Quantum Recurrent Architectures for Natural Language Processing

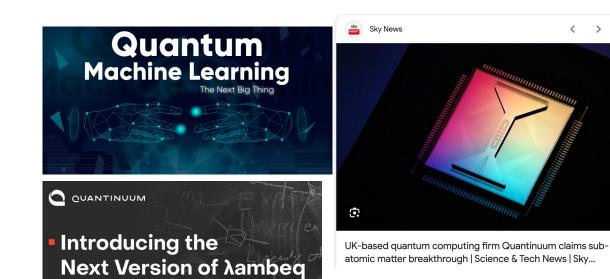
Stephen Clark

CLASP, University of Gothenburg

6 September 2023



Quantum Computing







SECURE TODAY, SECURE TOMORROW



Talk Outline

- 1. Introduction to quantum computing / quantum circuits
- 2. Application to sequence classification
 - a. our quantum RNN architectures
 - b. sentiment analysis experiments



The State of a Classical Bit

ψ _____0

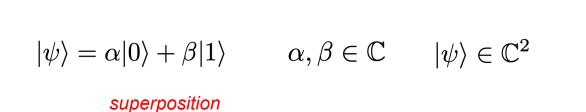


The State of a Classical Bit

ψ _____1



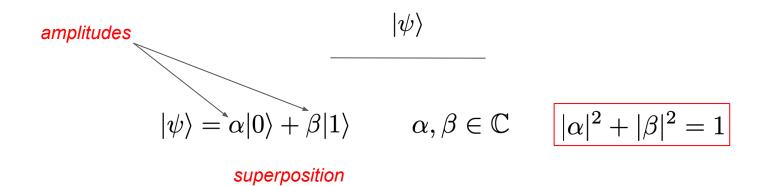
The State of a Qubit



 $|\psi\rangle$

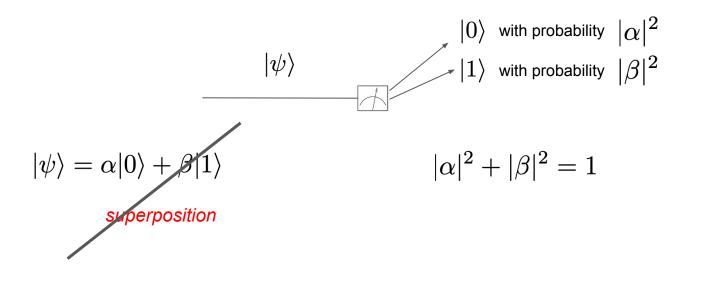


The State of a Qubit



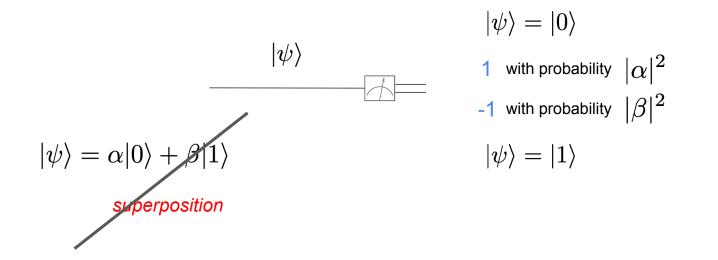


Measuring a Qubit



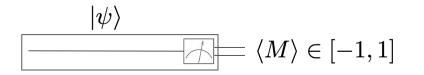


Measuring a Qubit (scalar output)



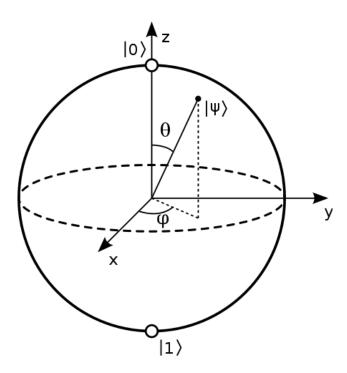


Measuring a Qubit (many times)





