

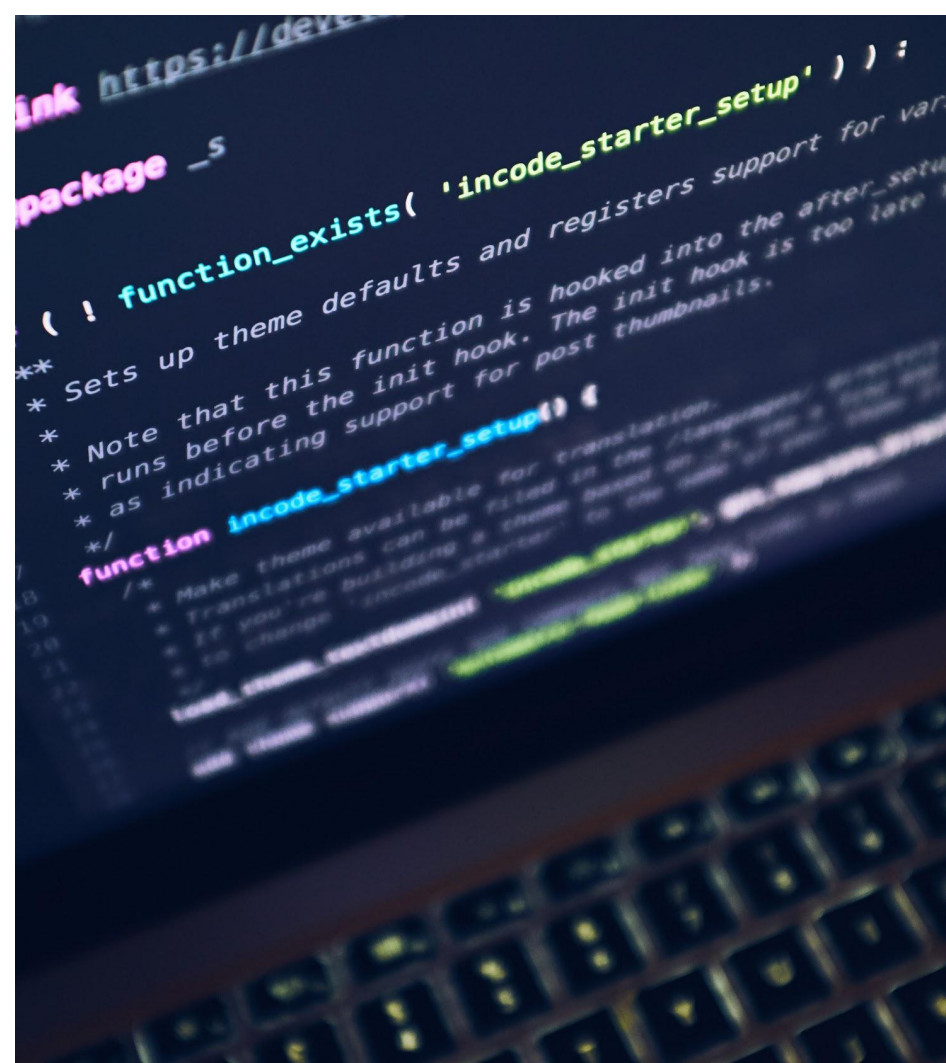
Machine Learning with full PyTorch & Qiskit capabilities

