



SC2001  
PROJECT 2

BY ATHENA,  
ARUN, AFREEN

Start

# CONTENT



1. Abstract
2. Dijkstra's pseudocode
3. Dijkstra's source code
4. Graph generation pseudocode
5. Dense vs sparse graph
6. Part A
7. Part B
8. Comparison

# ABSTRACT

This project aims to explore the impact of different graph representations and priority queue implementations on the time complexity of Dijkstra's algorithm

(a) Input graph  $G$  stored in an adjacency matrix and we use an array for the priority queue

(b) Input graph  $G$  is stored in an array of adjacency lists and we use a minimizing heap for the priority queue

(c) Comparing the 2 implementation in (a) and (b) and conclude which implementation is better in what circumstances

# DIJKSTRA'S PSEUDOCODE

Dijkstra\_ShortestPath(Graph G, Node source):

for each vertex v in G:

$d[v] = \text{infinity}$   $\longleftarrow$  Initialize distances to infinity

$p[v] = \text{null}$   $\longleftarrow$  Predecessor of each vertex is initially null

$S[v] = 0$   $\longleftarrow$   $S[v]$  indicates whether v is in the shortest path set (S)

$d[\text{source}] = 0$   $\longleftarrow$  Distance to the source is set to 0

$Q = \text{PriorityQueue}()$   $\longleftarrow$  Initialize priority queue with all vertices,  
where the priority is their distance  $d[v]$

for each vertex v:

$Q.\text{insert}(v, d[v])$

# DIJKSTRA'S PSEUDOCODE

while Q is not empty:

u = ExtractCheapest(Q) ← Get the vertex u with the minimum distance

S[u] = 1 ← Mark vertex u as part of the shortest path set

for each vertex v adjacent to u: ← Update distances for all adjacent vertices of u

if S[v] == 0 and d[v] > d[u] + w(u, v): ← Only update if vertex v is not in S

d[v] = d[u] + w(u, v) ← Update distance

p[v] = u ← Update predecessor

Q.update(v, d[v]) ← Update the priority of v in the priority queue

//End of while loop

# DIJKSTRA'S SOURCE CODE

```
1 // Dijkstra's algorithm for a graph represented using an adjacency matrix
2 void Dijkstra_ShortestPath(int graph[V][V], int source) {
3     int d[V];    // d[v] will hold the shortest distance from source to vertex v
4     int pi[V];   // pi[v] will hold the predecessor of vertex v in the shortest path
5     int S[V];    // S[v] is 1 if vertex v is in the shortest path set, otherwise 0
6
7     // Initialize all distances as INFINITE and predecessors as null (-1), S[] as false
8     for (int i = 0; i < V; i++) {
9         d[i] = INT_MAX;
10        pi[i] = -1; // -1 denotes no predecessor
11        S[i] = 0;  // Vertex v is not yet processed
12    }
13
14    // Distance from the source to itself is always 0
15    d[source] = 0;
16
17    // Put all vertices in the priority queue Q, While its not empty
18    for (int count = 0; count < V - 1; count++) {
19        // Extract the vertex u with the minimum distance from Q
20        int u = ExtractCheapest(d, S);
21
22        // If the minimum distance vertex is -1, then all reachable vertices have been processed
23        if (u == -1) break;
24
25        // Mark vertex u as processed (add it to set S)
26        S[u] = 1;
27
28        // For each vertex v adjacent to u
29        for (int v = 0; v < V; v++) {
30            // Check if there is an edge from u to v, v is not processed, and
31            // if the distance to v can be minimized through u
32            if (!S[v] && graph[u][v] && d[u] != INT_MAX
33                && d[u] + graph[u][v] < d[v]) {
34                d[v] = d[u] + graph[u][v]; // Update d[v]
35                pi[v] = u;                 // Update predecessor of v
36            }
37        }
38    }
39
40    // Print the shortest distances and predecessors
41    printSolution(d, pi);
42 }
43
```

```
// Function to find the vertex with the minimum distance value
int ExtractCheapest(int d[], int S[]) {
    int min = INT_MAX, min_index = -1;

    for (int v = 0; v < V; v++) {
        if (!S[v] && d[v] <= min) {
            min = d[v];
            min_index = v;
        }
    }

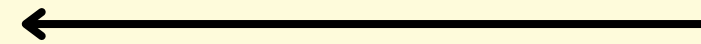
    return min_index;
}
```

# GENERATING GRAPHS - ALGO IMPLEMENTATION

FUNCTION generate\_random\_graph(V, E):

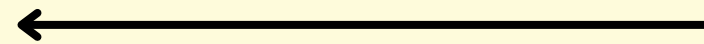
CREATE graph as a VxV matrix filled with zeros

Initialize a V x V matrix with zeros



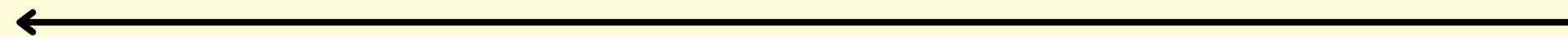
edges\_added = 0

Initialize edges counter to 0



WHILE edges\_added < E:

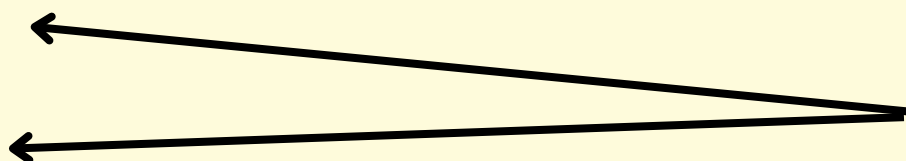
A while loop runs until the desired number of edges has been added. Each iteration attempts to add a new edge to the graph.



