

Comparing Quantum Encoding Methods in a Hybrid Deep Neural Network

Nidhi Munikote, Dr. Ang Li, Sam Stein, Dr. Chenxu Liu

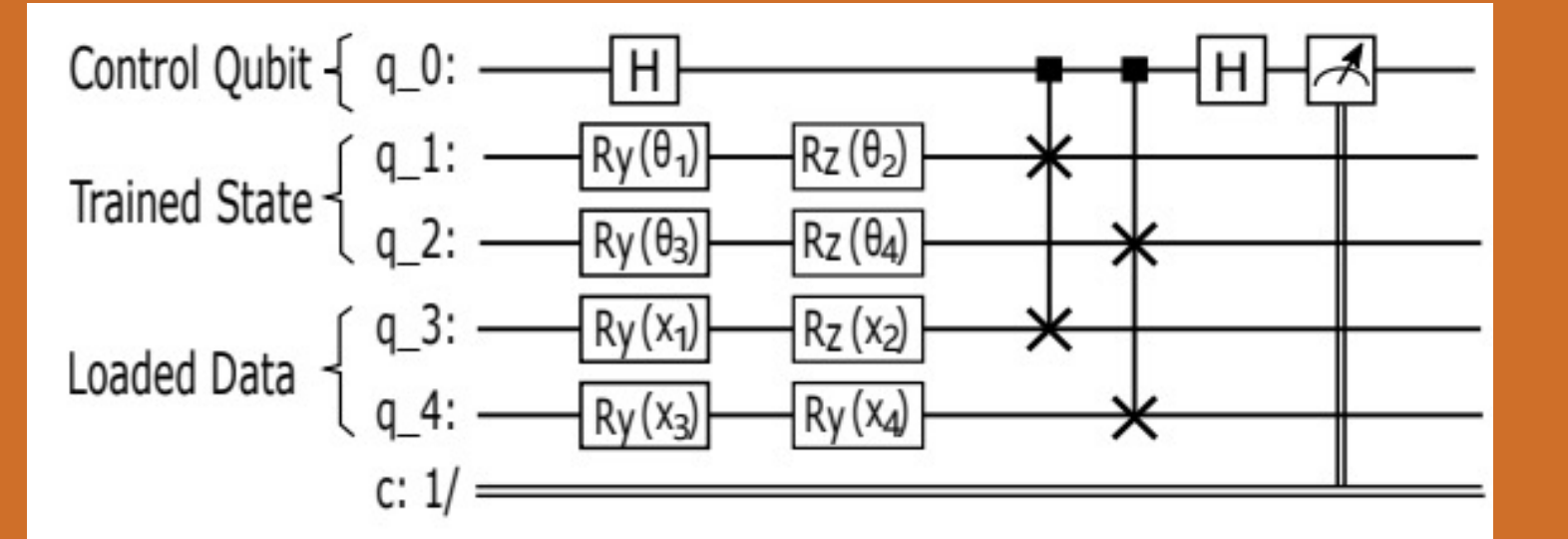
To utilize quantum computing for common tasks (such as machine learning) classical data must be converted into quantum states through **quantum encoding**. This study tests the three main quantum encoding methods on QuClassi, a quantum-classical hybrid neural network.

Methods:

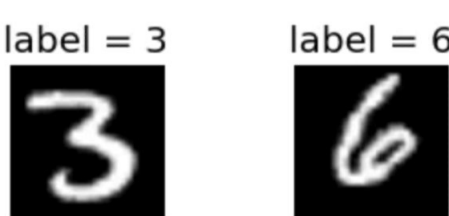
- Implemented the three encoding methods on QuClassi
- Tested different parameters to determine that 5 epochs and 0.01 learning rate are optimal for accuracy and time
- Tested the models (the mean values of three trials displayed in **average accuracy**)
- Trained models on simulator, tested on both simulator and hardware (displayed in **Simulator vs Hardware Accuracy**)
- Calculated Accuracy, loss (MSE), and entropy every epoch (displayed in graphs)

QuClassi:

Uses qubits as “neurons”, uses classical gradient optimization to recalculate parameters.