Team QizGloria

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IBM Coach: Christa Zoufal

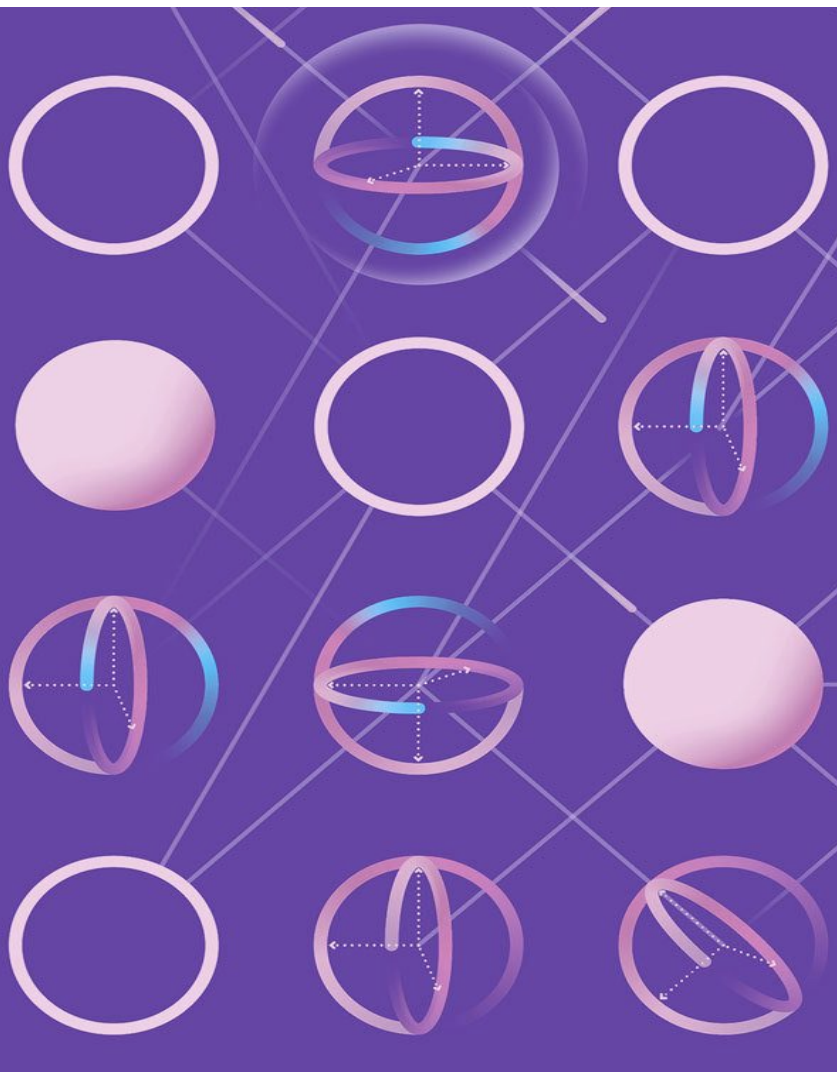
Our mission

- Closer integration of Pytorch & Qiskit beyond existing tools
- Enable seamless co-training of quantum circuits & neural networks
- Encourage classical ML engineers to use quantum nodes

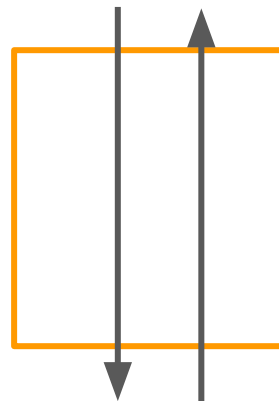
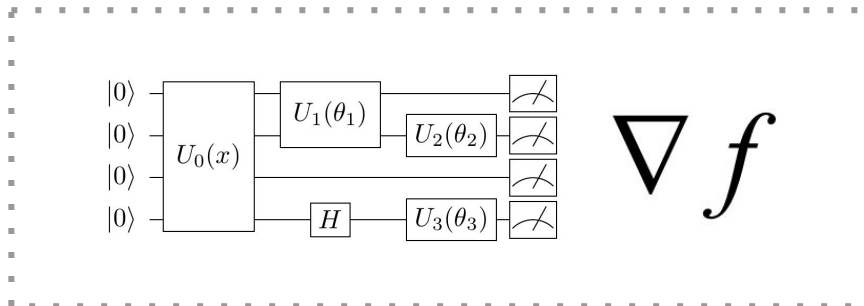


Why is this cool?

- PyTorch neural networks & tools
- More options for optimizers (RMSprop; Momentum; etc.)
- Full Qiskit capabilities (circuit definition, transpiler, Aqua,...)
- Back-end management by Qiskit (QPU, simulators)
- Can now parallelize your optimization code
- Bridges the gap between the QML and classical ML community