u = RANDOM integer between 0 and V-1

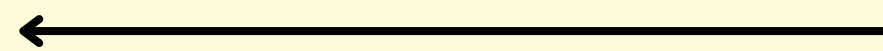
v = RANDOM integer between 0 and V-1

Two vertices u and v are randomly selected



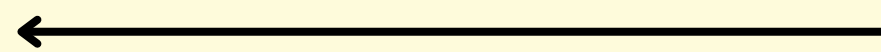
# GENERATING GRAPHS - ALGO IMPLEMENTATION

IF u is not equal to v AND graph[u][v] == 0:



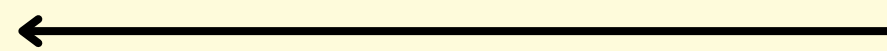
1ST test to ensure that the vertices u  
and v to be different i.e no self loop  
2ND test to ensure there isn't a already  
existing loop between vertices u and v

weight = RANDOM integer between 1 and 9



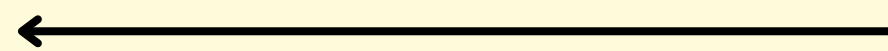
Randomly generate a weight to  
the edge (between 1 and 9)

graph[u][v] = weight



Add the weight

edges\_added += 1



Edge counter incremented

# Return the generated graph  
RETURN graph



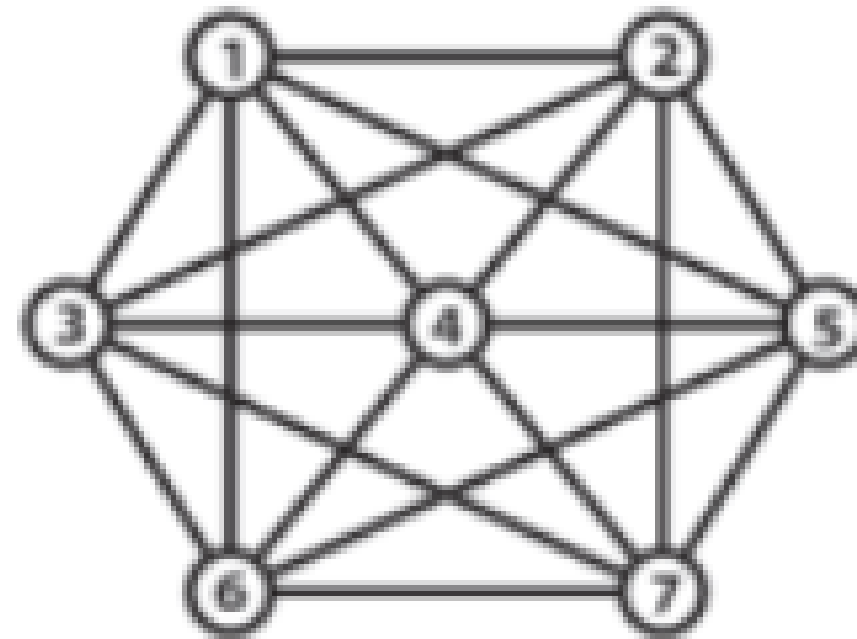
# DENSE VS SPARSE

$$E = \frac{v(v-1)}{2}$$

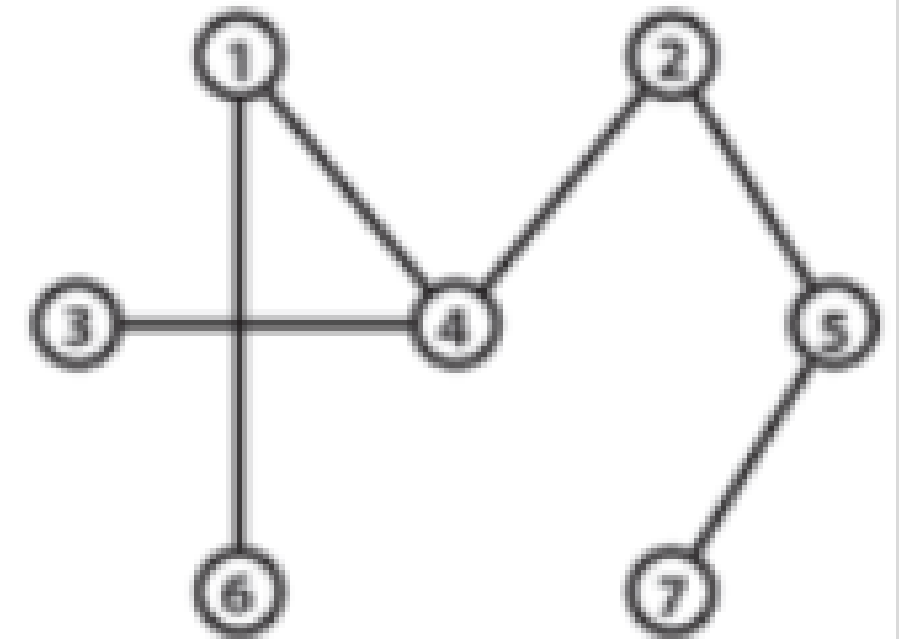
- For undirected graphs

$$E = v(v-1)$$

- For directed graphs



Dense



Sparse

Denser graphs has high connectivity compared to sparser graphs making denser graphs have density  $E/v^2$  closer to one

# ADJACENCY MATRIX AND ARRAY

## 1. Adjacency Matrix Representation:

A 2D array where the weight of the edge between any two vertices  $u$  and  $v$  is stored at `matrix[u][v]`.

## 2. Array-Based Priority Queue:

A simple array to store the distances. The vertex with the smallest distance is selected by scanning the entire array which takes  $O(|V|)$  time.

