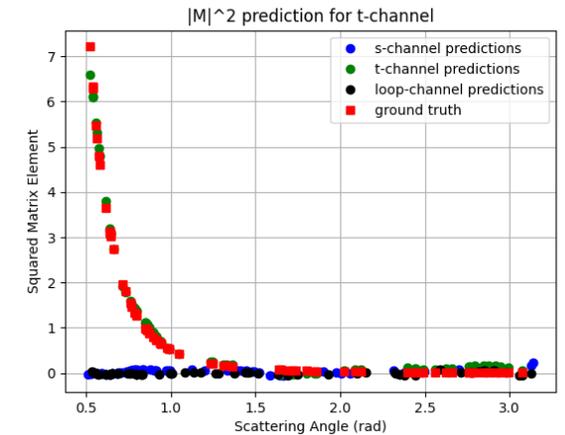
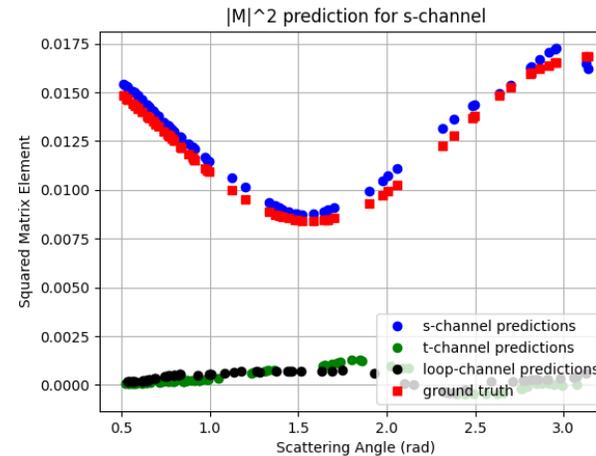
*phase-space factor*
*phase-space cuts*

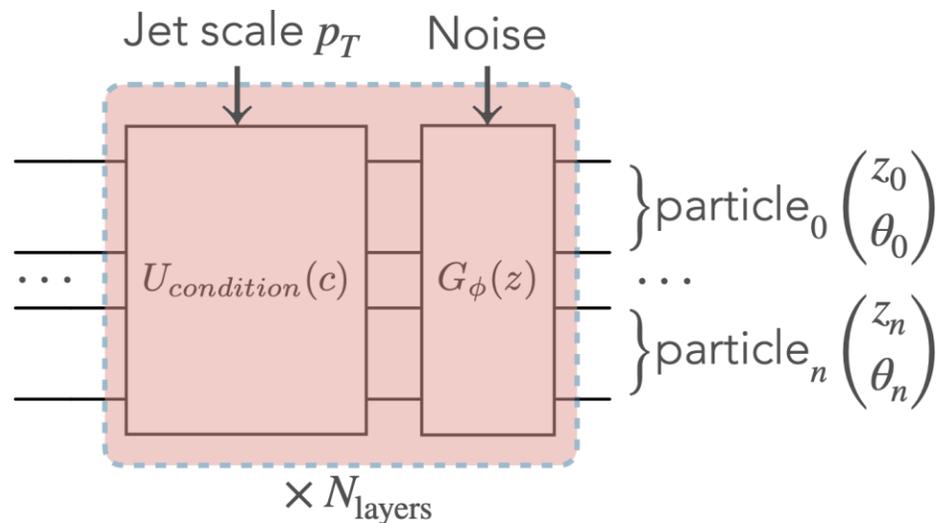
*matrix element*



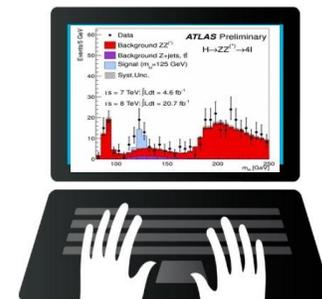
## Successful training:

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





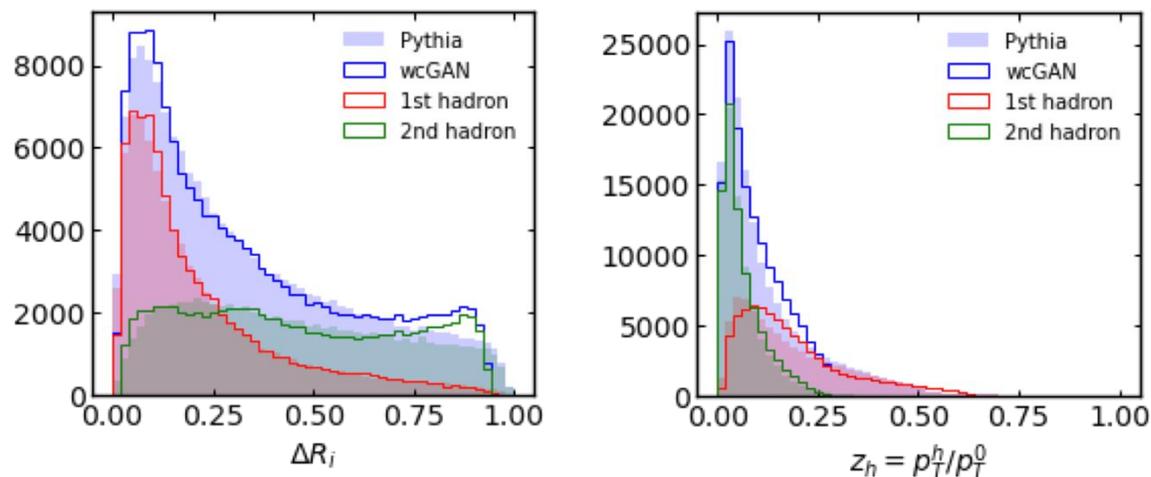
- Each jet constituent represented by two features:
  1. Momentum fraction  $z_i = p_T^i / p_T^{jet}$
  2. Angle with reference to the jet axis  $\theta_i = \Delta R_i / R$
- qubit  $\rightarrow$  1 feature:  $\mathbf{x}_{fake} = \{ \langle \sigma_Z^0 \rangle, \langle \sigma_Z^1 \rangle, \dots, \langle \sigma_Z^n \rangle \}$
- Style-based approach: The noise is inserted in every gate:  $R_{x,y,z} \rightarrow R_{x,y,z}(w \cdot z + b)$



## Style-based Hybrid QGAN

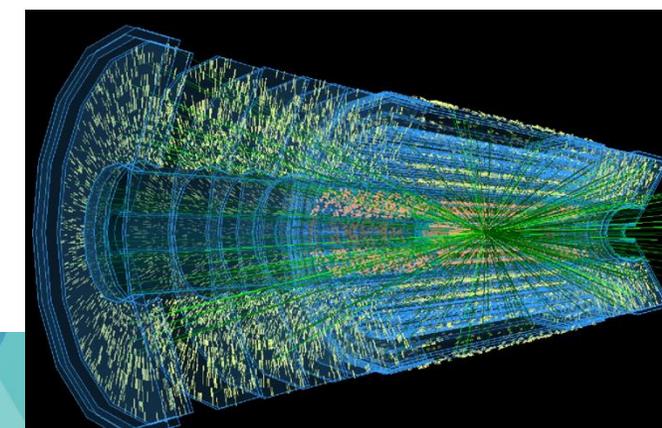
The Quantum GAN trained on Z+jets events generated by Pythia8.

