Grover

April 10, 2020

1 Grover's Search Algorithm

This script executes the Grover search algorithm on two qbits, with a randomized solution, using the devices at the IBM Quantum Experience. Two qbit tomography and readout calibration are used to find the resulting state, before calculating the fidelity of the output to the expected output.

```
[1]: #importing necessary modules
     import numpy as np
     from random import randint
     %config InlineBackend.figure_format = 'svg' # Saves plots in svg format
     # qiskit modules
     import qiskit
     from qiskit import ClassicalRegister, QuantumRegister, QuantumCircuit, Aer
     from qiskit.tools.visualization import plot_state_city, plot_bloch_multivector,_
     →plot_state_paulivec
     from qiskit.tools.monitor import job_monitor
     from qiskit.quantum_info import state_fidelity, DensityMatrix
     from qiskit.quantum_info.operators import Operator, Pauli
     # tomography functions
     from qiskit.ignis.verification.tomography import state_tomography_circuits,u
      \rightarrowStateTomographyFitter
     # readout calibration
     from qiskit.ignis.mitigation.measurement import complete meas_cal,_
      →CompleteMeasFitter
```

```
[2]: # importing IBMQ backends
from qiskit import IBMQ
ibmq_provider = IBMQ.load_account() # credentials stored on disk
device = ibmq_provider.get_backend('ibmqx2')

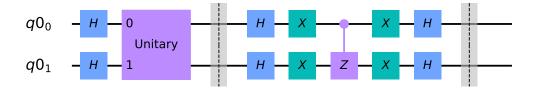
# necessary imports for noise modelling
from qiskit.providers.aer.noise import NoiseModel
noise_model = NoiseModel.from_backend(device)
coupling_map = device.configuration().coupling_map
basis_gates = noise_model.basis_gates
```

1.0.1 Defining the circuit

```
[3]: dim = 2
     qr= QuantumRegister(dim)
     qc = QuantumCircuit(qr)
     qc.h(range(dim))
     # creating a randomized Grover operator
     a = randint(0,2**dim-1)
     A = np.identity(2**dim)
     A[a,a] = -1
     print(A)
     Grov = Operator(A)
     qc.append(Grov,range(dim))
     qc.barrier()
     qc.h(range(dim))
     qc.x(range(dim))
     qc.cz(0,1)
     qc.x(range(dim))
     qc.h(range(dim))
     qc.barrier()
     qc.draw(output='mpl')
```

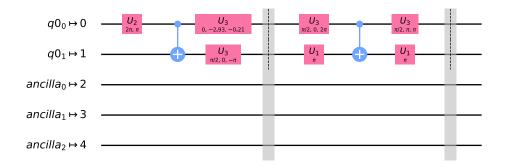
```
[[ 1. 0. 0. 0.]
 [ 0. -1. 0. 0.]
 [ 0. 0. 1. 0.]
 [ 0. 0. 0. 1.]]
```

[3]:



```
[4]: # what does this circuit look like on the device?
transpile_qc = qiskit.compiler.transpile(qc,device)
transpile_qc.draw(output='mpl')
```

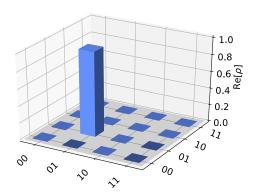
[4]:

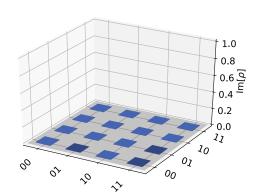


```
[5]: # visualizing the expected state
   job = qiskit.execute(qc, Aer.get_backend('statevector_simulator'))
   psi_expected = job.result().get_statevector(qc)
   print('The expected state is:', psi_expected)
   plot_state_city(psi_expected, title='Expected state')
```

The expected state is: [3.92523115e-17-2.6147292e-32j -1.00000000e+00+2.4492936e-16j 3.92523115e-17-2.6147292e-32j 1.11022302e-16-1.2246468e-16j]

