

Discovering New Physics with Quantum Computers

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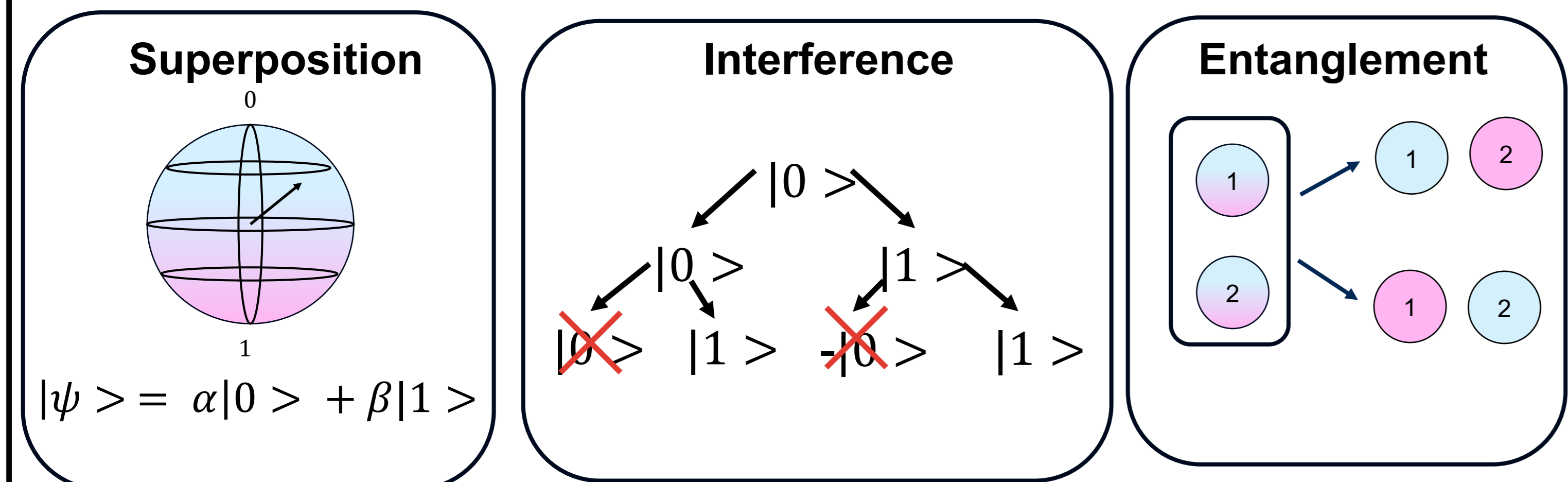
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1. What is Quantum Computing?

- A new form of computing exploiting quantum mechanics.
- Certain problems can be **solved exponentially faster** than classical machines.

Key Properties:

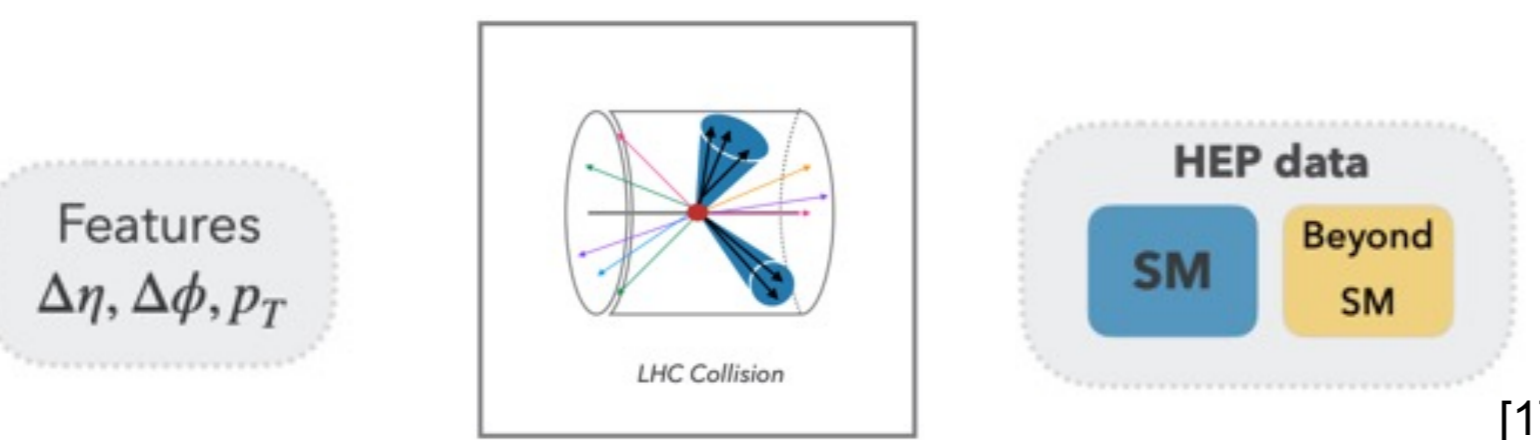


2. Potential Applications

- Simulations:**
- New battery technology.
 - Improved drug design.
 - Particle physics simulations.
- Machine learning:**
- Faster training on certain tasks.
 - Better prediction capabilities.
 - Learning with fewer samples.

3. Our Project

- Since the Higgs boson what's next?
- Can we find **new physics** we have never observed?
- We use Model agnostic searches.
- Extract anomalous events using kinematic data from collisions.

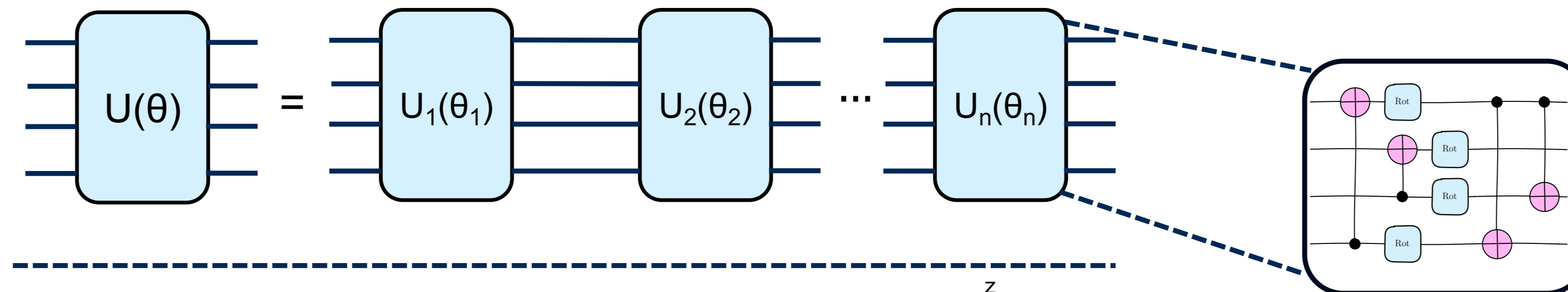


4. Variational Quantum Algorithms (VQA)

General form of cost function: $\theta^* = \text{argmin}_{\theta} C(\theta)$
 $C(\theta) = \sum_k f(\text{Tr}[O_k U(\theta) \rho_k U^\dagger(\theta)])$

