Lab-02: Quantum Encoding - Basis, Angle, Amplitude and Qsample/Arbitrary Encoding

Data Representation for classical machine learning, how to represent data numerically, it can be processed by classical machine learning algorithms?

How to represent and efficiently input the data into a quantum system such that it can be processed by a quantum machine learning algorithm? Quantum Data Encoding/Embedding for Quantum Machine Learning Algorithm.

Let's consider a classical dataset X consisting of M samples, each with N features $\$ \class{script-x}{\mathscr{X}} = \class{brace}{\{\}x^{(1)},\class{ellipsis}{\dots},\csid{_x-\lil-m}{x^{(m)}},\dots,x^{(M)}\class{brace}{\}}\\$\$

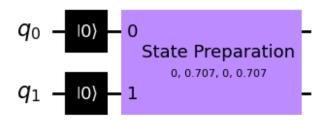
where $x^{(m)}$ is an N dimensional vector for $m=1,\ldots,M$. To represent this dataset in a qubit system, we can use various embedding techniques - from Qiskit Textbook

1. Basis Encoding

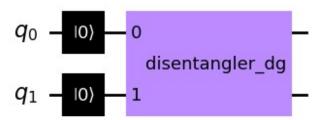
```
import math
from qiskit import QuantumCircuit
from qiskit.quantum info import Statevector
from qiskit.circuit.library import ZZFeatureMap
from qiskit.circuit.library import EfficientSU2
# data that we want to encode
state = [0, 1/math.sqrt(2), 0, 1/math.sqrt(2)]
display(state)
[0, 0.7071067811865475, 0, 0.7071067811865475]
# Let's create a gubit state vector and check is it valid state or not
s = Statevector(state)
s.is valid()
True
# Creating a quantum circuit with 2 qubit
qc = QuantumCircuit(2)
# initialize the quantum circuit's each qubit with the classical data.
qc.initialize(state, [0, 1])
gc.draw(output="mpl")
```

$$q_0 \cdot 0$$
 | ψ | $q_1 \cdot 1$ | $q_1 \cdot q_2 \cdot q_3 \cdot q_4 \cdot q_5 \cdot q_6 \cdot q_7 \cdot q_8 \cdot q_8$

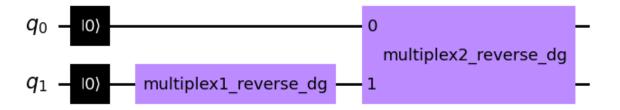
To decompose one level (shallow decompose).
Quantum Circuit with one level decompose.
qc.decompose().draw(output="mpl")



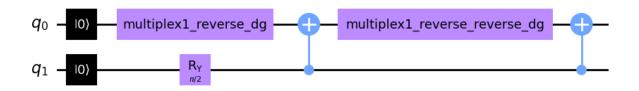
circuit.decompose().decompose(). can decompose specific gates
specific time
qc.decompose().decompose().draw(output="mpl")



qc.decompose().decompose().draw(output="mpl")



qc.decompose().decompose().decompose().draw(output="mpl")



Amplitude Encoding

Amplitude encoding encodes data into the amplitudes of a quantum state. It represents a normalised classical N-dimensional data point, x, as the amplitudes of a n-qubit quantum state, $i \psi_x$):

$$\left|\psi_{x}\right| = \sum_{i=1}^{N} x_{i} i$$

where $N = 2^n$, X_i is the i^{th} element of X and i i is the i computational basis state.

```
state = [
    1 / math.sqrt(15.25) * 1.5,
    0,
    1 / math.sqrt(15.25) * -2,
    1 / math.sqrt(15.25) * 3]
display(state)

[0.3841106397986879, 0, -0.5121475197315839, 0.7682212795973759]
s = Statevector(state)
s.is_valid()

True

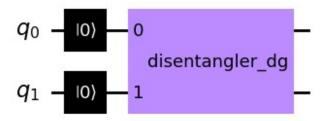
qc = QuantumCircuit(2)
qc.initialize(state, [0, 1])
qc.draw(output="mpl")
```

$$q_0 - 0$$
 $|\psi\rangle$
 $q_1 - 1$
 $[0.384, 0, -0.512, 0.768]$

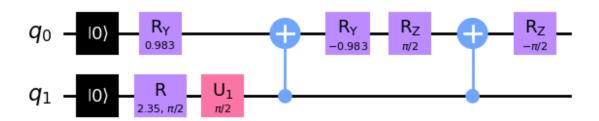
qc.decompose().draw(output="mpl")

$$q_0 - |0\rangle - 0$$
State Preparation
 $q_1 - |0\rangle - 1$
- State Preparation
 $q_1 - |0\rangle - 1$

qc.decompose().decompose().draw(output="mpl")



qc.decompose().decompose().decompose().decompose().draw(ou
tput="mpl")



