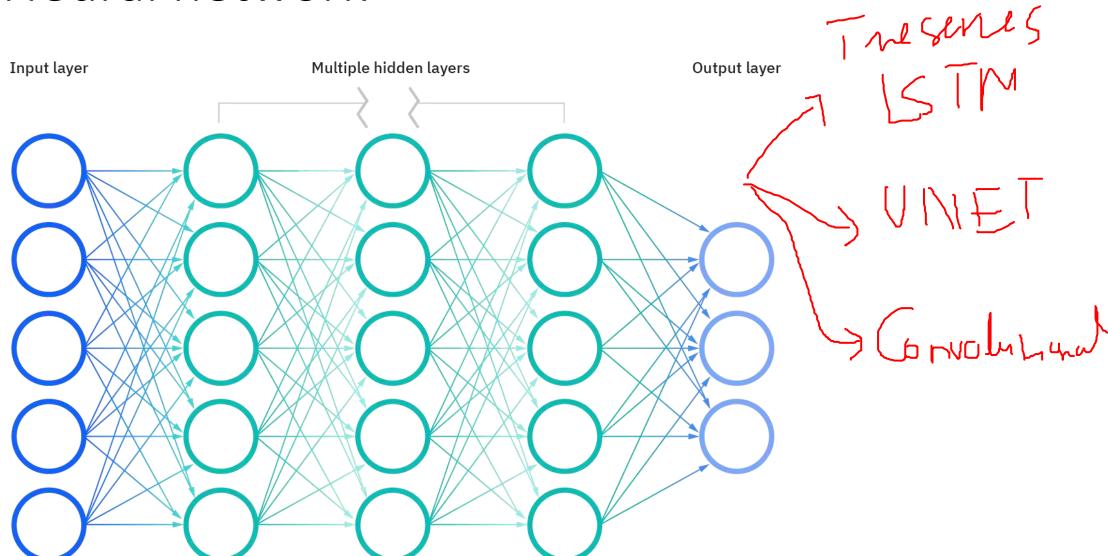
Quanvolutional neural network

20/5/2022

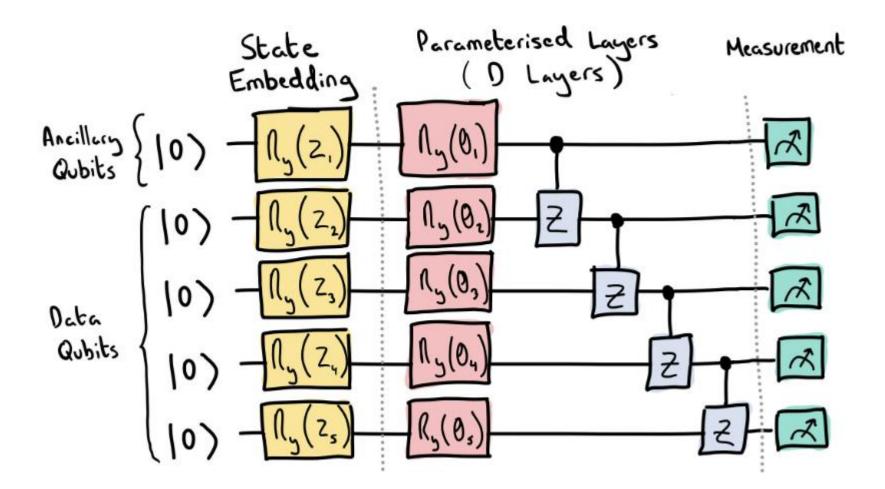
https://arxiv.org/abs/1904.04767

https://arxiv.org/abs/2108.00661.pdf

Neural network



Quantum neural network



Convolutional neural network feature maps feature haps feature maps feature maps input 14 x 14 32 x 32 28 x 28 output 5x5 convolution convolution subsampling 5x5 convolution subsampling classification feature extraction

Quantum convolutional neural network

FC (a) (b) (c) (MERA) QCNN MERA $|\psi\rangle$

Absence of Barren Plateaus in Quantum Convolutional Neural Networks, PRX, 11,041011 (2021)