[5]: Expected state



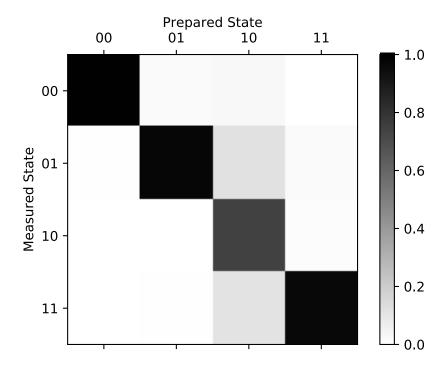


1.0.2 Readout correction

This is done by initializing the 2 qbits in 00, 01, 10, 11 and measuring the output

```
job_monitor(cal_job)
calres = cal_job.result()
meas_fitter = CompleteMeasFitter(calres, state_labels)
meas_fitter.plot_calibration()
```

Job Status: job has successfully run



```
[7]: # post-processing to find the readout error correction
     counts = [0,0,0,0]
     bits = ['00', '01', '10', '11']
     for i in range(4):
         counts[i] = calres.get_counts(cal_circuits[i])
         for bit in bits:
             if bit not in counts[i]:
                 counts[i][bit] = 0
     m_{IZ} = np.zeros(4)
     m_ZI = np.zeros(4)
     m_ZZ = np.zeros(4)
     for i in range(4):
         for num, bit in enumerate(bits):
             if num < 2:
                 m_IZ[i] += float(counts[i][bit]) / 8192
             if num > 1:
```

```
m_IZ[i] -= float(counts[i][bit]) / 8192
        if num % 2 == 0:
            m_ZI[i] += float(counts[i][bit]) / 8192
        if num % 2 == 1:
            m_ZI[i] -= float(counts[i][bit]) / 8192
        if num == 0 or num == 3:
            m ZZ[i] += float(counts[i][bit]) / 8192
        if num == 1 or num == 2:
            m ZZ[i] -= float(counts[i][bit]) / 8192
M = [[1, 1, 1, 1],
     [1, 1, -1, -1],
     [1, -1, 1, -1],
     [1, -1, -1, 1]
Minv = np.linalg.inv(M)
beta_IZ = np.dot(Minv, m_IZ)
beta_ZI = np.dot(Minv, m_ZI)
beta_ZZ = np.dot(Minv, m_ZZ)
b_vec = [beta_IZ[0], beta_ZI[0], beta_ZZ[0]]
B = [beta_IZ[1:], beta_ZI[1:], beta_ZZ[1:]]
Binv = np.linalg.inv(B)
```

1.0.3 State Tomography

In order to reconstruct the output state of the circuit, it is necessary to perform 2 qbit tomography to build an arbitrary state from the outcomes of a series of measurements that together give the full state. For 2 qbits, that is 9 (3^n) circuits.

```
[8]: # defining tomography circuits
qst_circuit = state_tomography_circuits(qc,qr)
```

```
for num, backend in enumerate(backends):
   state ={}
   for bas in basis:
       state.update({bas: {'00': 0, '01': 0, '10': 0, '11': 0}})
   if num == 0:
       job_sim = qiskit.execute(qst_circuit, Aer.
 tom = StateTomographyFitter(job_sim.result(), qst_circuit)
   if num == 1:
       job_sim = qiskit.execute(qst_circuit, Aer.

→get_backend('qasm_simulator'), shots=8192,
                                noise_model=noise_model,
                                coupling_map=coupling_map,
                                basis_gates=basis_gates)
       tom = StateTomographyFitter(job_sim.result(), qst_circuit)
   if num == 2:
       job_dev = qiskit.execute(qst_circuit, backend, shots=8192)
       job_monitor(job_dev)
       tom = StateTomographyFitter(job_dev.result(), qst_circuit)
   for pauli in paulis:
       for bit in bits:
           if bit not in tom.data[pauli]:
               tom.data[pauli][bit] = 0
   for y,bas in enumerate(basis):
       state.update({bas: {'00': state[bas]['00']+tom.data[paulis[y]][bits[0]],
                           '01': state[bas]['01']+tom.data[paulis[y]][bits[1]],
                           '10': state[bas]['10']+tom.data[paulis[y]][bits[2]],
                           '11': state[bas]['11']+tom.
 →data[paulis[y]][bits[3]]}})
   if num == 0:
       state_sim = state
       print('Simulator done.')
   if num == 1:
       state sim noise = state
       print('Noisy simulator done.')
   if num == 2:
       state_dev = state
       print('Device done.')
```