Angle Encoding

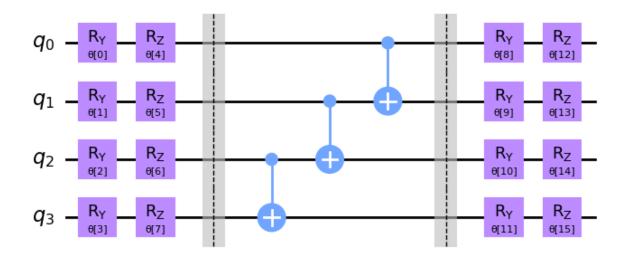
```
qc = QuantumCircuit(4)
qc.ry(0, 0)
qc.ry(2*math.pi/4, 1)
qc.ry(2*math.pi/2, 2)
qc.ry(2*math.pi, 3)
qc.draw(output="mpl")
```

$$q_0 - \frac{R_Y}{0} - \frac{R_Y}{0}$$

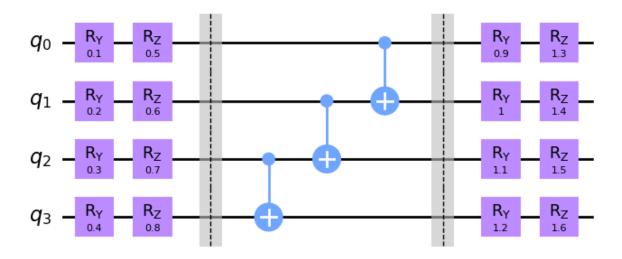
Arbitary Encoding

Arbitrary encoding encodes N features as rotations on N parameterized gates on n qubits, where $n \le N$. Like angle encoding, it only encodes one data point at a time, rather than a whole dataset.

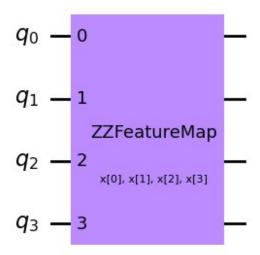
```
# Specifies how often the structure of a rotation layer followed by an
entanglement
# The number of qubits of the EfficientSU2 circuit.
# If True, barriers are inserted in between each layer. If False,
qc = EfficientSU2(num_qubits=4, reps=1, insert_barriers=True)
qc.decompose().draw(output="mpl")
```



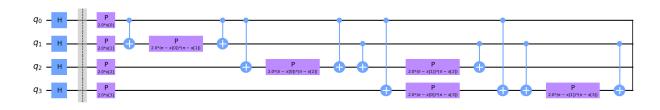
```
x = [0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1.0, 1.1, 1.2, 1.3, 1.4, 1.5, 1.6]
encode = qc.bind_parameters(x)
encode.decompose().draw(output="mpl")
```

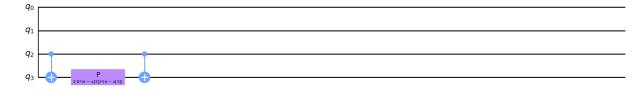


```
# Create a new second-order Pauli-Z expansion.
circuit = ZZFeatureMap(4, reps=1, insert_barriers=True)
circuit.draw(output="mpl")
```

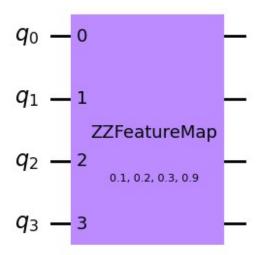


circuit.decompose().draw(output="mpl")

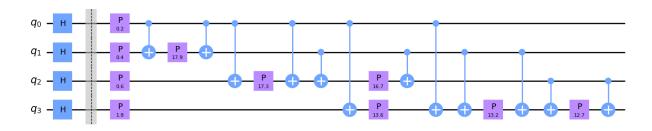




```
x = [0.1, 0.2, 0.3, 0.9]
encode = circuit.bind_parameters(x)
encode.draw(output="mpl")
```



encode.decompose().draw(output="mpl")



import qiskit.tools.jupyter
%qiskit_version_table

<IPython.core.display.HTML object>