IEG313 : Quantum Computing Course project

Indian Institute of Information Technology Kottayam

Quantum Medical Image Diagnostics using QML

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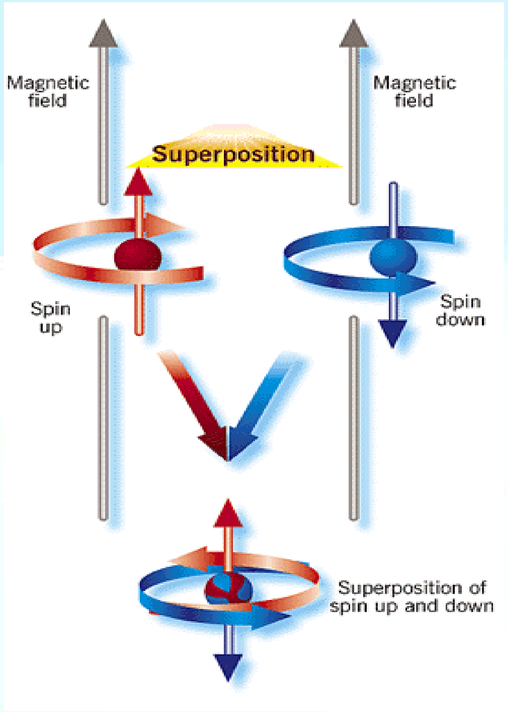
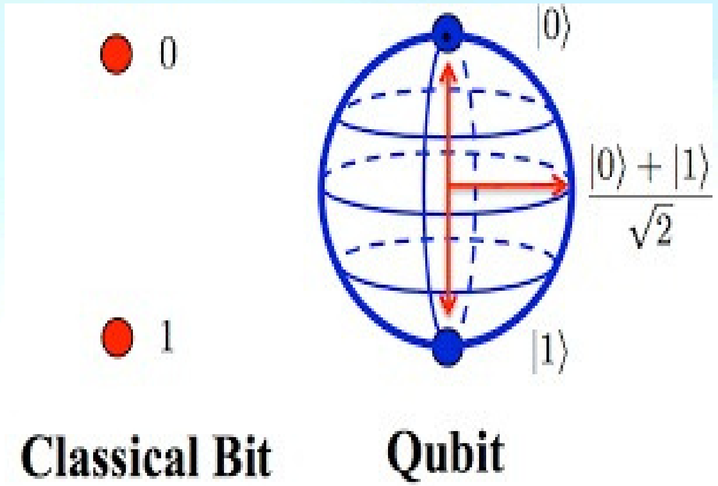
Problem Statement

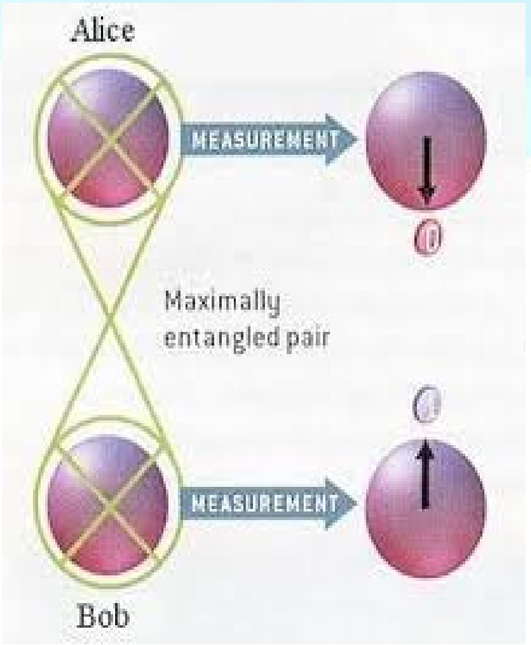
Our mission is to delve into the unique convergence of quantum computing and image analysis.

Specifically, we aim to implement a model classifying face image data set to predict disorders (i.e.,Parkinson’s, Autism, etc), And to unravel the mysteries of QCNN architecture and its application in classifying images.