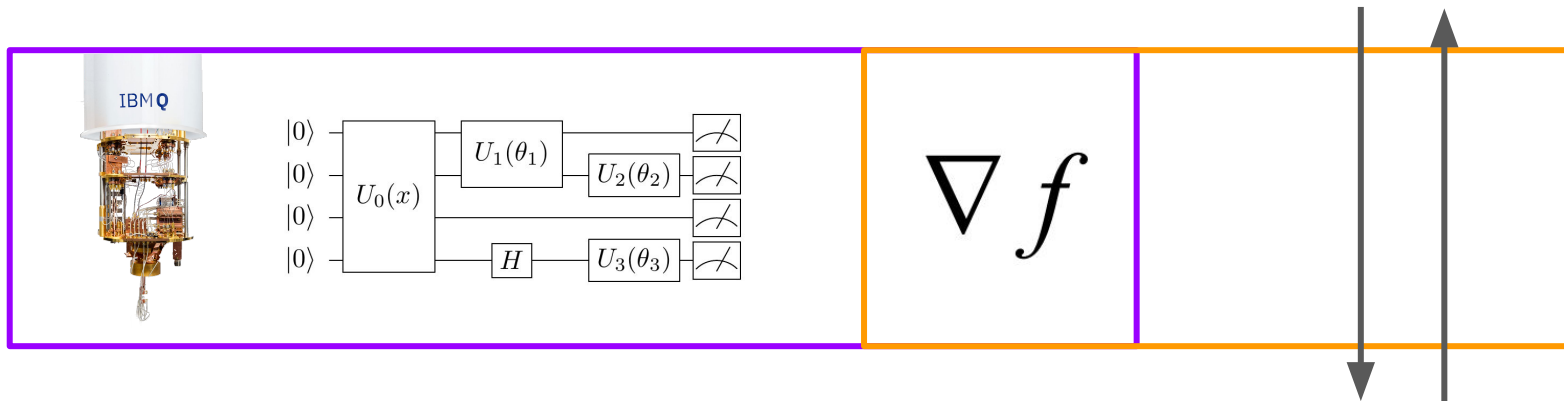


How does it work?



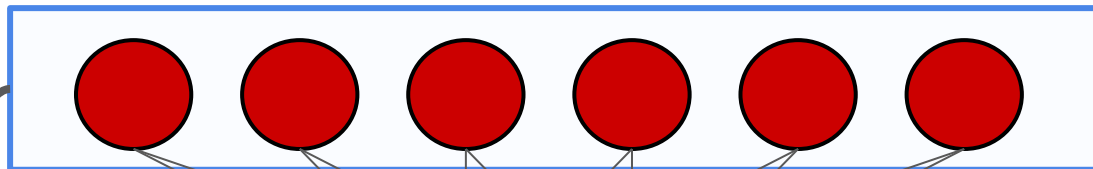
other frameworks:
blocked from Qiskit-tools!

How does it work?

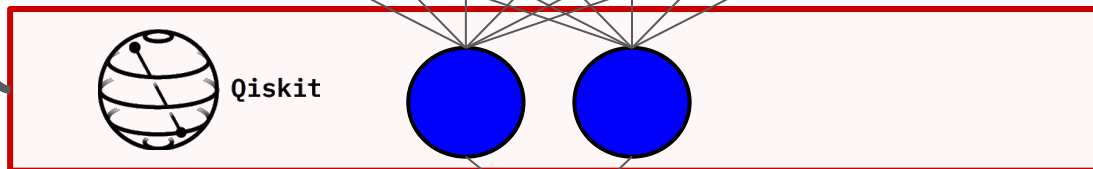


How does it work?

Classical node
(E.g. Pytorch neural network)

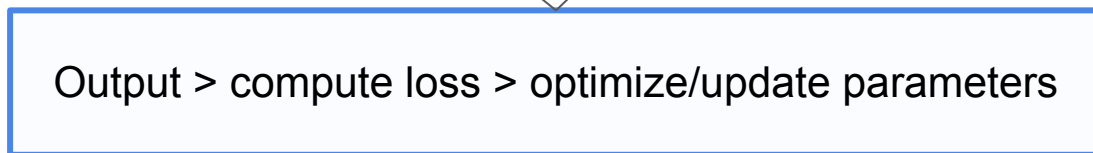


 PyTorch



Quantum node
(Qiskit circuit with trainable parameters; VQE, QAOA etc.)

Pytorch
Optimization
(Computes gradient of loss function wrt parameters)



 PyTorch

How does it work?