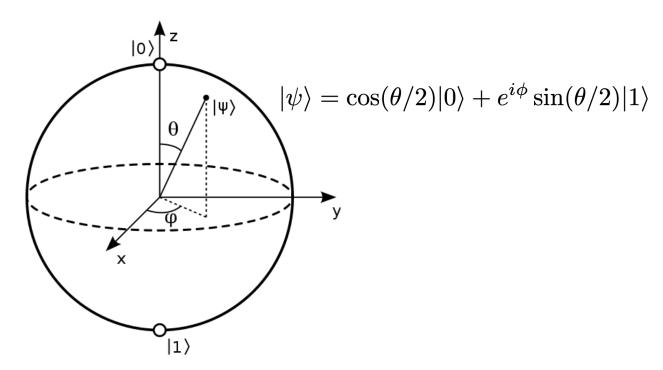
The Bloch Sphere Representation of a Qubit



 $|\psi\rangle = \alpha|0\rangle + \beta|1\rangle$

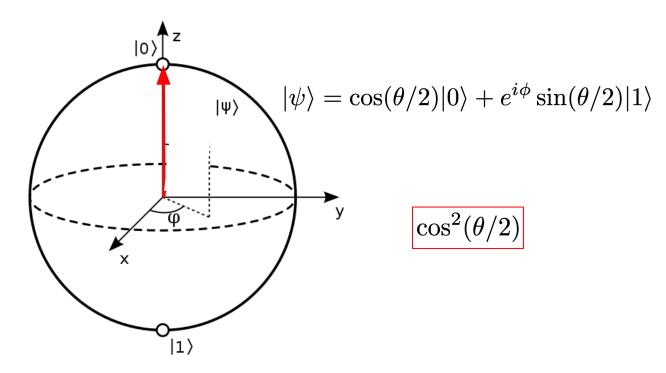


The Bloch Sphere Representation of a Qubit



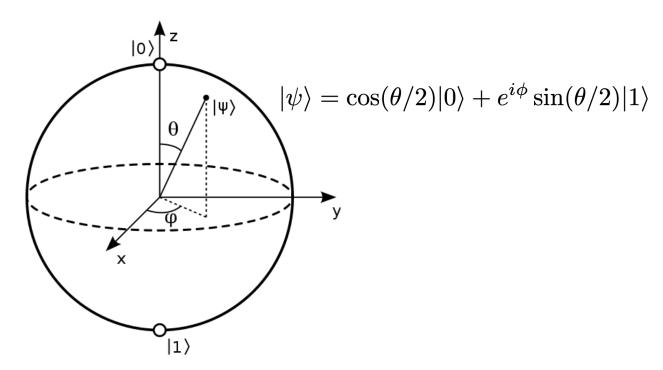


"The Collapse of the Wave Function"



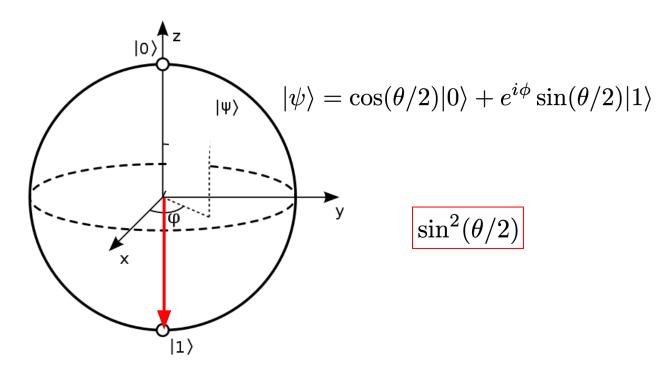


The Bloch Sphere Representation of a Qubit



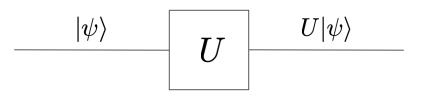


"The Collapse of the Wave Function"





