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# **Find If There is a Path of More Than $k$ Length From a Source**

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Term Project Report

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## **Abstract**

Fully Homomorphic Encryption (FHE) scheme is a method where calculations are done on the encrypted data without needing to decrypt. Calculated values are also encrypted. FHE enables the information owner (client) to have full privacy while benefiting remote(ex. cloud) calculations. In this report is about the term project focused on adapting a common conventional network problem "Find if there is a path of more than  $k$  length from a source" and its solution into a Fully Homomorphic Encryption (FHE) network.

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# 1 Introduction

The problem with today's technology is the poor privacy. Smartphones and a lot of smart devices are online nowadays and they constantly broadcast private and personal information to their servers for the "service". Those services usually where your data needed to be synced with other devices or computation for a outcome.

Conventionally those computations needs the information to be decrypted before the computational phase. Even if the information is encrypted on information sender (client) side, client is obligated to share its key with the server to be able to decrypt computational data.

Fully Homomorphic Encryption (FHE) is developed to change this workflow with a fully-encrypted one (Figure 1 ).

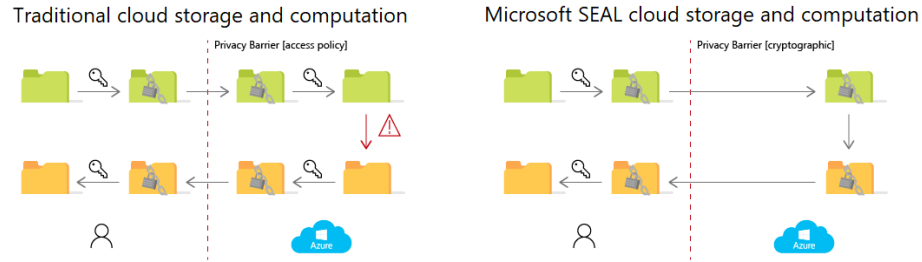


Figure 1 . SEAL NETWORK Workflow  
[1]

While improving privacy and security features, one must compromise on other features. In this case, SEAL compromised on operation features. One can only add, subtract, multiply, negate, shift right/left with the given encrypted values.

Limitations in operations comes with the real-world adaption challenges. Cloud computation developer needs to use only those basic operations to achieve the results.

In this project report, Section 2 will provide background information needed for the work done. Section 3 includes contributions done in this project to the given subject. Section 4 is where the project results and discussions will be presented. Conclusion of the project and future work will be found in Section 5.

## 1.1 Real Life Use Case

An example might be given for better understanding the reason for developers embrace such challenges to adapt the FHE to currently working systems.

For a real-case example scenario we might use the example "A medical researcher wants to compute descriptive statistics on a population of lung cancer patients at a hospital." given in an article at website inpher.io[2]

In this scenario the hospital is unable to share its private medical records with the researcher due to the HIPAA privacy rule. Normally, end of the research. However, with FHE technology, the hospital encrypts its sensitive data using a fully homomorphic encryption scheme, so that the data is protected while also able to be computed on.

What the researcher do is he encrypts the patient dataset and sends the data on the cloud computing servers without a key. The server only sees encrypted data and still it is able to conduct analytical computations for the researcher and the server sends back the results

back to the researcher, also encrypted. Only the researcher is able to decrypt and observe the results as he is the only key-holder. The Figure 2 displays a similar scenario data workflow.

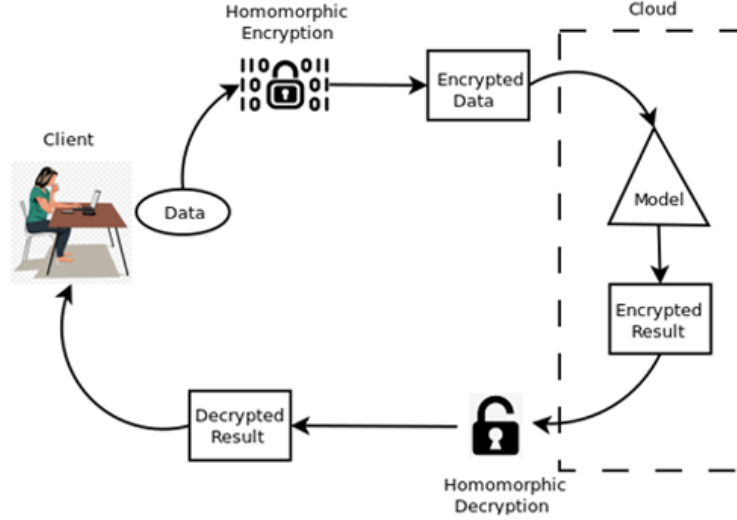


Figure 2 . FHE Encryption Flow  
[3]

