

Classical-Quantum Algorithms for Anomaly Detection



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University of Tokyo



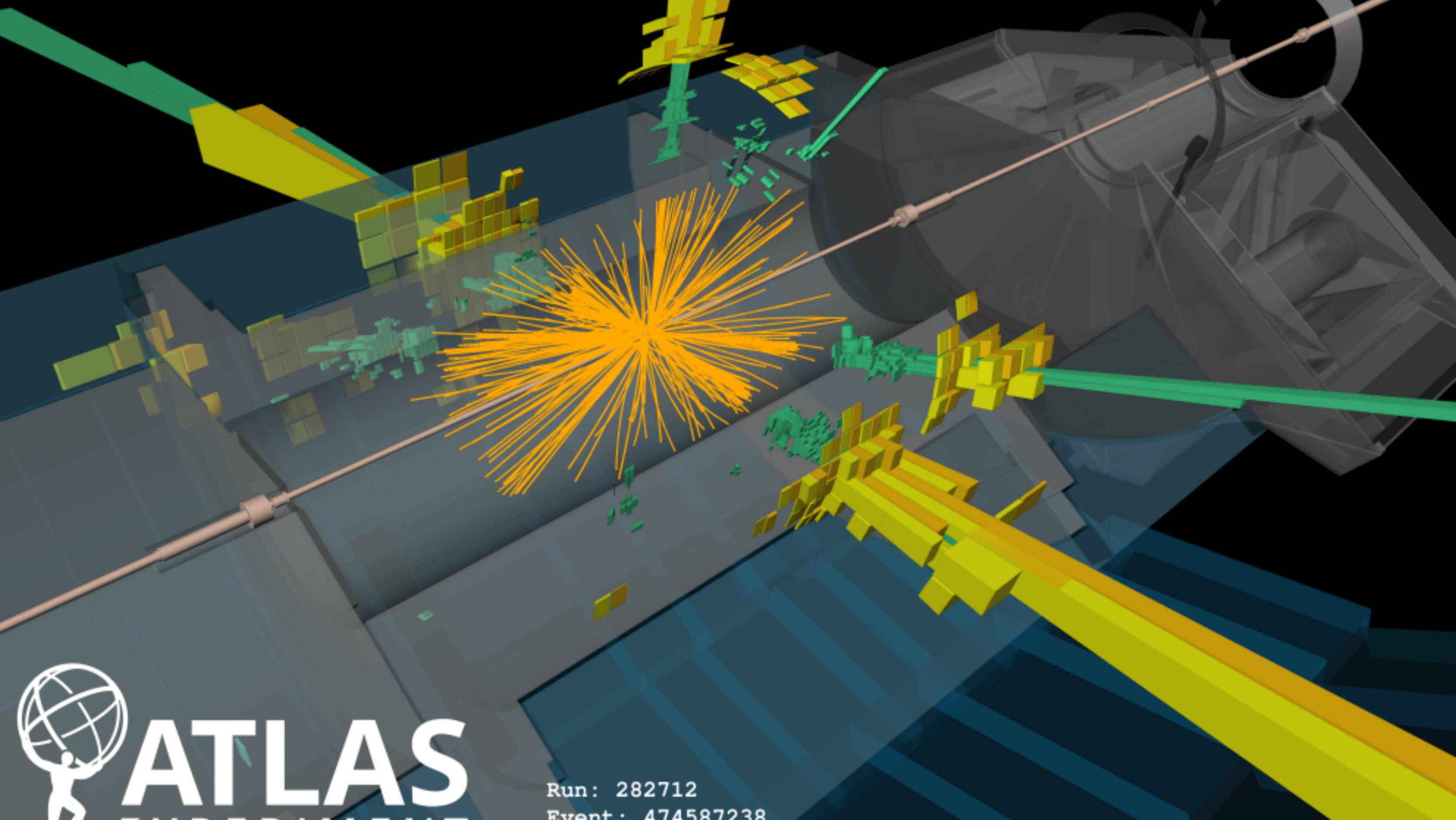
December 9, 2024 @ QUPosium, Tsukuba



World Premier International
Research Center Initiative



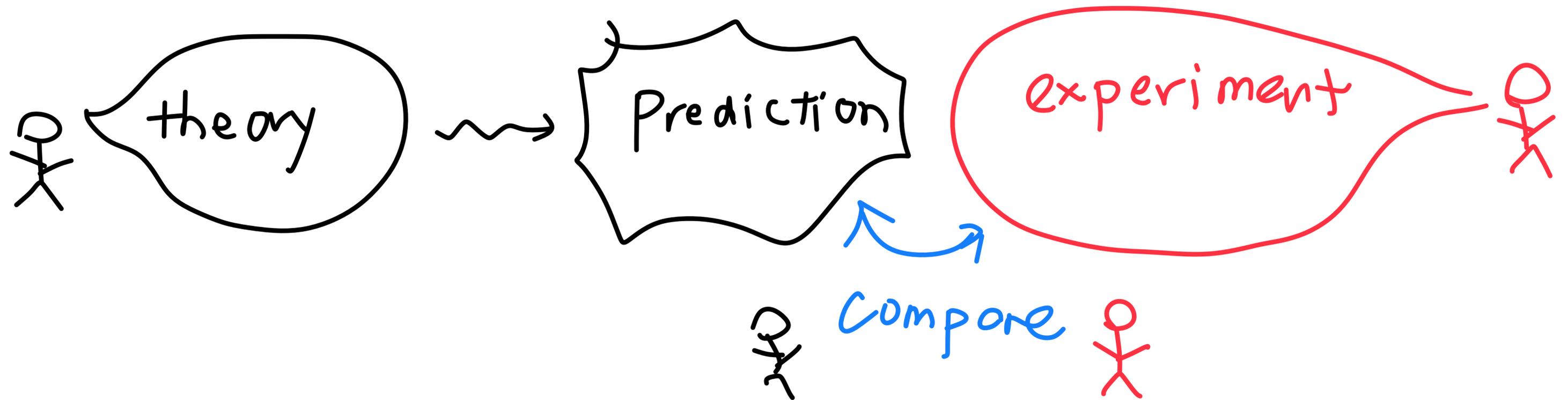
1. HEP



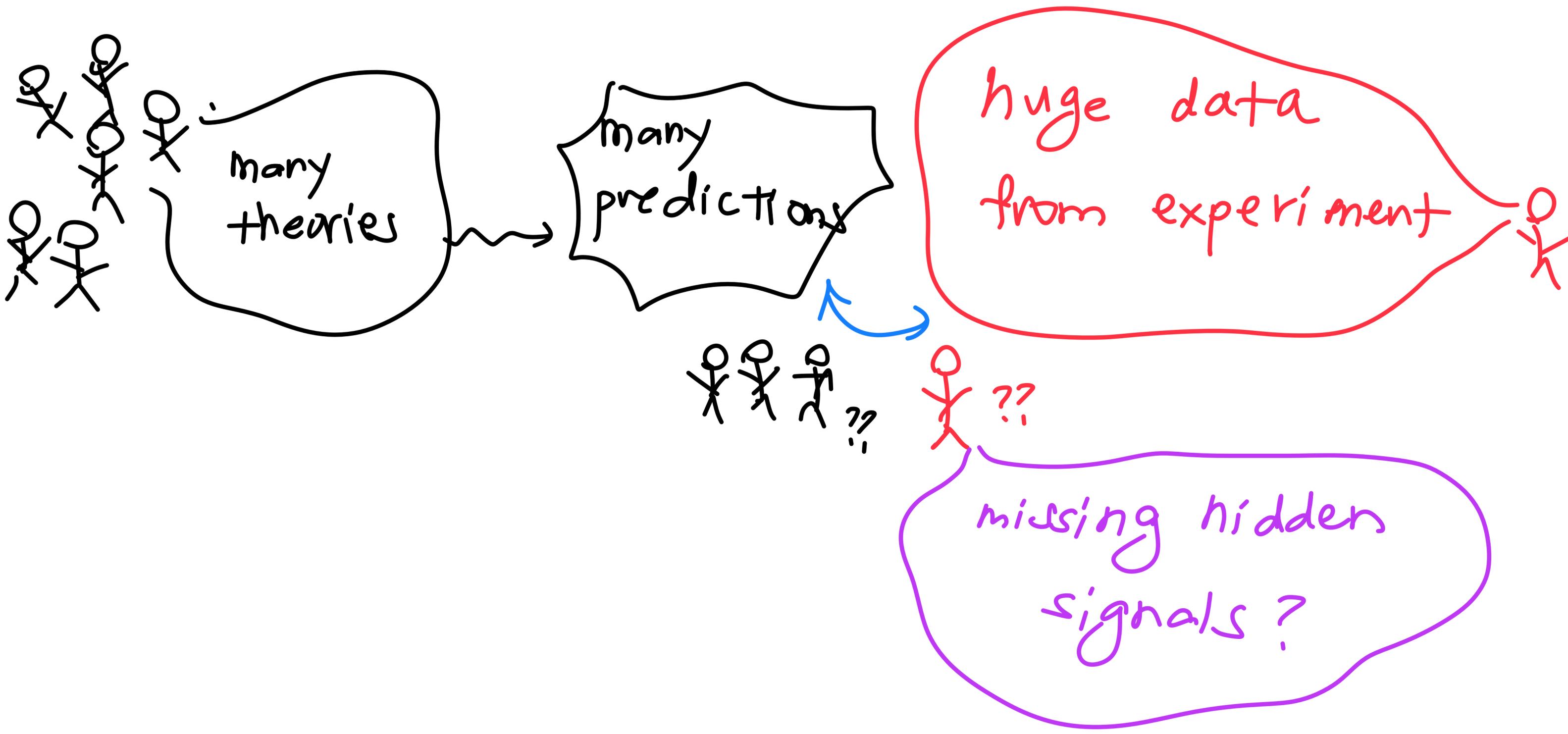
ATLAS

Run : 282712
Event : 474587238

Typical Strategy: Top-Down from Theory



Typical Strategy: Top-Down from Theory



⚠ ATLAS/CMS has implemented **model-agnostic searches**
(need to be careful with look-elsewhere effects)

⚠ Some proposal for density-based methods in phase space
(e.g. search for bumps)

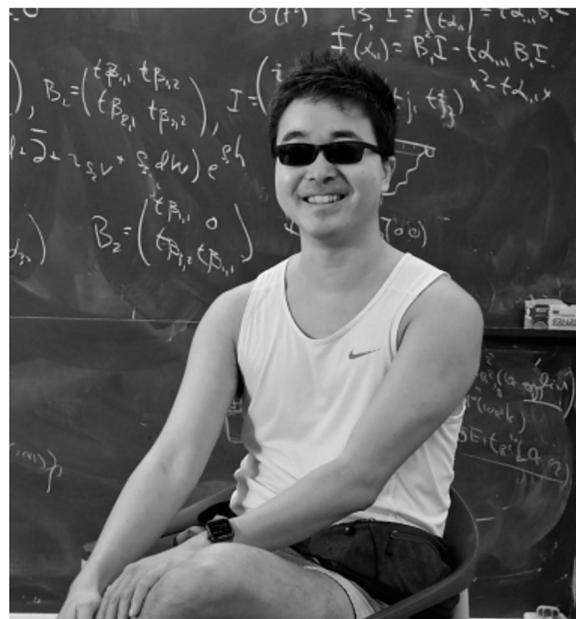
Heimel, Kasiieczka, Plehn, Thompson, 1808.08979