# THEORETICAL ANALYSIS WHEN $V$ IS FIXED AND $E$ IS VARIED

## 1. Main Loop:

- Iterates over all vertices:  **$O(|V|)$**
- Finding the minimum distance vertex (using an array-based queue):  **$O(|V|)$  per iteration.**
- Total for all vertices:  **$O(|V|^2)$**

## 2. Updating Neighbors:

- Each vertex checks  $O(|V|)$  neighbors in the adjacency matrix.
- Total:  **$O(|V|^2)$**

## 3. Overall Time Complexity:

- $O(|V|^2)$ : Constant regardless of the number of edges ( $|E|$ ) as long as the number of vertices ( $|V|$ ) is fixed.

## 4. Graph Density:

- This time complexity indicates that whether the graph is dense (many edges) or sparse (few edges), the runtime remains the same with a fixed  $|V|$  since it's dominated by vertex count, not edge count.

# THEORETICAL ANALYSIS WHEN $E$ IS FIXED AND $V$ IS VARIED

## 1. Total Time Complexity:

- Overall:  $O(|V|^2)$ .

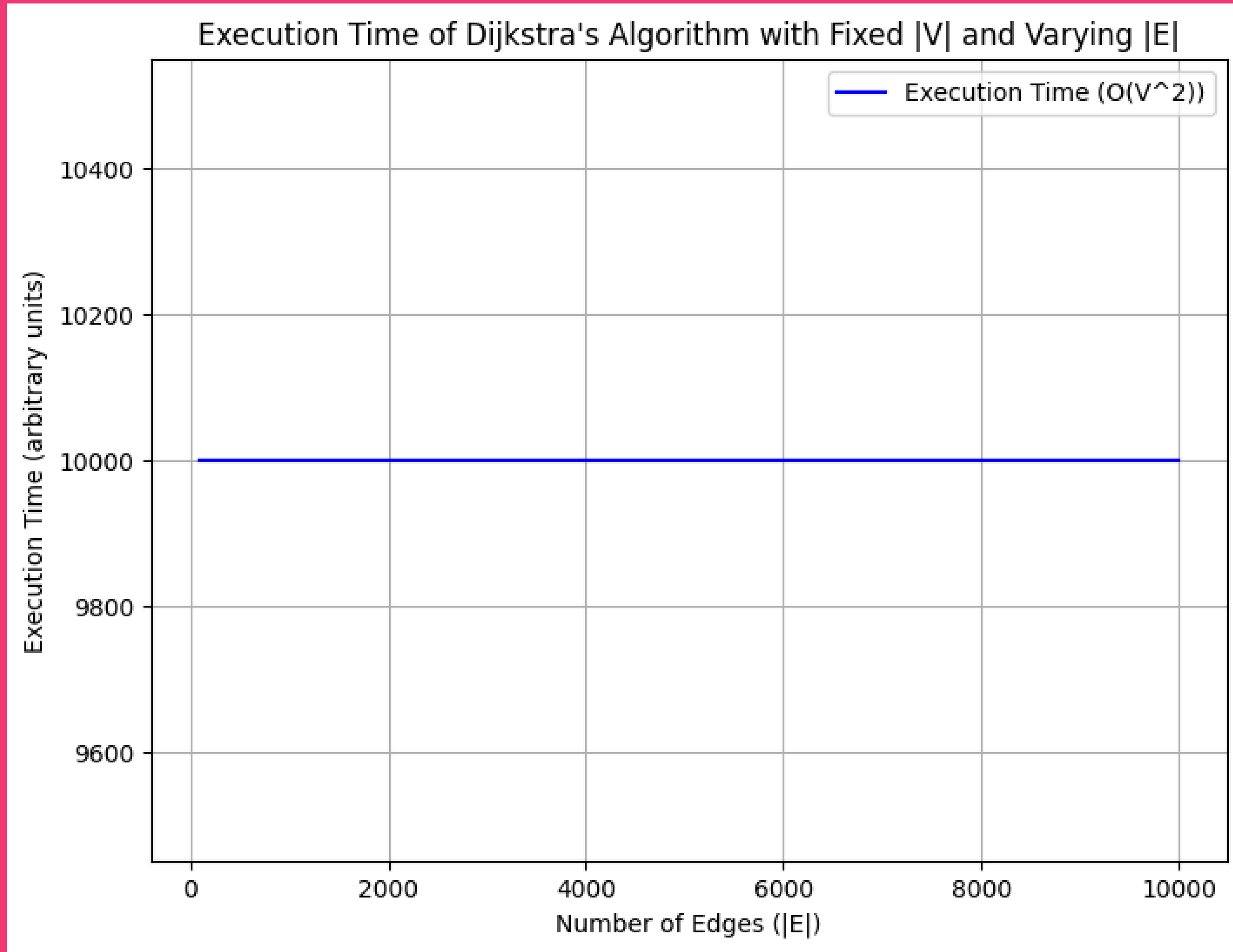
## 2. Impact of Fixed $|E|$ :

- With a fixed  $|E|$ , the complexity remains  $O(|V|^2)$ , indicating it mainly depends on the number of vertices  $|V|$ .
- The time complexity scales quadratically with  $|V|$ , not  $|E|$ .