## 2 Background Information

### 2.1 Homomorphic Encryption

Companies often send, receive, and store their cloud data in encrypted form. But to take advantage of cloud computing, companies must provide either unencrypted data or the keys to decrypt it. This practice puts company data at increased risk. Homomorphic encryption allows computation directly on encrypted data, making it easier to apply the potential of the cloud for privacy-critical data.

Traditional encryption schemes consist of three functionalities: key generation, encryption, and decryption. Symmetric-key encryption schemes use the same secret key for both encryption and decryption. It enables efficient encryption of large amounts of data for secure, outsourced cloud storage. Public-key encryption schemes use a public key for encryption and a separate, secret key for decryption. Anyone who knows the public key can encrypt data, but only someone who knows the secret key can decrypt and read the data. Public-key encryption enables secure online communication, but is typically less efficient than symmetric-key encryption.

You can use traditional encryption for secure storage and communication, but outsourced computation has required the removal of encryption layers. Cloud services that provide outsourced computation must implement access policies to prevent unauthorized access to the data and keys. Data privacy relies on the access control policies that are imposed by the cloud provider and trusted by the customer.

## 2.2 SEAL

With SEAL homomorphic encryption, cloud providers never have unencrypted access to the data they store and compute on. Computations can be performed directly on encrypted data. The results of such encrypted computations remain encrypted, and can be decrypted only by the data owner by using the secret key. Most homomorphic encryption uses public-key encryption schemes, although the public-key functionality may not always be needed. [1]

## 2.3 EVA

EVA is a compiler for homomorphic encryption, that automates away the parts that require cryptographic expertise. This gives you a simple way to write programs that operate on encrypted data without having access to the secret key.

Think of EVA as the "C compiler" of the homomorphic world. Homomorphic computations written in EVA IR (Encrypted Vector Arithmetic Intermediate Representation) get compiled to the "assembly" of the homomorphic encryption library API. Just like C compilers free you from tricky tasks like register allocation, EVA frees you from encryption parameter selection, rescaling insertion, relinearization. [4]

## 2.4 The Problem and The Conventional Solution to It

### 2.5 The problem

Finding if there is a path of more than  $k$  length from a source (node)

Brief explanation: Given a graph, a source node in the graph and a number  $k$ , find if there is a simple path (without any cycle) starting from given source and ending at any other node.

One important thing to note is, simply doing BFS or DFS and picking the longest edge at every step would not work. The reason is, a shorter edge can produce longer path due to higher weight edges connected through it.

For example:

### 2.6 The Conventional Solution with Python

The idea is to use Backtracking. We start from given source, explore all paths from current node. We keep track of current summation of distance from source. If distance becomes more than  $k$ , we return true in order to use a more efficient algorithm. If a path does not produces more than  $k$  distance, we backtrack and proceed another branch.

How does one can be sure that the path is simple and do not loop in a cycle? The solution is to keep track of current path vertices in an array. Whenever we add a vertex to path, we check if it already exists or not in current path. If it exists, we ignore the edge.