Farina, Nakai, Shih, 1808.08992

⚠ We discuss **similarity learning** (self-supervised ML)

Dillon, Favaro, Feiden, Modak, Plehn, 2301.04660

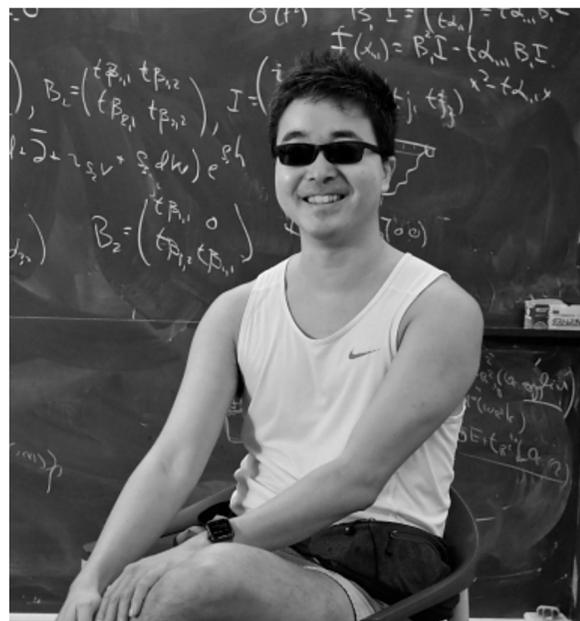
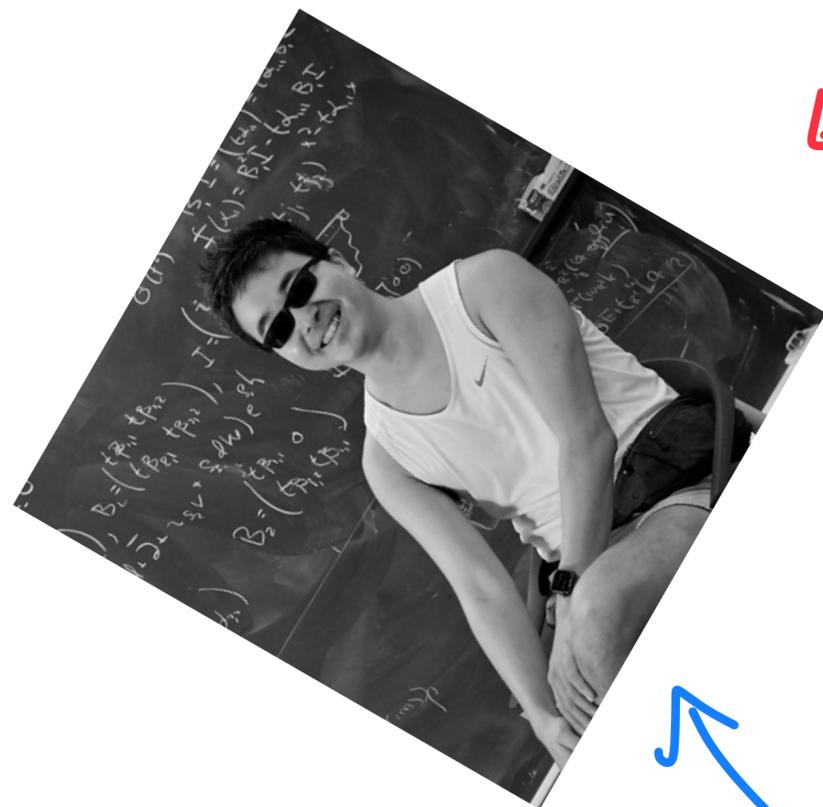
Similarity Learning



different

Similarity Learning

augmentation



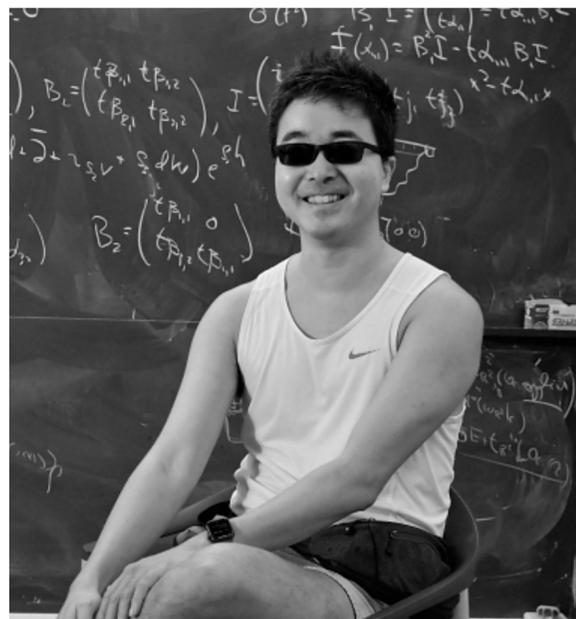
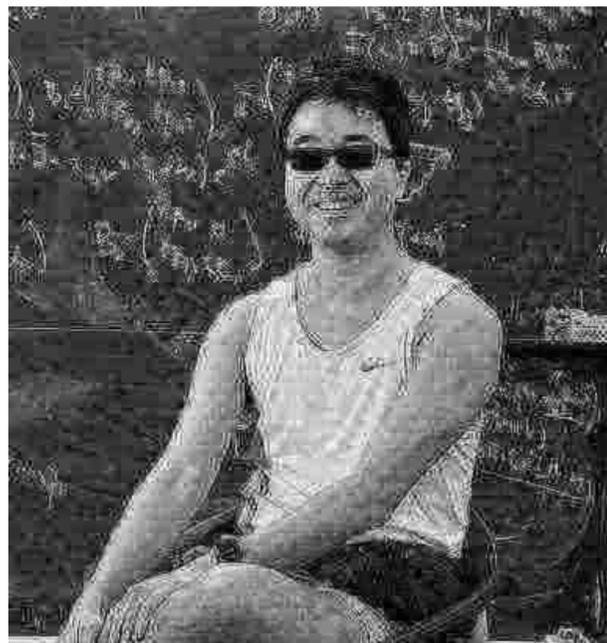
Similar



different

Similarity Learning

augmentation

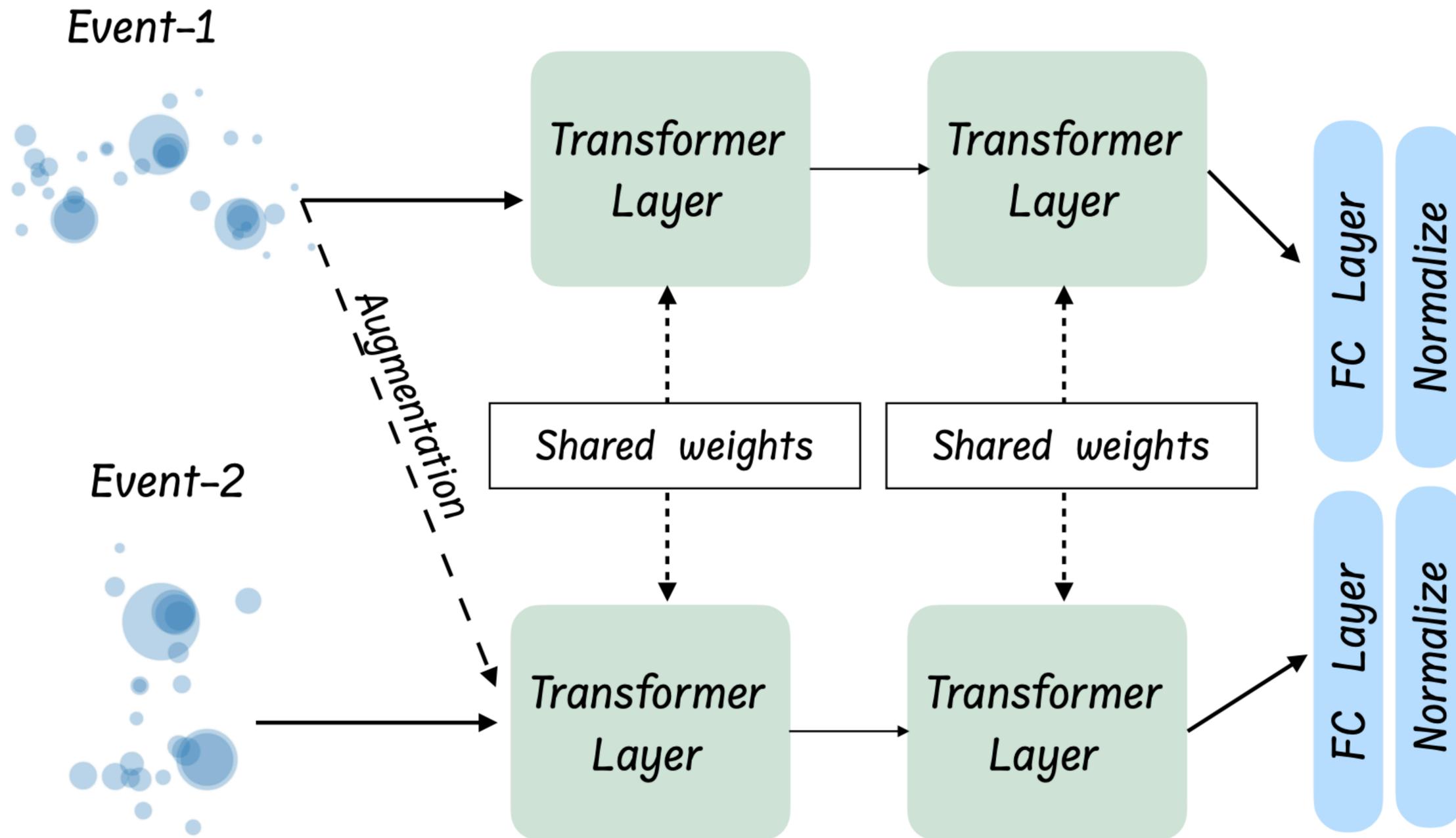


Similar



different

Similarity Learning with Transformers



2. Quantum x HEP

My interest: **Quantum Algorithms**

for **HEP Experiment Data Analysis?**

(e.g. anomaly detection)