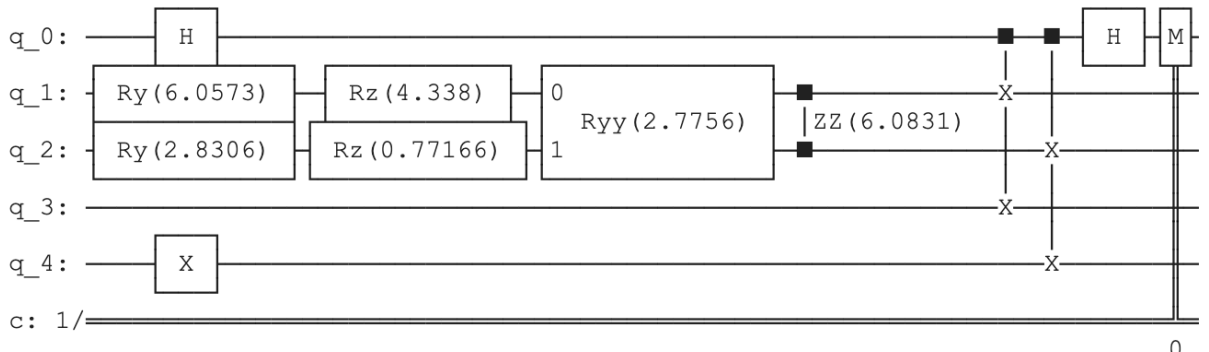
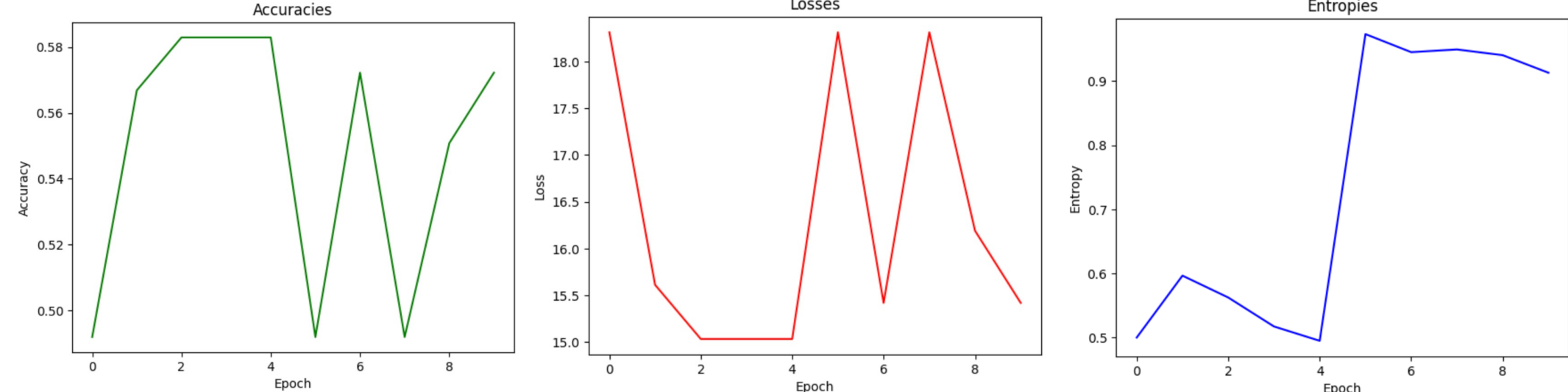
Goal: binary classification between handshapes from MNIST dataset



Basis Encoding

n data points of m binary digits encoded into n*m qubits

$$x = b_{n-1}b_{n-2} \dots b_0 \rightarrow |x\rangle = |b_{n-1}b_{n-2} \dots b_0\rangle$$