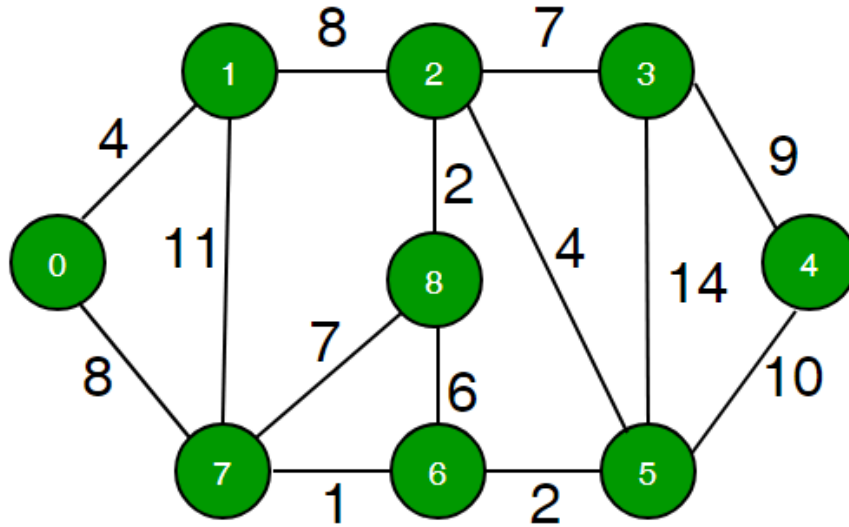


Figure 3 . Example network for the problem  
[5]

Example solution code using python can be given as shared by [5]

```

1
2 # Program to find if there is a simple path with
3 # weight more than k
4
5 # This class represents a dipathted graph using
6 # adjacency list representation
7 class Graph:
8     # Allocates memory for adjacency list
9     def __init__(self, V):
10         self.V = V
11         self.adj = [[] for i in range(V)]
12
13     # Returns true if graph has path more than k length
14     def pathMoreThanK(self,src, k):
15         # Create a path array with nothing included
16         # in path
17         path = [False]*self.V
18
19         # Add source vertex to path
20         path[src] = 1
21
22         return self.pathMoreThanKUtil(src, k, path)
23
24     # Prints shortest paths from src to all other vertices
25     def pathMoreThanKUtil(self,src, k, path):
26         # If k is 0 or negative, return true
27         if (k <= 0):
28             return True
29
30         # Get all adjacent vertices of source vertex src and
31         # recursively explore all paths from src.
32         i = 0
33         while i != len(self.adj[src]):
34             # Get adjacent vertex and weight of edge
35             v = self.adj[src][i][0]
36             w = self.adj[src][i][1]
37             i += 1
38
39             # If vertex v is already there in path, then

```

```

40         # there is a cycle (we ignore this edge)
41         if (path[v] == True):
42             continue
43
44         # If weight of is more than k, return true
45         if (w >= k):
46             return True
47
48         # Else add this vertex to path
49         path[v] = True
50
51         # If this adjacent can provide a path longer
52         # than k, return true.
53         if (self.pathMoreThanKUtil(v, k-w, path)):
54             return True
55
56         # Backtrack
57         path[v] = False
58
59         # If no adjacent could produce longer path, return
60         # false
61         return False
62
63     # Utility function to an edge (u, v) of weight w
64     def addEdge(self, u, v, w):
65         self.adj[u].append([v, w])
66         self.adj[v].append([u, w])
67
68 # Driver program to test methods of graph class
69 if __name__ == '__main__':
70
71     # create the graph given in above figure
72     V = 9
73     g = Graph(V)
74
75     # making above shown graph
76     g.addEdge(0, 1, 4)
77     g.addEdge(0, 7, 8)
78     g.addEdge(1, 2, 8)
79     g.addEdge(1, 7, 11)
80     g.addEdge(2, 3, 7)
81     g.addEdge(2, 8, 2)
82     g.addEdge(2, 5, 4)
83     g.addEdge(3, 4, 9)
84     g.addEdge(3, 5, 14)
85     g.addEdge(4, 5, 10)
86     g.addEdge(5, 6, 2)
87     g.addEdge(6, 7, 1)
88     g.addEdge(6, 8, 6)
89     g.addEdge(7, 8, 7)
90
91     src = 0
92     k = 62
93     if g.pathMoreThanK(src, k):
94         print("Yes")
95     else:
96         print("No")
97
98     k = 60
99     if g.pathMoreThanK(src, k):
100         print("Yes")
101     else:
102         print("No")

```

## 3 Main Contributions

### 3.1 Trying to Implement the Known Algorithm

First of all, the example network and algorithm as given in "tutorialspoint.dev [5]" tried to be implemented in FHE scheme for the project.

After many tries, it was concluded that an algorithm that calculates longest path on a given cyclic network graph with only simple arithmetic operations and vectors were not easy enough to be implemented in EVA in project time.

### 3.2 Implementation of FHE friendly algorithm

In order to solve the problem in a FHE network, a non-cyclic graph decided to be used in this project.