Unitary Transformations of a Qubit

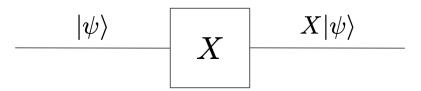


$$U:\alpha|0\rangle+\beta|1\rangle\mapsto\alpha'|0\rangle+\beta'|1\rangle$$

$$|\alpha'|^2+|\beta'|^2=1$$



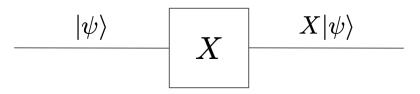
quantum Not gate



 $X: |0
angle \mapsto |1
angle \ X: |1
angle \mapsto |0
angle$



quantum Not gate acts linearly



$X:\alpha|0\rangle+\beta|1\rangle\mapsto\alpha|1\rangle+\beta|0\rangle$

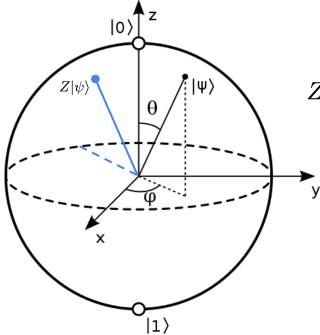


Pauli Z Gate $|\psi angle = Z |\psi angle$

$Z:\alpha|0\rangle+\beta|1\rangle\mapsto\alpha|0\rangle-\beta|1\rangle$



Pauli Z Gate rotates about the Z axis



 $Z:\alpha|0\rangle+\beta|1\rangle\mapsto\alpha|0\rangle-\beta|1\rangle$

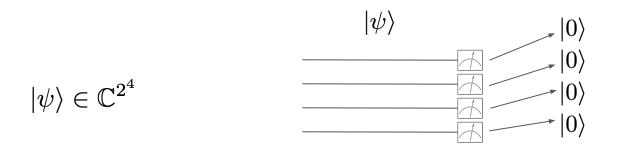


The State of Many Qubits



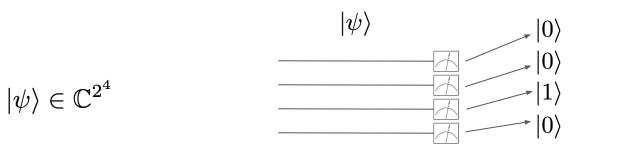
 $|\psi\rangle = \alpha_{0000}|0000\rangle + \alpha_{0001}|0001\rangle + \alpha_{0010}|0010\rangle + \dots + \alpha_{1111}|1111\rangle$





$$|\psi\rangle = \alpha_{0000} |0000\rangle + \alpha_{0001} |0001\rangle + \alpha_{0010} |0010\rangle + \dots \alpha_{1111} |1111\rangle$$
$$|\alpha_{0000}|^2$$

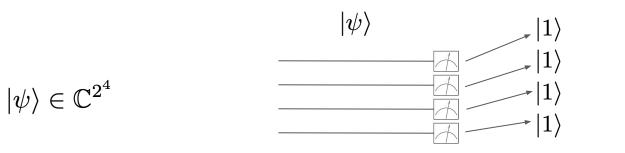




 $|\psi\rangle = \alpha_{0000}|0000\rangle + \alpha_{0001}|0001\rangle + \alpha_{0010}|0010\rangle + \dots + \alpha_{1111}|1111\rangle$

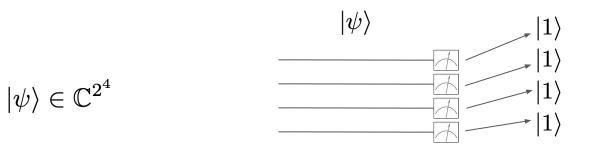
 $|lpha_{0010}|^2$

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 $|\psi\rangle = \alpha_{0000}|0000\rangle + \alpha_{0001}|0001\rangle + \alpha_{0010}|0010\rangle + \dots \alpha_{1111}|1111\rangle$ $|\alpha_{1111}|^2$