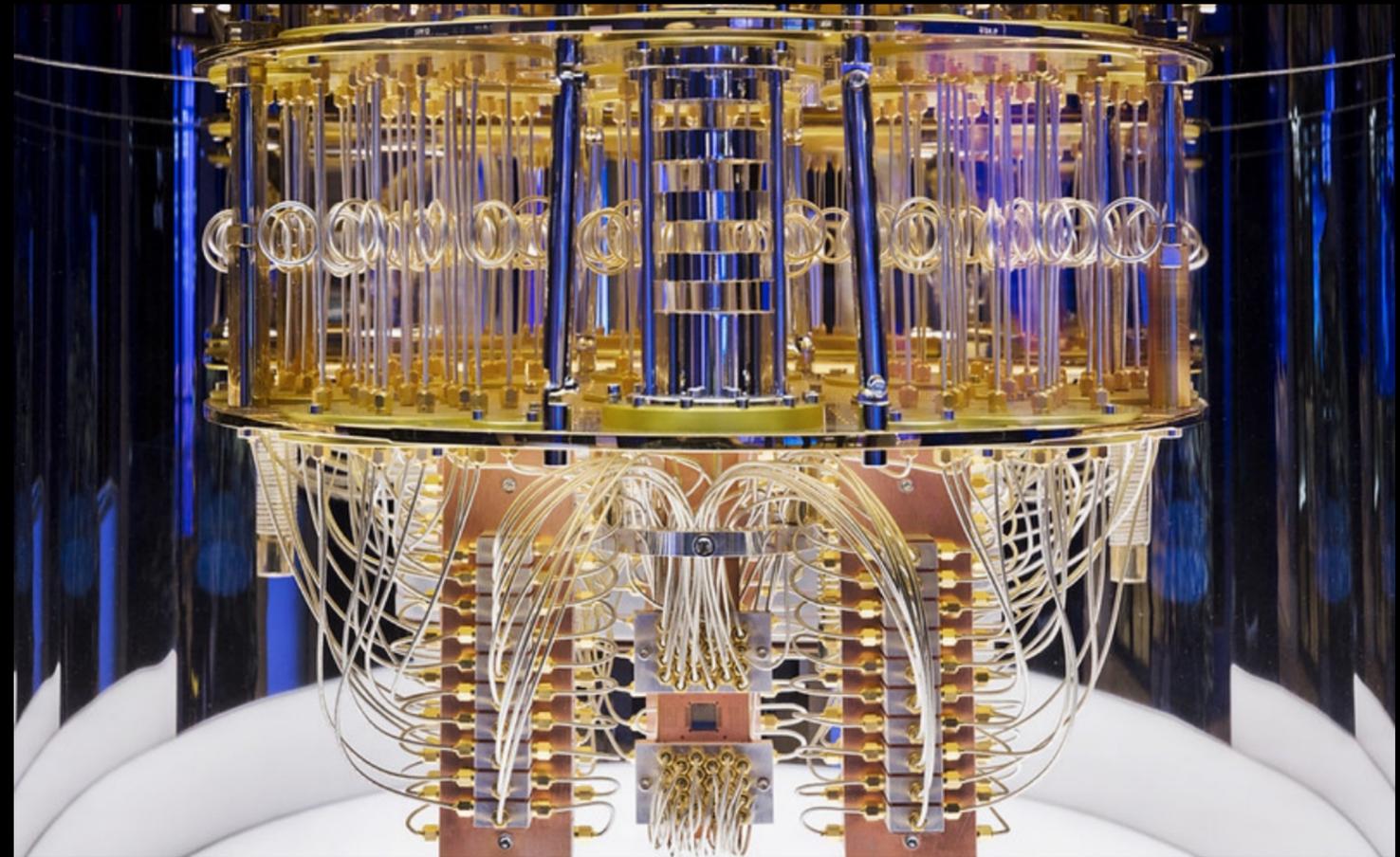
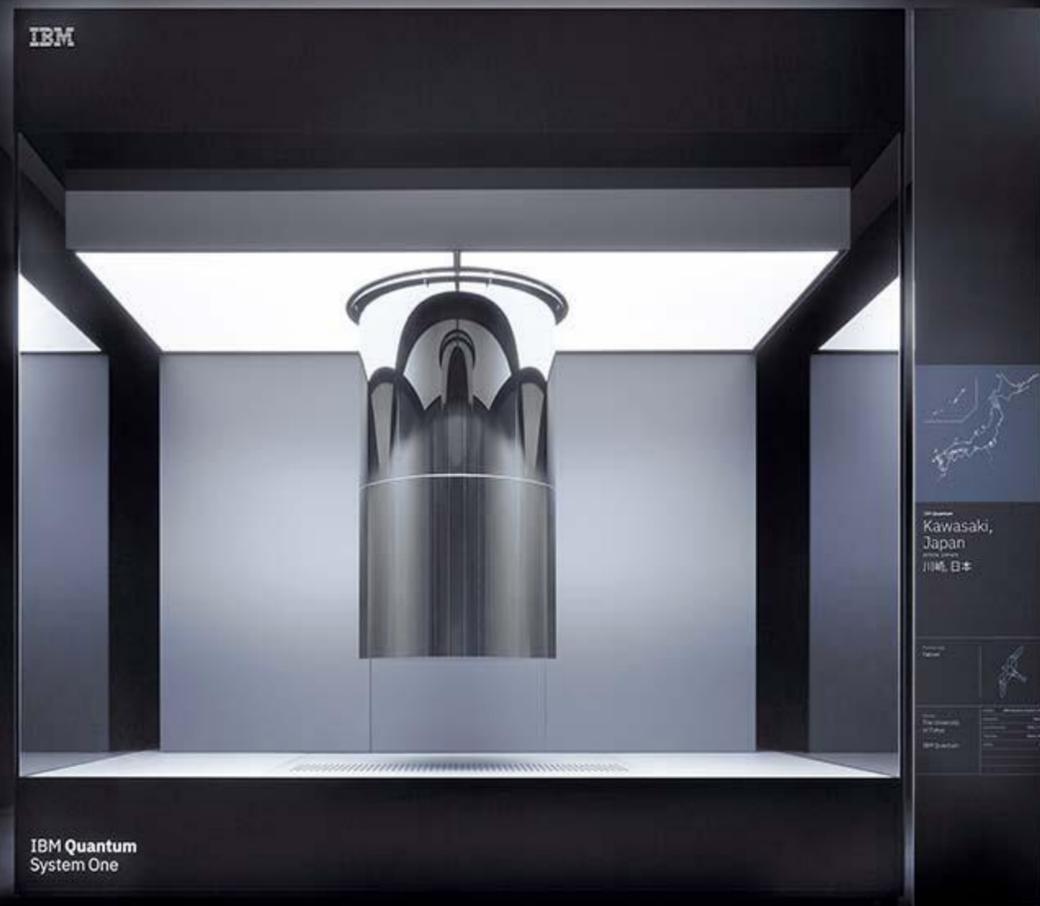


+



= ??

Quantum Computers have Arrived



127 qubit machine @ Kawasaki, Japan

☆ Quantum computers today are **noisy**

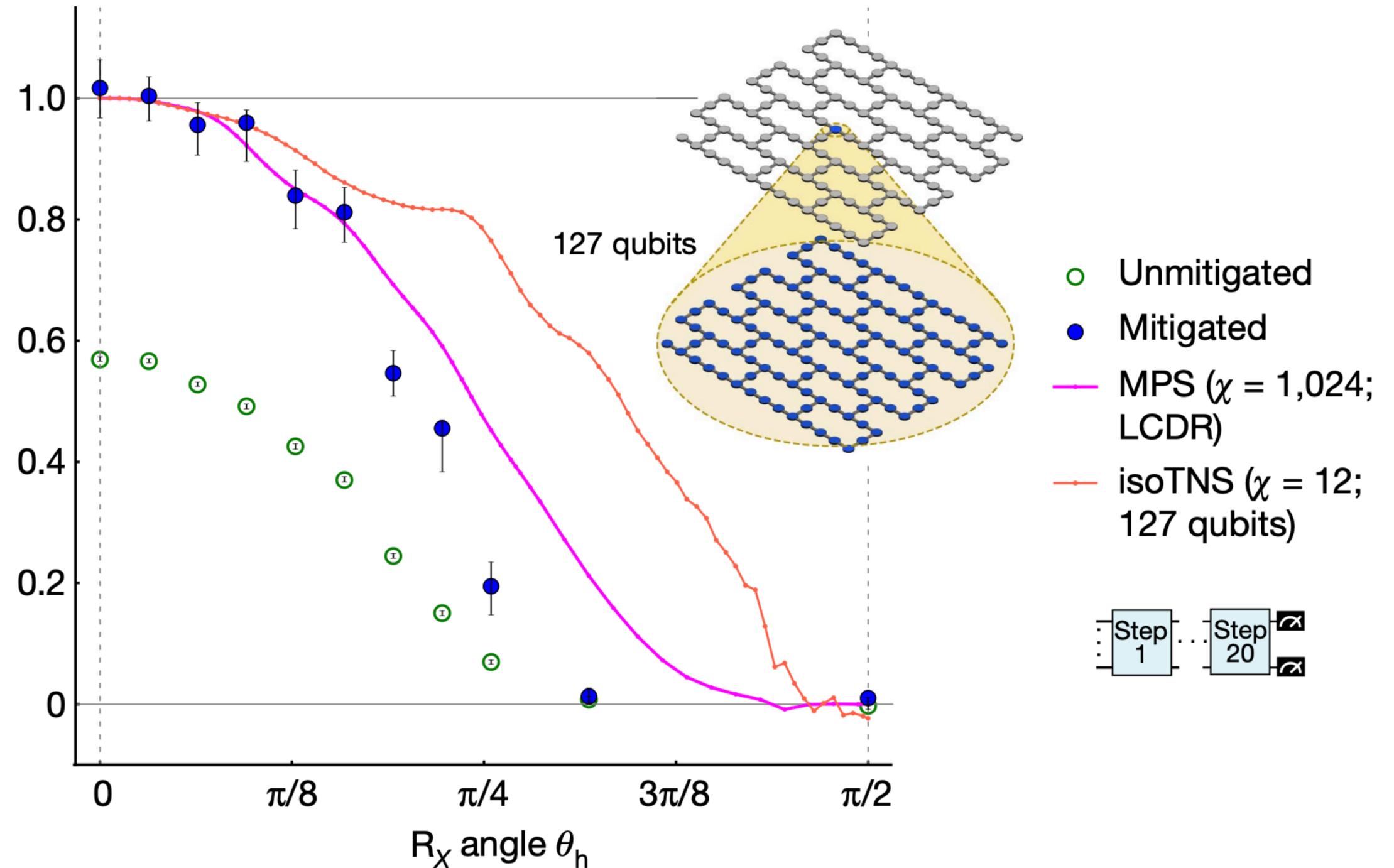
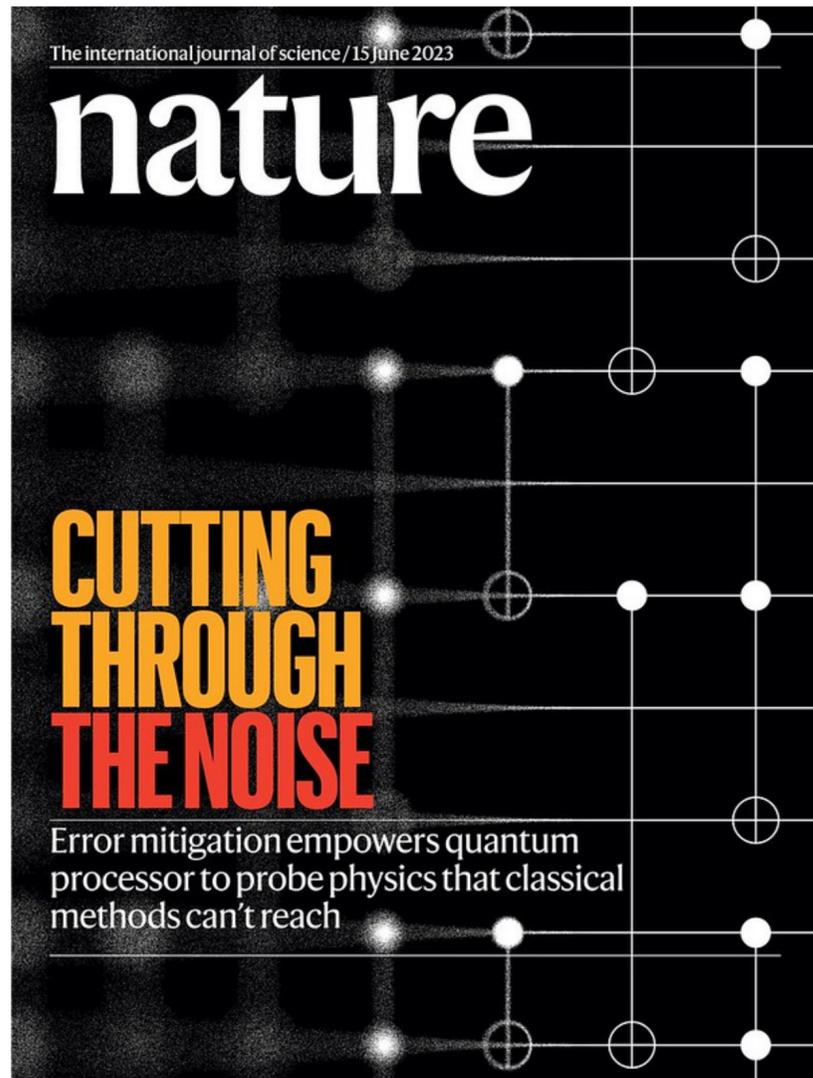