## 3. Graph Density (Sparse vs. Dense):

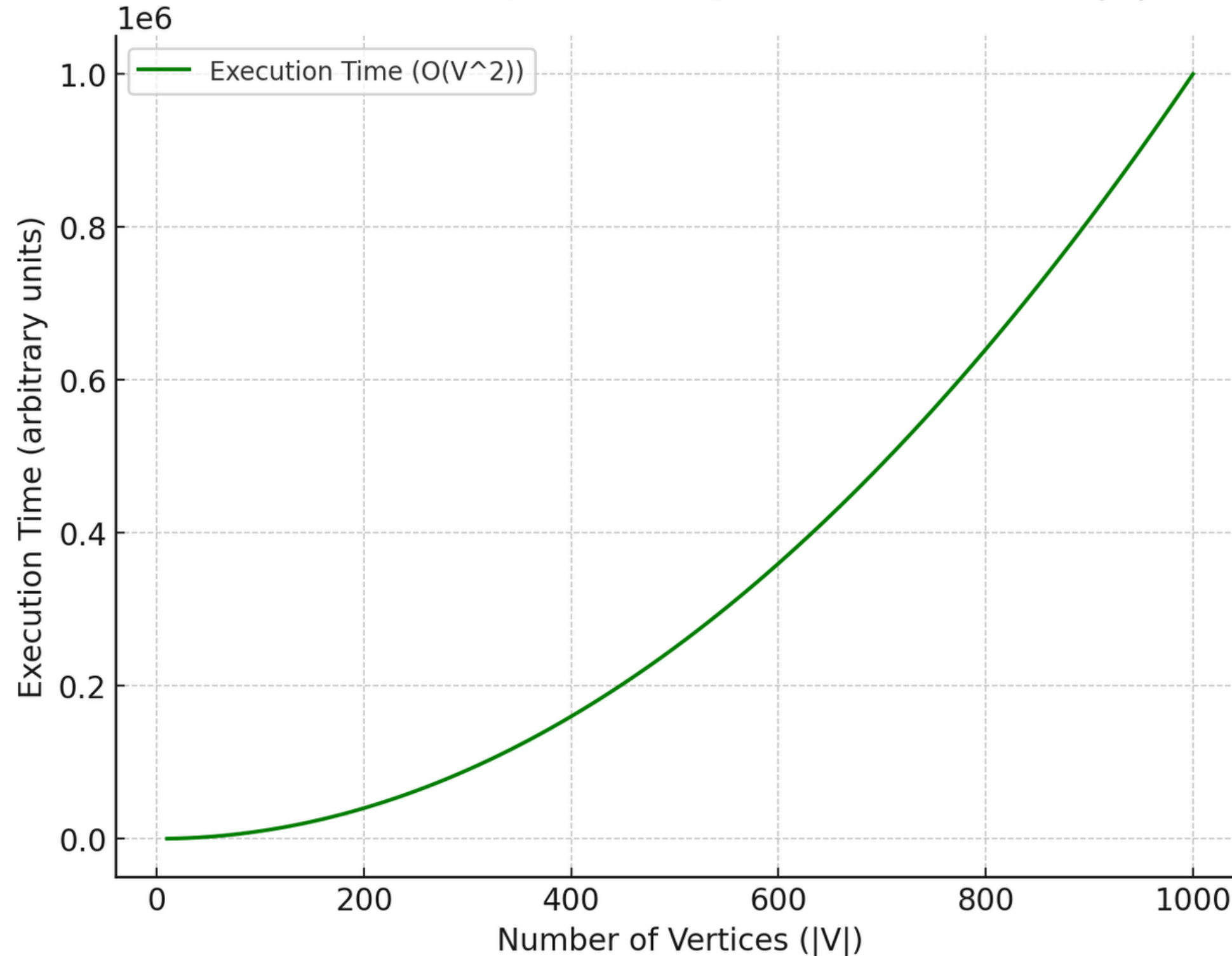
- Sparse Graph: Few edges relative to vertices—runtime remains  $O(|V|^2)$  since it's dominated by vertex count, not edge count.
- Dense Graph: Many edges relative to vertices—runtime still  $O(|V|^2)$  as the adjacency matrix approach focuses on vertices, making it unaffected by edge density.

# EMPIRICAL ANALYSIS WHEN $V$ IS FIXED AND $E$ IS VARIED



# EMPIRICAL ANALYSIS WHEN E IS FIXED AND V IS VARIED

Empirical Execution Time of Dijkstra's Algorithm with Fixed  $|E|$  and Varying  $|V|$



# ARRAY OF ADJACENCY LISTS AND MINIMISING HEAP

## Array of Adjacency Lists

- An array of adjacency lists represents a graph by using an array where each index corresponds to a vertex, and each element contains a list of its neighboring vertices and the edge weights connecting them.
- It allows fast traversal of the neighbors of each vertex, making it ideal for algorithms like Dijkstra's.

# ARRAY OF ADJACENCY LISTS AND MINIMISING HEAP

## Minimising Heaps

- A special tree-based data structure that is used to implement a priority queue.
- Allows for efficient extraction of the smallest element, which is essential in algorithms like Dijkstra's where you repeatedly select the vertex with the smallest tentative distance.



# THEORETICAL ANALYSIS

Graph Representation (Adjacency List): Each edge  $(u,v)$  is listed once, resulting in a space complexity of  $O(V+E)$ .

Min Heap:

- Push —  $O(\log V)$
- Pop —  $O(\log V)$

Overall Time Complexity:

- Initialization:  $O(V)$
- Vertex Extraction (Pop): Each vertex is extracted once;  $O(V \log V)$
- Relaxation: Each edge is relaxed at most once, and for each relaxation, insert into the heap;  $O(E \log V)$ .