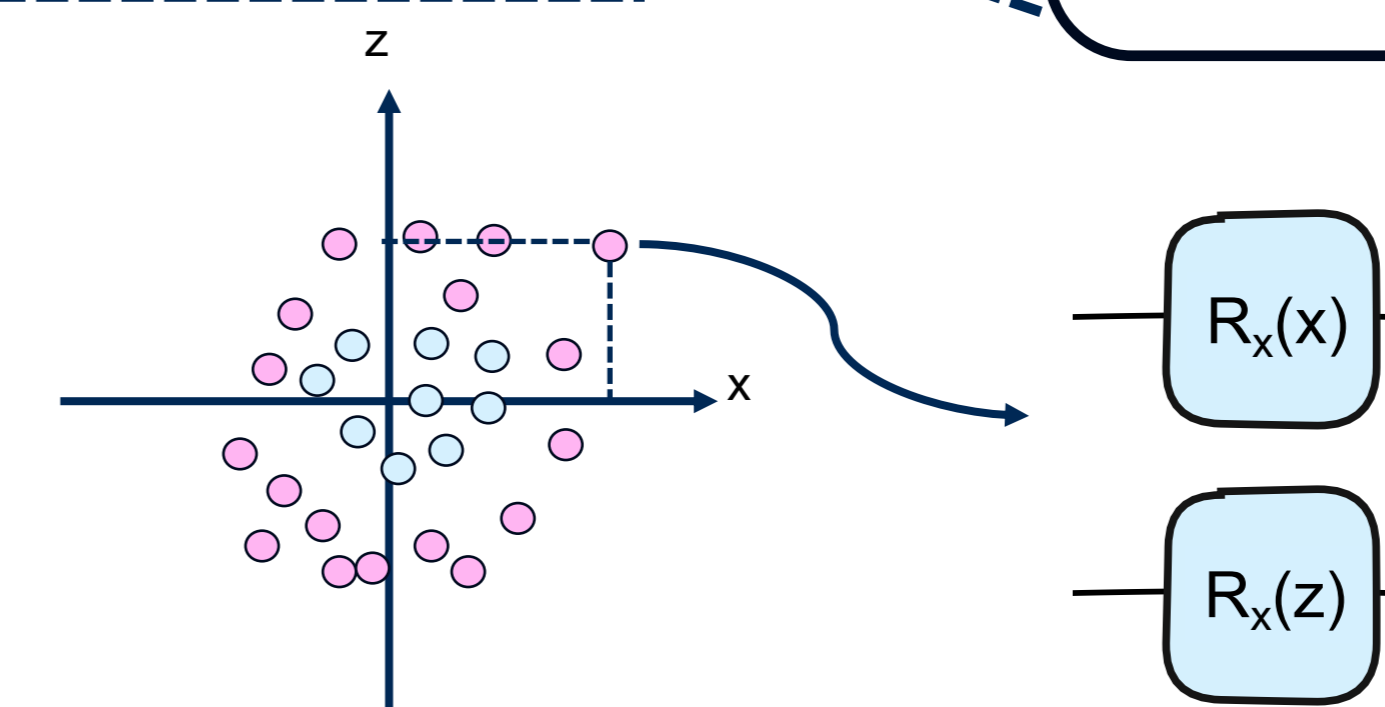
Ansatz:

- Structure determines the **inductive biases**, parameters θ are trained to minimize cost.
- A series of rotation and entangling gates, the structure can be general or problem specific.

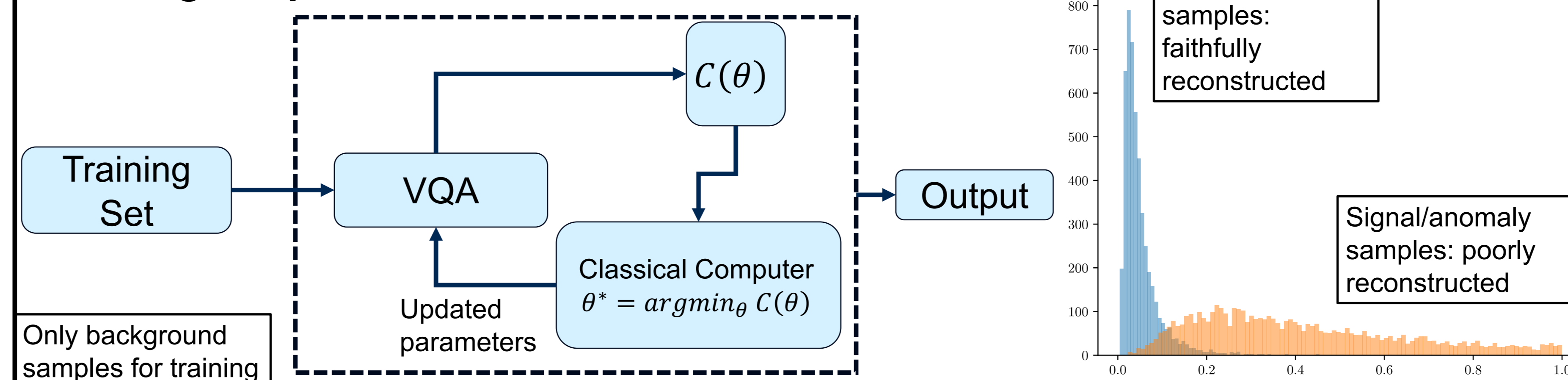


Feature map:

- Maps input vectors to a Hilbert space.
- Encoded data can be fed into an ansatz.



