Project Idea



Quantstellar

We aim to demonstrate the quantum enhancement of QSVM for big data applications on a novel domain of astronomical object classification using the <u>SDSS</u> (Sloan Digital Sky Survey) dataset. We will benchmark our implementation against a classical SVM algorithm comparing time complexity and preciseness.



The research topic is highly relevant and a potential contribution to an <u>actual project</u> (Fink)

In the astronomy alert system, a lot of research has been conducted to prepare for one of the most ambitious astronomical projects, the so-called Legacy Survey of Space and Time (<u>LSST</u>), to be launched in 2022. LSST aims to chart the universe on a wider and deeper scale than all previous surveys combined. It is projected to capture more than 20 Terabytes of raw data and produce more than 10 million alerts every night.



Quantum enhancements are a solution to the big data challenges in astronomy and beyond

Astronomical alert systems often face a prominent challenge of exponential growth in data volume and algorithm complexity, making it harder to scale and implement complex classification algorithms. These challenges are not unique to astronomy but other fields such as genome sequency or high-frequency trading.

Our project aims to address the challenges of working with big data by leveraging quantum machine learning properties and optimizing QSVM for big data training and classification. We will validate our work using the high dimensional SDSS dataset and have many opportunities of development beyond the time period of the hackathon

Impact and Goals



Quantstellar

Short-term goals, minimum to be implemented within the hackathon timeframe



Big data solution

Show exponential quantum speedup in big data training and classification



Qiskit tutorial

Apply Qiskit's QSVM module to a novel real-life use case, with a ready to contribute Oiskit tutorial



Webapp

Build a simple click-and-run webapp to apply trained model for new predictions

Long-term goals and opportunity for huge impact



LSST 2022

Goal of enabling quantum implementation in the analysis of the LSST 2022 run



Qiskit

Extend Qiskit's QML capability by adding improved functionality for big data training and classification



Quantum plug-in

Extend prototype as a quantum plug-in for current alert system pipelines



Big data beyond astronomy

Big data algorithms can be applied to other projects both within and beyond the field of astronomy

Implementation Plan



Quantstellar

Descriptive Analytics

Load and understand the <u>SDSS</u> dataset

Skillset - data analysis Key Team - Rossy, Sathvik Resources – Kaggle, GitHub repo



Benchmark

Performance testing using classical SVM as a benchmark.

Skillset – Python, Qiskit, QML Key Team – Calum, Zarreen, Rita Resources – SVM repository



Develop

Finalise code for Qiskit Tutorial and develop project beyond SDSS example

 Improve general big data functionality for Qiskit



Quantum ML

Encode data into quantum space, design quantum feature maps and train OSVM model on SDSS dataset

Skillset – Python, Qiskit, QML **Key Team** – Calum, Sathvik, Zarreen **Resources** – <u>Qiskit</u> AER and ML (Aqua), <u>Research paper</u> on <u>QSVM</u>, <u>ML in</u> Astronomy



Prototype

Demo prototype on webapp

Skillset – Python, Streamlit Key Team – Rossy, Rita Resources – Streamlit



The team



Rossy Ng
Software development
Games



Calum Holker

QML

Data Science



Rita Abani Physics Computer Science



Sathvik Lakkaraju
Physics
OMI



Zarreen Reza Software Dev OMI

Resources

- Qiskit QSVM Library -https://qiskit.org/documentation
- OSVM for big data https://arxiv.org/pdf/1307.0471.pdf
- ML in Astronomy https://arxiv.org/pdf/1904.07248.pdf
- SDSS Dataset https://www.sdss.org
- Supervised learning with quantum enhanced feature spaces https://arxiv.org/pdf/1804.11326.pdf
- Streamlit https://streamlit.io