$$O((V+E) \log V)$$

# THEORETICAL ANALYSIS

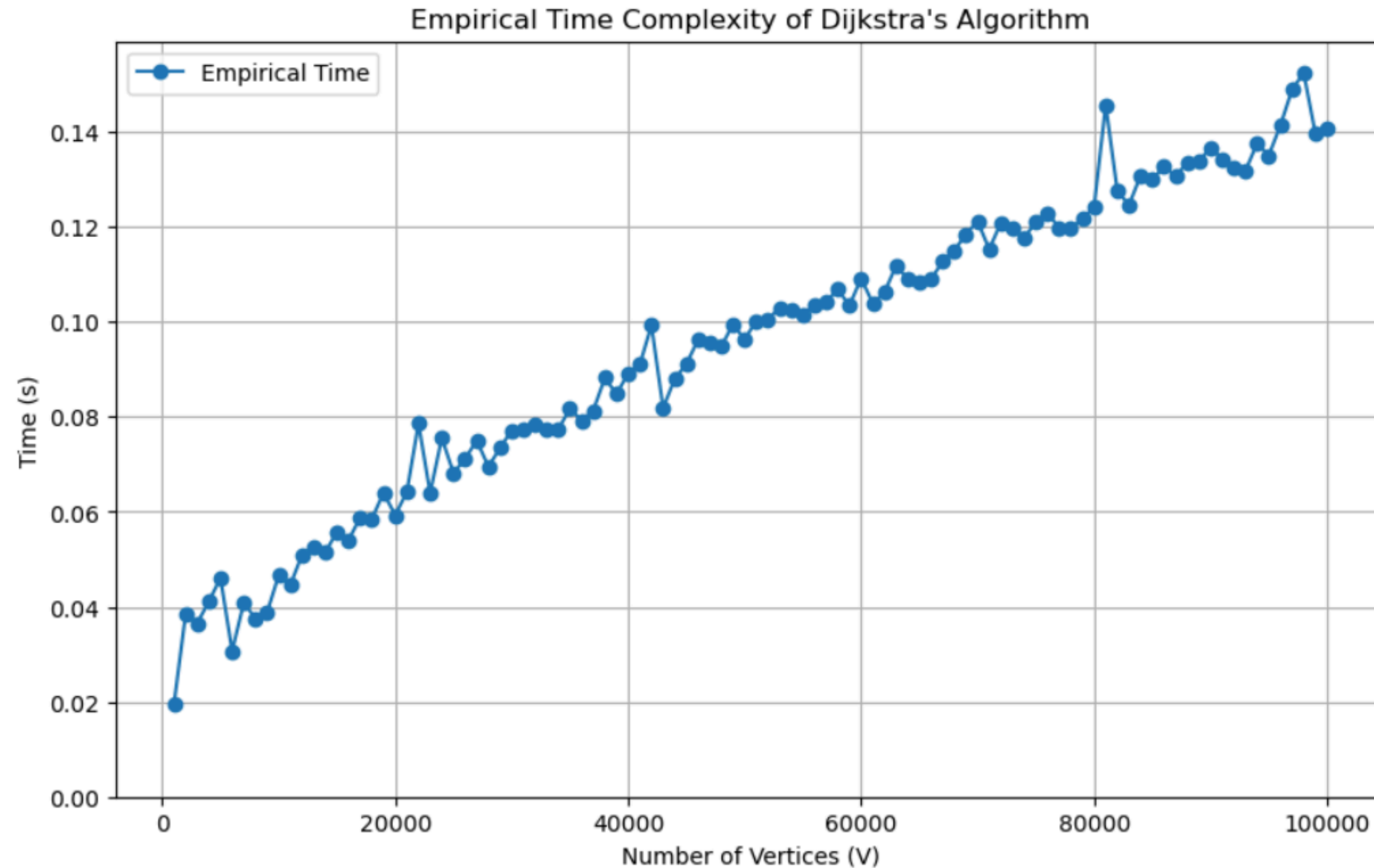
## Dense

- As the graph gets denser, the number of edges  $E$  approaches  $V^2$  (the maximum possible number of edges in a complete graph). In this case, the time complexity approaches  $O(V^2 \log V)$