Our code generates series of nodes connected to each other with randomly generated path (edge) weight.

```
1 def generateGraph(n):  
2     ws = nx.path_graph(n)  
3     return ws
```

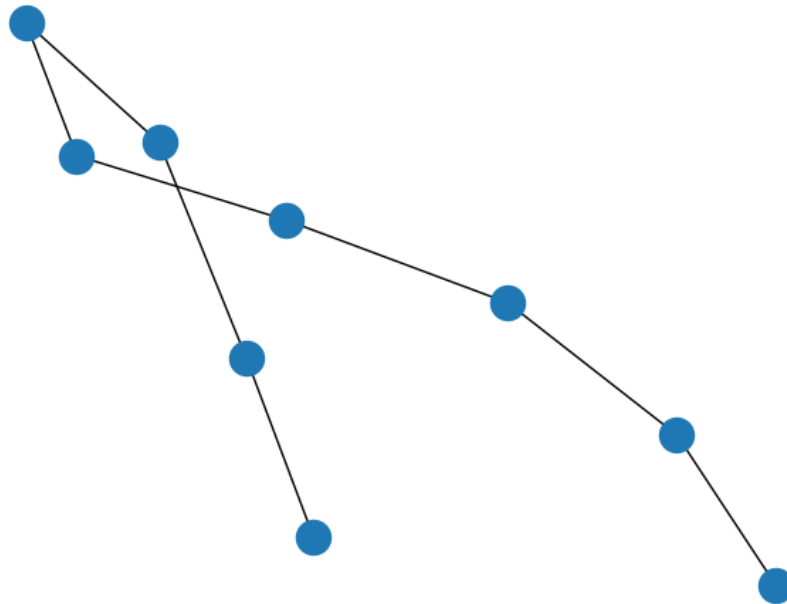


Figure 4 . Network type used for the project (n=9)

Then this graph is converted to a python dictionary with the key format of 'p\_nodeIndexFrom\_nodeIndexTo' = pathWeight in order to be able to feed into EVA program.

```

1 def serializeGraphZeroOne(GG,vec_size):
2     n = GG.size()
3     print("n:",n)
4     graphdict = {}
5     g = []
6     for nodeindex in range(n):
7
8         weight = randrange(1,10)
9
10        g.append(weight)
11        key = 'p_' + str(nodeindex) + '_' + str(nodeindex+1)
12        graphdict[key] = [weight,0] # EVA requires str:listoffloat
13    # EVA vector size has to be large, if the vector representation of the graph is
14    # smaller, fill the eva vector with zeros
15    for i in range(vec_size - n):
16        g.append(0.0)
17    print(len(g))
18    return g, graphdict

```

An example output dictionary result:

```

1 graphdict: {'p_0_1': [4, 0], 'p_1_2': [5, 0], 'p_2_3': [4, 0], 'p_3_4': [6, 0], '
    p_4_5': [4, 0], 'p_5_6': [9, 0], 'p_6_7': [2, 0], 'p_7_8': [2, 0]}

```

Notice there is a 0 appended on each path weight vector because EVA needs the input vector size to be power of 2. So a dummy 0 is added.

Then the built input dictionary fed to EVAProgramDriver:

```

1 grap_find_max_path_length = EvaProgramDriver("grap_find_max_path_length", vec_size=m,
    n=n)
2     with grap_find_max_path_length:
3         input = {'zero': Input('zero', is_encrypted=True)}
4         for i in range(n):
5             key = f"p_{i}_{i+1}"
6             input[key] = Input(key)
7         path_sum_left,path_sum_right = pathsumprogram(input, n ,start_node_index)
8         Output('path_sum_left', path_sum_left)
9         Output('path_sum_right', path_sum_right)

```

Using the given keys of EVA inputs, a function makes the summation to both direction of starting node index.

```

1 def pathsumprogram(path,nodecount,start_node_index):
2     matrix = np.zeros(shape=(1,2), dtype=Expr)
3     sum_left = 0
4     sum_right = 0
5     print(range(start_node_index,nodecount,1))
6     print(range(0,start_node_index,1))
7
8     for path_index in range(start_node_index,nodecount-1,1):
9         sum_right += path[f"p_{path_index}_{path_index+1}"]
10    for path_index in range(0,start_node_index-1,1):
11        sum_left += path[f"p_{path_index}_{path_index+1}"]
12    return sum_left,sum_right

```

To visualize an example, let's say the starting node is 2. In Figure 5 it can be seen the function's path while doing summation.