Data Generation



The Quantum GAN captures the distributions of the first and second emissions, reproduce their dependence with the jet scale

MG, Y. Haddad, V. Croft, C. Tuszynski in preparation



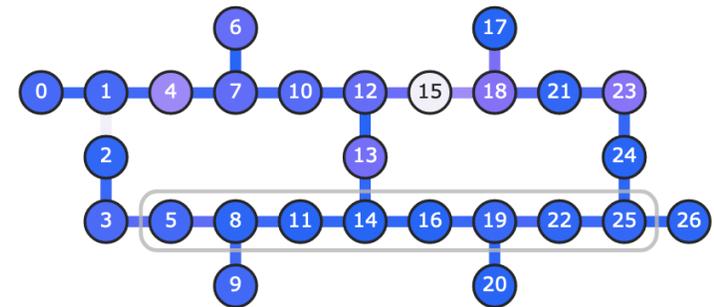
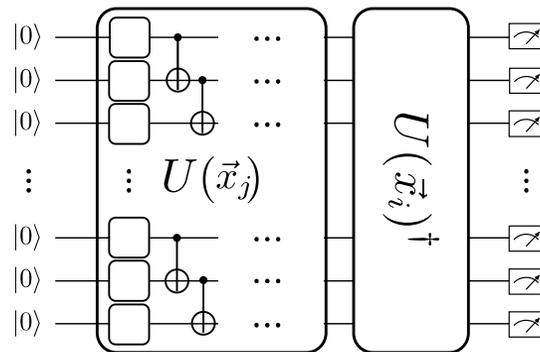
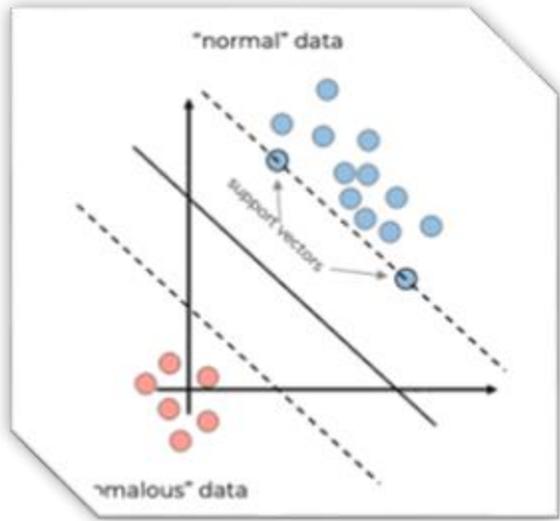
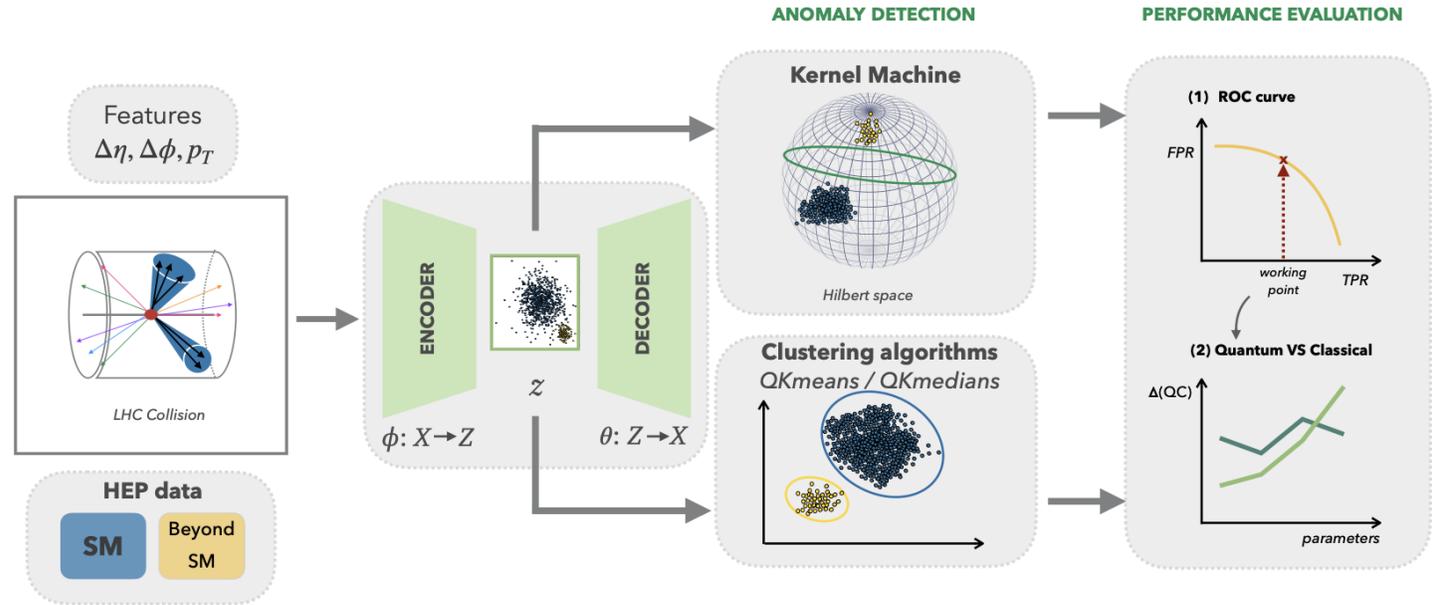