→ We need “**error mitigation**” by classical computers,
at least before the era of full error correction

☆ One should not expect “practical” results too easily

(classical AI/ML already good)

The Era of “Quantum Utility”



“Quantum” as a State-of-Art

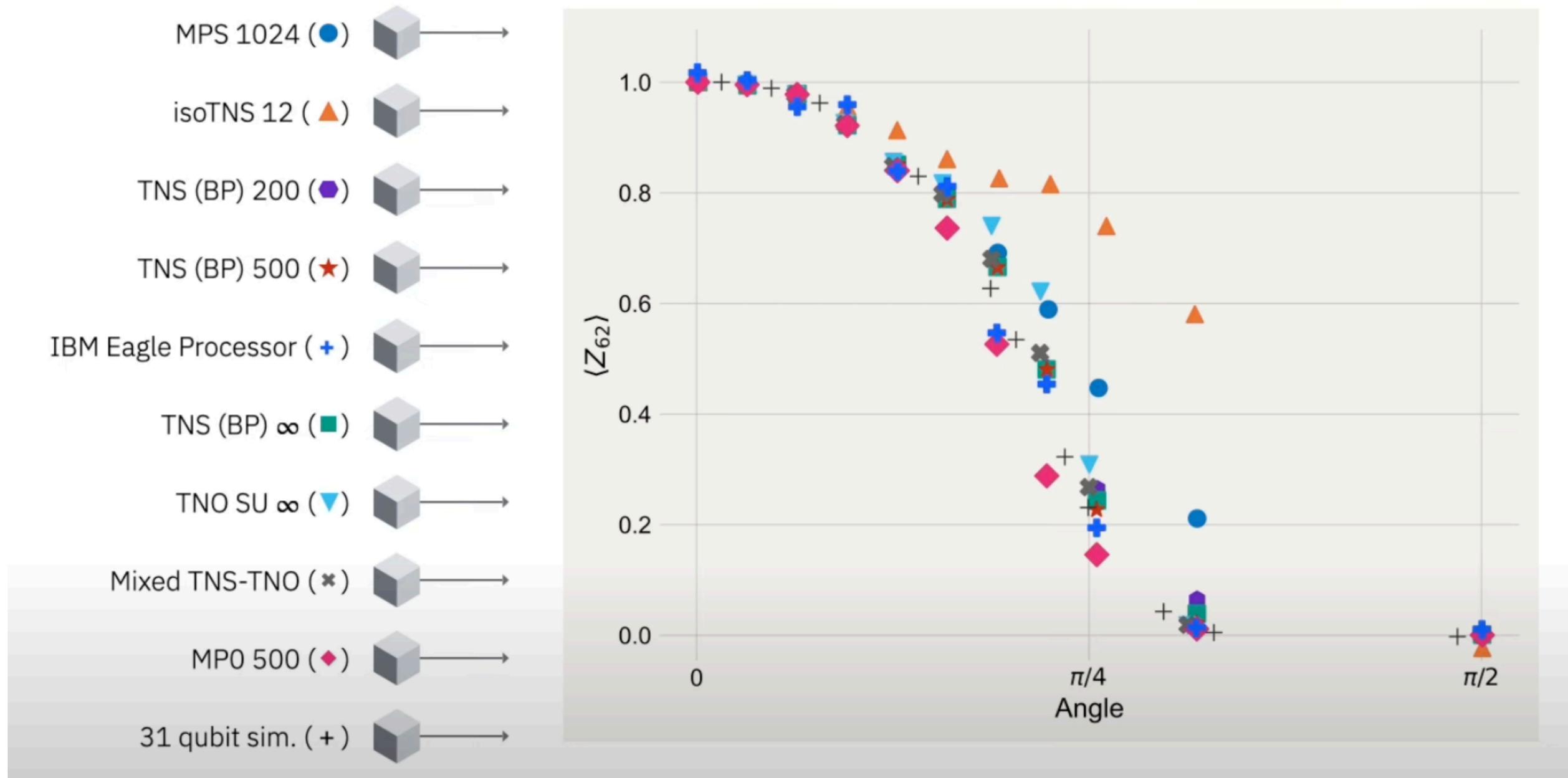
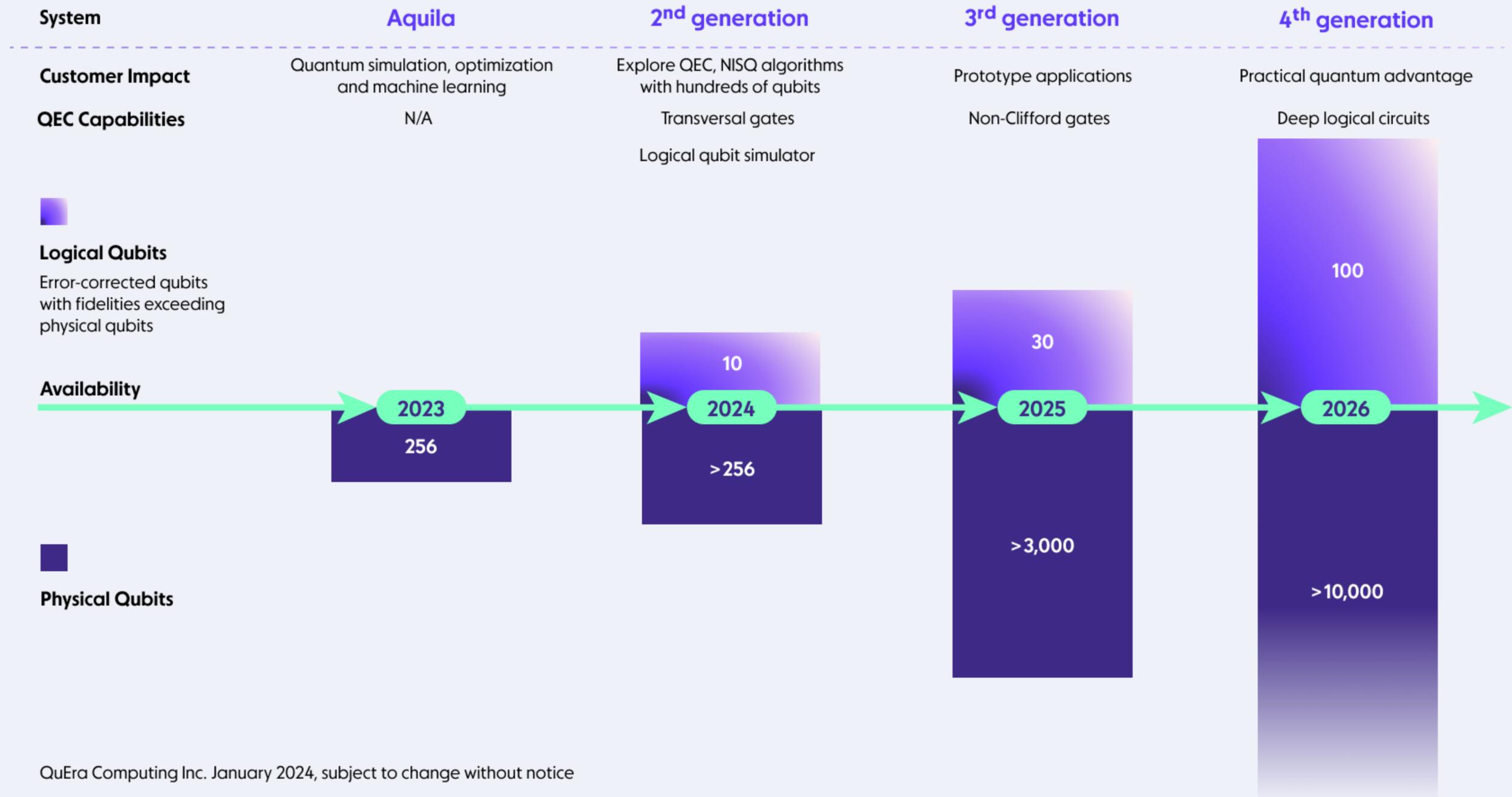


Figure from IBM QCD24 conference

Rapid Progress Expected



Error-Corrected Quantum Computing Roadmap



✧ Quantum computers will NOT replace
classical computers anytime soon



→ we should aim for **quantum-classical hybrid**

✧ We can try to **enhance classical AI/ML by**

quantum algorithms (quantum-classical hybrid)