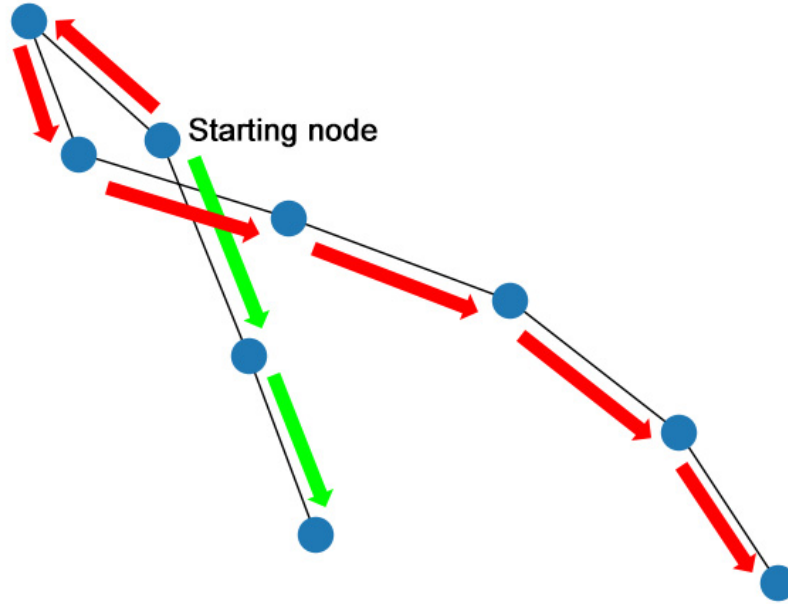


Figure 5 . Summation flow

After the results are decrypted by the client, client-side can see if either path length is longer than the given length  $k$  or not.

## 4 Results and Discussion

### 4.1 Methodology

In this project a rather simple implementation was chosen because of the complexity of the other network types.

Therefore the methodology is also straightforward.

**Client side:** There is a generated network with number  $n$  nodes and each has maximum of 1 paths to others. This network is generated by networkx library on python.

This network was vectorized and put in a python dictionary format with the connecting node's indexes and the weight of the connecting path.

Afterwards this vector was fed to EVA.

EVA encrypted the given vector and used the encrypted version to calculate both left hand side and right hand side total path length.

After the outputs generation, there is an observable calculation error caused by doing the encrypted operations and conversions. In the result, those errors will be addressed.

Table 1. Average Time and Error

compile time (ms)	1.41
key generation time	20.93
encryption time	48.01
execution time	10.91
decryption time	6.52
reference execution time	0.14
mean square error	9.43e-19

## 4.2 Results

After 1000 tries, method seems to be consistently able to calculate the correct results every time with similar amounts of error.

For a randomly picked results of one of the runs, Paths on the left hand side was calculated as 4.00000000144591 in the FHE method, without FHE the result was 4.0, this means  $\%3.6e - 8$  error. Paths on the right hand side was calculated as 26.999999998706993 instead of 27.0 which yields  $\%4.8e - 11$  error.

Averages for 9 nodes input can be given as:

## 4.3 Discussion

Although this implementation works as intended on non-cycle networks, FHE network using EVA was not capable of solving networks with cycle paths in ease. It can be a result of lack of knowledge or limited time for this project. However, it was fairly easy to implement as done on the project.

Another issue needs to be addressed is that execution times are high enough to hinder calculations for networks with high number of nodes. This was expected as it was a known downside of FHE networks from the start. However, promisingly FHE improves by both performance and capability every day and getting ready to be used in wider scenarios.

Finally, calculation errors caused by the conversions and encryption/decryption were so low for this project that they could be neglected as rounding them to the nearest integer.

## 5 Conclusion

In this term project, an implementation of FHE into a known network problem "Finding if there is a path of more than k length from a source" was done. FHE approach was applied successfully to solve the problem for non-cycle networks using Microsoft SEAL and EVA compiler via python. Client side generated network was converted to vector and then python dictionary format and fed to the EVA Program. Values were encrypted on client side and sent to server for calculations to be done as encrypted. Left and right hand side available path length summations were done and results were sent back to client still-encrypted. Client was able to decrypt and get the desired results. Further work is needed to implement on networks with cycles.

## References

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