Conditions for Quantum Computation



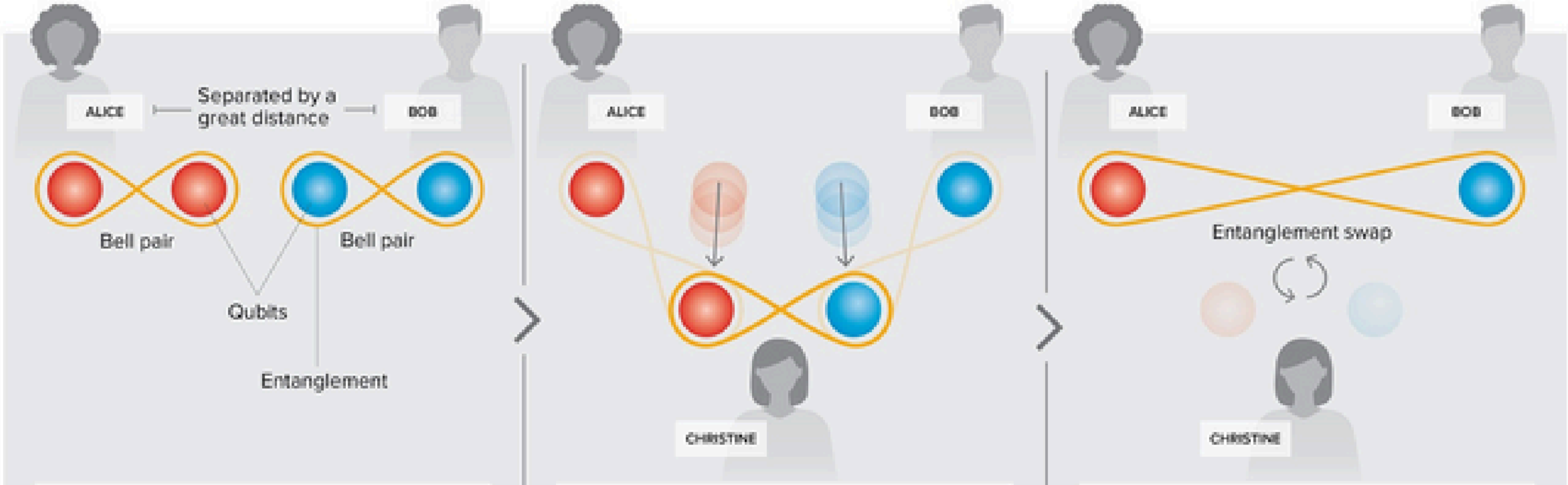




QUANTUM ENTANGLEMENT SWAPPING

Argonne NATIONAL LABORATORY

Building a quantum information network will rely on entangling qubits across great distances



Alice and Bob each have a pair of qubits in a **Bell state**, the simplest and maximally entangled state.

Each gives a qubit to Christine, who performs a **Bell projection**, cleaning up the noise and entangling her newly acquired qubits.

An **entanglement swap** occurs, linking Alice and Bob's qubits.