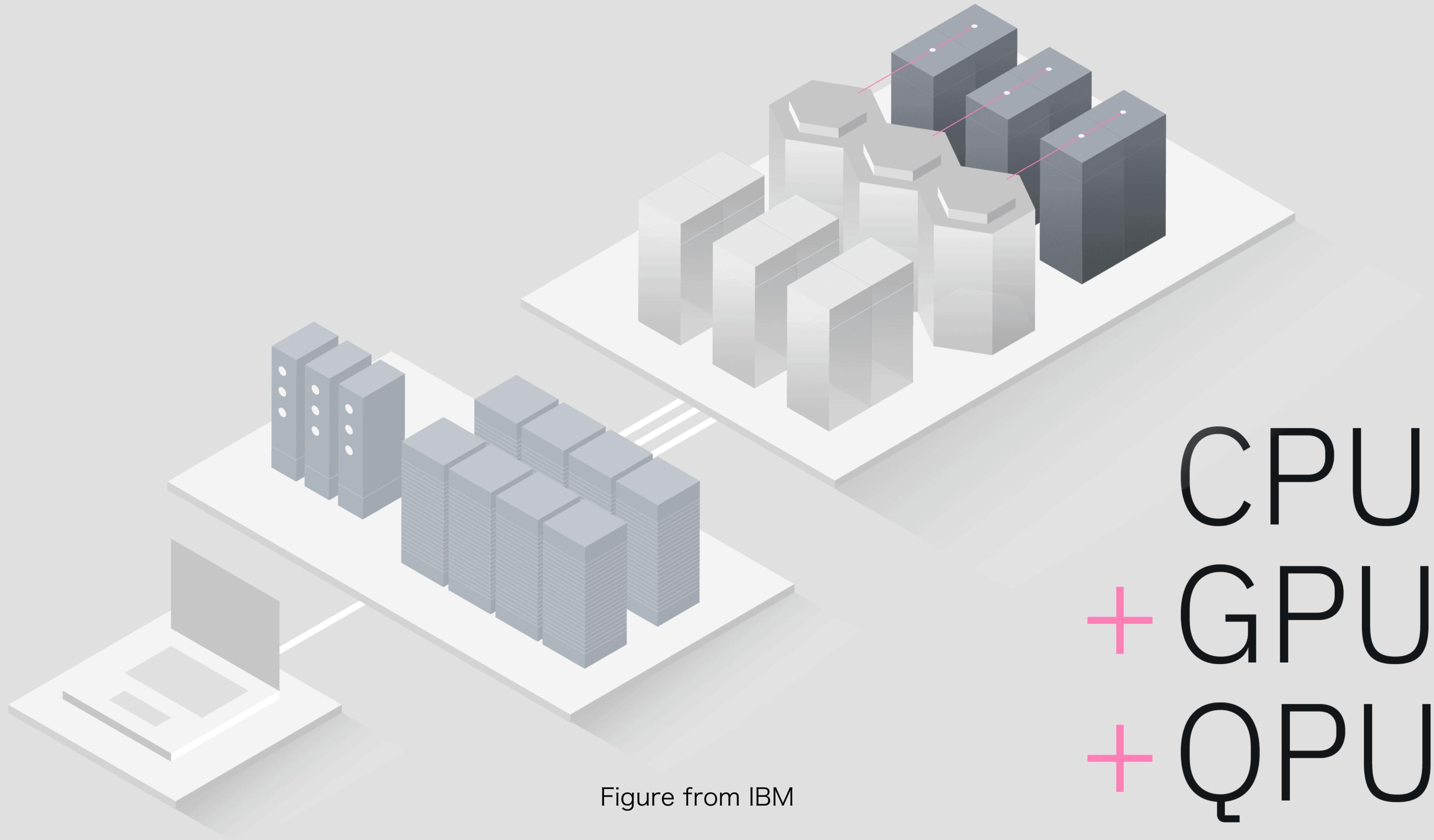


Figure from IBM

3. Classical/Quantum Similarity Learning

“Quantum Similarity Learning for Anomaly Detection”

arXiv: 2411.09927 [hep-ph]

collaborators

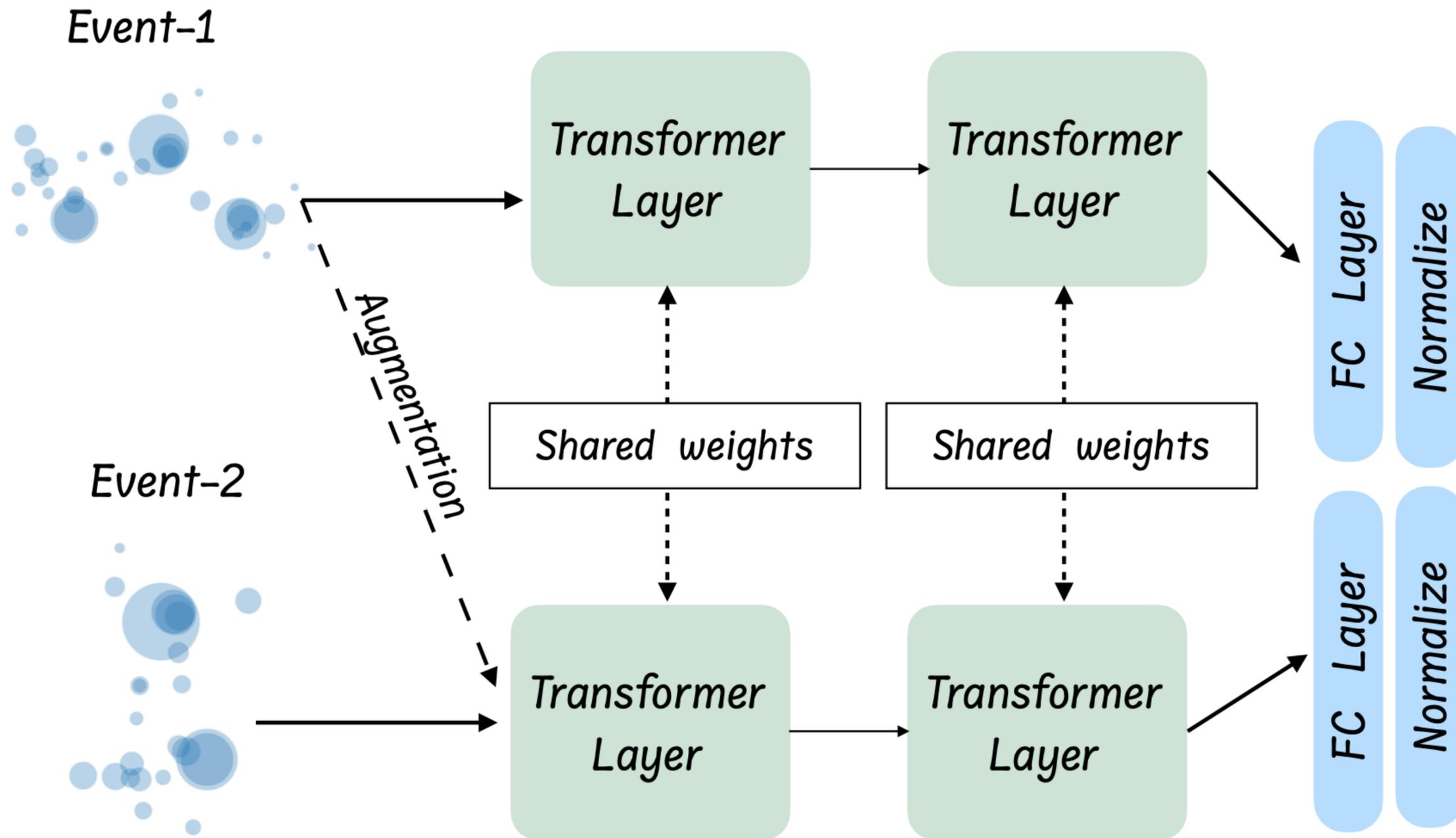


A. Hammad (KEK)

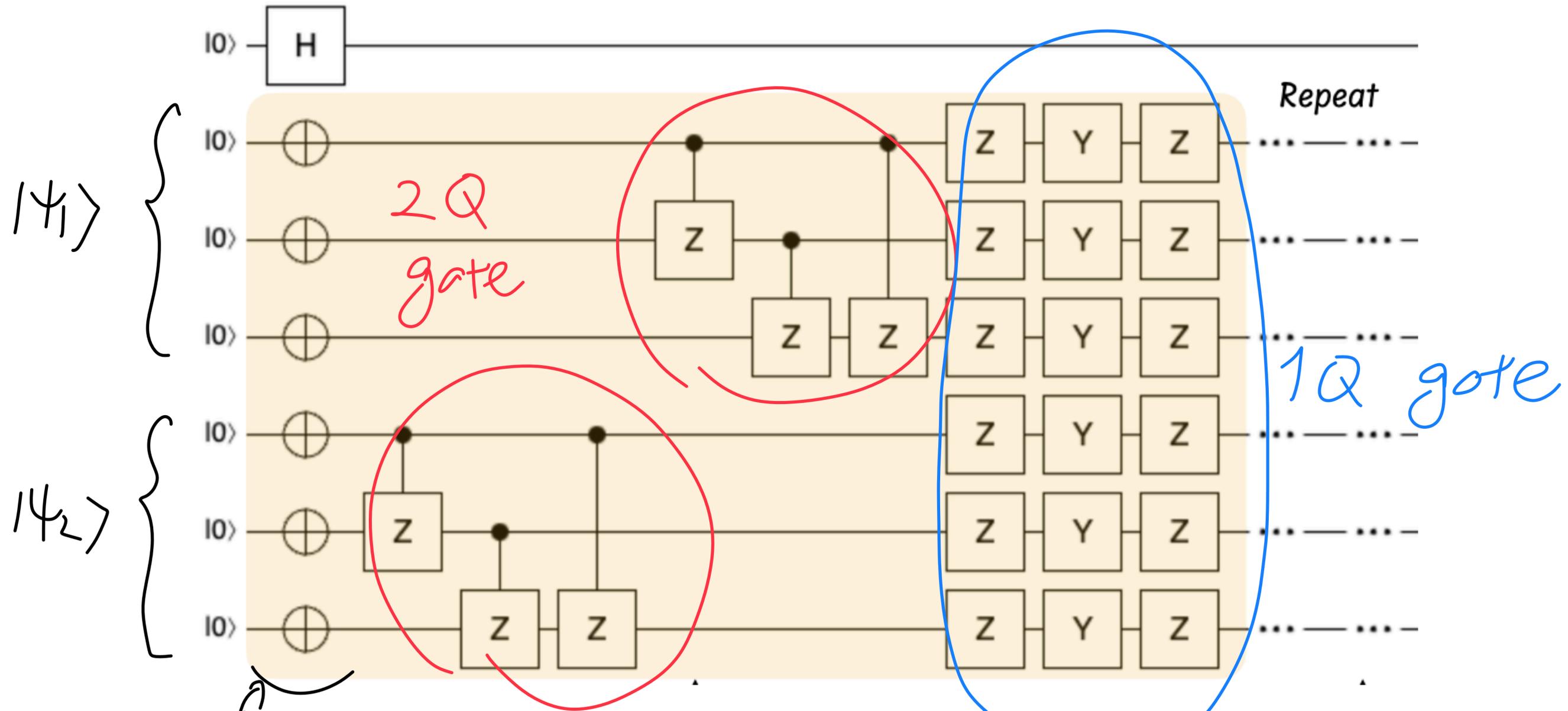


M. Nojiri (KEK)

Classical Part (Re)