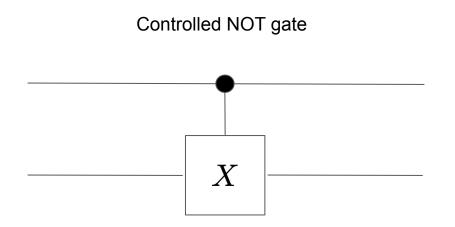




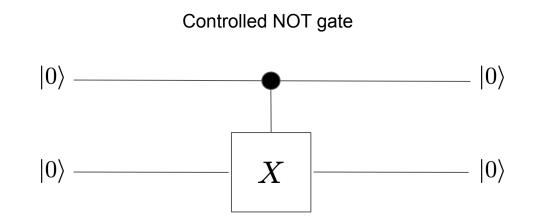
 $|\psi\rangle = \alpha_{0000}|0000\rangle + \alpha_{0001}|0001\rangle + \alpha_{0010}|0010\rangle + \dots \alpha_{1111}|1111\rangle$

$$\sum_{b \in \{0,1\}^4} |\alpha_b|^2 = 1$$

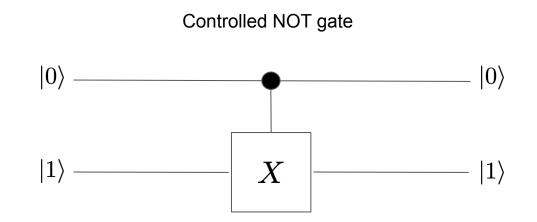




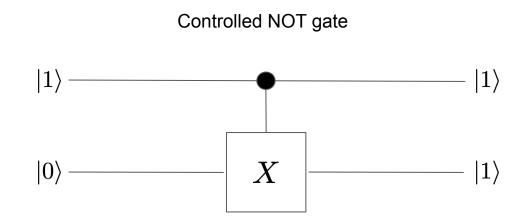




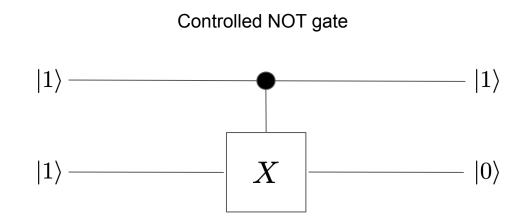




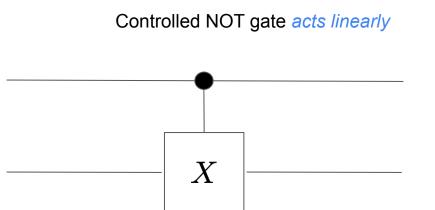








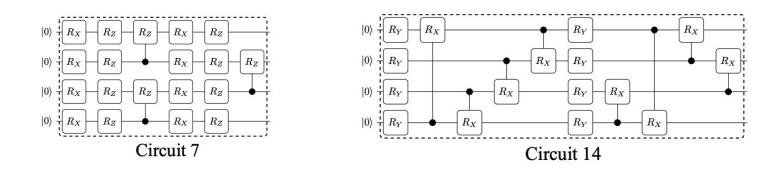




 $CX: \alpha_{00}|00\rangle + \alpha_{01}|01\rangle + \alpha_{10}|10\rangle + \alpha_{11}|11\rangle \mapsto \alpha_{00}|00\rangle + \alpha_{01}|01\rangle + \alpha_{10}|11\rangle + \alpha_{11}|10\rangle$



Quantum Circuits

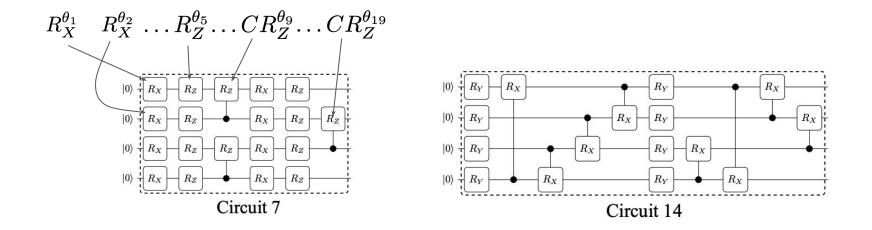


 $\begin{array}{c} \text{Expressibility and entangling capability of parameterized quantum circuits for hybrid} \\ \text{quantum-classical algorithms} \end{array}$

