## Tic Tac Toe Machine - Using Quantum Computation Contributors: Hemanth Kumar J Anish Sharma

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A qiskit implementation of the classic game of Tic Tac Toe using quantum simulators which can be migrated to run on quantum computers.

Initialization

In [132...

qc.h(q)

In [133...

qc.h(target) # Actual diffuser function that applies the diffuser to the circuit

def diffuser(qc):

for i in range(3):

diffuser(circuit)

circuit.barrier(list(range(10)))

circuit.barrier(list(range(10)))

The circuit used for this algorithm is drawn below.

circuit.append(PhaseOracle(expression), list(range(9)))

for qubit in range(10): qc.h(qubit) qc.x(qubit)

for qubit in range(10): qc.x(qubit) qc.h(qubit)

multicontrolled\_z(qc, list(range(9)), 9)

# Expression defining all the winning states expression = '(x1 & x2 & x3) ^ (y1 & y2 & y3) ^ (z1 & z2 & z3) ^ (x1 & y1 & z1) ^ (x2 & y2 & z2) ^ (x3 & y3 & z initialize\_s(circuit, 10)

# Output circuit.measure(list(range(9)), list(range(9))) #circuit.measure([9], [8])

Out[134...

In [135... circuit.draw(output='mpl') Out[135...

Output We finally run the quantum circuit on the "aer\_simulator" provided by qiskit to simulate a quantum circuit and observe the output using

In [136...

Out[136...

plot\_histogram(counts) 0.0060

0.0045

0.0030

In [137...

Out[137... 300 250 200

150

100

50

208 000110111 352 100000100 318 011001010 158 011100110 391 110000011 494 101000011 [495 rows x 2 columns] In [139... df1 = df.copy(deep = True) df1 = df1[df1[1] > N/1000]print(df1) df1.hist()

260 110101110 458 202 111001100 456 105 101010100 454 194 111110100 450 [192 rows x 2 columns] array([[<AxesSubplot:title={'center':'1'}>]], dtype=object) Out[139... 50 40 30 20 10

0

577

563

563

462

111010011

114 011111000 165 100010001 239 001110101

208 000110111

460

Hence, the problem is solved

3

111100000

In [131... # importing Qiskit from qiskit import Aer, execute from qiskit import QuantumCircuit #importing the Oracle circuit builder from qiskit.circuit.library import PhaseOracle # import basic plot tools from qiskit.visualization import plot\_histogram

> def initialize\_s(qc, n): for q in list(range(n)):

Diffuser Grovers algorithm works by applying an Oracle reflection and a diffuser amplitude amplification around  $\sqrt{n}$  times to search for a particular

We start with initializing the qubits with all states having equal amplitudes using Hadamard Gates.

We define the diffuser function here which amplifies the negated state that the Oracle has prepared. # Multicontrolled z-gate subroutine that applies the gate to the circuit based on the control and target qubits def multicontrolled\_z(qc, control, target): qc.h(target) qc.mct(control, target)

Oracle and the completed circuit We continue to build the Oracle using the PhaseOracle method, using the winning clause expressions to determine the winning state. All these pieces are finally put together to run the Grover's algorithm 3 times(since we use 9 qubits, 1 for each space on the board) In [134... circuit = QuantumCircuit(10, 9)

<qiskit.circuit.instructionset.InstructionSet at 0x170dfc5d2b0>

4Phase Oracle  $q_0$  $\alpha$  $q_1$  $q_2$  $q_3$ 4Phase Oracle  $q_5$  $q_6$  $q_7$  $q_8$ 

N = 100000 #Since we encounter random outputs due to the nature of the algorithm, we average the outcomes of N

0.006

Probabilities 0.0015 0.0000 To further analyse the output we received, we plot the histogram as a bar graph. The results containing the higher outcomes represent the winning states while the ones having lower/zero outcomes represent the losing/tied states. import pandas as pd df = pd.DataFrame(counts.items()) df = df.sort\_values(by=1, ascending=False) df.hist() array([[<AxesSubplot:title={'center':'1'}>]], dtype=object)

simulator = Aer.get\_backend('aer\_simulator')

counts = results.get\_counts()

results = execute(circuit, backend=simulator, shots=N).result()

In [138... print(df) 0 1 111010011 577 8 011111000 165 100010001 239 001110101 1 1 1 1

500

600