Exercises for Introduction to Quantum Computing

Name: Pugazharasu Anancia Devaneyan (s6puanan) Matriculation number: 3300280

In []: #Importing the required Libraries
import matplotlib.pyplot as plt
import numpy as np
from qiskit import *

1 Grover's algorithm

b) We know from the way the Grover's agorithm is designed, the probability of finding the item after t iterations is given by,

$$P(ext{ success }) = \left|\left\langle x^* \mid \psi
ight
angle
ight|^2 = \sin^2\!\left(rac{2t+1}{\sqrt{N}}
ight)$$

We are to compute the largest value of $N=2^n$ such that P(success)=1 for a single iteration i.e. t=1, thus we are to solve

$$P(ext{ success }) = \left| \left\langle x^* \mid \psi
ight
angle
ight|^2 = \sin^2 \! \left(rac{3}{\sqrt{N}}
ight) = 1$$

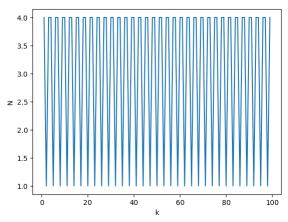
for N. Solving this, we find,

$$N = 4$$

This is exemplified by the following plot for N vs k where k signifies the multivalued nature of the sine function.

```
In [ ]: n = np.arange(1,100, dtype = int)
   plt.xlabel("\n")
   plt.ylabel("\n")
   plt.plot(n,[1/(np.sin((np.pi * (i-0.5) )/3))**2) for i in n])
```

Out[]: [<matplotlib.lines.Line2D at 0x18364b177c0>]



2 Error mitigation

a) The Hamiltonian.

$$H=X_0\otimes Y_1$$

when exponentiated, takes the form,

$$U = e^{-iH\delta t} = (\mathbb{I}_n \otimes \mathbb{I}_n) \cdot \cos(\delta t) - i\sin(\delta t) \cdot (X_0 \otimes Y_1)$$

Computing the expectation value

$$\langle \psi \left| Z_0 \otimes Z_1 \right| \psi \rangle$$

for $\delta t=1$ and

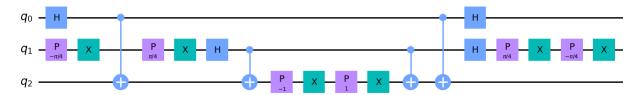
$$|\psi\rangle=e^{-iH\delta t}|00\rangle$$

, we have,

$$\langle \psi \, | Z_0 \otimes Z_1 | \, \psi \rangle = 1$$

b) Running the code in the given Juypter notebook, we find that probabilistic error cancellation gives the best result (see below for details).

```
In [ ]: # e^i theta Z
def Rz(qc, theta,i):
    qc.p(theta,i)
    qc.x(i)
    qc.p(-theta,i)
    qc.x(i)
                         return qc
                 def expiHdt(circuit,dt, r):
    qc=QuantumCircuit(3)
    qc.h(0)
                         qc.cx(0,2)
                         qc = Rz(qc, -np.pi/4,1)
qc.h(1)
                         qc.cx(1,2)
                         qc = Rz(qc, -dt, 2)
                         ac.cx(1,2)
                         qc.k(1,2)
qc.h(1)
qc = Rz(qc, np.pi/4,1)
qc.cx(0,2)
                         qc.h(0)
                         circuit.append( qc, [0,1,2])
                         return circuit
                 from qiskit_ibm_runtime import QiskitRuntimeService, Sampler, Estimator, Session, Options
service = QiskitRuntimeService(channel="ibm_quantum")
backend = "ibmq_qasm_simulator"
                 backend = "ibmq_qasm_simulator"
from qiskit.providers.fake_provider import FakeManila
from qiskit_aer.noise import NoiseModel
                # Make a noise model
fake_backend = FakeManila()
noise_model = NoiseModel.from_backend(fake_backend)
                 from qiskit.quantum_info import SparsePauliOp
ZZ = SparsePauliOp("IZZ")
                 qc=QuantumCircuit(3)
                qc=expiNdt(qc,1,0)
qc = qc.remove_final_measurements(inplace=False)
display(qc.decompose().draw('mpl'))
                 noise_options = Options()
noise_options.simulator = {
                         "noise_model": noise_model,
"basis_gates": fake_backend.configuration().basis_gates,
"coupling_map": fake_backend.configuration().coupling_map,
"seed_simulator": 42
                noise options.execution.shots = 1000
                with Session(service-service, backend=backend):
    noise_options.resilience_level = 0
    estimator = Estimator(options=noise_options)
    job = estimator.run( circuits=[qc],
    # parameter_values=individual_phases,
                        # parameter_values=individual_phases,
observables=[Z] )
print("level 0: ",job.result() )
noise_options.resilience_level = 1
estimator = Estimator(options-noise_options)
job = estimator.run( circuits=[qc],
    # parameter_values=individual_phases,
    observables=[ZZ] )
print("level 1: ",job.result() )
                         noise_options.resilience_level = 2
                        noise options.resilience level = 3
```



level 0: EstimatorResult(values=array([0.822]), metadata=[{'variance': 0.3243160000000005, 'shots': 1000}])
level 1: EstimatorResult(values=array([0.8407225]), metadata=[{'variance': 1.5997098182629468, 'shots': 1008, 'readout_mitigation_num_twirled_circuits': 16, 'readout_mitigation_shots_calibration': 8192}])
level 2: EstimatorResult(values=array([0.60933333]), metadata=[{'zne': {'noise_amplification': {'noise_amplifier': "<TwoQubitAmplifier:{'noise_factor_relative_tolerance':
0.01, 'random_seed': None, 'sub_folding_option': 'from_first')', 'noise_factors': [1, 3, 5], 'values': [0.52, 0.788, 0.52], 'variance': [0.7296, 0.3790559999999995, 0.729
6], 'shots': [1000, 1000, 1000]), 'extrapolation': {'extrapolation': \linearExtrapolator'}}}])]
level 3: EstimatorResult(values=array([0.99423044]), metadata=[{'standard_error': 0.00780675664913258, 'confidence_interval': [0.9083033453137671, 1.0801575320869066], 'confidence_level': 0.95, 'shots': 157184, 'samples': 1228, 'sampling_overhead': 1.2289498202368314, 'total_mitigated_layers': 4}])