



# Conditional Born machine for Monte Carlo events generation

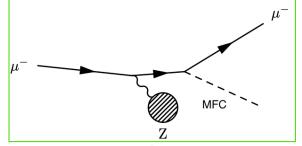
Michele Grossi, Enrique Kajomovitz, Oriel Kiss and Sofia Vallecorsa

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# **Muonic Force Carriers**



- MFCs appear in different theoretical hypothesis, as a constituent of dark matter and could explain the anomalous magnetic dipole moment of the muon or the anomaly in the measurement of the proton radius.
- 2 use cases: a) muon fixed-target collision (FASER).
  b) muon interactions in the ATLAS<sup>5</sup> calorimeter.
- We are interested to generates following features for the outgoing muon and MFC: energy (E), transversal momentum (pt) and pseudorapidity (η).

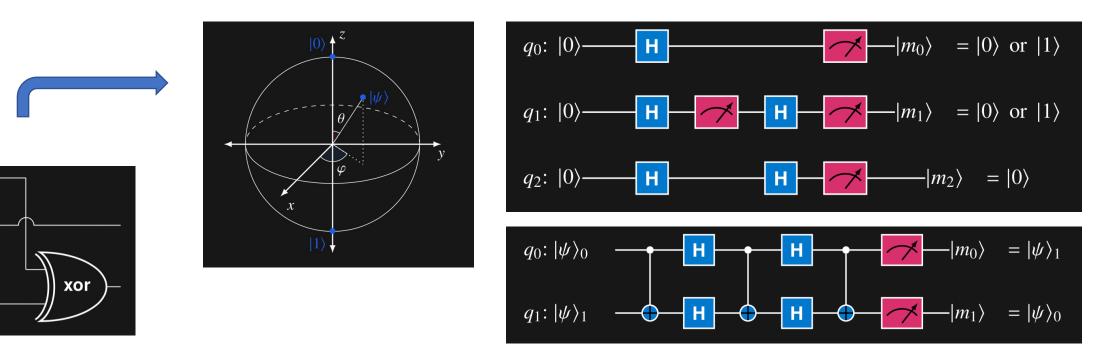
[5] Galon, I, Kajamovitz, E et al. "Searching for muonic forces with the ATLAS detector". In: Phys. Rev. D 101, 011701 (2020)



### Quantum computing uses essential ideas from quantum mechanics

# • Classical logical circuits use operations like and, or, not, nand, and xor. We also call these gates.

• Quantum circuits use reversible gates that change the quantum states of one, two, or more qubits.



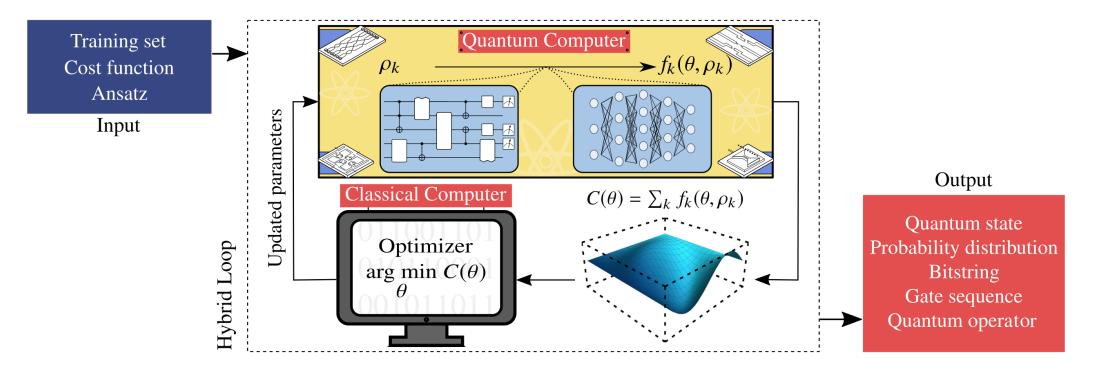


and

**Gates / operations** 

### **Quantum machine learning models**

### Quantum circuits as heuristic machine learning models.

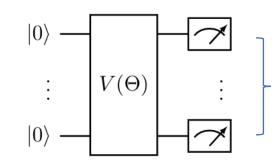


Cerezo et al., Nat Rev Phys 3, 625-644 (2021)



### **Quantum Circuit Born machine (QCBM)**

**1.** Sample from a variational pure state  $|\psi(\theta)\rangle$  by projective measurement with probability given by the Born rule:  $p_{\theta}(x) = |\langle x | \psi(\theta) \rangle|^2$ .



n dimensional binary strings map to 2<sup>n</sup> bins of the discretized dataset.

- KL divergence
- Adversarial (QGAN)
- In the phase space.
- Delgado and Hamilton, arXiv:2203.03578 (2022). Zoufal, et al., *npj Quantum Inf* **5**, 103 (2019).
- Kyriienko, et al., arXiv: 2202.08253 (2022).
- Maximum Mean Discrepancy

 $\mathsf{MMD}(\mathsf{P},\mathsf{Q}) = \mathbb{E}_{\substack{X \sim P \\ Y \sim P}} [K(X,Y)] + \mathbb{E}_{\substack{X \sim Q \\ Y \sim Q}} [K(X,Y)] - 2\mathbb{E}_{\substack{X \sim P \\ Y \sim Q}} [K(X,Y)]$ 

- 3. Why the *Maximum Mean Discrepancy* MMD?
  - Resource efficient for NISQ devices.
  - Stable.
  - However, empirically less performant than adversarial.

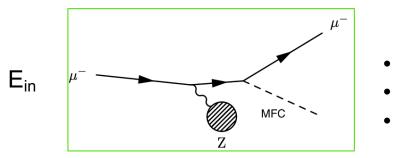


*Training* (Hybrid loop):

2.

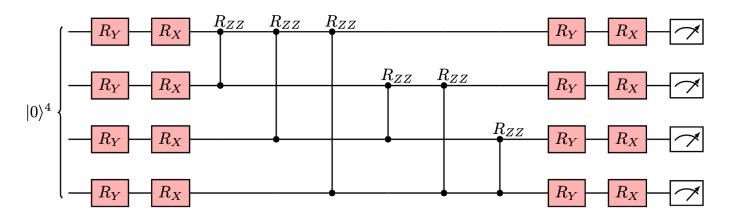
### **Generation of 1-D Monte Carlo samples**

#### MC sampling: time consuming



- Energy E
- Transversal momentum *p<sub>t</sub>*
- Pseudorapidity  $\eta$

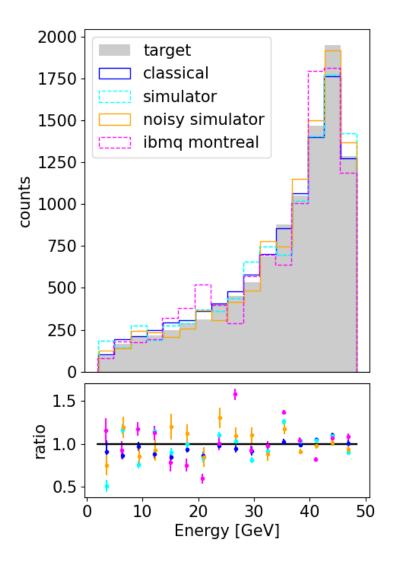
#### **QCBM:** hardware efficient ansatz (22 parameters)



