

## Project plan

### Warm up

- Change from 0&1 → 3 & 6
- Change from 9x1 input to 3x3

## 1 Image encoding

Using circuit structure in lecture notes with fixed layers (=2)

Adjoint diff method + ADAM optimiser

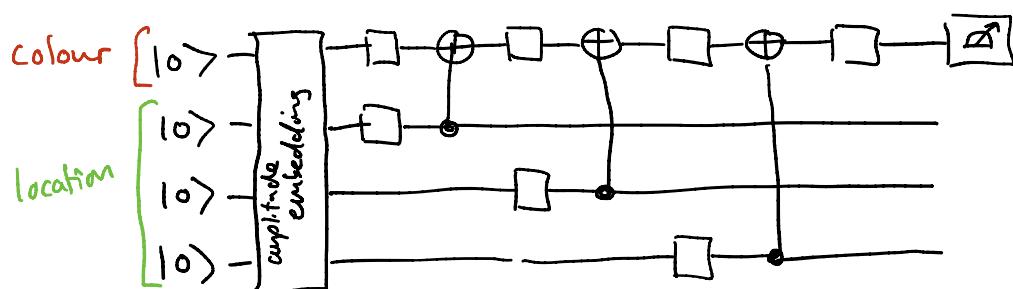
Binary classification 0 & 1 and 3 & 6

- Basis encoding (9 qubits - 9x1 image)
- Amplitude encoding (10 qubits - full image using amplitude embedding)
- FRQI (11 qubits - full image using amplitude encoding).

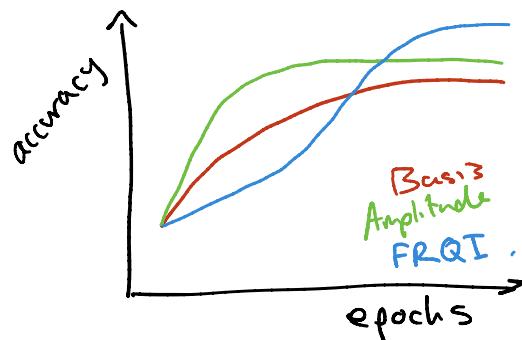
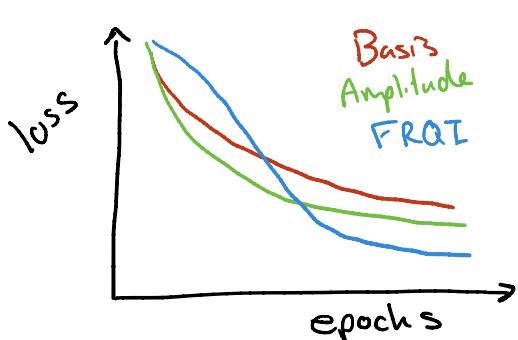
For FRQI, there is a choice for the measured qubit.

Could be colour qubit, or could be some other.

e.g.



Compare performance (final value + optimisation speed)



## 2 Different ansatz

Adjoint diff method + ADAM optimiser

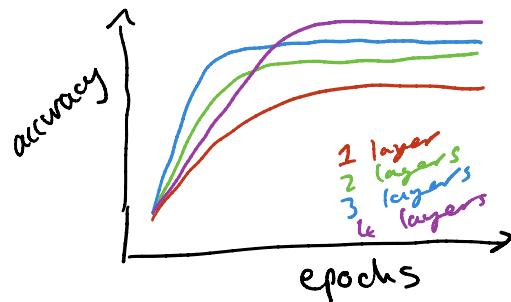
Binary classification 0 & 1

Start by fixing encoding (to best from previous)

Bonus: Can also compare different encodings!

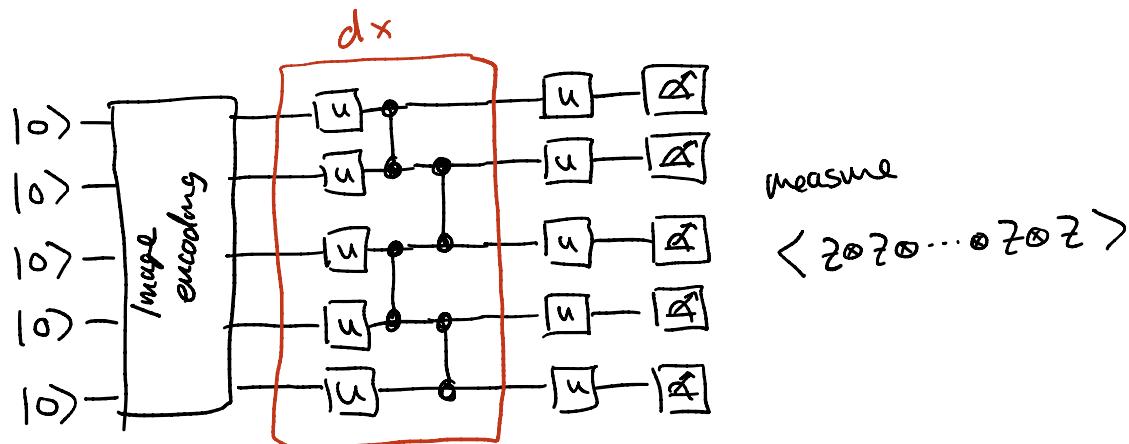
- Ansatz from lecture notes.