QiskitCircuit

circuit definition (Terra, Aqua)
parameter binding
expectation value evaluation
back-end management

TorchCircuit

tensorization
parallelization
forward pass
backward pass (finite diff, aqgd)

Seamless integration of Pytorch and Qiskit

```
class Net(nn.Module):
    def __init__(self):
        super(Net, self).__init__()
        self.conv1 = nn.Conv2d(1, 10, kernel_size=5)
        self.conv2 = nn.Conv2d(10, 20, kernel_size=5)
        self.conv2_drop = nn.Dropout2d()
        self.fc1 = nn.Linear(320, 50)
        self.fc2 = nn.Linear(50, 3)

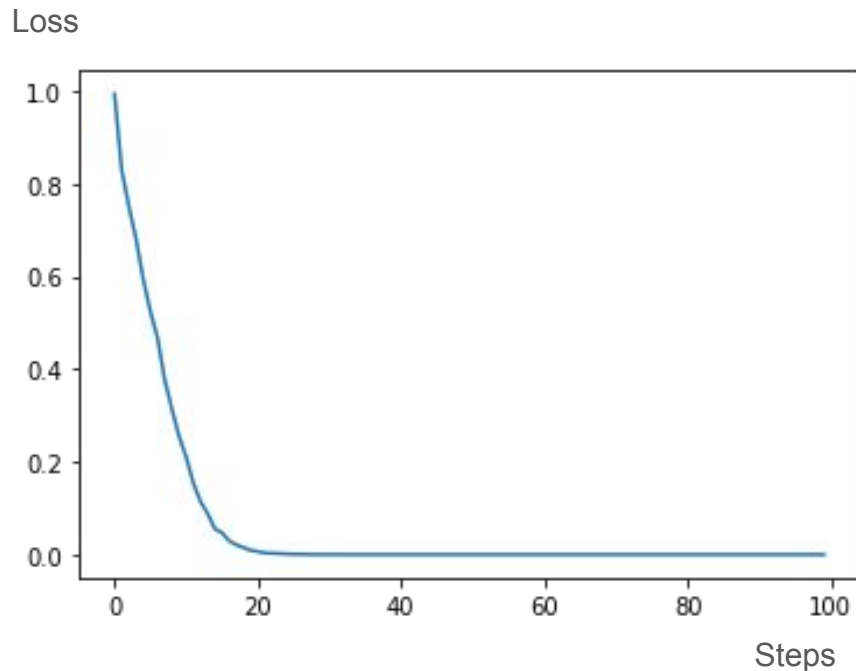
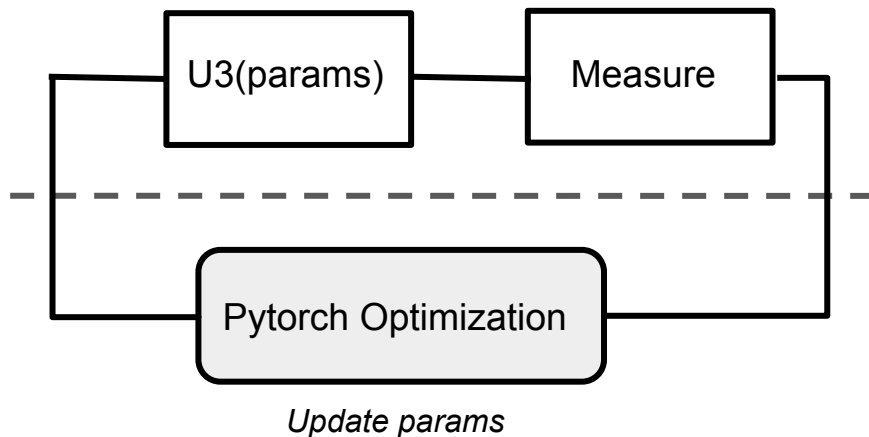
    def forward(self, x):
        x = F.relu(F.max_pool2d(self.conv1(x), 2))
        x = F.relu(F.max_pool2d(self.conv2_drop(self.conv2(x)), 2))
        x = x.view(-1, 320)
        x = F.relu(self.fc1(x))
        x = F.dropout(x, training=self.training)
        x = self.fc2(x)
        x = qc(x)
        return x
```


A wide-angle photograph of a majestic mountain range. The foreground and middle ground are dominated by steep, snow-covered slopes with some rocky outcrops. In the background, a series of jagged, snow-capped peaks stretch across the horizon under a clear, pale blue sky. The overall scene conveys a sense of vastness and potential.