# Quantum Anomaly Detection

Belis V., GM, et al – COMMSPHYS-23-1149C



Data Analysis

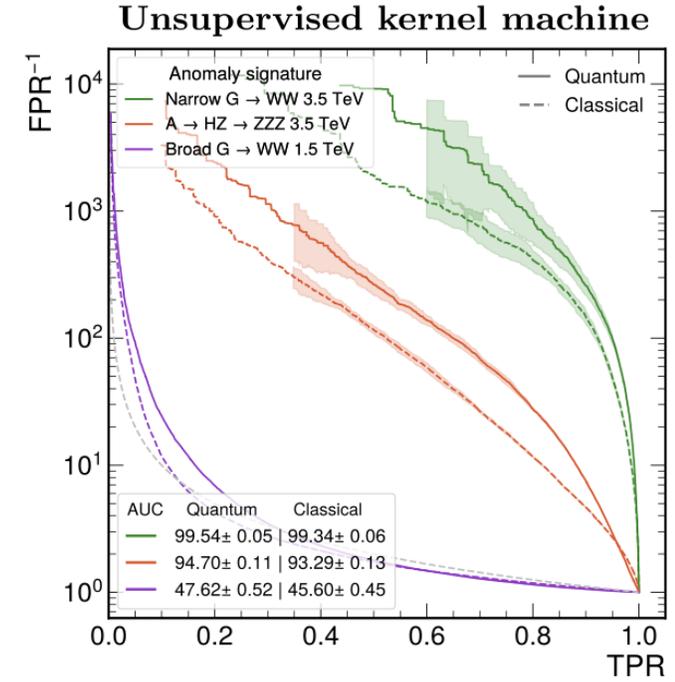
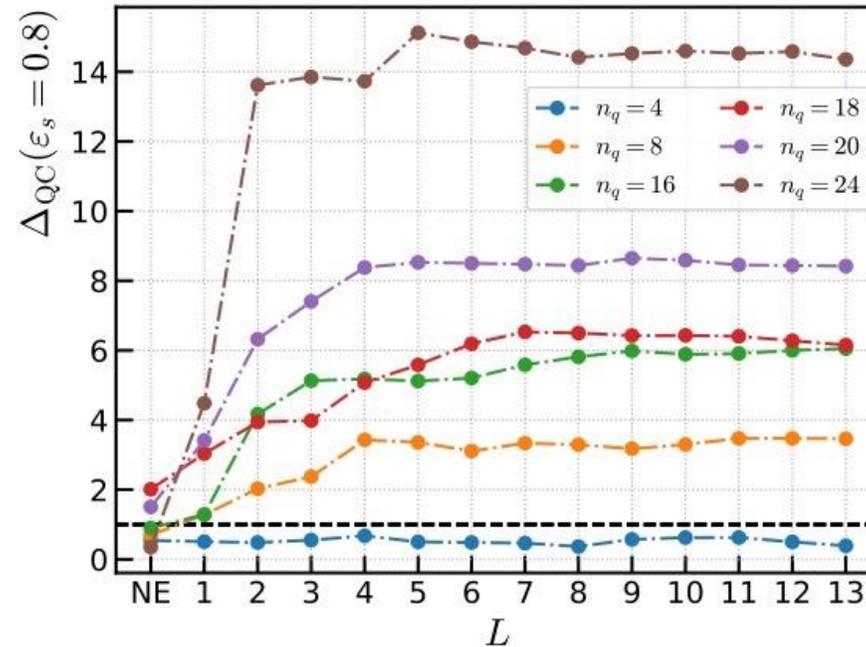
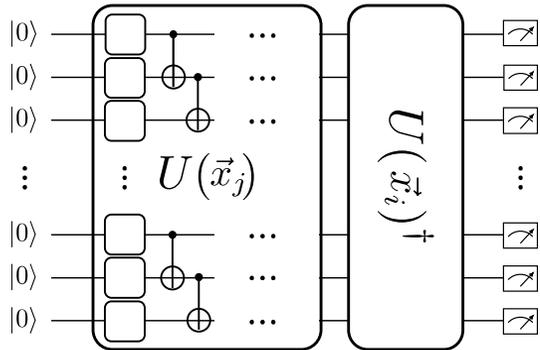