U.S. DEPARTMENT OF ENERGY
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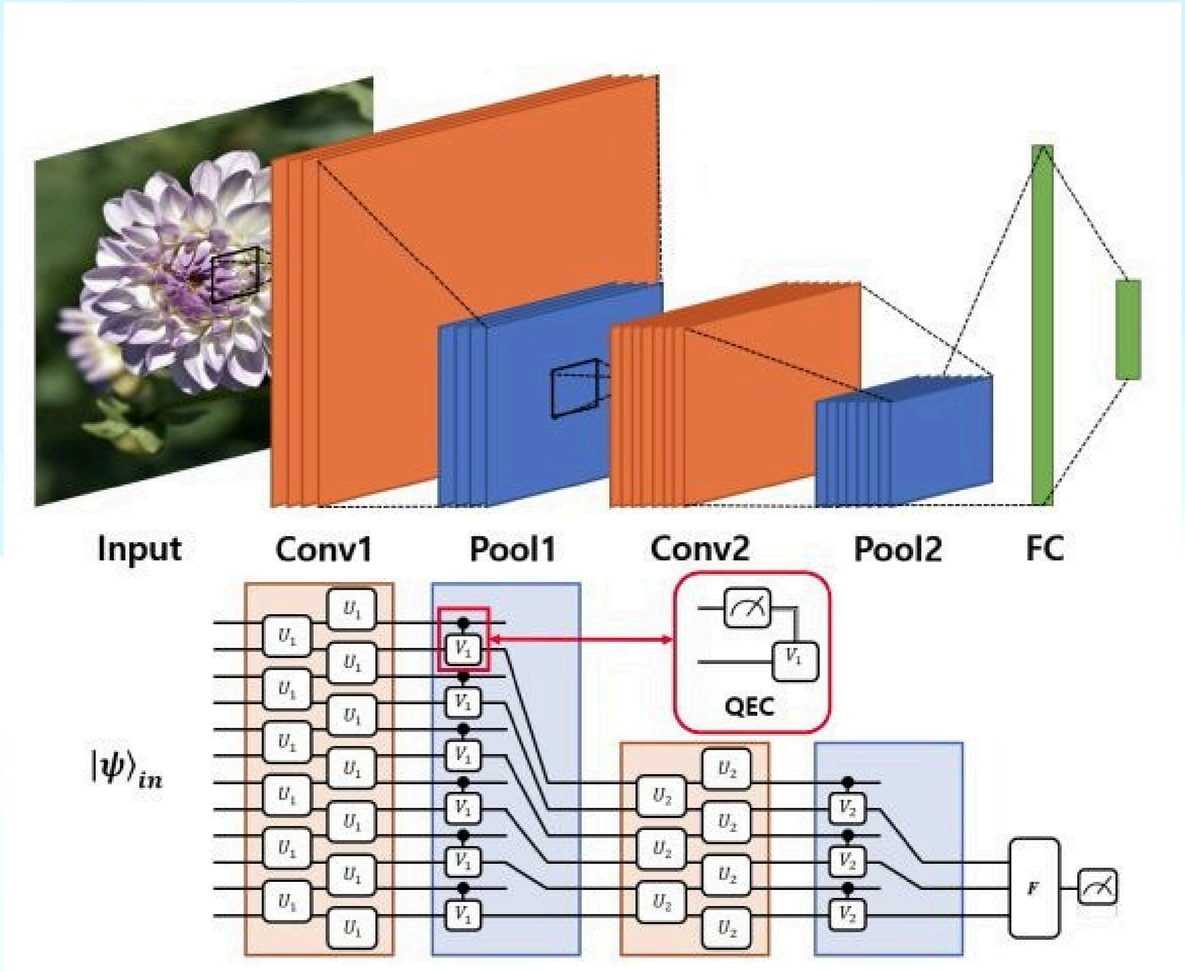