The options are endless with what you can do!

Hello World!

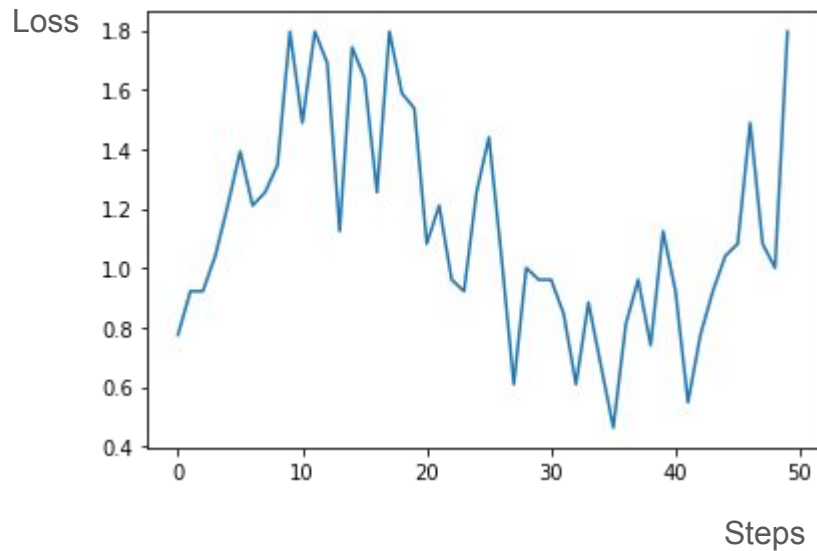
- Learn how to rotate 1 qubit to get a defined σ_z expectation
- Used U3 rotation in qiskit



Details:

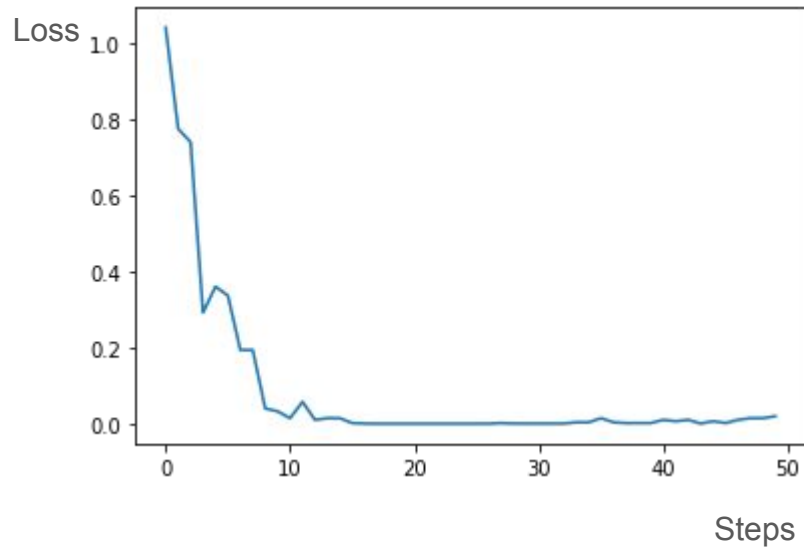
Finite difference gradient estimation; shots = 10 000

Hello World! - analytical gradients



Details:

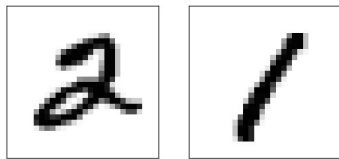
*Finite difference gradient estimation; shots
= 100*