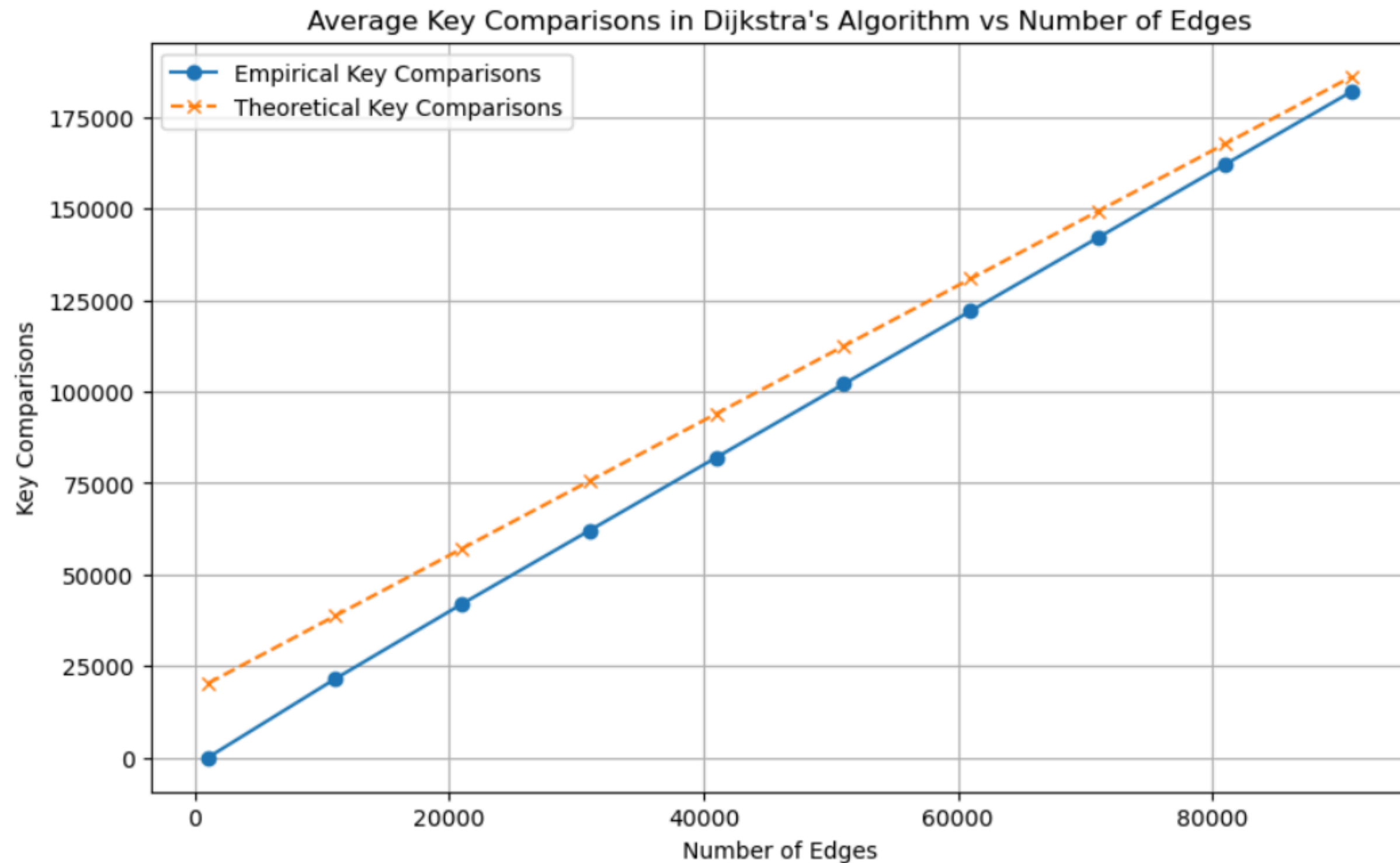
## Sparse

- When the graph is sparse,  $E$  is much smaller than  $V^2$ . The complexity remains close to  $O(V \log V)$ .

# EMPIRICAL ANALYSIS WHEN $E$ IS FIXED AND $V$ IS CHANGED

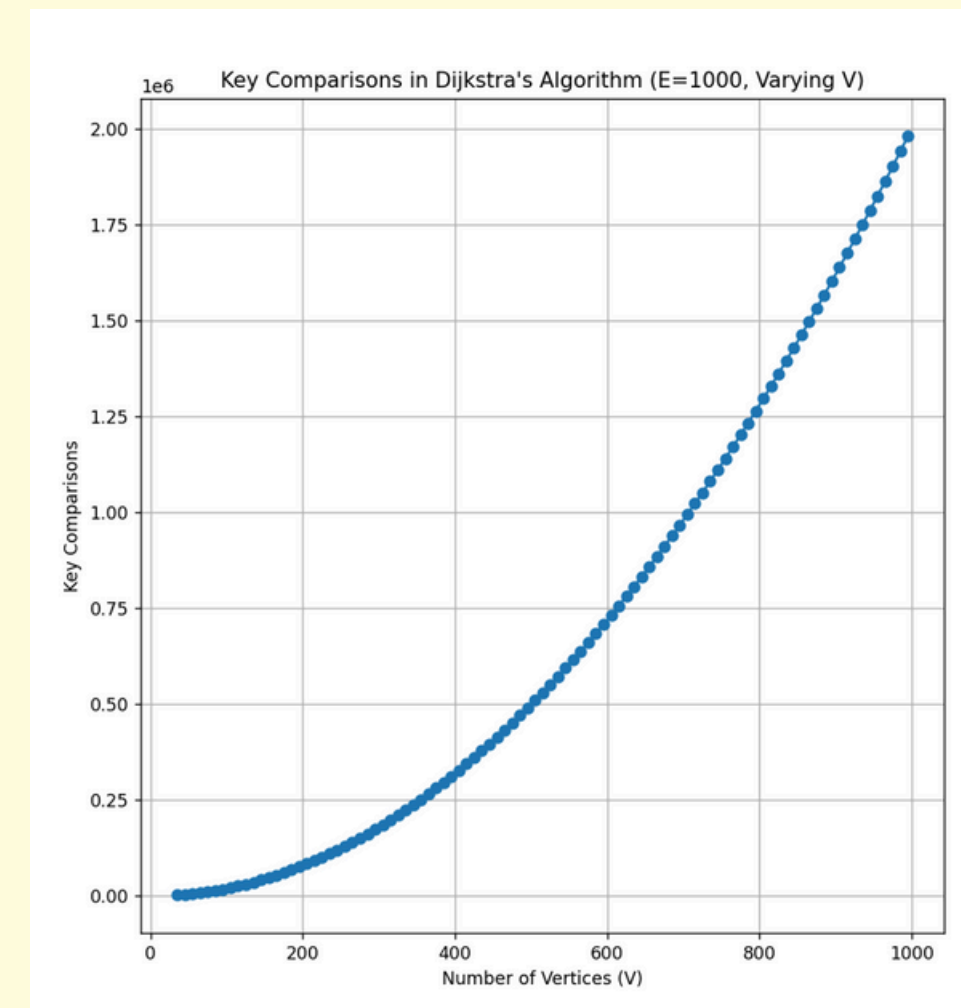
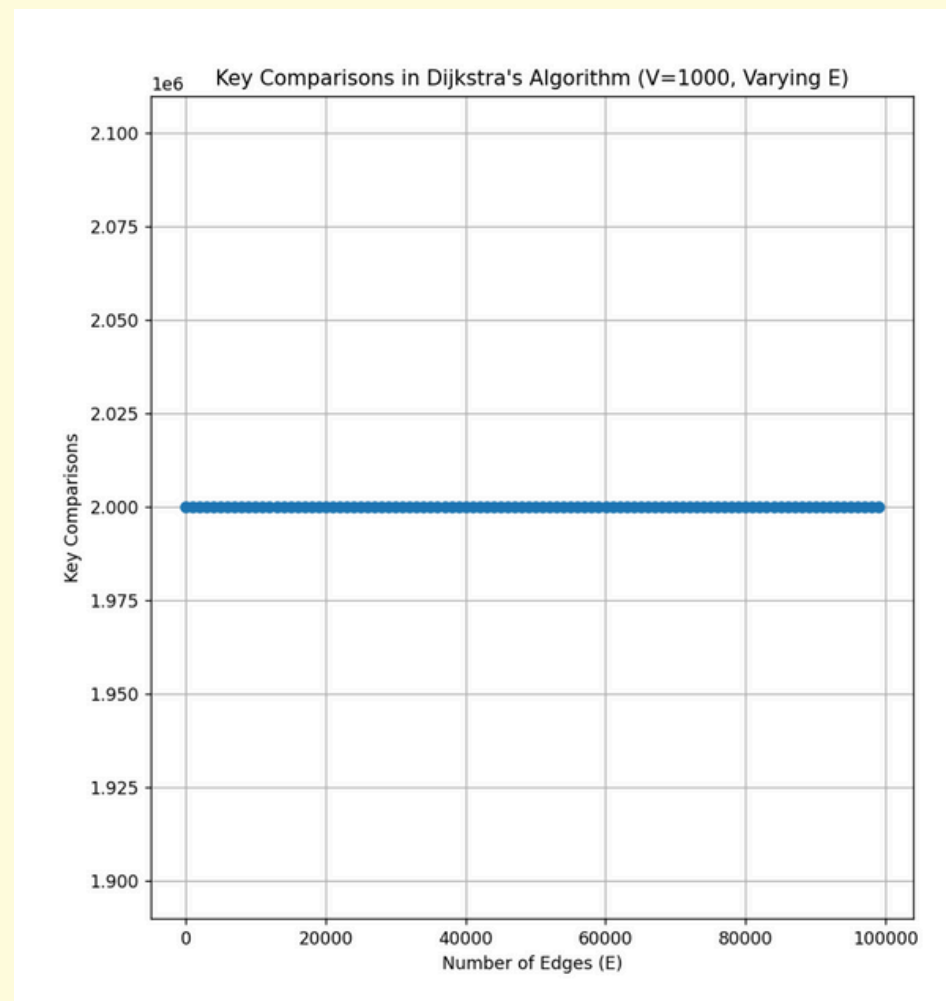