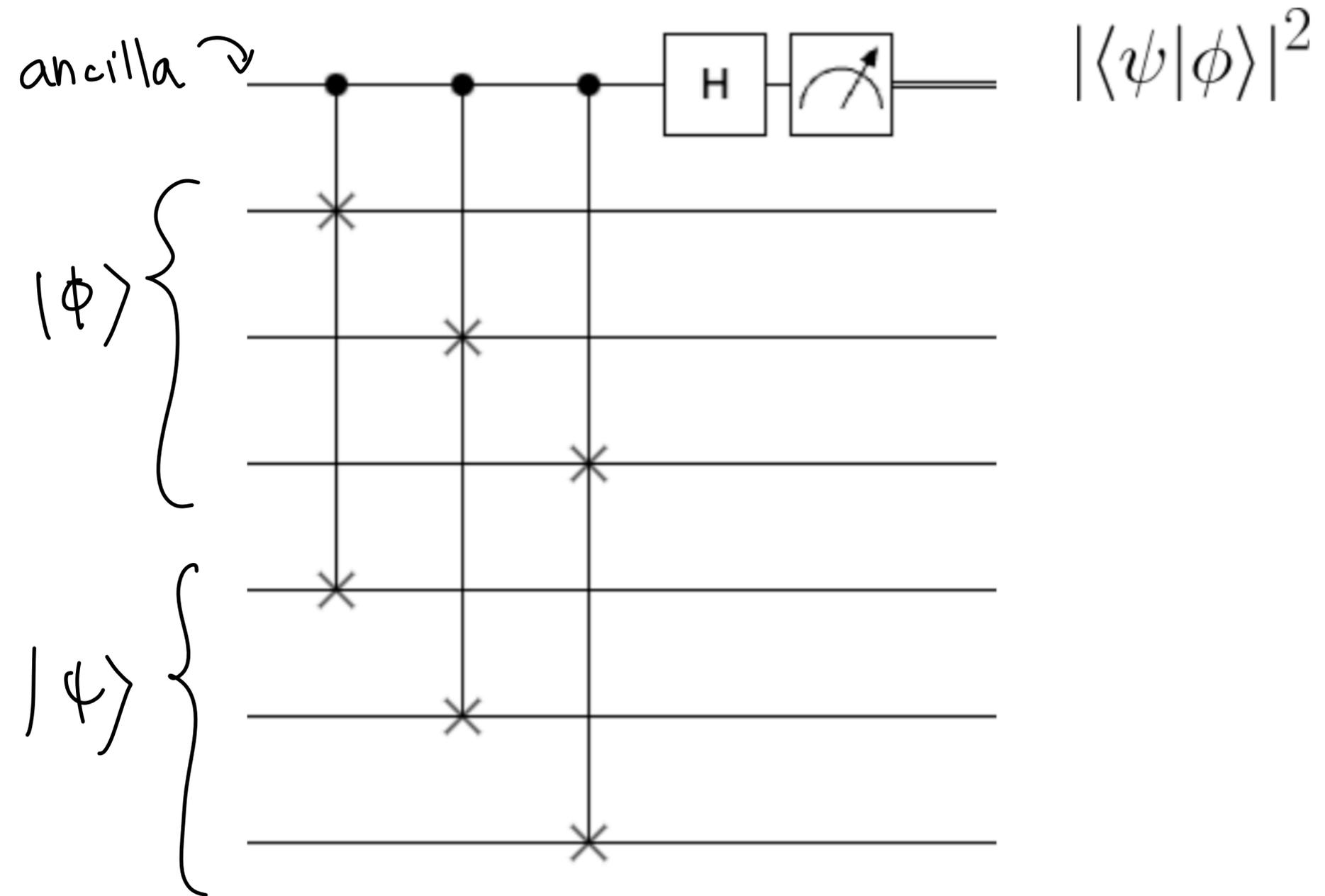


Variational Encoding Layers

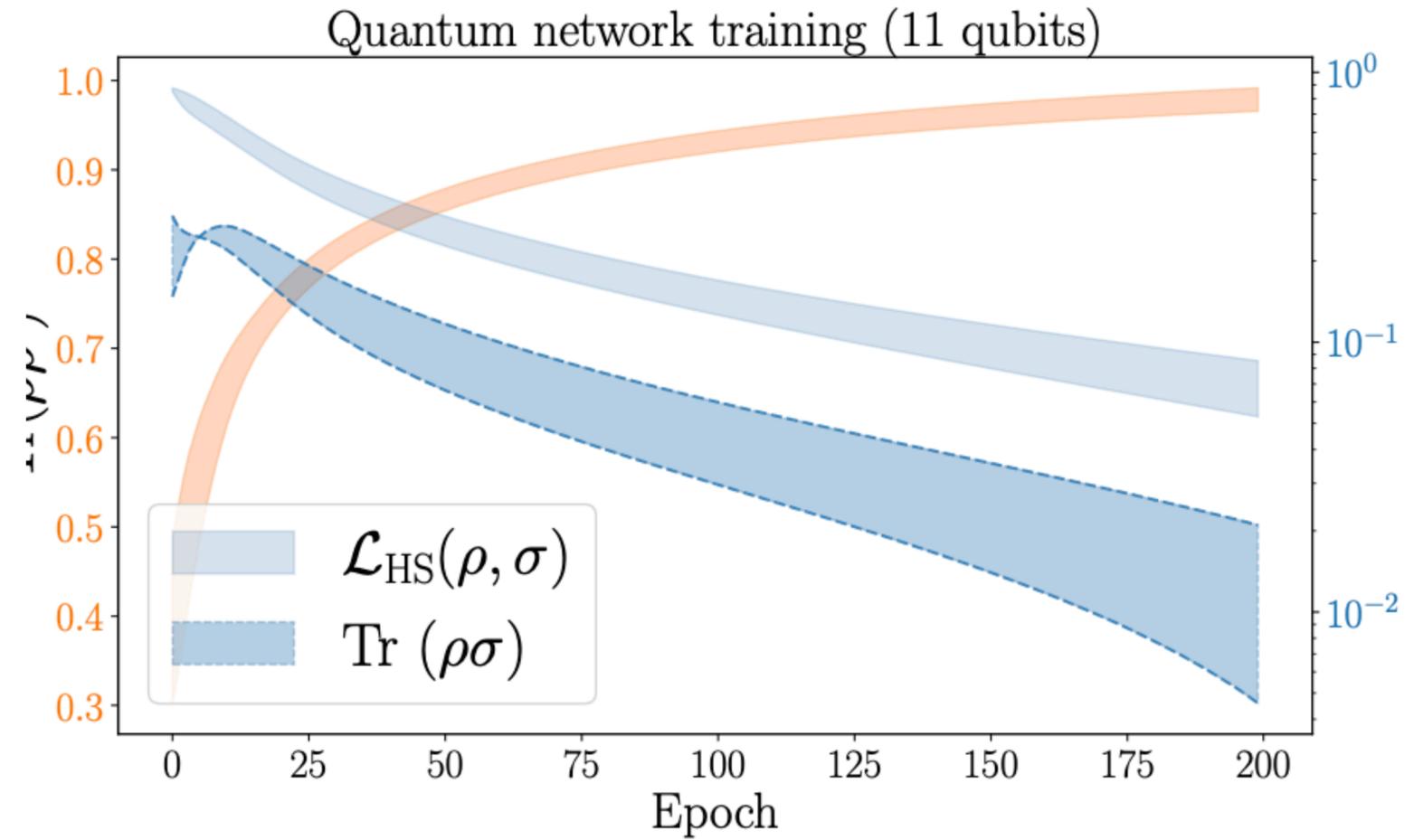
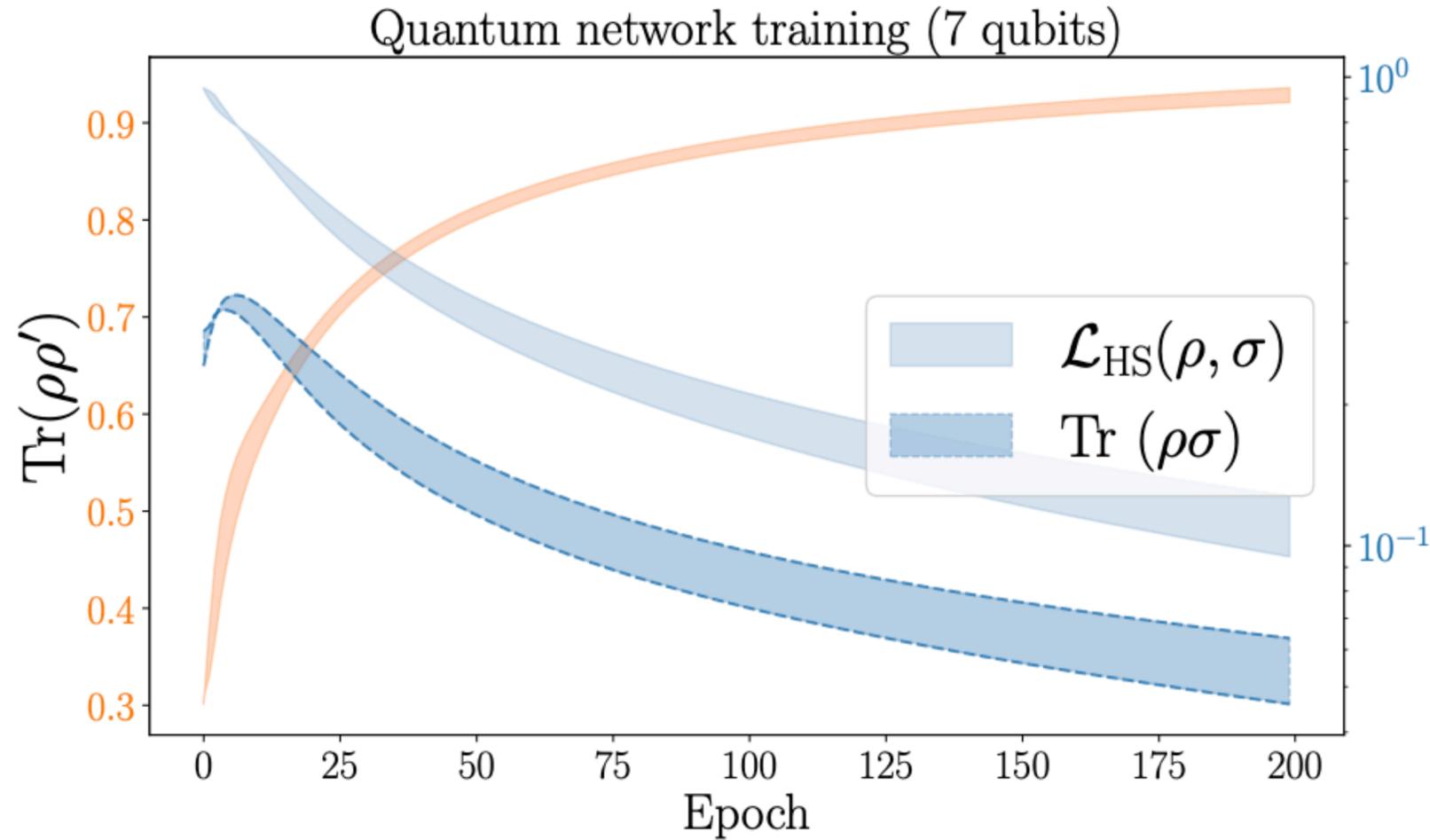


$$|\psi(\theta, x)\rangle = \left(\prod_i R_z(x_i + \theta_i) R_y(x_j + \theta_j) R_z(x_k + \theta_k) \right) |0\rangle^{\otimes n}$$

SWAP Test

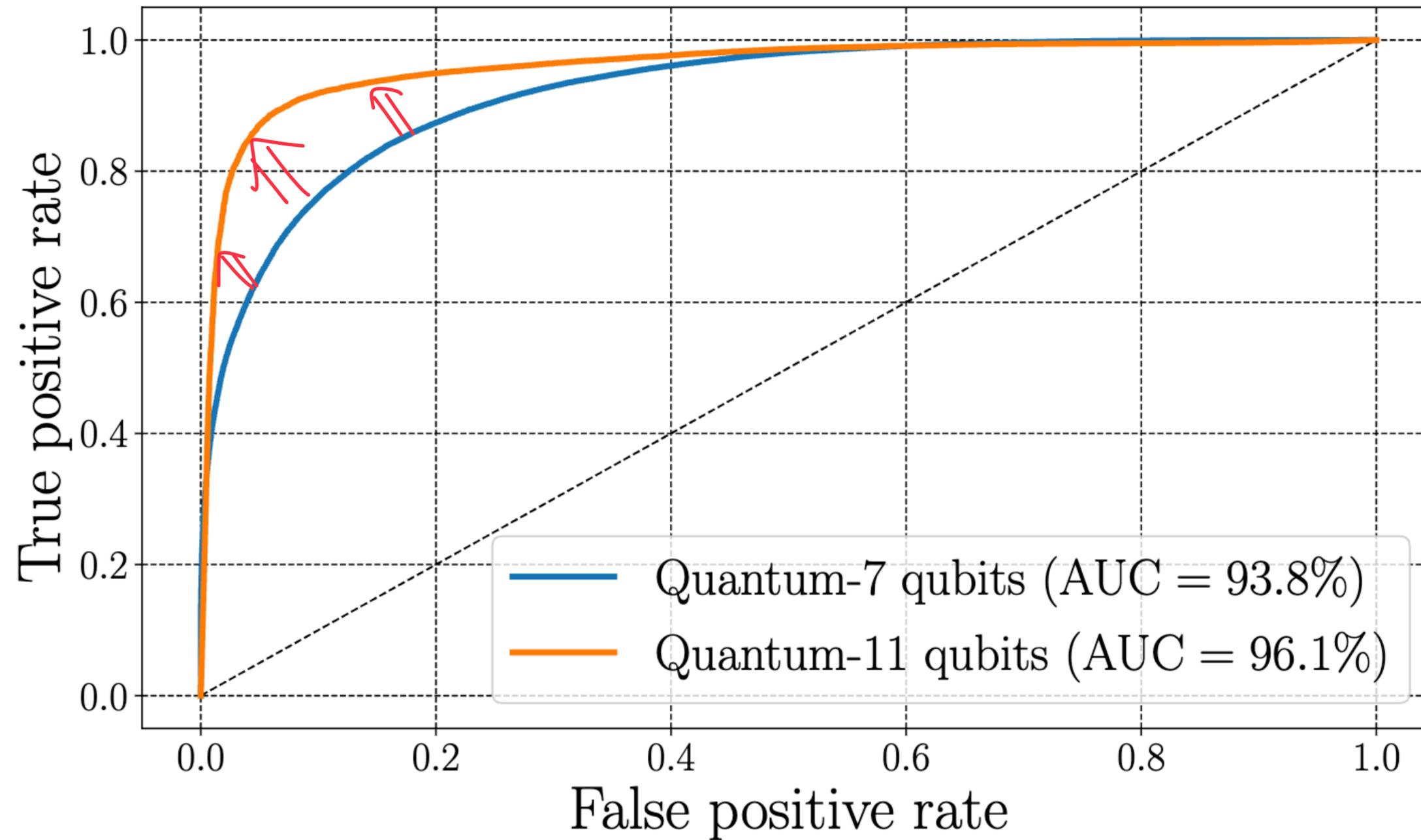


Training Metrics



ρ' ← augment $\rho = \frac{1}{M} \sum_{i \in \text{batch}} |\psi^1\rangle\langle\psi^1|$ and $\sigma = \frac{1}{M} \sum_{j \neq i \in \text{batch}} |\psi^2\rangle\langle\psi^2|$

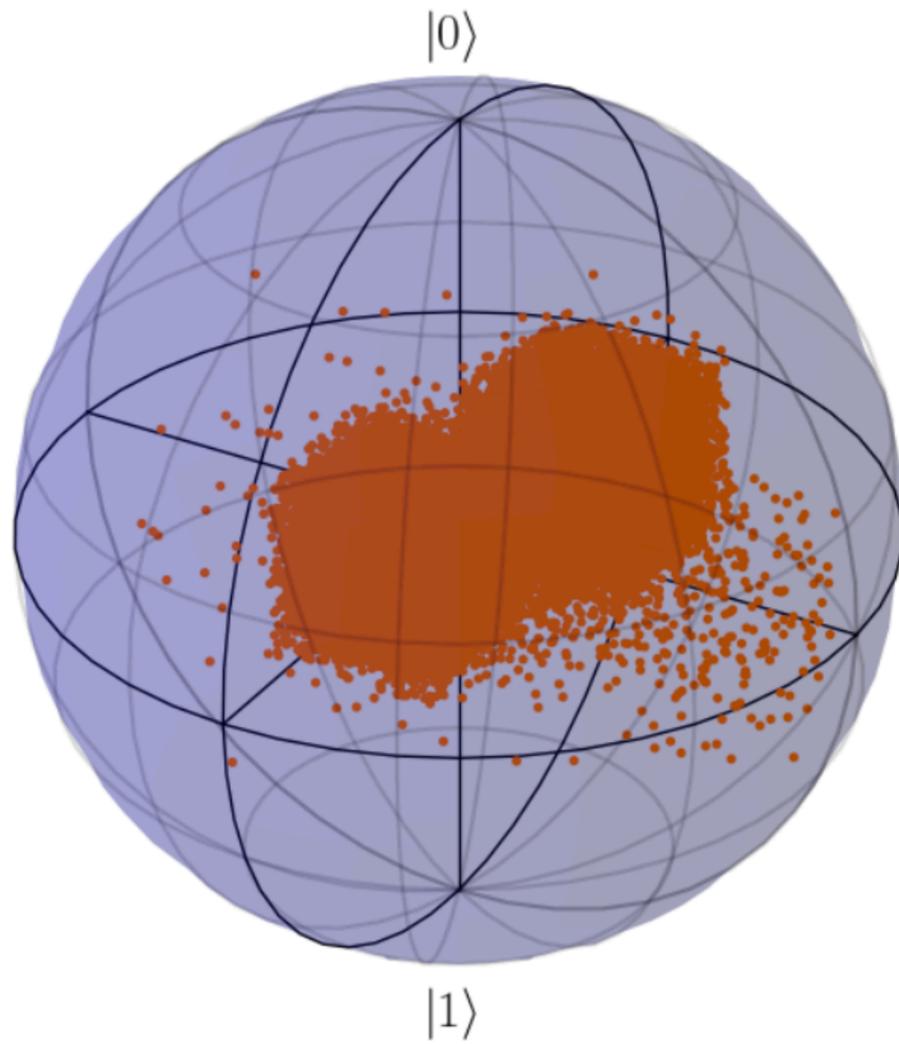
Receiver Operating Characteristics (ROC) Curve



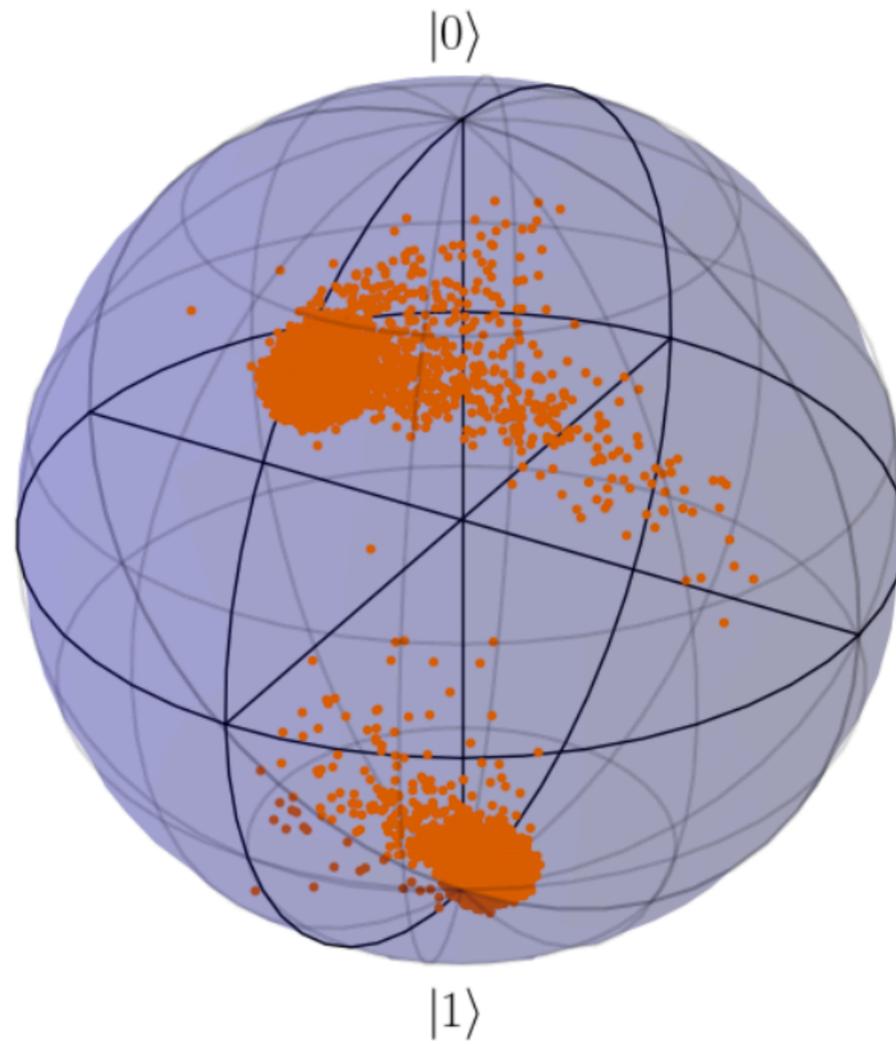
Our results are on classical simulators,
but real quantum devices today are **noisy** ...