Simulator done.
Noisy simulator done.
Job Status: job has successfully run
Device done.

```
[10]: # defining the matrices to reconstruct the density matrices
      I_matrix = np.matrix('1, 0; 0, 1')
      X_matrix = np.matrix('0, 1; 1, 0')
      Y_matrix = np.matrix('0, 0-1j; 0+1j, 0')
      Z_{matrix} = np.matrix('1, 0; 0, -1')
      Iden = np.kron(I_matrix,I_matrix)
      IXPaul = np.kron(I_matrix,X_matrix)
      IYPaul = np.kron(I_matrix,Y_matrix)
      IZPaul = np.kron(I matrix, Z matrix)
      XIPaul = np.kron(X_matrix,I_matrix)
      YIPaul = np.kron(Y_matrix,I_matrix)
      ZIPaul = np.kron(Z_matrix,I_matrix)
      XXPaul = np.kron(X_matrix,X_matrix)
      XYPaul = np.kron(X_matrix,Y_matrix)
      XZPaul = np.kron(X_matrix,Z_matrix)
      YXPaul = np.kron(Y_matrix,X_matrix)
      YYPaul = np.kron(Y_matrix,Y_matrix)
      YZPaul = np.kron(Y_matrix,Z_matrix)
      ZXPaul = np.kron(Z_matrix,X_matrix)
      ZYPaul = np.kron(Z_matrix,Y_matrix)
      ZZPaul = np.kron(Z_matrix,Z_matrix)
[11]: # there are 4 density matrices to calculate for each of sim, sim_noise, dev and__
      \rightarrow readout cal
      # XX, XY, XZ etc are 9 total r_xx etc terms to calculate
      r exp = np.zeros(9)
      # there are three IX type terms and each can be calculated three ways (from XX, __
      \hookrightarrow YX, ZX for IX)
      r_{IP} = np.zeros(9)
      r_PI = np.zeros(9)
      r_IPavg = np.zeros(3)
      r_PIavg = np.zeros(3)
      m = np.zeros((9,3))
      r_{corr} = np.zeros((9,3))
      r_IPavg_corr = np.zeros(3)
      r_PIavg_corr = np.zeros(3)
      state_tot=[state_sim,state_sim_noise,state_dev]
      rho = [0,0,0,0]
```

```
for stat, states in enumerate(state_tot):
    for j in range(9):
        r = \exp[j] = (states[basis[j]]['00']*1 + states[basis[j]]['10']*(-1) + ____
 \hookrightarrowstates[basis[j]]['01']*(-1) + states[basis[j]]['11']*1)/
 →(states[basis[j]]['00'] + states[basis[j]]['10'] + states[basis[j]]['01'] + L
 ⇔states[basis[j]]['11'])
        r_{IP}[j] = (states[basis[j]]['00']*1 + states[basis[j]]['10']*(1) + ___
 \hookrightarrowstates[basis[j]]['01']*(-1) + states[basis[j]]['11']*(-1))/
 \hookrightarrow (states[basis[j]]['00'] + states[basis[j]]['10'] + states[basis[j]]['01'] +

states[basis[j]]['11'])

        rPI[j] = (states[basis[j]]['00']*1 + states[basis[j]]['10']*(-1) + LL
 \hookrightarrowstates[basis[j]]['01']*(1) + states[basis[j]]['11']*(-1))/
 →(states[basis[j]]['00'] + states[basis[j]]['10'] + states[basis[j]]['01'] + __
 ⇔states[basis[j]]['11'])
    for j in range(3):
             r_{IPavg[j]} = (r_{IP[j]} + r_{IP[j+3]} + r_{IP[j+6]})/3
             r_{PIavg[j]} = (r_{PI[3*j]} + r_{PI[3*j+1]} + r_{PI[3*j+2]})/3
    rho[stat] = (1/4)*(Iden + r_exp[0]*XXPaul + r_exp[1]*XYPaul +_{\square}
 \rightarrowr_exp[2]*XZPaul
                            + r \exp[3]*YXPaul + r \exp[4]*YYPaul + r \exp[5]*YZPaul
                            + r_{exp}[6]*ZXPaul + r_{exp}[7]*ZYPaul + r_{exp}[8]*ZZPaul
                            + r_IPavg[0]*IXPaul + r_PIavg[0]*XIPaul
                            + r_IPavg[1]*IYPaul + r_PIavg[1]*YIPaul
                            + r_IPavg[2]*IZPaul + r_PIavg[2]*ZIPaul)
for num, bas in enumerate(basis):
    m[num,:] = [r_IP[num], r_PI[num], r_exp[num]]
    r_corr[num] = np.dot(Binv, np.subtract(m[num,:],b_vec))
for j in range(3):
        r_{IPavg\_corr[j]} = (r_{corr[j]}[0] + r_{corr[j+3]}[0] + r_{corr[j+6]}[0])/3
        r_PIavg_corr[j] = (r_corr[3*j][1] + r_corr[3*j+1][1] + 
 \rightarrowr_corr[3*j+2][1])/3
rho[3] = (1/4)*(Iden + r_{corr}[0][2]*XXPaul + r_{corr}[1][2]*XYPaul +_{\sqcup}
 \rightarrowr_corr[2][2]*XZPaul
                            + r_corr[3][2]*YXPaul + r_corr[4][2]*YYPaul +
\rightarrowr_corr[5][2]*YZPaul
                            + r_corr[6][2]*ZXPaul + r_corr[7][2]*ZYPaul + L
 \rightarrowr_corr[8][2]*ZZPaul
                            + r_IPavg_corr[0]*IXPaul + r_PIavg_corr[0]*XIPaul
                            + r_IPavg_corr[1]*IYPaul + r_PIavg_corr[1]*YIPaul
                            + r_IPavg_corr[2]*IZPaul + r_PIavg_corr[2]*ZIPaul)
```