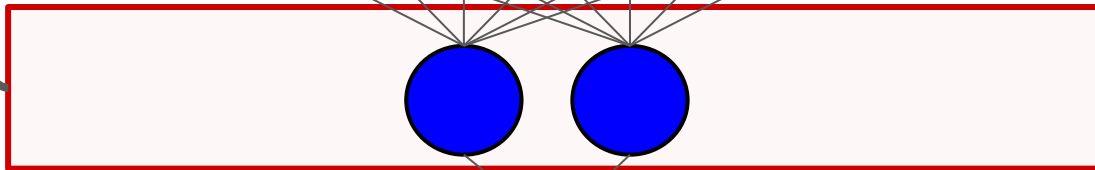
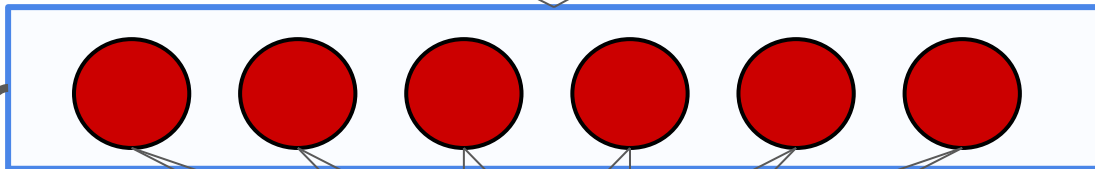
Details:

*Analytical gradient;
shots = 100*

MNIST



Classical node
ConvNet



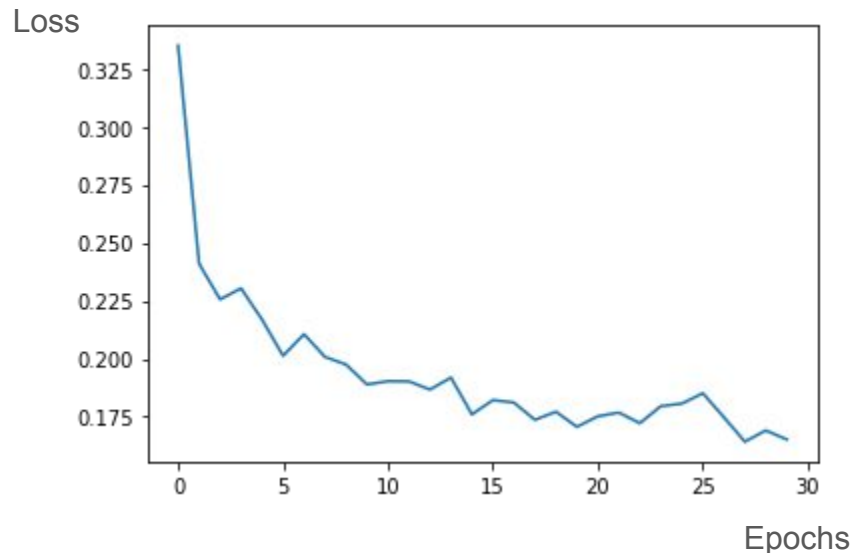
Quantum node
(Qiskit circuit: Rx & Ry rotations)

Pytorch
Optimization
(Computes gradient of
loss function wrt two
parameters)

Output > compute loss > optimize/update parameters

Sigma z exp > NLL/cross entropy loss > Pytorch Adam
optimizer

MNIST



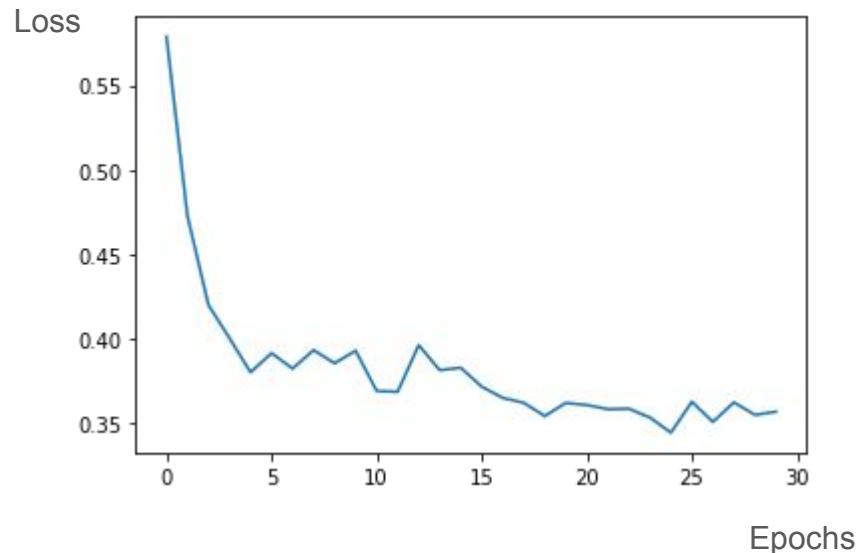
Details:

Analytical gradients

Negative-log-likelihood-loss

Shots = 100

(200 data samples per epoch)



Details:

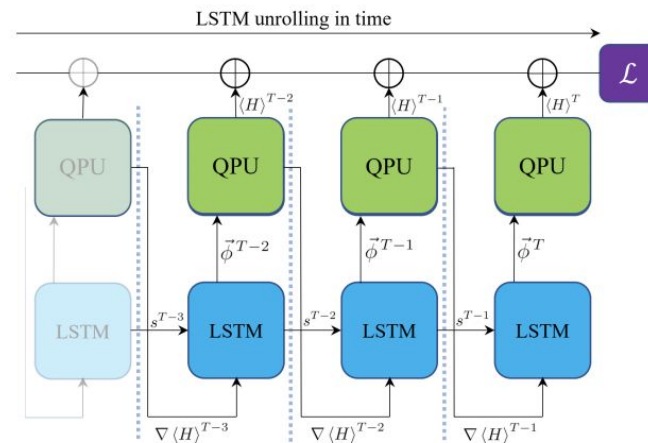
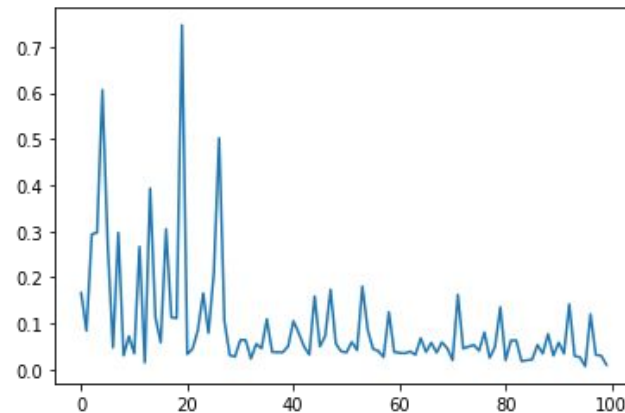
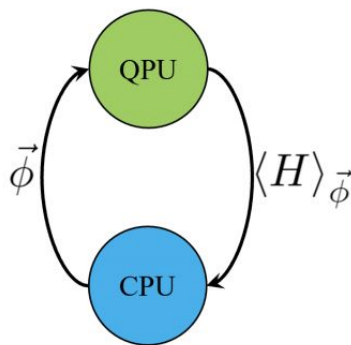
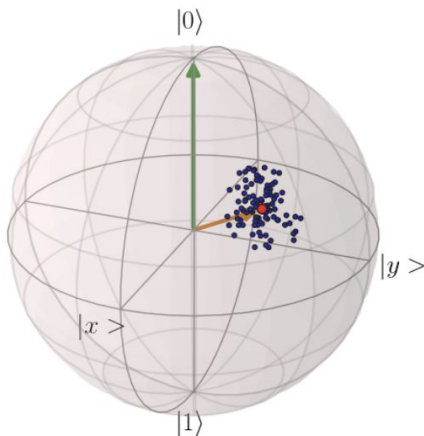
Finite difference gradient estimation