# EMPIRICAL ANALYSIS WHEN $V$ IS FIXED AND $E$ IS CHANGED



## Implementation(a):

- Just like the execution time, the key comparisons also showed very similar trends
- This implementation is mostly dependent on V rather than on E.



### **Implementation(b):**

- This implementation is mostly dependent on  $E$  and a varying  $V$  doesn't affect its performance much.
- Even as  $V$  increases against a fixed  $E$ , the execution time only increases fairly making the overall performance still remain stable.

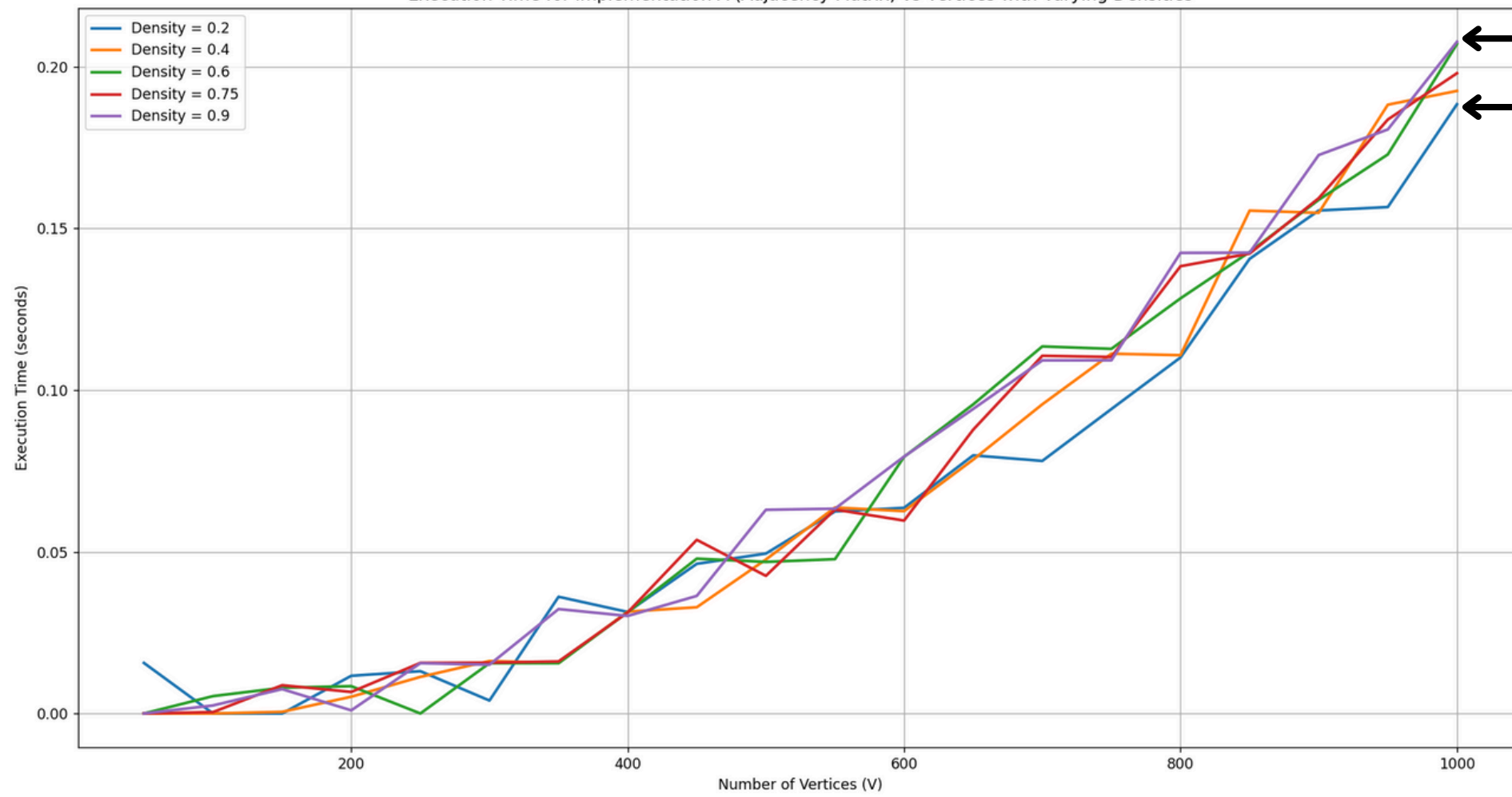
- Implementation A is more suitable for denser graphs( $E$  is large) where the relationship between vertices and edges is high

