Quantum Generative Adversarial Networks on $t\bar{t}H(b\bar{b})$ Process Data Generation

CERN Summer Student Session 2021





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github.com/eraraya-ricardo/CERN-QGAN

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Introduction

Dataset

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Outline

Introduction:

- Background
- Goal & Related Works

Dataset:

- Dataset Introduction
- A Closer Look Into the Dataset

Current Progress:

Classical Benchmark

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The Quantum Models

Introduction: Background

HL-LHC upgrades at CERN will require enormous computing resources^[1]

01

Quantum computing has potential in improving performance of data processing and ML^[2]

02

Can it improves HEP simulation and data analysis?



Projected LHC performance through 2038 the amount of data will increase at least 10x more luminosity = produce more data^[1]

[1] Burkhard Schmidt 2016 *J. Phys.: Conf. Ser.* 706 022002.[2] Biamonte J, et al. *Nature* 2017;549.

Introduction

Project's Goal

Explore Quantum Generative Adversarial Networks (QGAN) performance on high-energy physics data simulation, compared to the classical GAN.

Related Works

- Classical GAN to simulate LHC QCD dijet events has been explored^[3]
- Quantum Classifiers had been explored to classify the $t\bar{t}H(b\bar{b})$ dataset, with performance comparable to the classical counterparts (SVM, Random Forest, AdaBoost)^[4]

Dataset: Introduction



Feynman diagram of the signal process in red and the dominant background process in green

- Higgs Boson is produced in association with $t\bar{t}$ via gluon fusion and it decays to $b\bar{b}$
- We focused only on the simulation of two b-jets from the Higgs
- We also focused only to simulate the positive event (not the background)

Dataset: A closer look



- The dataset is from numerical Monte Carlo simulations
- We limit the problem to just 4 features for each b-jet (from 8)
- Normalized into either [-1, 1] or [0, 1] range
- The features that we select are:
- transverse momentum Pt
- pseudo-rapidity η
- azimuthal angle ϕ
- energy E

Classical Benchmark

(Modified) DijetGAN^[5]



Classical Benchmark

Best Result (so far)

- Filter size for Conv2D and Upsample: 3x3, with stride 1x1
- Cross-entropy loss
- Adam with $lr = 10^{-5}$, $\beta_1 = 0.5, \beta_2 = 0.9$
- Wasserstein distance: 0.040866



Next Work Plans, The Quantum (fun) Part

Near-term Quantum-Classical Associative Adversarial Networks (QAAN)^[6]





Variational Quantum Circuit for the QCBM^[7]

[6] Eric R. Anschuetz and Cristian Zanoci. Phys. Rev. A 100, 052327.[7] Rudolph, M. S., et al. arXiv:2012.03924v2

Next Work Plans, The Quantum (fun) Part

Variational Quantum Generators^[8]



[8] Romero, J. and Aspuru-Guzik, A. (2021), Adv. Quantum Technol., 4: 2000003.



Thank You! Any Questions?