An Introduction to Quantum Natural Language Processing (QNLP)

Part 2:

Basics of Quantum Machine Learning

Brief Introduction to Machine Learning

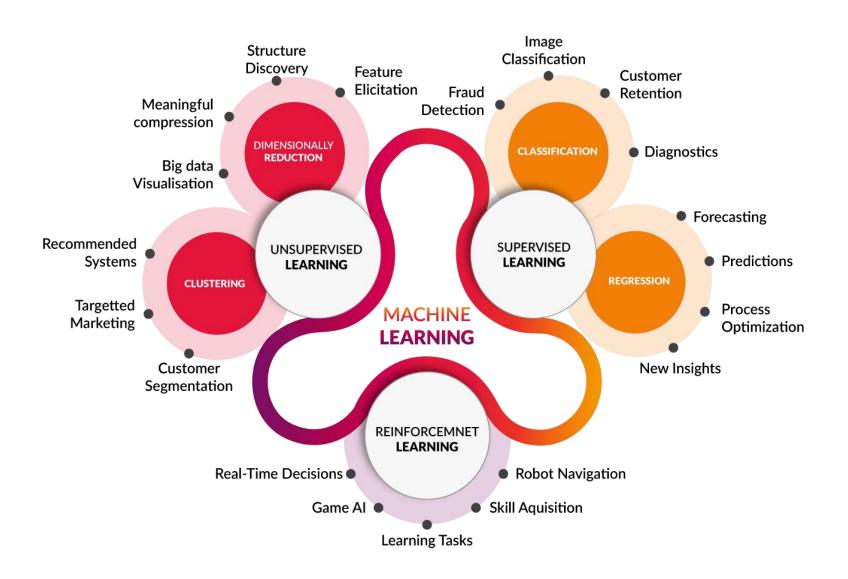
- Brief Introduction to Machine Learning
- Neural Network Basics

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- Quantum Neural Networks Briefly

Brief Introduction to Machine Learning

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Brief Introduction to Machine Learning Machine Learning Algorithms

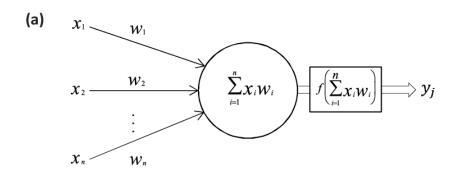
Brief Introduction to Machine Learning

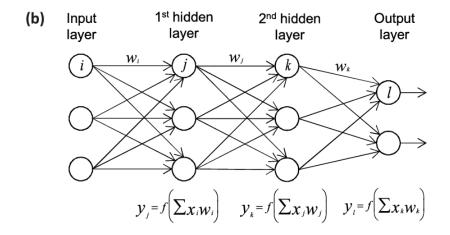
Machine Learning Algorithms

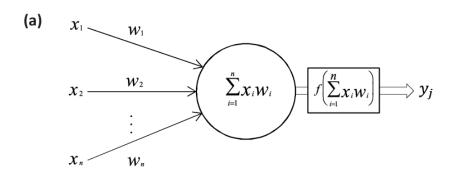
Data Type/Algorithm Type	Unsupervised	Supervised
Continuous	Clustering & Dimensionality Reduction ★ K-Means ★ PCA	Regression ★ Linear ★ Polynomial Decision Trees Random Forests Neural Networks
Categorical	Association Analysis ★ Apriori ★ FP-Growth Hidden Markov Model	Classification ★ KNN ★ Decision Trees ★ Logistic Regression ★ Naive Bayes ★ SVM ★ Neural Networks

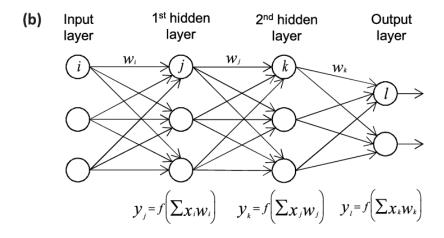
Brief Introduction to Machine Learning

Introduction to Machine Learning Concludes

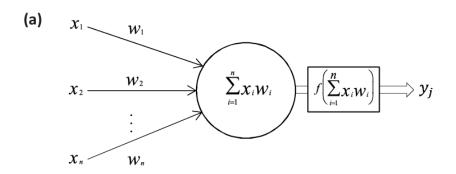


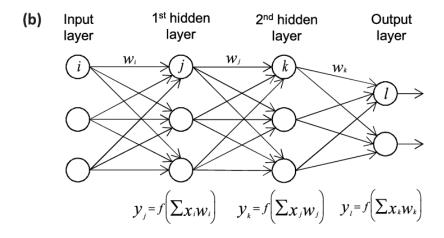






➤ Neural networks have modifiable or trainable (variational) weights





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- ➤ These weights are optimized by an optimizer and learned by the model

Neural Network Basics Concludes

Type of Data/Type of Algorithm	Classical Algorithm	Quantum Algorithm
Classical Data	CC	CQ
Quantum Data	QC	QQ