1.0.4 Calculating Fidelities

```
[12]: Fidelity = np.zeros(4)
for j in range(4):
    Fidelity[j] = state_fidelity(psi_expected, rho[j], validate=False)

print('Simulator Fidelity:', Fidelity[0])
print('Noisy Simulator Fidelity:', Fidelity[1])
print('Device Fidelity:', Fidelity[2])
print('Corrected Device Fidelity:', Fidelity[3])
```

Simulator Fidelity: 0.9999999999998

Noisy Simulator Fidelity: 0.9326171874999998

Device Fidelity: 0.9318847656249998

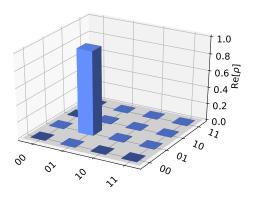
Corrected Device Fidelity: 1.188190598752761

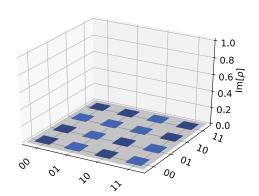
1.1 Visualizing Output States

```
[13]: plot_state_city(rho[0], title='Simulator Density Matrix')
```

[13]:

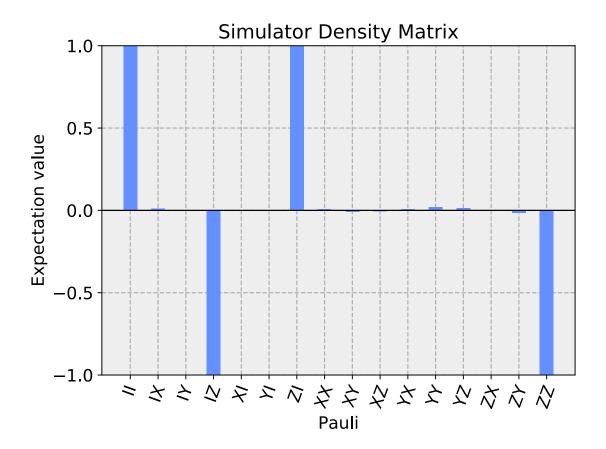
Simulator Density Matrix

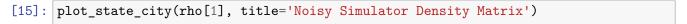




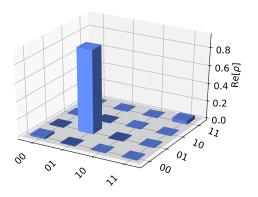
```
[14]: plot_state_paulivec(rho[0], title='Simulator Density Matrix')
```

