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# Quantum Walk: A Different Approach to Searching Solution Spaces

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**Darwin Vargas, Yu Qing Peng**  
Rensselaer Polytechnic Institute  
Troy, NY  
vargad@rpi.edu, pengy12@rpi.edu

## Abstract

1 Search problems involve taking a computational task for which all the solutions can  
2 be mapped to a graph and traversed. The search for an optimized way to traverse the  
3 solution space/graph is an important field of study for its vast applications in many  
4 other fields. This project demonstrates the results obtained by using a Quantum  
5 Walk to traverse a solution space for the max-cut problem (QW). QW's leverage  
6 the properties of superposition, entanglement, and interference to traverse solution  
7 spaces differently than classical walks. As a result, we obtain different probability  
8 distributions and higher standard deviation growth from the origin [8]. In practice  
9 (using small three and five-node graphs), we get the expected distribution of states  
10 but continue looking for methods to favor binary sequences of high cut values.  
11 Finally, we create an informative presentation, accompanied by an implementation  
12 with slight heuristic adjustments.

## 13 1 The Problem (Background)

### 14 1.1 Unweighted Max-cut

15 For the max-cut problem we define an undirected graph  $G = (V, E)$  where we seek to determine a  
16 bipartition of vertices  $V$  into two sets  $V_1, V_2$  to maximize the number of edges between the two sets  
17 (the cut size). Figure 1 is an example maxcut [2].

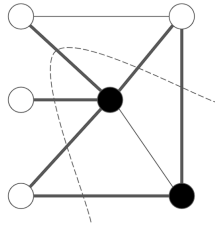


Figure 1: Partitions the black and white nodes into 2 sets [2].

18 Each solution is represented as a binary sequence of length  $N$ -nodes. The cut value is evaluated using  
19 the following equation:

$$\text{Cut}(G) = \frac{1}{2} \sum_{(i,j) \in E} (1 - \sigma_i \sigma_j)$$

## 20 1.2 Classical Random Walks:

21 Before we go into QWs, it is good to understand how a classical system would traverse a graph.  
 22 The intuition is that a particle transitions between vertices starting out in some initial position  
 23 (vertex) on a graph  $G = (V, E)$  and transitions to neighboring vertices probabilistically, e.g. a pseudo-  
 24 random number generator or a stochastic matrix for time evolution. The following equation portrays  
 25 mathematically how a particle's position would be updated at each iteration according to a stochastic  
 26 matrix [1, 4].

$$\mathbf{p}(t+1) = M\mathbf{p}(t) \quad (1)$$

27 Discrete-time classical walk: vector  $\mathbf{p}$  is updated via stochastic matrix  $M$ .

28 This approach tends to get stuck at local maxima and only evaluates one solution at a time.

## 29 2 Motivation

30 Discrete-time QW's maintain a superposition of states starting at the origin and expanding with every  
 31 unitary operator that is applied to the current state. Depending on the coin used, applying it can flip  
 32 the phase of a qubit and cause another state to collapse via destructive interference. In contrast, the  
 33 classical approach transitions between states, one by one, and cannot be efficiently used to see a  
 34 bigger picture within the solution space. Thus, it is interesting to ask if adjustments can be made to a  
 35 QW circuit such that high probabilities are attributed to max-cut solutions with higher cuts. Our team  
 36 is motivated to leverage these properties of quantum states to create such a circuit.

## 37 3 Methodology: Discrete-Time Quantum Walk

38 Given an initial position  $|n\rangle$ , the Discrete-Time QW uses a coin followed by a shift operator to  
 39 transition between nodes. A typical coin (not skewed) would be the Hadamard gate since it gives  
 40 equal probability to the  $|0\rangle$  and  $|1\rangle$  states. Here is an example shift gate that moves the  $n$ th position  $n$

$$\begin{aligned} S|0\rangle|n\rangle &= |0\rangle|n+1\rangle, \\ S|1\rangle|n\rangle &= |1\rangle|n-1\rangle. \end{aligned}$$

41 These operations are combined into the unitary operator  $U^t$  where  $t$  is the total amount of time  
 42 iterations. The probability distribution in this case is not Gaussian like the classical approach, it is a  
 more volatile superposition of all the previous states.