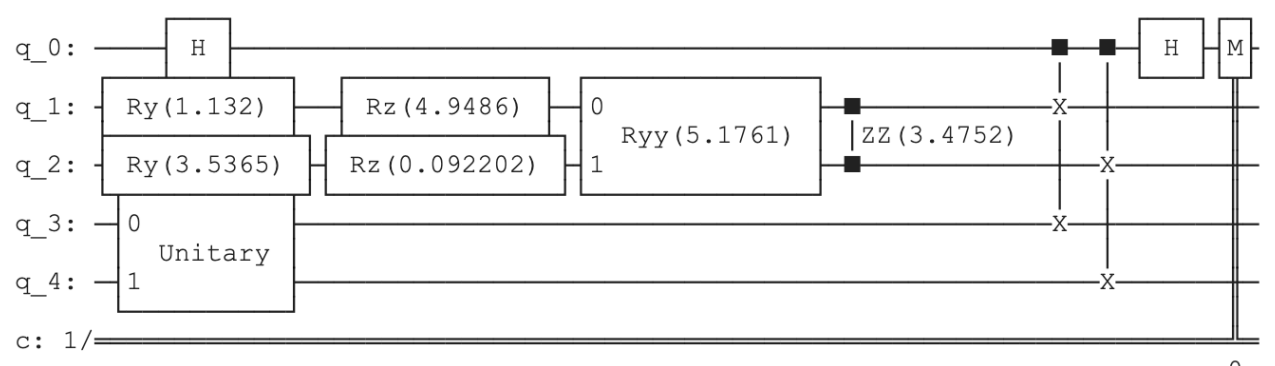
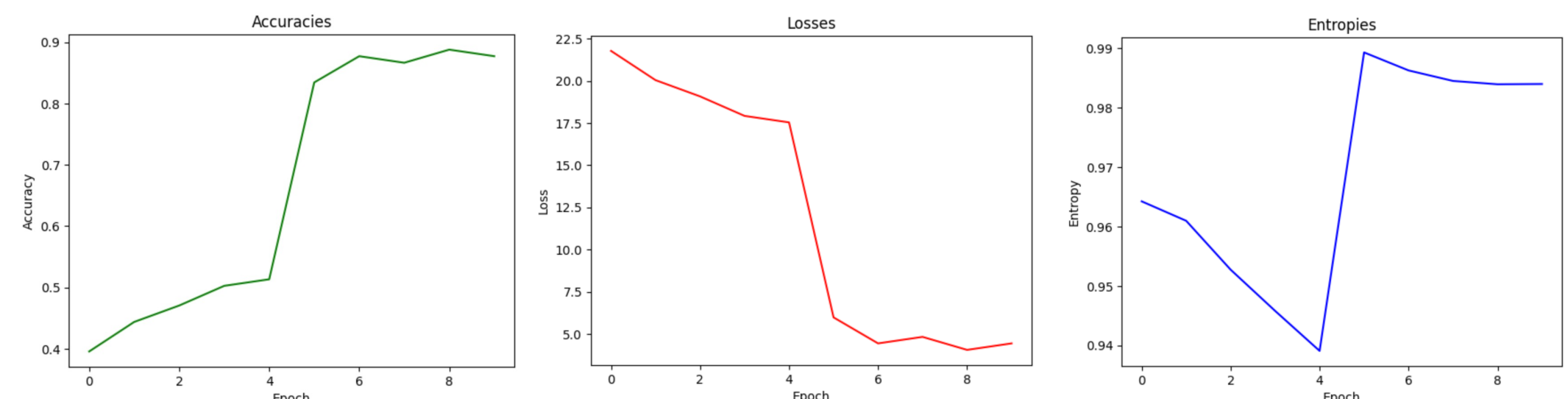


Average Accuracy on Simulator: **54.36%**
Simulator vs Hardware Accuracy:
56.95% vs 55.61%

Amplitude Encoding

n data points encoded into log(n) qubits

$$X=[1.2,2.7,1.1,0.5], \quad |\psi\rangle = \frac{1.2}{\sqrt{10.19}}|00\rangle + \frac{2.7}{\sqrt{10.19}}|01\rangle + \frac{1.1}{\sqrt{10.19}}|10\rangle + \frac{0.5}{\sqrt{10.19}}|11\rangle$$