- Vary number of layers ( $d = 1, 2, 3, 4$ )



- New ansatz / measurement.

e.g.



for varying layers  $d = 1, 2, 3, 4$ . (plot accuracy / loss)

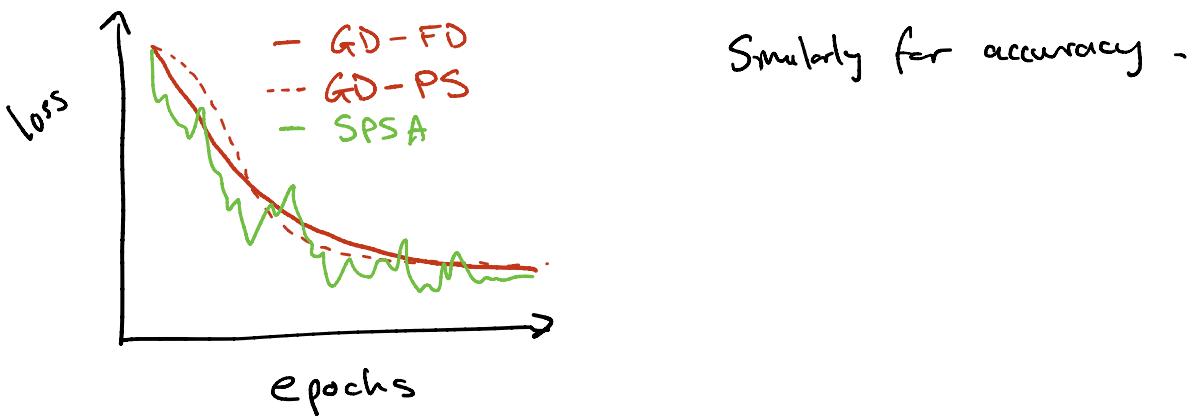
### 3 Change optimization

Binary classification.

Choose a circuit ansatz (Bonus: could also compute)

Choose an image encoding (Bonus: could also compute)

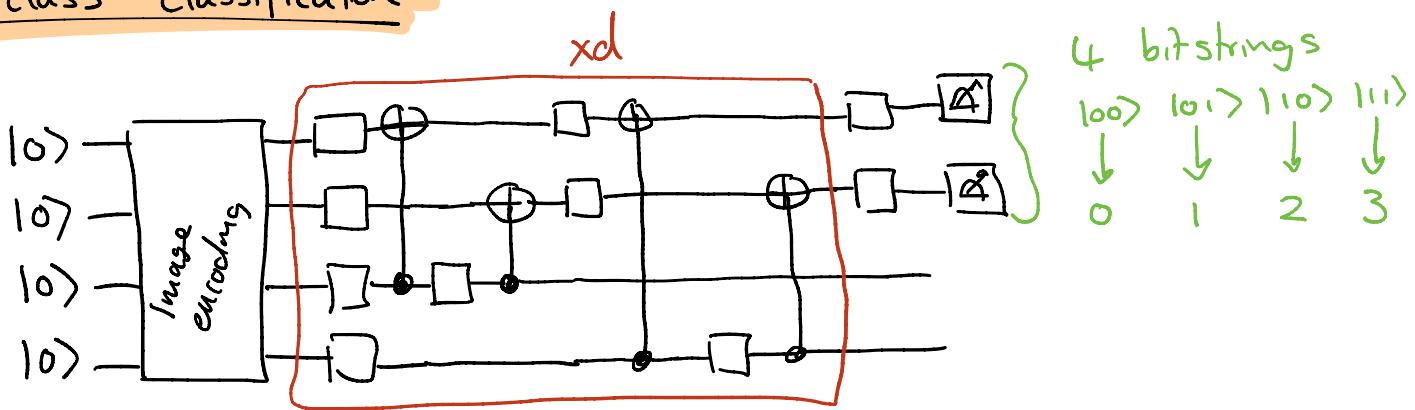
- Gradient descent. (pennylane Gradient Descent Optimizer)
  - diff-method = "finite-diff"
  - diff-method = "parameter-shift"
  - Start with default step size. You may want to increase.
- SPSA (pennylane SPSA Optimizer)
  - doesn't use gradients.
  - Start with default parameters.



Similarly for accuracy -

## 4 Multi-class classification

E.g.



## 5 Fashion MNIST

Repeat steps using FashionMNIST