# Quantum Anomaly Detection

Belis V., GM, et al – COMMSPHYS-23-1149C



Data Analysis





Quantum Motivation & Challenge  
CERN examples  
**Discussion**

# QC research directions in HEP



## 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?

## Experimental data is shaped by physics laws

- Can we leverage them to build better algorithms?
- Can we train the loss on a classical device, and sample on quantum (GENERATIVE MODELS)
- Quantum Error Mitigation is the way, waiting for scalable **ERROR CORRECTION**

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

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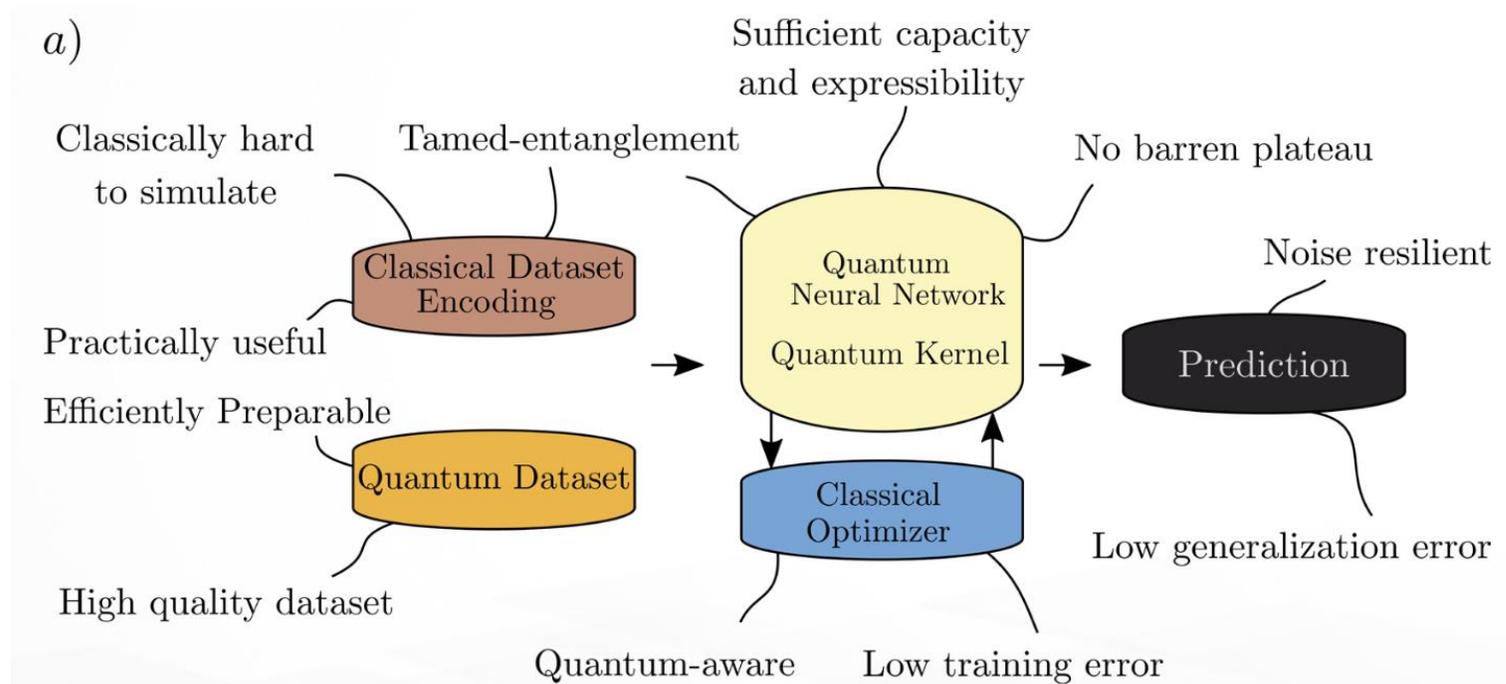
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# Variational Quantum Algorithms – Summary