Sukin Sim,^{1,2,*} Peter D. Johnson,² and Alán Aspuru-Guzik^{2,3,4,5,†}



Parameterised Quantum Circuits (PQCs)



Expressibility and entangling capability of parameterized quantum circuits for hybrid quantum-classical algorithms

Sukin Sim,^{1,2,*} Peter D. Johnson,² and Alán Aspuru-Guzik^{2,3,4,5,†}

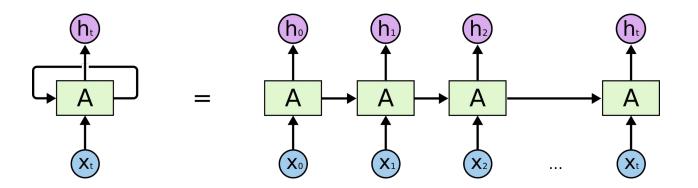


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Recurrent Neural Networks (RNNs)

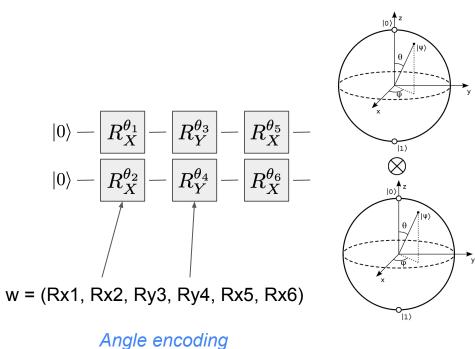


 $h_t = f(x_t \mathbf{U} + h_{t-1} \mathbf{W})$

From Colah's blog: https://colah.github.io/posts/2015-08-Understanding-LSTMs/

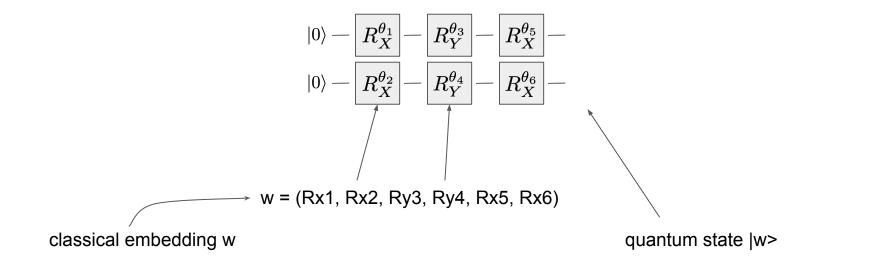


Angle Encoding for Words

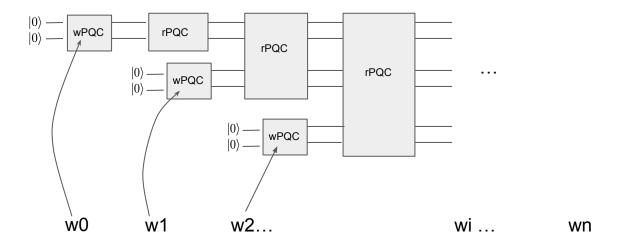




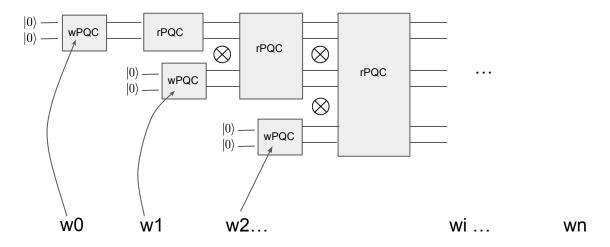
Angle Encoding for Words







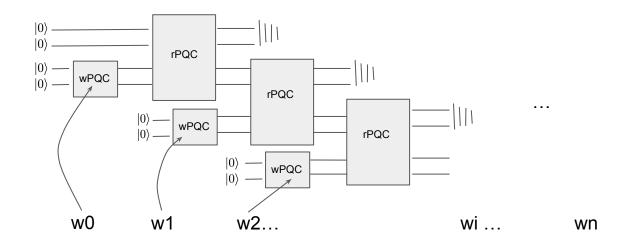




tensor product

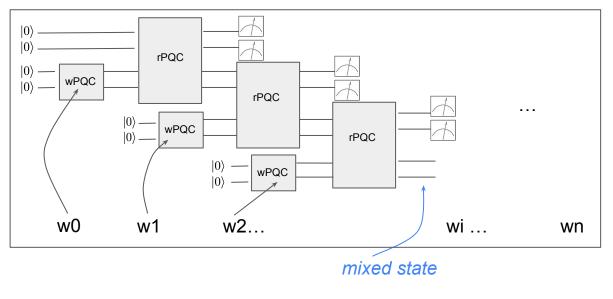


discarding



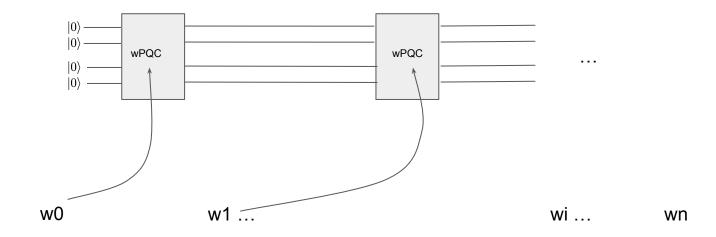


discarding - "measure and ignore"