Training Loop:



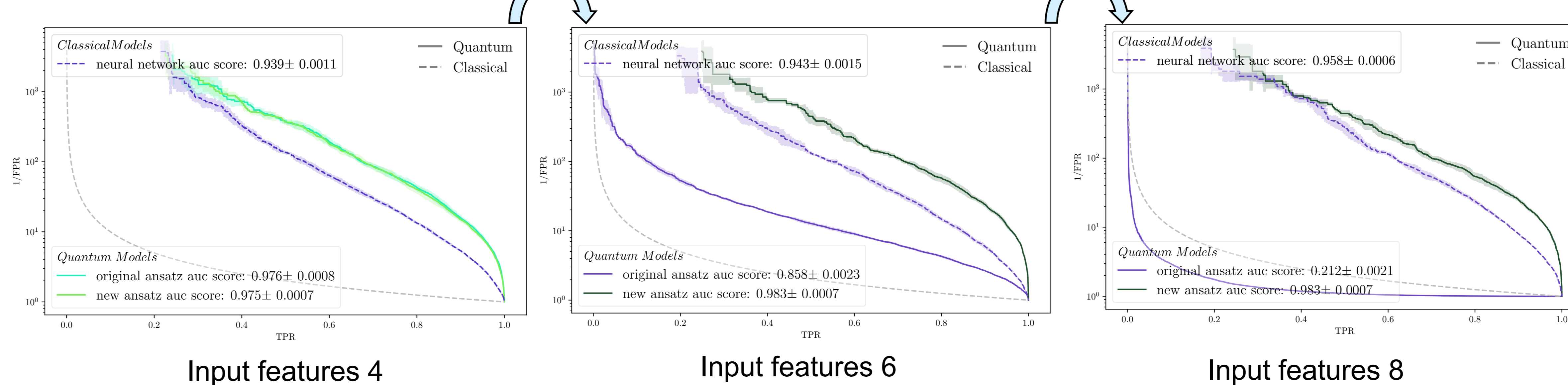
9. Results

- Identifying BSM Higgs amongst QCD events.
- AUC score: measures predictive performance.

New Ansatz!

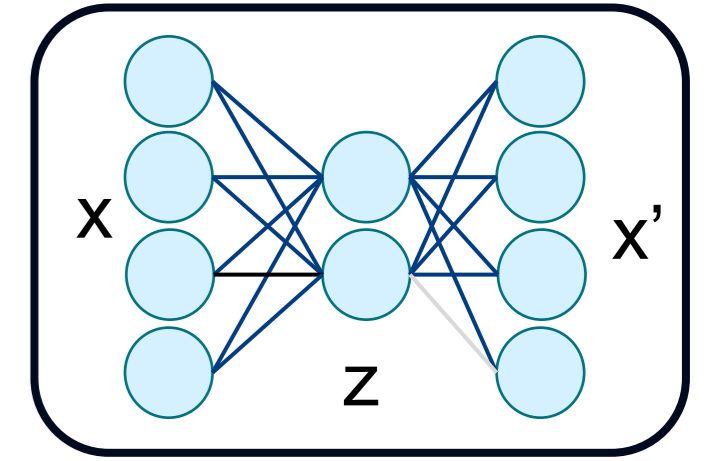
- Scales better with problem size
- Fewer parameters than CAE
- Fewer CNOT gates
- Algorithm specific

Heavy Higgs Search



5. Classical Autoencoder (CAE)

- Input vectors x are compressed to a representation z .
- z is decoded into x' .
- x and x' are compared with a cost function to be minimized.



