



August 26, 2024

## The state-of-the-art quantum technology

Applications to High Energy Physics problems

Abdualazem Fadol Mohammed



#### The Beijing Super Cloud Computing Center

We got two free accounts for testing:
 • NC-N40: GeForce RTX 4090 (Spent)
 • BSCC-N32-E: problem installing packages
 The test was done using four jet types:

 $^{O}H \rightarrow b\bar{b}, H \rightarrow c\bar{c}, H \rightarrow gg$ , and  $t \rightarrow bqq'$ 

- Each class has 100k events: 20%
- ☐ The default configuration for the network was used.
- Running time for 50 epochs: 2d, 3h, 20m for QParT.



2

#### The Beijing Super Cloud Computing Center

![](_page_2_Figure_1.jpeg)

## Quantum Transformer

![](_page_3_Figure_1.jpeg)

This is still experimental work towards building a pure-quantum transformer inspired by ArXiv: 2205.05625.

![](_page_3_Picture_3.jpeg)

![](_page_3_Picture_4.jpeg)

## Quantum Transformer

Wrapping the quantum self-attention with the following class

```
import torch
import torch nn as nn
import numpy as np
from Modules.QuantumSelfAttention import QuantumSelfAttention
You, 7 days ago | 1 author (You)
class BinaryClassifier(nn.Module):
   def ___init__(self, input_dim):
        super(BinaryClassifier, self).__init__()
        self.n_qubits = input_dim // 2
        self.fc = nn.Linear(self.n_qubits * 3, self.n_qubits * 3, bias=True)
        #self.sigmoid = nn.Sigmoid()
   def forward(self, x):
        output = self.fc(x)
        #output = self.sigmoid(output)
       output = output.view(x.size(0), -1)
        return output
You, 3 weeks ago | 1 author (You)
class CombinedModel(nn.Module):
   def ___init___(self, input_dim, seed, shots, device="cpu", L=1):
        super(CombinedModel, self).___init___()
        self.attention = QuantumSelfAttention(input_dim, seed, shots, device, L)
        self.classifier = BinaryClassifier(input_dim)
   def forward(self, x):
        attention_scores = self.attention(x)
        output = self.classifier(attention_scores)
        return output
```

This was tested using the CEPC data. However, the performance is not good. The reason might be from:

![](_page_4_Figure_5.jpeg)

We need to refine the wrapper.

![](_page_4_Figure_7.jpeg)

![](_page_4_Picture_8.jpeg)

# Pennylane package

- □ I heard that Pennylane is faster than Qiskit packages in simulation.
- We re-write the quantum self-attention using the Pennylane package.
- However, the performance is worse for now than for Qiskit.
- ☐ The time difference between the two packages is not big. Both are a round three days for 100 epochs.

![](_page_5_Picture_5.jpeg)