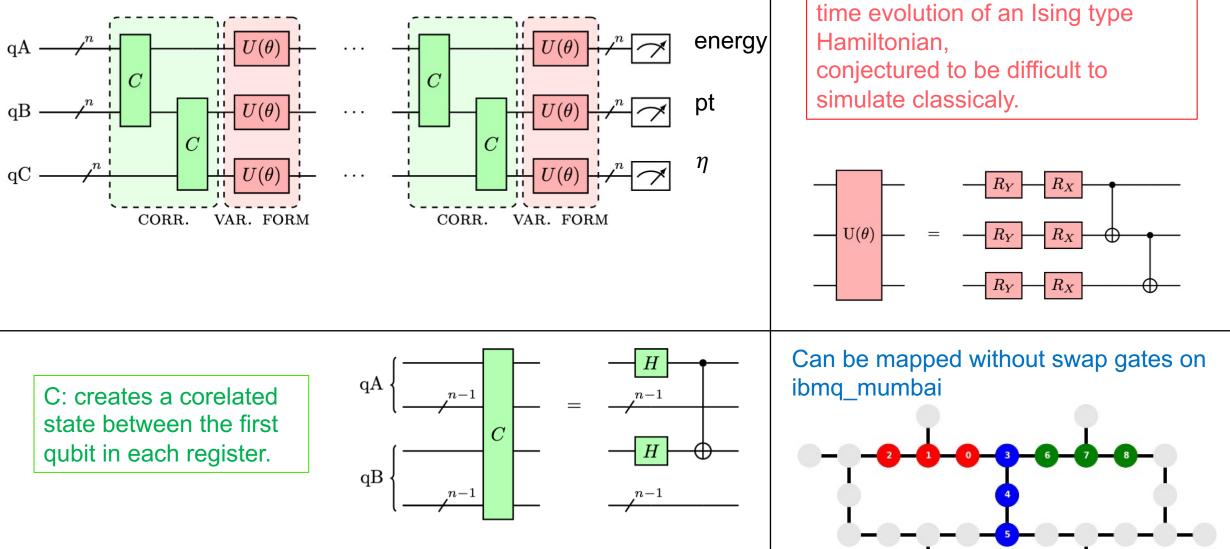
Kiss, Grossi, et al., Phys. Rev. A 106, 022612 (2022)

QUANTUM

TECHNOLOGY NITIATIVE



### **Multivariate probability distribution**



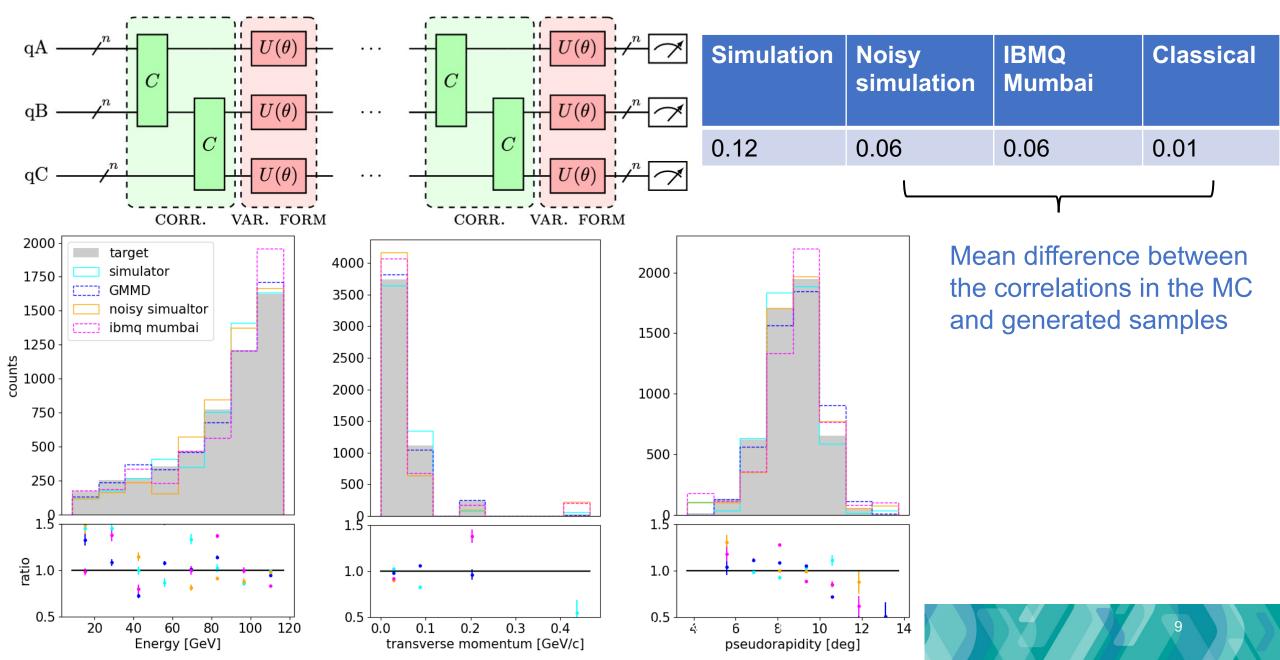
Kiss, Grossi, et al., Phys. Rev. A 106, 022612 (2022)