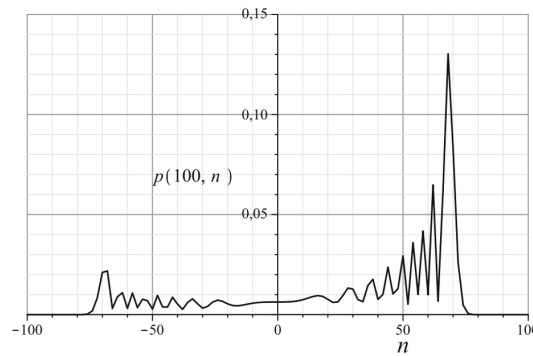


Figure 2: Probability distribution of Hadamard quantum walk [9].

## 4 Implementation

We used the 1-D QW implementation in [7] from the qiskit community. The code was out of date and required adapting to the most recent version of qiskit. We used this new version to implement the quantum walk on a simple five-node graph. This involved applying Hadamard gates followed by the shift operator. The shift operation was separated into an increment and decrement set of gates to increment  $n$  in the system. These gates generalize to  $U^t$ , which we applied between five to twelve steps adjusting heuristically to optimize the outcomes. Some other heuristic improvement involved applying gates in between the shift operations, specifically controlled Z-gates.

link: <https://colab.research.google.com/drive/1nkfcaAJ7gNTaxkeanalJKvniF7-lcKJ3?usp=sharing>

## 5 Milestone

Our target was to achieve specific bit sequences that would result in the highest cut values. The target bit sequences we looked for were {00101, 11010, 01010, 10101}. The QW initially does not favor any solution over another, making heuristic adjustments necessary. We were only able to find one of the four optimal solutions with a cut of five on the five-node graph. More experimentation is required to find gate placements and additions that favor better cut solutions.

## 6 Deliverables

The final deliverable of this project includes the Qiskit code formulation of the Discrete-time QW and heuristic adjustments. Additionally, we created an introductory presentation detailing the max-cut problem, classical limitations for searching solution spaces, and example coin and shift applications using Dirac notation. We discovered that between five and twelve iterations, the unitary operation ( $U^t$ ) produced good results that ran within a few seconds. The probability distribution we obtained aligns with the expected distribution shown in Figure 2 except for the center value. This is likely due to the heuristic adjustments made to the circuit.

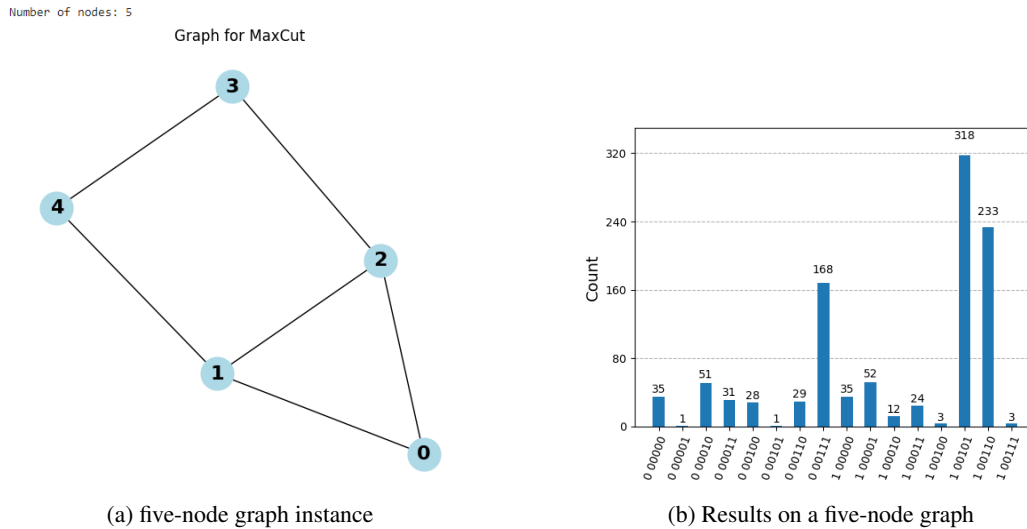


Figure 3: Comparison of results on a five-node graph.

## 7 Expected Results & Take-Home Message

More experimentation is required to create a circuit that collapses predominantly to solutions of high cut values. These results defy our initial expectations for the QW. Precisely because of the misconception that we already know something about the optimal solutions or that classical adjustments can be made so that the circuit gives preference to better cut solutions. The challenge is that the QW

72 is in many states at a time, and one cannot simply measure midway through the circuit to see how  
73 the circuit is performing because the Qubits would collapse upon measurement. The differences  
74 between the QW and the classical random walk, present a unique opportunity for experimentation  
75 and discovery, considering that much more work can be done to improve how solution spaces are  
76 searched.

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