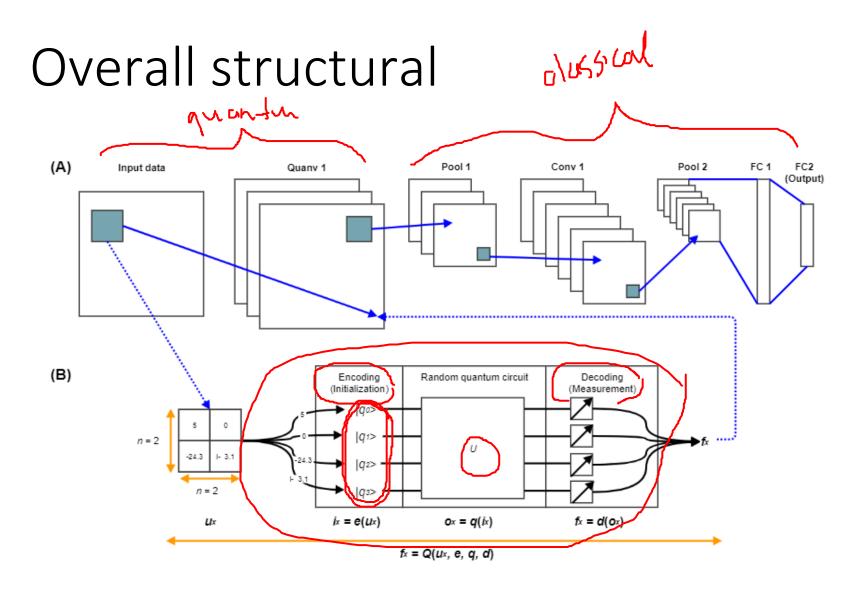
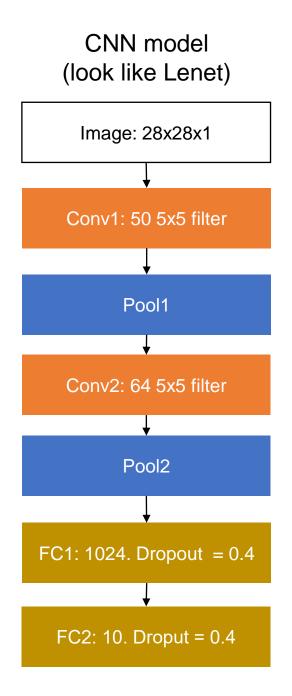
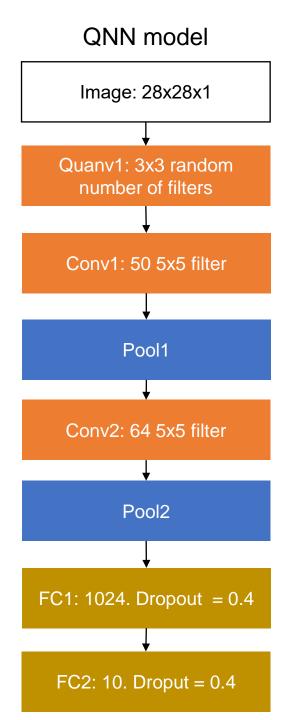


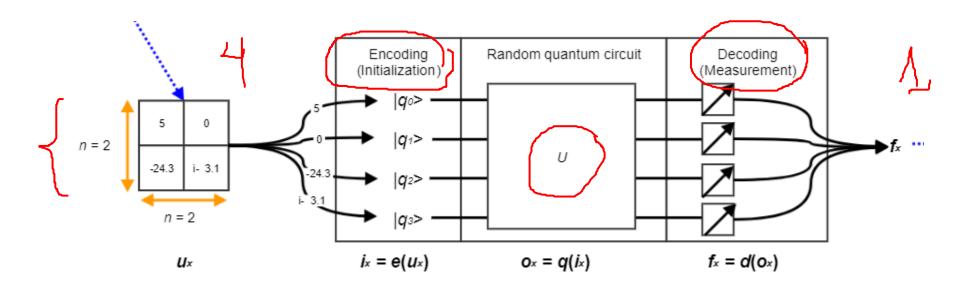
Fig. 1.: A. Simple example of a quanvolutional layer in a full network stack. The quanvolutional layer contains several quanvolutional filters (three in this example) that transform the input data into different output feature maps. B. An in-depth look at the processing of classical data into and out of the random quantum circuit in the quanvolutional filter.

Models





Quanvolutional layer



Quanvolutional layer = $f(u_x, e, q, d)$: $\mathbb{R}^{nxn} \to \mathbb{R}$ where:

- u_x : patch
- *e*: encoder
- *q*: random quantum circuit
- *d*: decoder

Encoding & Decoding

 $e \colon \mathbb{R}^{n \times n} \to |\psi\rangle$ $\log N$ qubits but exponentially number of gates N qubit, linear number of gate

Threshold encoding: if pixel value is less than threshold t, the according qubits will be $|0\rangle$ and vice versa.

$$d: |\psi\rangle \to \mathbb{R}$$

Quantum circuit

The number of qubits is 9.

Chose (0 \rightarrow 2 n^2) random 1 qubit gate.