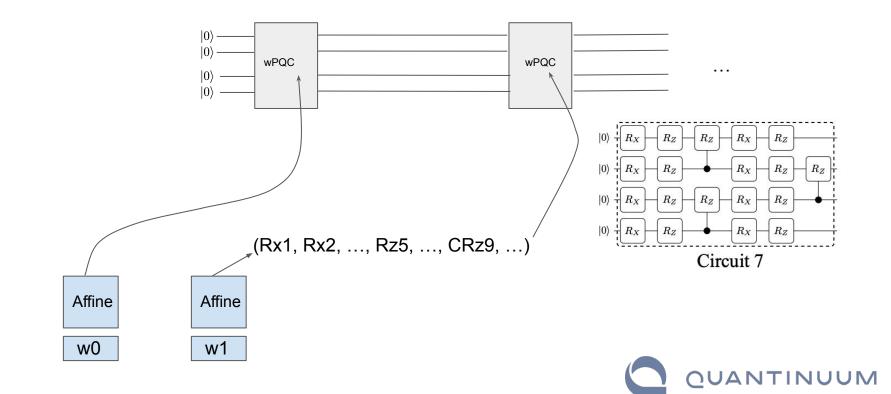


qRNN Take Two

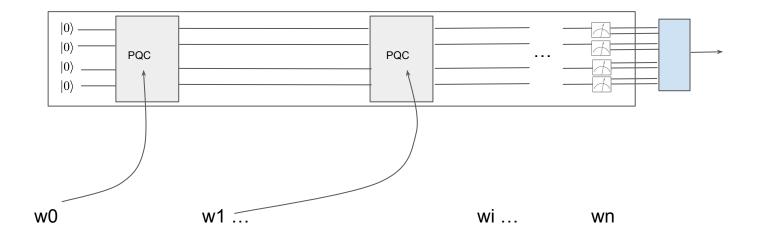




qRNN Take Two

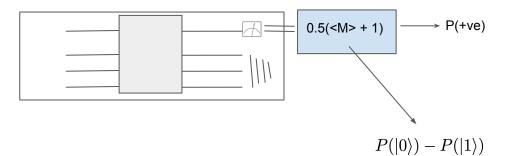


Output



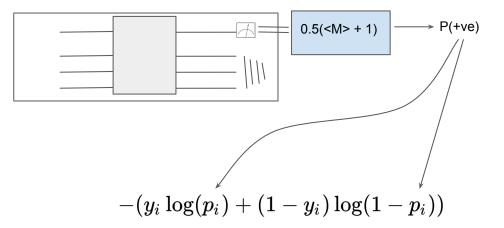


Probabilistic Output



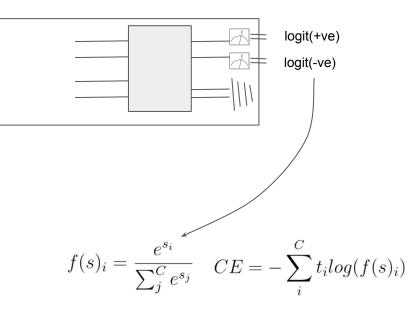


Probabilistic Output for Training



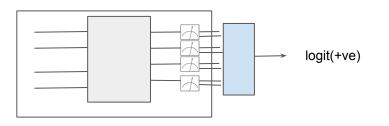


Logits Output





Neural Output





The Task

- Sentiment analysis (Rotten Tomatoes dataset)
- 8,530 training examples (well balanced); 1,066 dev examples
- Simple binary classification task

if you sometimes like to go to the movies to have fun , wasabi is a good place to start . emerges as something rare , an issue movie that's so honest and keenly observed that it doesn't feel like one .

simplistic , silly and tedious . it's so laddish and juvenile , only teenage boys could possibly find it funny . 0



Baseline / Goal