- VQA can't be trusted any more than **classical machine learning**
- VQA requires linear algebra and python
- Some success has been achieved for small problem sizes ( $N < 30$  qubits)
- We do not yet have the hardware required to **test these algorithms at scale**



Perspective: *Challenges and opportunities in quantum machine learning*, M. Cerezo, et al., Nature Comp. Sc., 2, 567 (2022).

# Quantum Algorithms – Summary

## Conventional quantum algorithms

- come with provable guarantees
- require significant knowledge of quantum information, group theory, physics, etc.

## Query complexity: classical versus quantum

	Determ. machine (worst case)	Quantum computer
Deutsch	2	1
Deutsch–Jozsa	$2^n/2 + 1$	1
Bernstein–Vazirani	$n$	1
Grover	$2^n - 1$	$O(\sqrt{2^n})$
Simon	$2^n/2 + 1$	$O(n)$
Period finding	$O(r)$	$O(1)$

# Conclusion

NISQ → ISQ

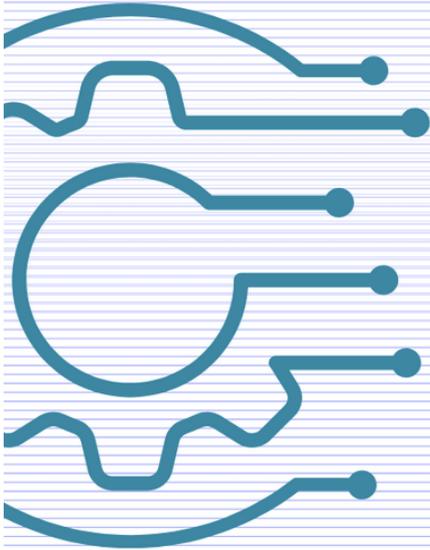
- Complexity & learning theory mostly gives us insights into **worst-case behavior**
  - ML: Learning theory predicted deep neural networks to not be trainable
  - Optimization: The travelling salesperson problem is NP-complete. An instance with 85900 cities was solved in 2006. Exponential complexity does not imply infeasibility
- **Benchmarking** can help us to understand the behavior on specific instances
- We need to make a comparison of **computational cost** - may lead to poly advantages!
- Change the goal: quantum advantage will be unlikely in many cases **BUT** we can identify promising paths for **hybrid computational advantages**
- We can train the loss on a classical device, and sample on quantum (GENERATIVE MODELS)
  - larger devices for high-quality data?
- What's the role of data?
- Community goal is bridging the gap between near-term and fault-tolerant quantum machine learning

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