$$E = \frac{v(v-1)}{2} \quad \text{- For undirected graphs}$$

$$E = v(v-1) \quad \text{- For directed graphs}$$

- Implementation B is more suitable for sparse graphs where the number of edges are limited relative to the number of vertices.

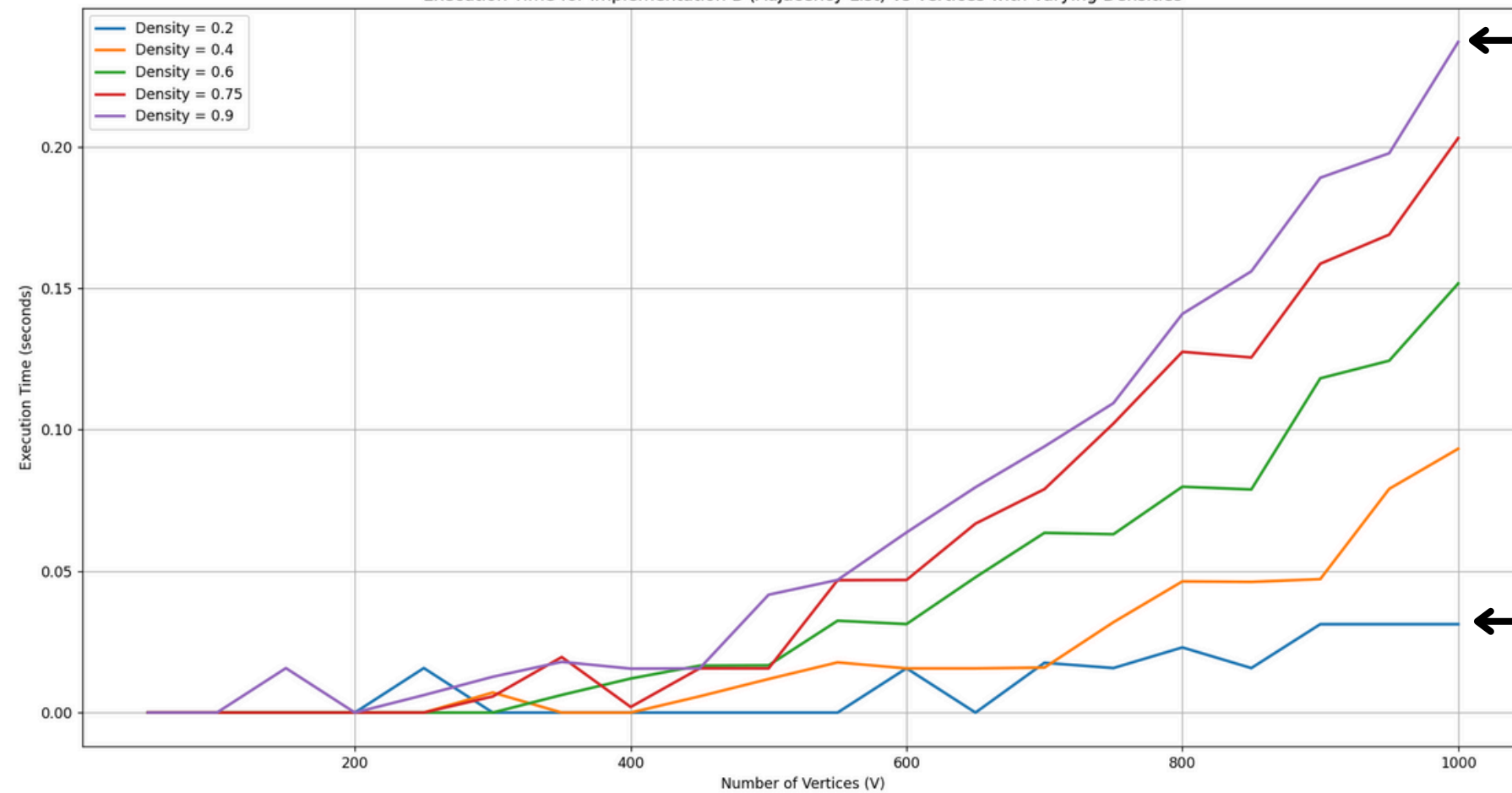
Execution Time for Implementation A (Adjacency Matrix) vs Vertices with Varying Densities



0.9 density : ~0.21

0.2 density : ~0.18

Execution Time for Implementation B (Adjacency List) vs Vertices with Varying Densities



0.9 density : ~0.24

0.2 density : ~0.03



**THANK YAAAAA**