- Goal is *not* to beat the s-o-t-a
- Goal (at this stage) is to be competitive with a classical vanilla RNN



Hybrid Toolkits





Hybrid Toolkit

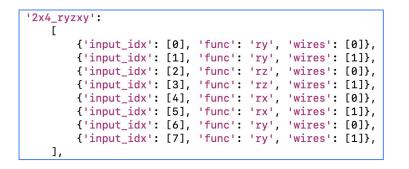
- Requirements for classical simulation:
 - easily interfaces with PyTorch (or TensorFlow, JAX, ...)
 - fast to train on real-world datasets
 - accommodates batching
 - essentially PyTorch ML library with complex number linear algebra

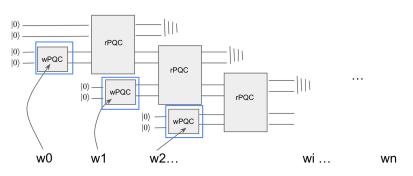




"Stairs" Architecture in Practice

- We added density matrices to TorchQuantum (for mixed states)
- Choice of PQC:

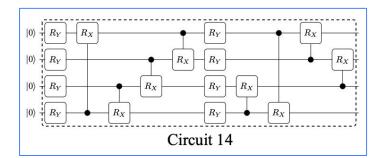


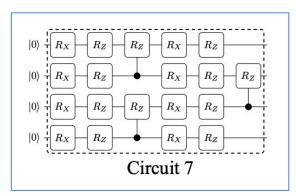


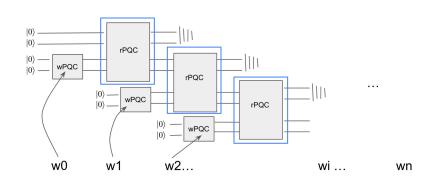
discarding



"Stairs" Architecture in Practice





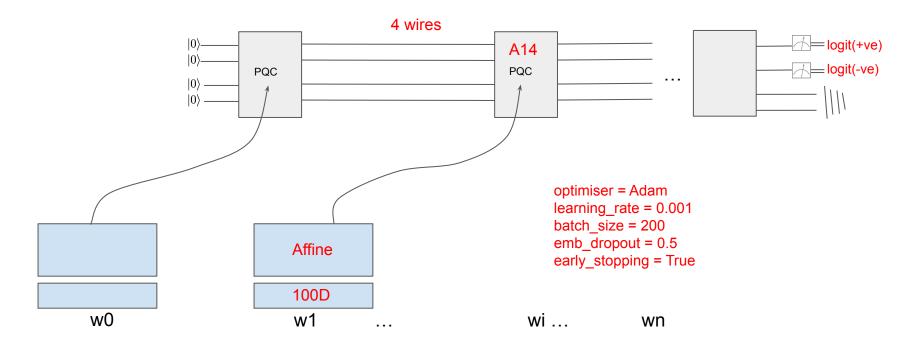


Expressibility and entangling capability of parameterized quantum circuits for hybrid quantum-classical algorithms