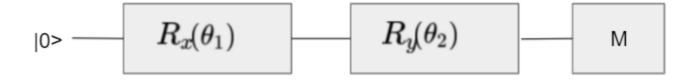
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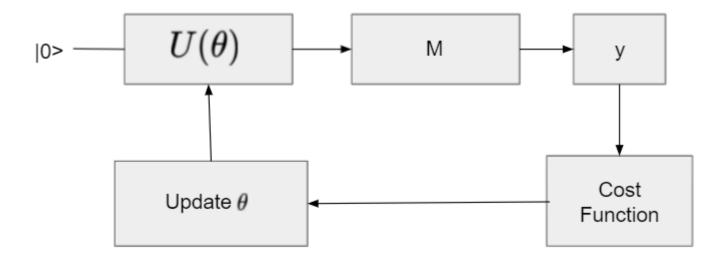
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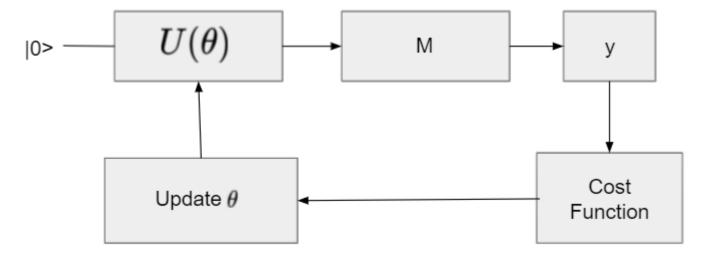
In this circuit θ_1 and θ_2 are the variational parameters. Rx and Ry are rotations around X and Y axis respectively. θ_1 and θ_2 are analogous to trainable weights of a neural network

The basic QML architecture is very similar to that of the classical machine learning models.

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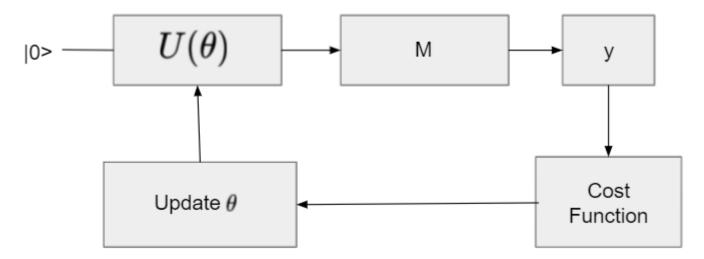


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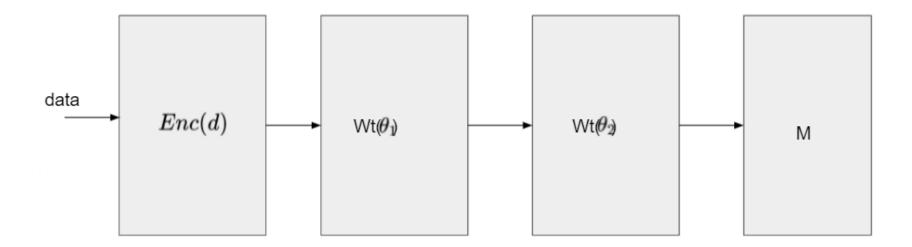


- > Non linearity in a quantum setting can come from the measurements
- \blacktriangleright The parameters θ are learned by the model and optimized by a classical optimizer

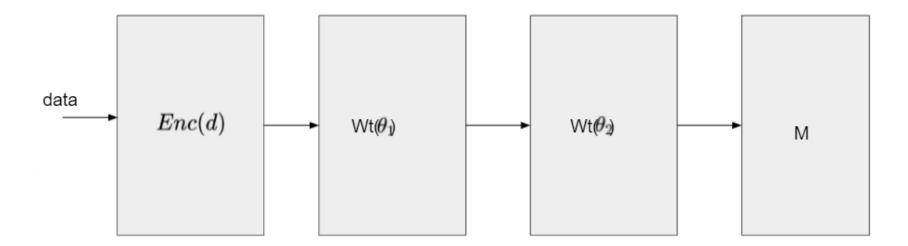
Quantum Machine Learning (QML) Concludes

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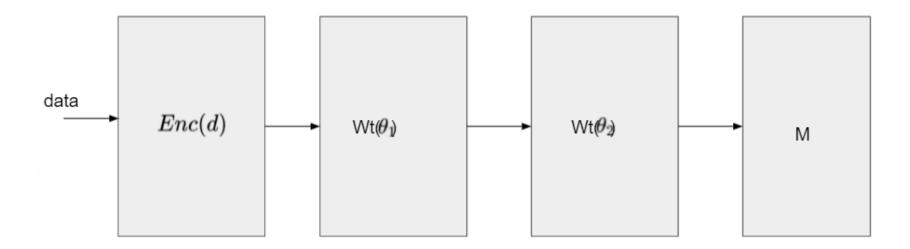


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- $ightharpoonup Wt(\theta_1)$ & Wt(θ_2) are variational circuits analogous to weights present in hidden layers of classical neural networks.

Quantum Neural Networks (QNNs) Concludes

References

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Thank you so much!