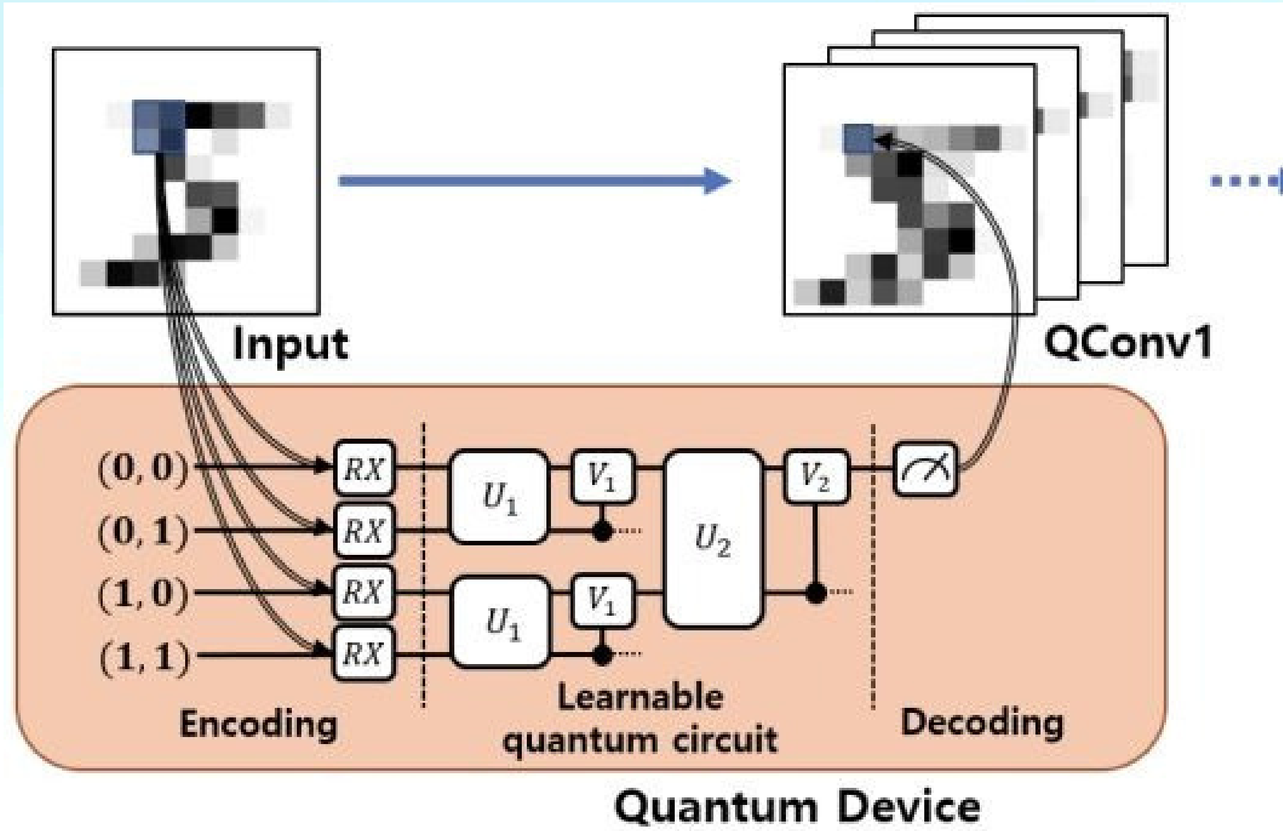
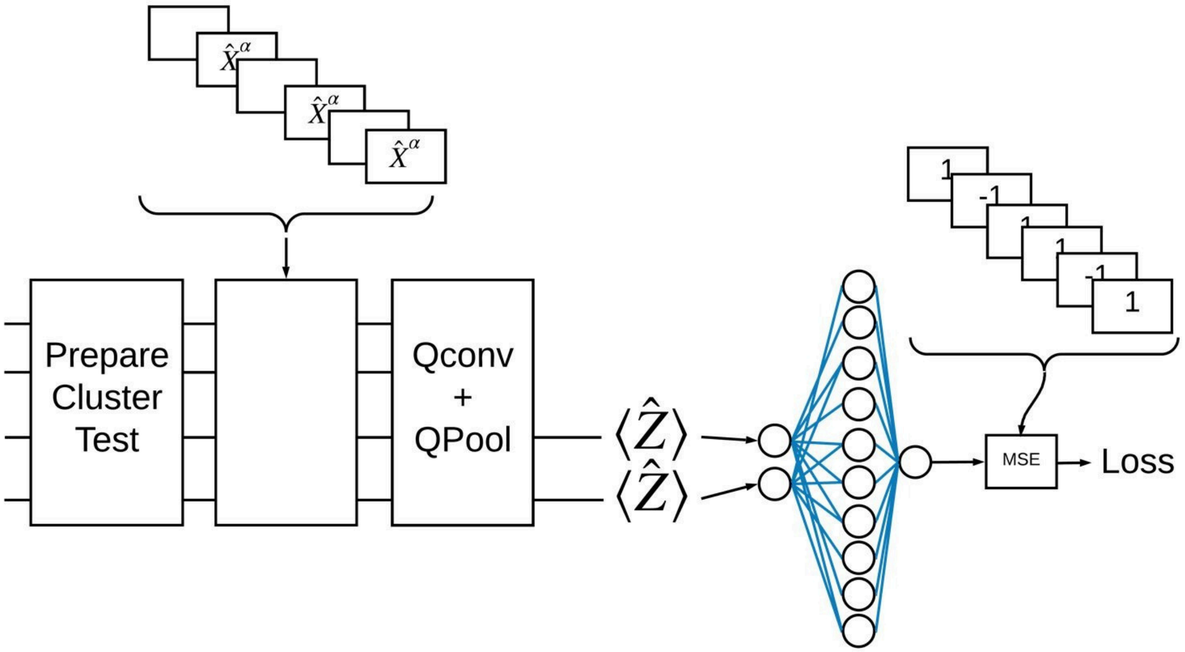
QCNN Architecture