Sukin Sim,^{1, 2, *} Peter D. Johnson,² and Alán Aspuru-Guzik^{2, 3, 4, 5, †}



Example Experimental Settings



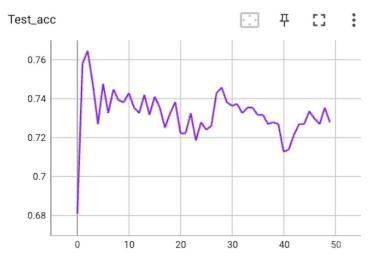


Results on RT Test Set

4 wires	Acc
Classical RNN	77.2
"Flat"	77.9
"Stairs"	79.5
Classical GRU	78.7
Classical LSTM	79.4
LSTM (Dai and Le, 2015)	79.7



Learning Curve

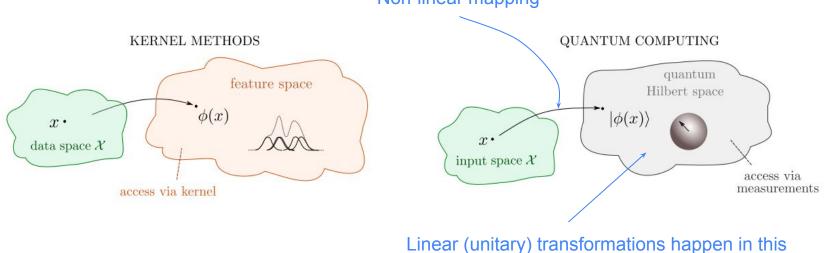


NVidia A30 GPU, PyTorch 1.12:

~5 secs / epoch for 1 wire (pure state) ~11 secs / epoch for 2 wires (pure state) ~14 secs / epoch for 4 wires (pure state) ~26 secs / epoch for 8 wires (pure state)



Where's the (Potential) Advantage?



Non-linear mapping

(potentially very large) space

Supervised quantum machine learning models are kernel methods

Maria Schuld Xanadu, Toronto, ON, M5G 2C8, Canada

So What's the Current State of Quantum Hardware?

Quantinuum H-Series quantum computer accelerates through 3 more performance records for quantum volume: 2¹⁷, 2¹⁸, and 2¹⁹

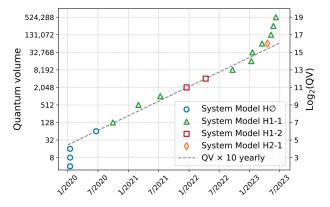


Figure 1: H-series progress quantum volume improvement trajectory

June 30, 2023

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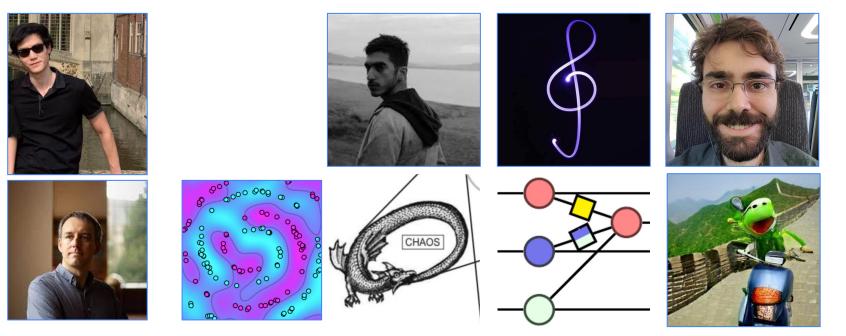


Future Work

- Apply the models to more tasks
 - sequence labelling, language modelling, translation, ...
- Apply pre-training / fine-tuning paradigm
- Develop more hybrid architectures
 - based on CNNs (e.g. MERA-like), transformers, ...
- Run on quantum hardware



The Oxford Hybrid NLP Team



Wenduan Xu, Konstantinos Meichanetzidis, Douglas Brown, Gabriel Matos, Charlie London, Richie Yeung, Carys Harvey, Nikhil Khatri, Stephen Clark



The Future is (Almost) Here