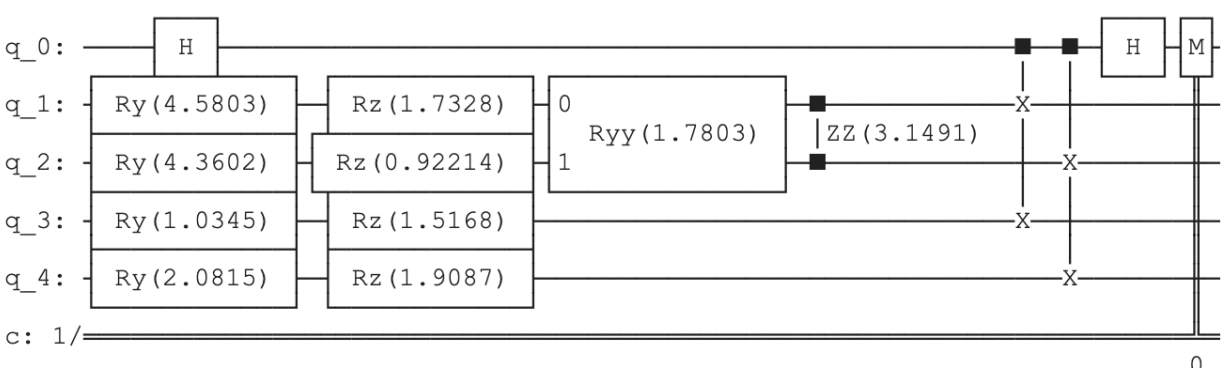
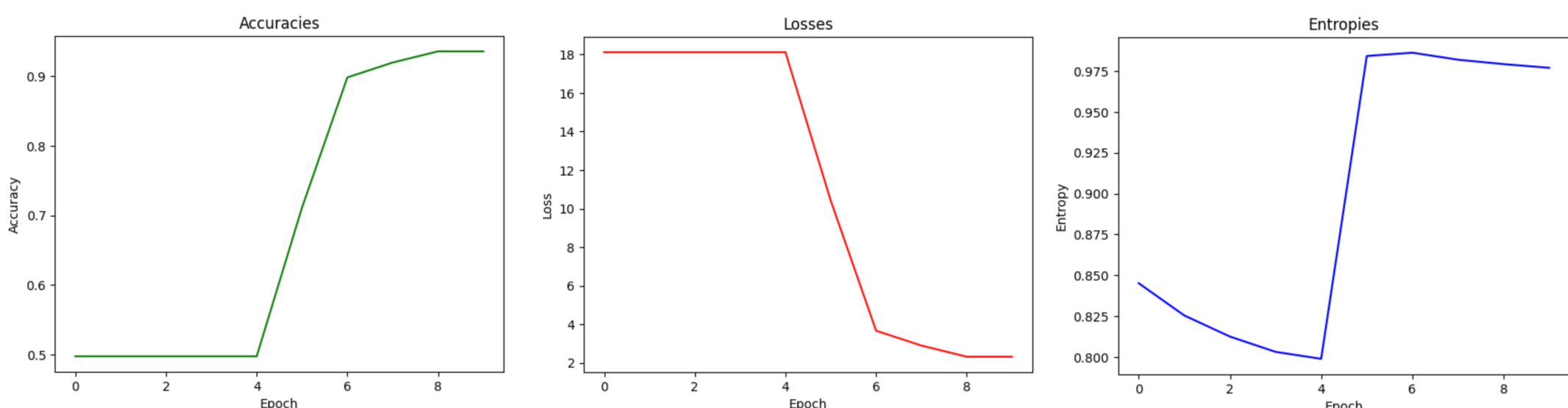


Average Accuracy on Simulator: **86.45%**
Simulator vs Hardware Accuracy:
88.72% vs 52.94%

Rotation Encoding

n data points encoded into n qubits

$$|\psi(\theta)\rangle = R_y(\theta) |0\rangle + e^{i\phi} R_y(\theta) |1\rangle$$



Average Accuracy on Simulator: **85.91%**
Simulator vs Hardware Accuracy:
96.25% vs 54.01%

Future Directions: test other proposed encoding techniques, test on other neural network architectures.

Sources

Stein, S. A., Baheri, B., Chen, D., Mao, Y., Guan, Q., Li, A., Xu, S., & Ding, C. (2022). *QuClassi: A Hybrid Deep Neural Network Architecture based on Quantum State Fidelity*. arXiv preprint arXiv:2103.11307. Retrieved from <https://arxiv.org/abs/2103.11307>

Rath, M., & Date, H. (2023). *Quantum Data Encoding: A Comparative Analysis of Classical-to-Quantum Mapping Techniques and Their Impact on Machine Learning Accuracy*. arXiv preprint arXiv:2311.10375. Retrieved from <https://arxiv.org/abs/2311.10375>