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Local: learns the individual PDFs,

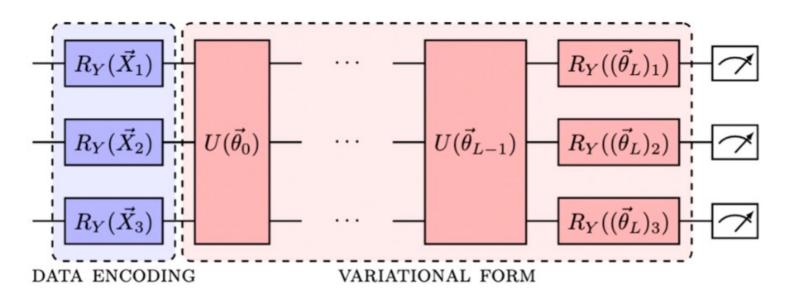
### **Multivariate probability distribution**



# **Conditional probability distribution**

green: fixed gates blue: data encoding gates red: trainable gates

**Data encoding**: via data-parametrized rotations **Input**: binning energy E (scaled between [0,1] and preprcoessed with arcsin()) **Interpolation**: train only on certain energy bins and the model should learn to predict in between.



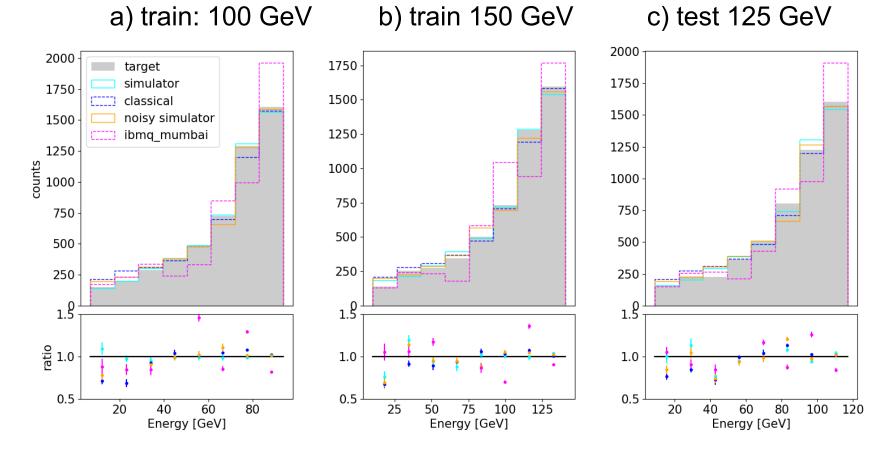
We want to modelize p(y|x) where x is the incoming energy  $E_{in}$ .



## **Conditional probability distribution**

**Goal: Interpolation** 

- 1. Data re-uploading does not improve the sampling.
- 2. Training on hardware is important to assimilate the noise.





# **Conclusion:**

- 1. Test of a quantum circuit Born machine to generate MFC events.
- 2. The Born machine is currently able to handle multivariate and conditional distributions.
- 3. Training on quantum hardware is important in order to assimilate the noise.
- 4. Futur work is devoted to scale the QCBM, in terms of number of qubits and register, and also extension to more difficult cases.



# Thank you

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