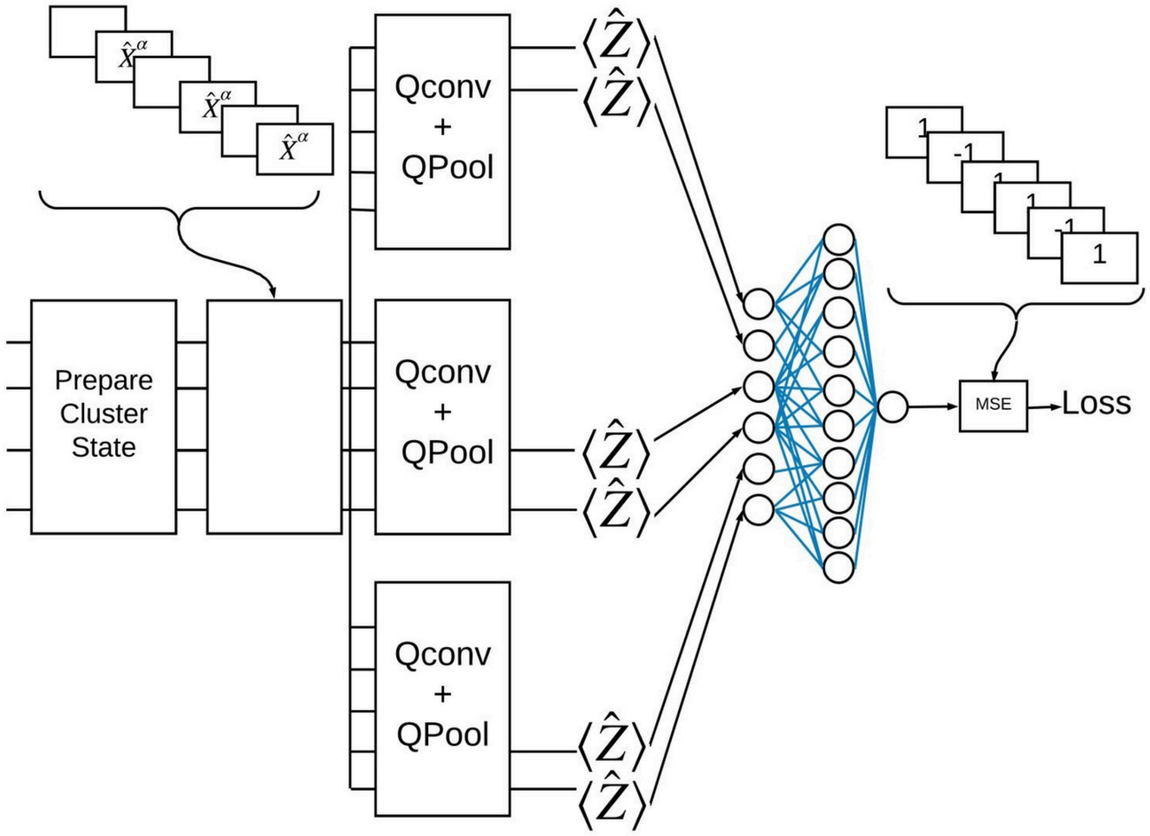




Understanding CNN with Quantum circuits, Image credits : Pennylane

Overview


$\hat{X}^\alpha$
$\hat{X}^\alpha$
$\hat{X}^\alpha$
Prepare
Cluster
Test
Qconv
+
QPool
$\langle\hat{Z}\rangle$
$\langle\hat{Z}\rangle$
1
-1
1
1
-1
1
MSE
Loss


$\hat{X}^\alpha$
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$\hat{X}^\alpha$
Prepare
Cluster
State
Qconv
+
QPool
Qconv
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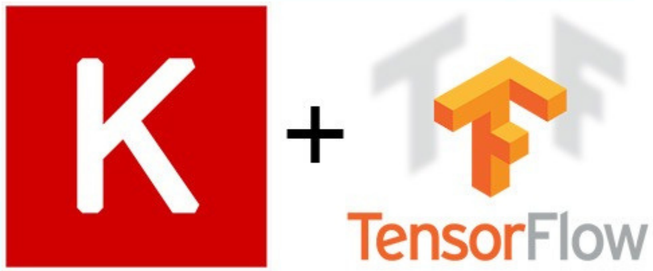
MSE
Loss

Libraries used for code



Preparation of Dataset’s

Medical image dataset of Autism

We have collected medical image dataset which represents patient with and without Autism.

