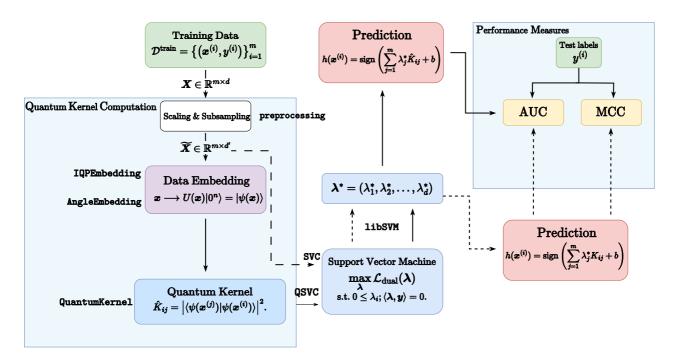
README.md 2024-12-16

# A quantum-based approach for supervised learning (bachelor thesis)

This is the implementation / codebase for comparing classical kernel models with their quantum-hybrid counterparts. We've trained and evaluated these models on real-world biomedical data to investigate whether the quantum-hybrid versions can achieve a better performance than their classical counterparts, potentially demonstrating a quantum advantage.

#### Welcome to QMLab!

The package created is called QMLab - written entirely in Python, using Pennylane for the simulation of a quantum mechanical backend and Sklearn for the classical backend. The main class we use is QSVC - short for Quantum Support Vector Classifier. It inherits from the SVC class of Scikit-Learn and extends its functionality by allowing a quantum kernel as well as classical ones.



Get started with the Jupyter-Tutorial-Series here!

## Information about the data 🧬

We're dealing with 9 different biomedical datasets in the thesis. The collection is coming from the study of *Beinecke & Heider [1]* that compares different augmentation methods for dealing with the imbalance in biomedical data. All of the datasets don't consist of a large number of examples \$m\$, but have the high class imbalance and a large number of features \$d\$.

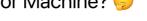
	_
19	
9	
9	_
	9

README.md 2024-12-16

NAME	\$m\$	Cases (+1)	Controls (-1)	\$d\$
WPDC	198	47	151	32
Haberman	306	81	225	3
HCV	546	20	526	12
WDBC	569	212	357	30
CCRF	761	17	744	7
Heroin	942	97	845	11
СТС	1831	176	1655	22

In the original paper different algorithms for data augmentation in order to deal with the imbalance have been compared. We have implemented a quantum kernel classifier based on Havlivcek et al. ~[2] and compared its performance to common kernel classifiers known from traditional machine learning, like the rbf or polynomial kernel.

## What is a Quantum Support Vector Machine? 🧐



Short answer: A quantum-classical hybrid algorithm, where only the kernel is computed on a quantum computer and the rest of the algorithm is completely managed classically.

Long answer: Since real quantum hardware is noisy and prone for errors a whole field submerged of quantum hybrid algorithms which are based on classical learning algorithms but outsource suitable parts to a quantum computer.

The traditional support vector machine algorithm can only classify data that is linearly separable. It becomes way more powerful when we introduce a class of positive semi-definite functions called kernels. Kernels allow us to map our data in a higher, perhaps infinite dimensional space where it is linearly separable. Therefore we are able to classify non-linearly separable data with a linear model.

A quantum kernel is a bivariate, positive semi-definite function that takes on the form

\$\$ k(\boldsymbol{x}, \boldsymbol{x}') = \left|\left\langle \psi(\boldsymbol{x}) | \psi(\boldsymbol{y}) \right\rangle \right|^2.\$\$

The parameterized quantum states are obtained by a data embedding \$\hat{\phi}\$, i.e. a mapping from classical data domain to a quantum hilbert space:

\$\$ \hat{\phi}: \boldsymbol{x} \longrightarrow \Ket{\psi(\boldsymbol{x})}. \$\$

#### Installation / Setup 🥦

For all the following commands we will assume that you opened a shell environment in the qmlsupervised folder.

For Unix OS

README.md 2024-12-16

Create a new python environment using python-command:

```
python —m venv myvirtualenv
```

Activate that environment via

```
source myvirtualenv/bin/activate
```

and then install all the required packages via:

```
pip install -r requirements.txt
```

#### For Windows OS

You need this additional package (Unix users do not need to have this installed apparently)

```
pip install virtualenv
```

Note that this command will install virtualenv globally for your python binary. Then create a virtual environment for installment of the required packages via

```
virtualenv --python C:\Path\To\Python\python.exe myvirtualenv
```

Note that you should replace the path with the location of your global python binary in the command above. Then activate the virtual environment with

```
.\myvirtualenv\Scripts\activate
```

Finally install all of the required packages via

```
pip install -r requirements.txt
```

## Tests 🔵 🤞

After installation you should check that all tests of the project run (Add the -v flag for verbose output):

```
pytest -v
```

README.md 2024-12-16

Additionally run all tests with coverage:

```
pytest -v --cov=src/qmlab tests/
```

#### Please note:

- Code reproducibility has been tested with Docker but let me know if there are any issues!
- Opening a notebook in Google Colab the same steps in installation as for your local machine!

torben.heckes@uni-duesseldorf.de