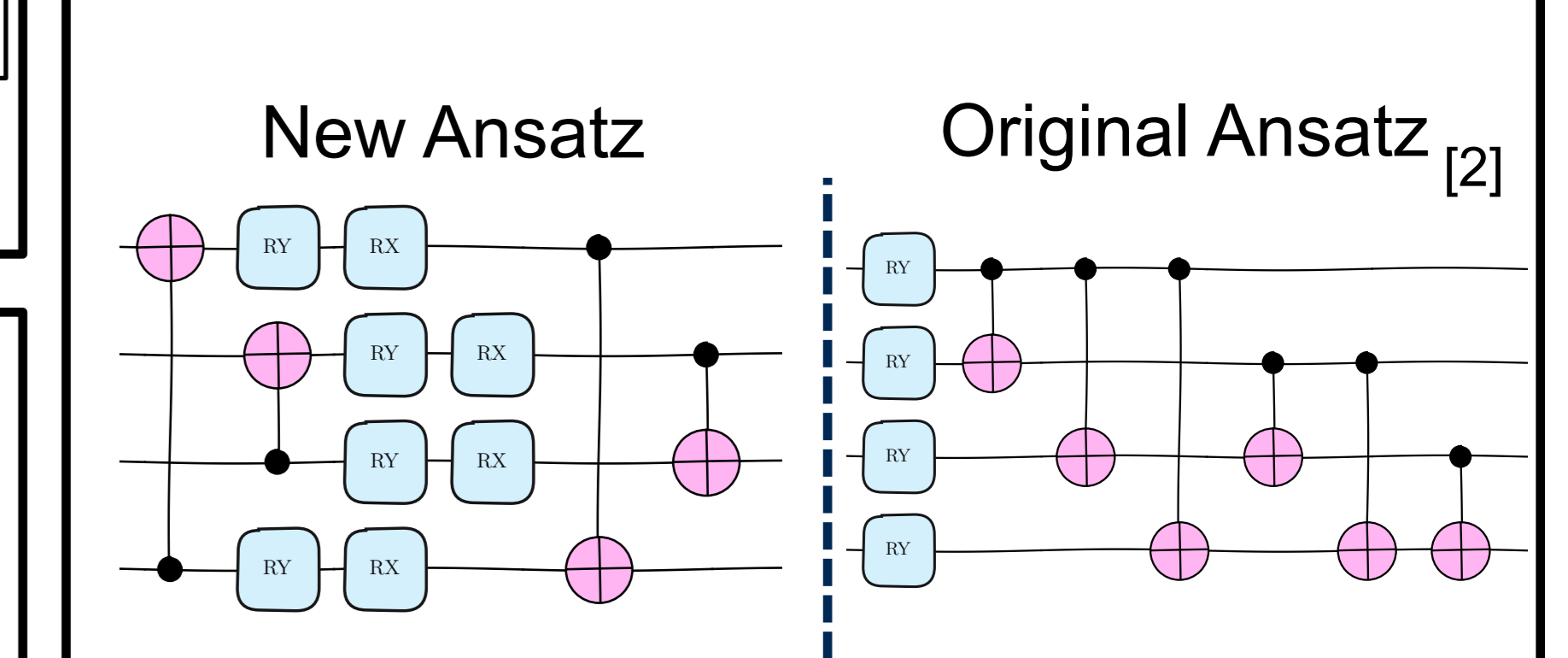
6. Quantum Autoencoder (QAE)

- Feature map $F(x)$ Ansatz $U(\theta)$ Latent Trash
- Ideal output state: $|\psi\rangle = |0\rangle^n \otimes |\phi\rangle$.
 - Cost function aims to maximize the number of zero states in the trash.

7. Compared to CAE, can QAE

- Learn with fewer samples?
- Make better predictions?
- Learn with fewer parameters?

8. New Ansatz Design



10. Conclusions and Outlook

- Potential for **better performance with fewer parameters** compared to CAE.
- Further analysis needed on a variety of classical autoencoders (pruned CAEs).
- Further study into the scaling of QAE with problem size and number of parameters.