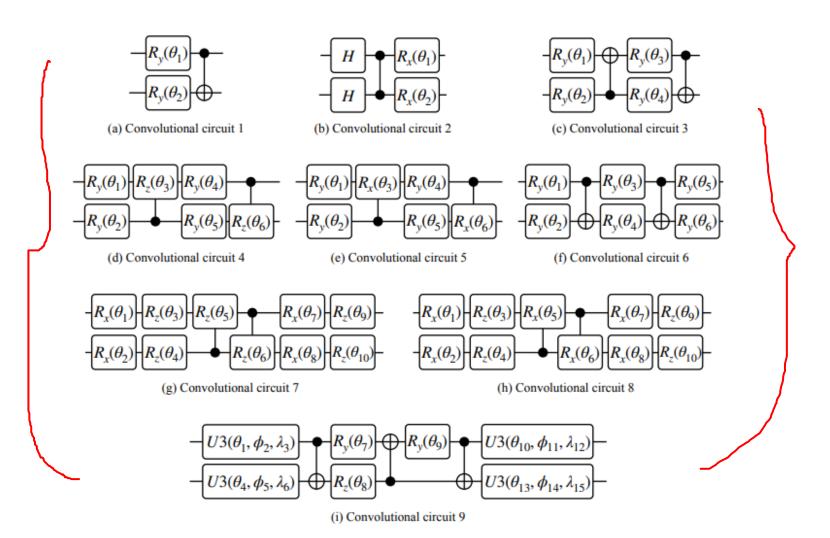
$$X(\theta), Y(\theta), Z(\theta), U(\theta), P, T, H$$

 θ is random in $[0,2\pi]$

And random 2 qubits gate (to make entanglement)

→ The set of gates is suffled => One quanvolutional layer

Quantum circuit: Example



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Quantum advantages

Quantum computers can access kernel functions in highdimensional Hilbert spaces much more efficiently than classical computers.

→ Coverage faster

Disadvatanges

- Number of measurements
- Number of quanvolutional filters

Dataset

• 70k 28x28 gray MNIST samples (60k train, 10k test)

Results

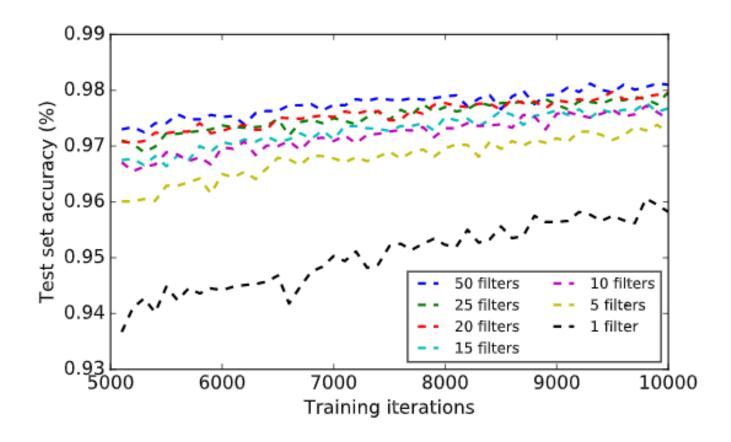
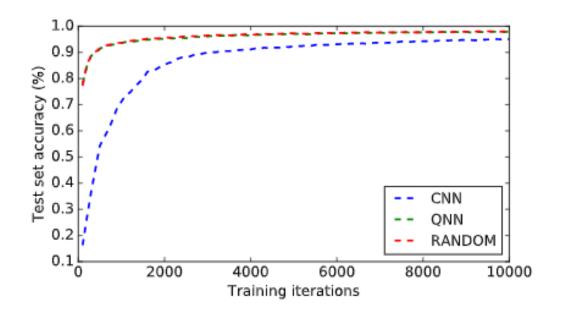


Fig. 2.: QNN MODEL test set accuracy results using a variable number of quanvolutional filters.

Results



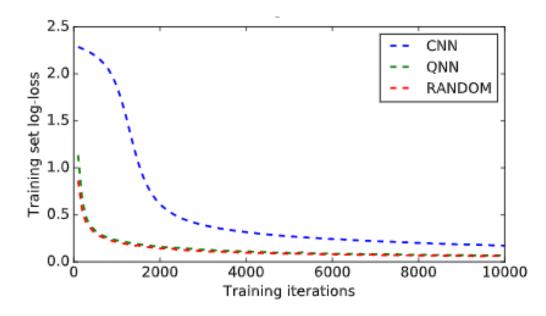


Fig. 3.: QNN MODEL performance, in terms of (A) test set accuracy and (B) training log-loss, compared to both CNN MODEL and RANDOM MODEL.

Future works

- Test on another dataset and deeper QNN.
- Make quanvolutional layer trainable or change by time by the evolutional / genetic algorithm. => Use QNG in quanvolutional layer

Thanks for listening!