Example of Noise: Shot Noise

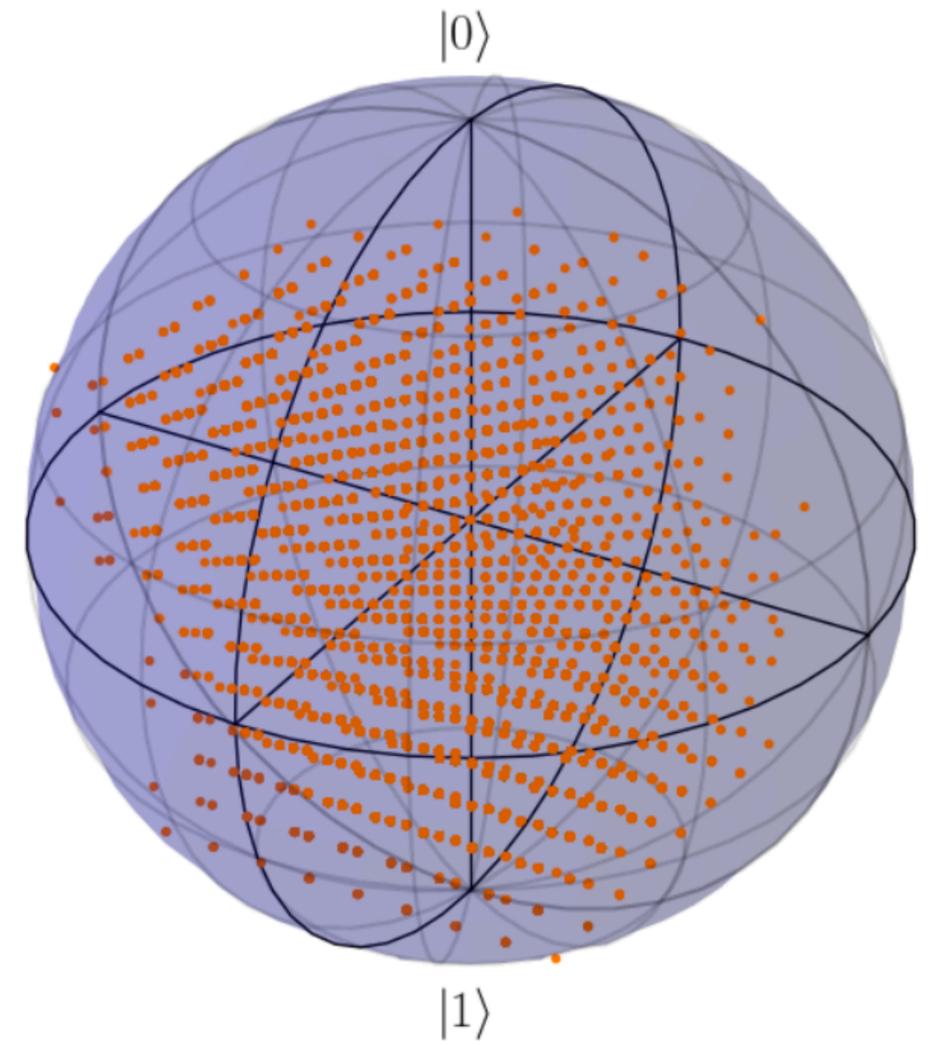
Epoch 0 with 10k shots



Epoch 200 with 10k shots



Epoch 200 with 10 shots



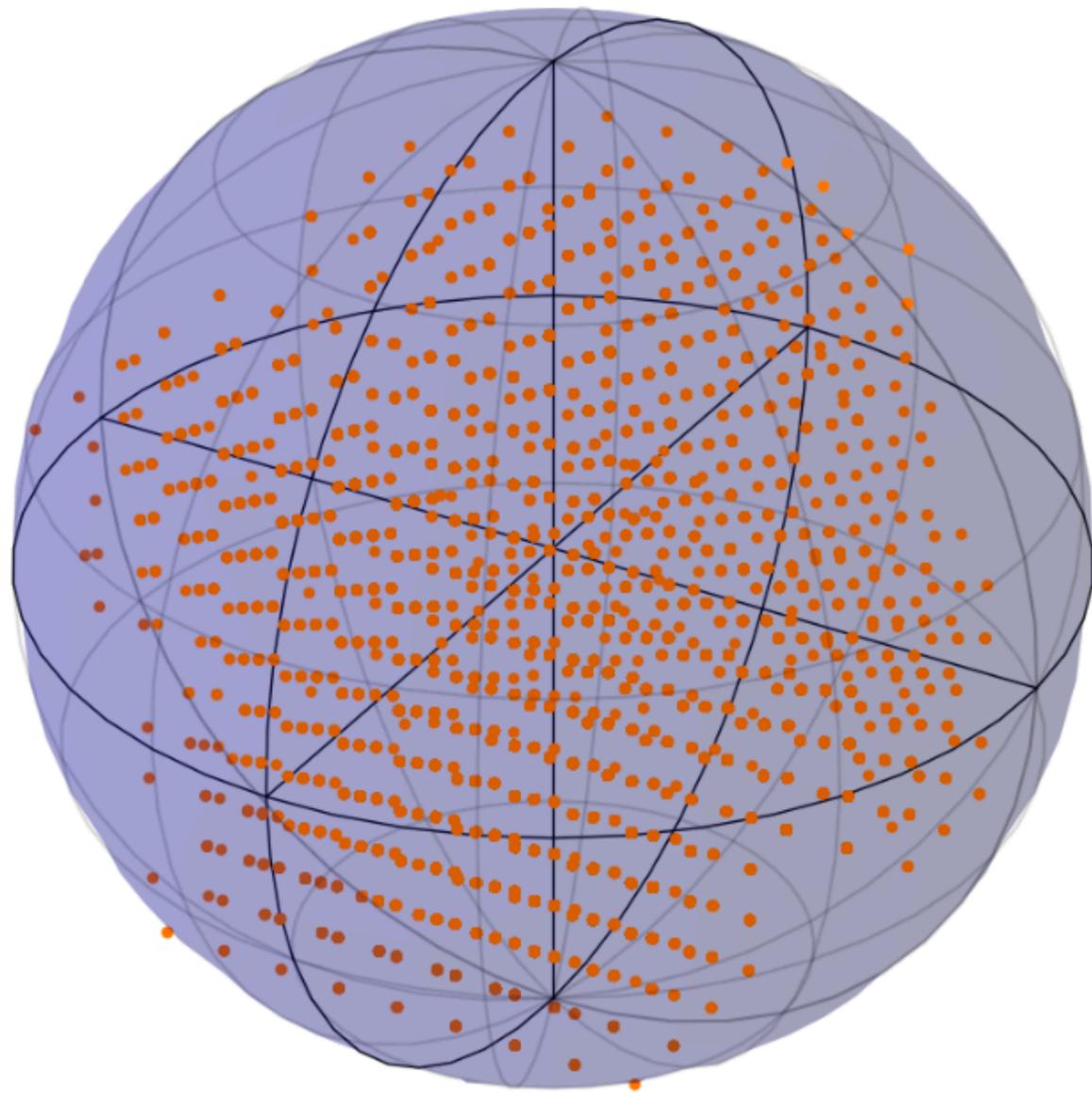
20,000 training events

Utility of Clustering

e.g. HDBSCAN = Hierarchical Density-Based Spatial Clustering of Applications with Noise

Before HDBSCAN

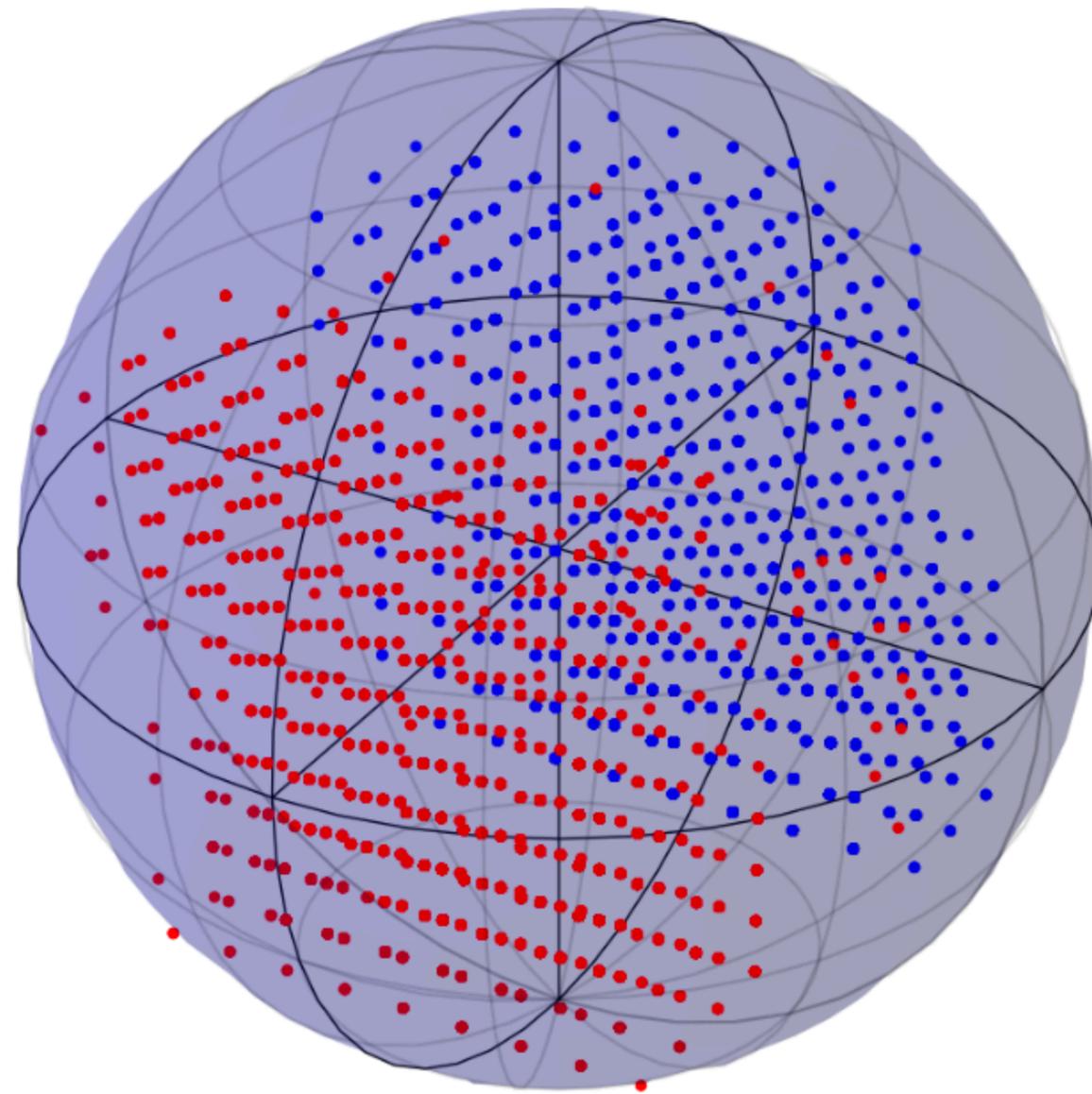
$|0\rangle$



$|1\rangle$

After HDBSCAN

$|0\rangle$



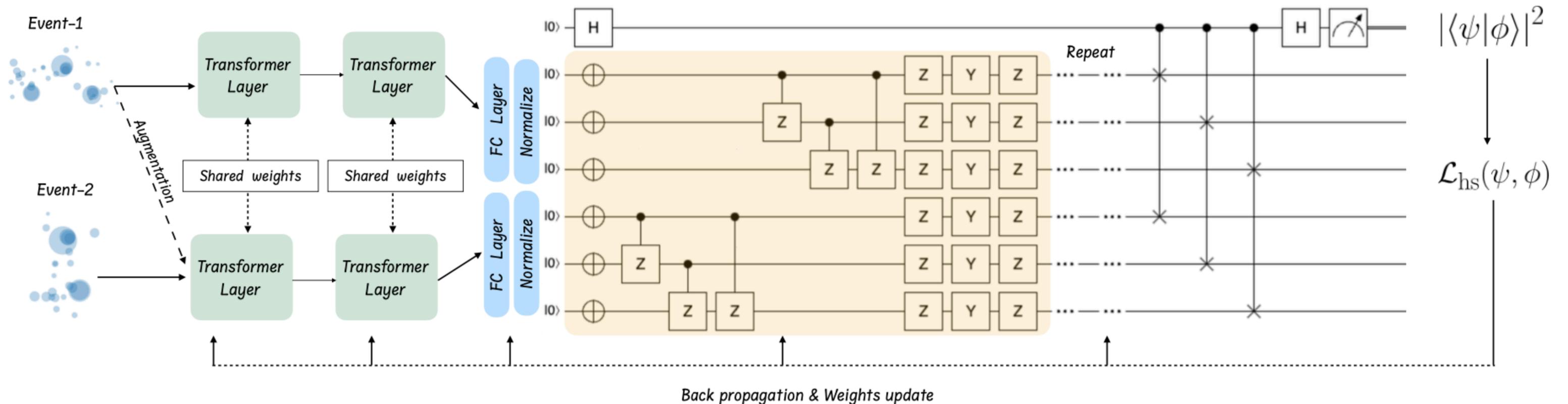
$|1\rangle$

Summary

🔗 **QPU** will be indispensable for future HPC

🔗 We discussed applications of **classical-quantum hybrid algorithm**

for **anomaly detection in colliders**



Summary