Cross-entropy loss for MNIST

Shots = 10 000

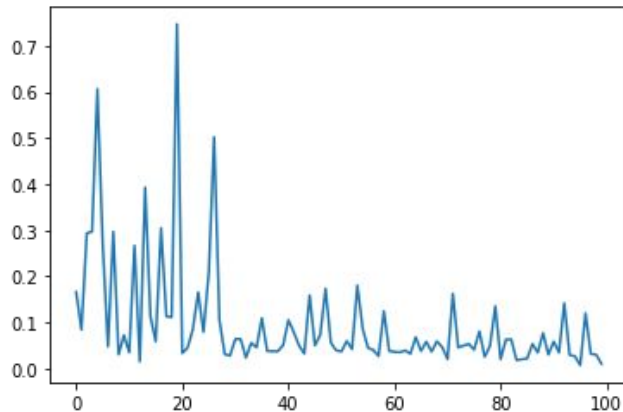
Other implementations we did

Meta-learning for neural optimizer for single qubit rotation

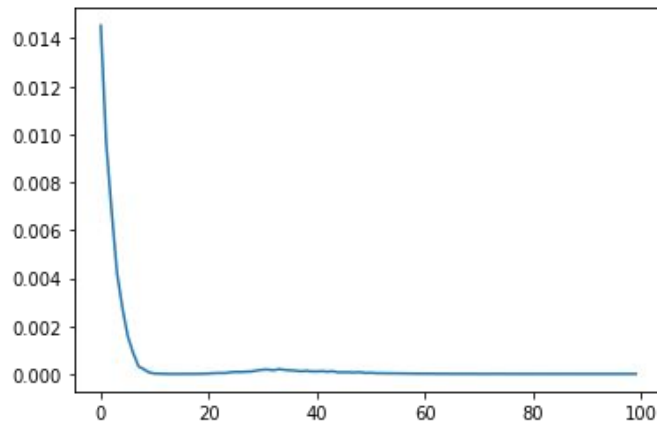


Other implementations we did

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2 qubit example: QAOA with Pytorch optimizers

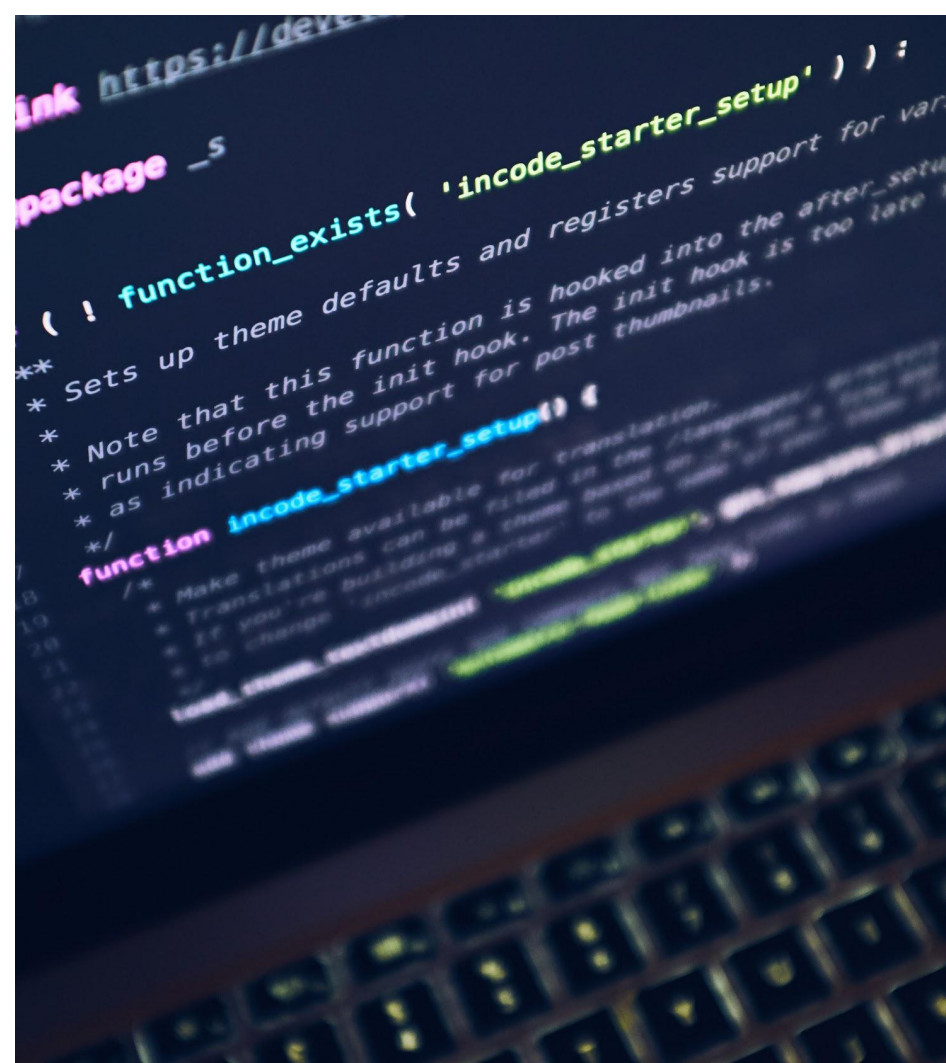


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