Dataset from Kaggle

Design and Implementation

- Quantum circuit as a convolution kernel
- Quantum pre-processing of the dataset
- Hybrid quantum-classical model

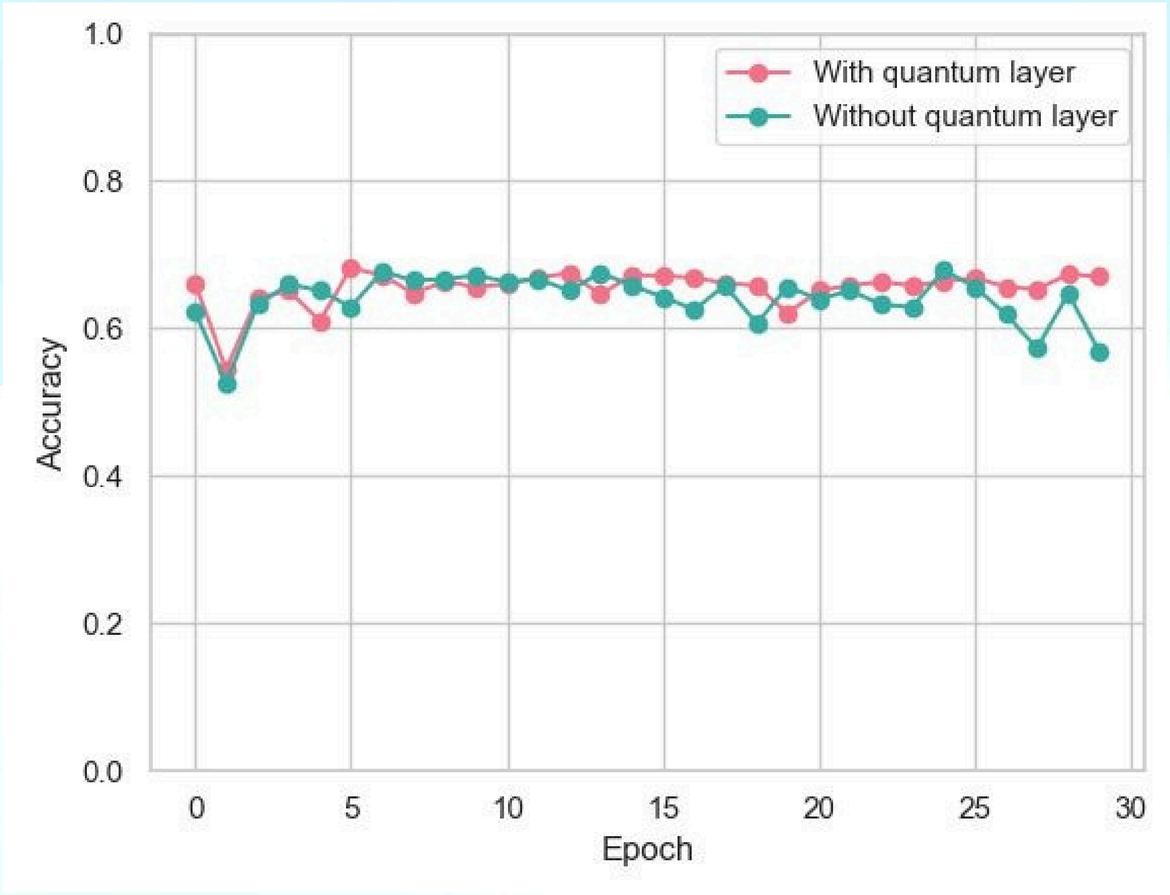
Training the dataset

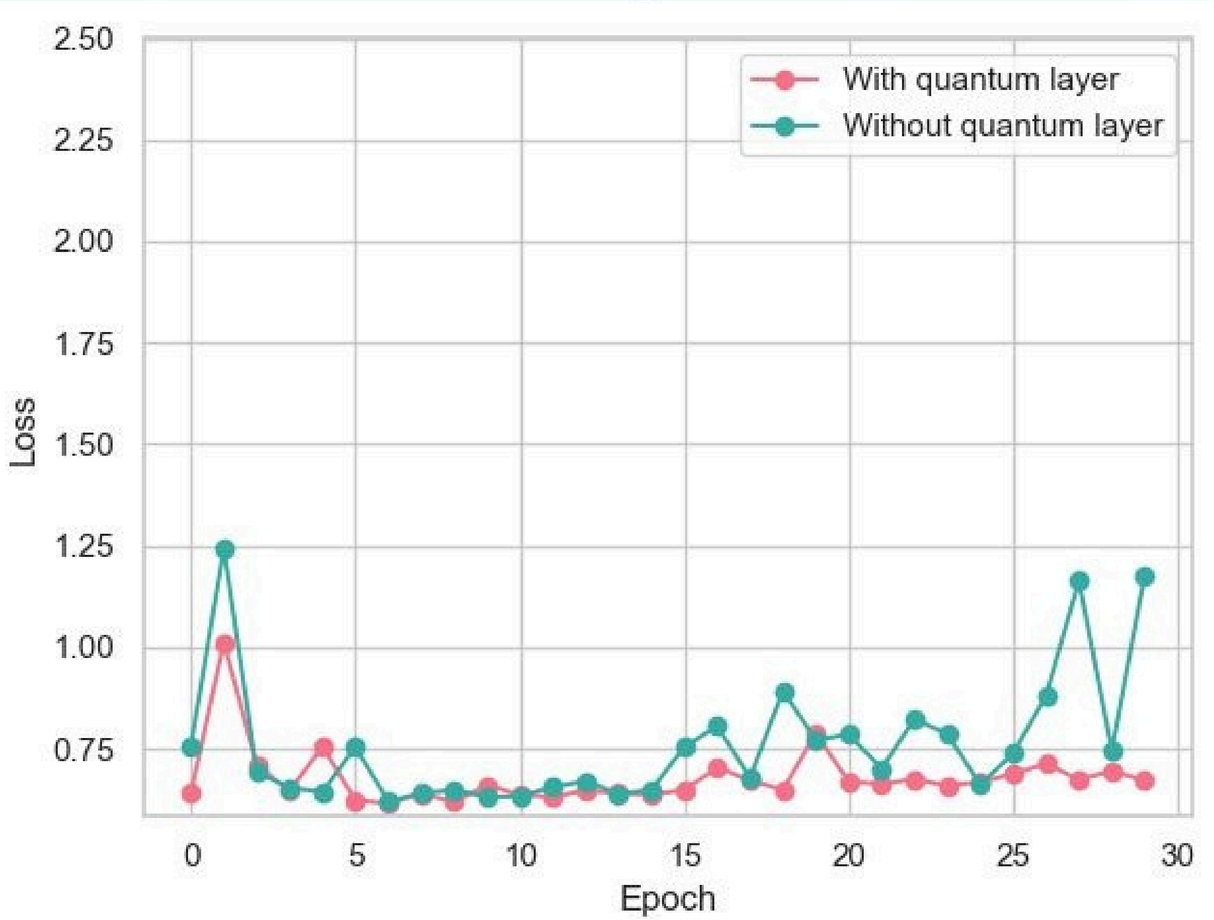
```
# Quantum pre-processing of train images
q_train_images = []
print("Quantum pre-processing of train images:")
for idx, img in enumerate(train_images):
    print("{}/{}        ".format(idx + 1, len(train_images)), end="\r")
    q_train_images.append(quanv(img))
q_train_images = np.asarray(q_train_images)

# Quantum pre-processing of test images
q_test_images = []
print("\nQuantum pre-processing of test images:")
for idx, img in enumerate(test_images):
    print("{}/{}        ".format(idx + 1, len(test_images)), end="\r")
    q_test_images.append(quanv(img))
q_test_images = np.asarray(q_test_images)

# Save pre-processed images
SAVE_PATH = "your_save_path/"
if not os.path.exists(SAVE_PATH):
    os.makedirs(SAVE_PATH)

np.save(SAVE_PATH + "q_train_images.npy", q_train_images)
np.save(SAVE_PATH + "q_test_images.npy", q_test_images)
```

File display

- 80% datasets were used for training and rest were used for testing.

Result and Observation





References

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- [2] Maxwell Henderson, Samriddhi Shakya, Shashindra Pradhan and Tristan Cook.Quantum Convolu
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- [3] Rajat Raina, Alexis Battle, Honglak Lee, Benjamin Packer, and Andrew Y Ng. Self-taught learning
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