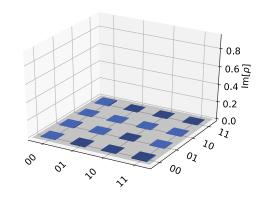
[14]:





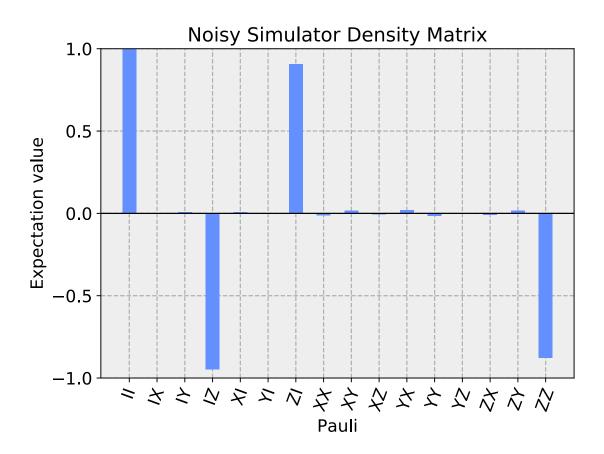
[15]: Noisy Simulator Density Matrix





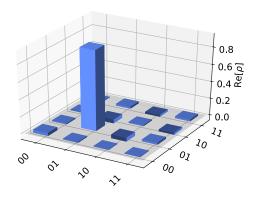
```
[16]: plot_state_paulivec(rho[1], title='Noisy Simulator Density Matrix')
```

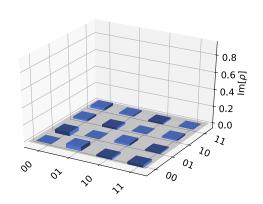
[16]:





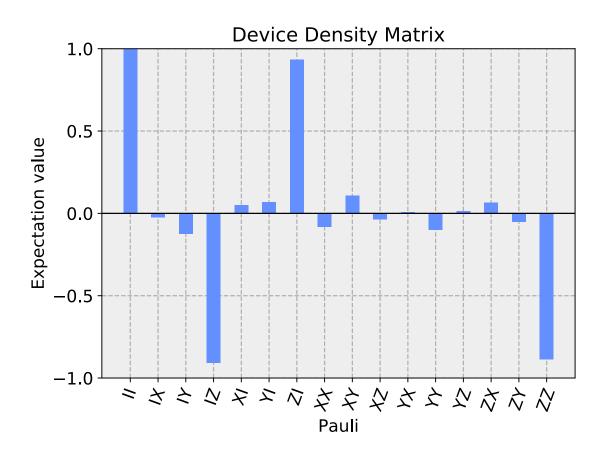
[17]: Device Density Matrix

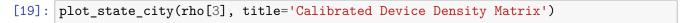




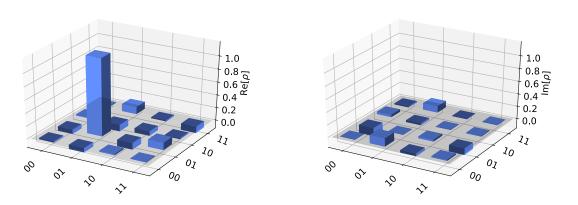
```
[18]: plot_state_paulivec(rho[2], title='Device Density Matrix')
```

[18]:





[19]: Calibrated Device Density Matrix



[20]: plot_state_paulivec(rho[3], title='Calibrated Device Density Matrix')

[20]:

