

Evaluation of algorithms for finding shortest paths in a network

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ABSTRACT

The paper evaluates three different algorithms for computing all-pairs shortest paths. We compare the well-known Floyd-Warshall algorithm with two simple modifications of it described in paper Modifications of the Floyd-Warshall algorithm by Andrej Brodnik, Marko Grgurovič and Rok Požar. The key difference lies in the fact that the relaxations are done in a smarter way. We evaluate the algorithms on three different graph models - uniform Erdős-Rényi, binomial Erdős-Rényi, and Albert-Barabási. Based on the results, we can observe that both modified algorithms outperform the Floyd-Warshall algorithm.

KEYWORDS

Three algorithm, Hourglass algorithm, Floyd-Warshall algorithm, network, all-pairs shortest path.

1 INTRODUCTION

Graph theory has long grappled with the timeless challenge of finding the shortest paths within graphs, making it a classic problem in algorithmic studies. The key idea revolves around navigating a (directed) graph, where each arc carries a specific weight, in search of paths that minimize the sum of these arc weights. This fundamental problem finds applications in various real-world scenarios, including bioinformatics, logistics, telecommunications and so on [1].

Two prominent variations of this problem are the single-source shortest path and the all-pairs shortest path (APSP) problems. The single-source variant focuses on discovering paths from a fixed starting vertex to all other vertices in the graph, while the APSP entails finding the shortest path between every possible pair of vertices [3].

In this work, focus will be solely on the APSP variant, aiming to evaluate three different algorithms which offer effective solution. Generally, the APSP problem can be tackled using the technique of relaxation. The core concept of relaxation involves evaluating whether we can enhance the weight of the shortest path from vertex u to v by routing it through vertex w , updating it whenever necessary. Among the well-known relaxation-based solutions, one of the most straightforward approaches is the dynamic programming Floyd-Warshall algorithm which is among algorithms being evaluated. This algorithm boasts a time complexity of $O(n^3)$ when dealing with graphs containing n vertices. While its ease of implementation is commendable, two alternative algorithms will be evaluated which to potentially improve efficiency and scalability in solving the APSP problem [3].

In this article, two modifications of the Floyd-Warshall algorithm are presented and evaluated, which are named the Tree algorithm and the Hourglass algorithm. A fundamental difference of both modifications in relation to the Floyd-Warshall algorithm is a smarter way to perform the relaxations. This is done by introducing a tree structure that allows to skip relaxations that do not contribute to the result. The worst-case time complexity of both algorithms remains $O(n^3)$, however, their expected running time is substantially better for a class of complete graphs with weights drawn randomly from a uniform distribution on $[0, 1]$ [2].

Algorithms are evaluated on graphs generated by using Erdős-Rényi and Albert-Barabási method. Specifically graphs are directed with uniformly distributed arc weights on $[0,1]$.

2 PRELIMINARIES

A digraph (or directed graph) G is a pair (V, A) , where V is a non-empty finite set of vertices and $A \subseteq V \times V$ a set of arcs. We assume $V = \{v_1, v_2, \dots, v_n\}$ for some n [2].

A path P in a digraph G from $v_{P,0}$ to $v_{P,m}$ is a finite sequence $P = v_{P,0}, v_{P,1}, \dots, v_{P,m}$ of pairwise distinct vertices such that $(v_{P,i}, v_{P,i+1})$ is an arc of G , for $i = 0, 1, \dots, m-1$. The length of a path P , denoted by $|P|$, is the number of vertices occurring on P . A k -path is a path in which all intermediate vertices belong to the subset $\{v_1, v_2, \dots, v_k\}$ of vertices for some k [2].

A weighted digraph is a digraph $G = (V, A)$ with a weight function $w : A \rightarrow \mathbb{R}$ that assigns each arc $a \in A$ a weight $w(a)$. A shortest path from u to v , is a path in G whose weight is infimum among all paths from u to v . The distance between two vertices u and v , is the weight of a shortest path from u to v in G [2].

3 ALGORITHMS

3.1 The Floyd-Warshall algorithm

The Floyd-Warshall algorithm is a well-known algorithm which uses simple dynamic programming approach to solve APSP on a graph $G(V, A)$ represented by a weight matrix W , where $W_{ij} = w(v_i, v_j)$ if $(v_i, v_j) \in A$ and ∞ otherwise. Its running time is $O(|V|^3)$ due to three nested for loops and does not depend on number of arcs. The pseudocode of Floyd-Warshall algorithm is given in Algorithm 1 [2, 3].

3.2 The Tree algorithm

The Tree algorithm is the first version of the modified Floyd-Warshall algorithm. Consider the k -th iteration, and let OUT_k represent a shortest path tree originating from vertex v_k . It is based on the observation that the relaxation in lines 4-5 would not always succeed

```

1 for k := 1 to n do
2   for i := 1 to n do
3     for j = 1 to n do
4       if  $W_{ik} + W_{kj} < W_{ij}$  then
5          $W_{ij} := W_{ik} + W_{kj}$ ;
6       end
7     end
8   end
9 end

```

Algorithm 1: Floyd-Warshall(W)

in lowering the value of W_{ij} . Instead of simply looping through every vertex of V in line 3, we perform the depth-first traversal of OUT_k . This permits us to skip iterations which provably cannot lower the current value of W_{ij} [2].

The pseudocode of the modified Floyd-Warshall algorithm augmented with the tree OUT_k , named the Tree algorithm, is given in Algorithm 2. First the tree OUT_k is constructed using ConstructOUT given in Algorithm 3 and then depth-first search is performed [2].

```

1 Initialize  $\pi$ , an  $n \times n$  matrix, as  $\pi_{ij} := i$ ;
2 for k := 1 to n do
3    $OUT_k := ConstructOUT_k(\pi)$ ;
4   for i := 1 to n do
5     Stack := empty;
6     Stack.push( $v_k$ );
7     while Stack  $\neq$  empty do
8        $V_k := Stack.pop()$ ;
9       forall children  $v_j$  of  $v_x$  in  $OUT_k$  do
10        if  $W_{ik} + W_{kj} < W_{ij}$  then
11           $W_{ij} := W_{kj} + W_{kj}$ ;
12           $\pi_{ij} := \pi_{kj}$ ;
13          Stack.push( $v_j$ );
14        end
15      end
16    end
17  end
18 end

```

Algorithm 2: Tree(W)

In Algorithm 2 vertices of OUT_k are visited in DFS order, which is facilitated by using the stack. We can avoid pushing and popping of each vertex by precomputing two read-only arrays dfs and $skip$ to support the traversal of OUT_k . The array dfs contains of OUT_k vertices as visited in the DFS order. Conversely, the array $skip$ is used to skip OUT_k subtree when relaxation does not succeed [2].

3.3 The Hourglass algorithm

The Tree algorithm can be further improved by using another tree. The second tree, denoted by IN_k is similar to shortest path tree, except that it is a shortest path "tree" for paths from v_i to v_k for each $v_i \in V \setminus \{v_k\}$. Precisely IN_k is not a tree, but if direction of arcs is reversed, it shifts it into a tree with v_k as the root. This

```

1 Initialize n empty trees:  $T_1, T_2, \dots, T_n$ ;
2 for k := 1 to n do
3    $T_1.Root := v_1$ ;
4 end
5 for i := 1 to n do
6   if  $i \neq k$  then
7     Make  $T_i$  a subtree of the root of  $T_{ki}$ .
8   end
9 end
10 return  $T_k$ 

```

Algorithm 3: Construct $OUT_k(\pi)$

is actually a substitute of the for loop on variable i in line 2 of Algorithm 1 and in line 4 of Algorithm 2. This modification of Floyd-Warshall algorithm is named the Hourglass algorithm, the name comes from placing IN_k tree atop the OUT_k tree, which gives it an hourglass-like shape, with v_k being at the neck. The pseudocode of the Hourglass algorithm is given in Algorithms 4 and 5. Additional algorithm constructs IN_k similarly to the construction of OUT_k , except that the matrix ϕ_{ik} is used instead [2].

```

1 Initialize  $\pi$ , an  $n \times n$  matrix, as  $\pi_{ij} := i$ ;
2 Initialize  $\phi$ , an  $n \times n$  matrix, as  $\phi_{ij} := i$ ;
3 for k := 1 to n do
4    $OUT_k := ConstructOUT_k(\pi)$ ;
5    $IN_k := ConstructIN_k(\phi)$  forall children  $v_i$  of  $v_k$  in  $IN_k$ 
6   do
7     RecurseIN( $W, \pi, IN_k, OUT_k, v_i$ );
8   end

```

Algorithm 4: Hourglass(W)

In practice, the algorithm's efficiency can be enhanced by employing an additional stack instead of recursion. This optimization significantly speeds up the implementation process. It's important to highlight that the worst-case time complexity of the Hourglass (and Tree) algorithm remains $O(n^3)$. This scenario is evident when all shortest paths are direct arcs themselves, resulting in a tree structure where all leaves are children of the root, and this configuration remains unchanged throughout the algorithm's execution [2].

4 EVALUATION

4.1 Testing environment

All algorithms and graph generators were implemented in C and C++ and compiled using gcc version 6.3.0. The tests were ran on an AMD Ryzen 7 5700U with Radeon Graphics 1.80 GHz with 16GB RAM running on Windows 11 64-bit.

4.2 Graph generation

Three different variant of random graphs were generated using Erdős-Rényi model and Albert-Barabási model. At the beginning of generating each variant seed was set to 1 and was incremented

```

1 Stack := empty;
2 Stack.push( $v_k$ );
3 while Stack  $\neq$  empty do
4    $v_x :=$  Stack.pop();
5   forall children  $v_j$  of  $v_x$  in  $OUT_k$  do
6     if  $W_{ik} + W_{kj} < W_{ij}$  then
7        $W_{ij} := W_{ik} + W_{kj}$ ;
8        $\pi_{ij} := \pi_{kj}$ ;
9        $\phi_{ij} := \phi_{ik}$ ;
10      Stack.push( $v_j$ );
11    else
12      Remove the subtree of  $v_j$  from  $OUT_k$ ;
13    end
14  end
15 end
16 forall children  $v_i$  of  $v_k$  in  $IN_k$  do
17   RecurseIN( $W, \pi, \phi, IN_k, OUT_k, v_i$ );
18 end
19 Restore  $OUT_k$  by reverting changes done by all iterations of
   line 12;

```

Algorithm 5: *RecurseIN*(W, π, IN_k, OUT_k, v_i)

by +1 for each new graph. All graphs generated underwent a DFS search to ensure strong connectivity¹.

Firstly, using binomial Erdős-Rényi model the input values for creating random graphs were the number of vertices, denoted as n , and probability, denoted as p . Each arc out of $n * (n - 1)$ in a directed graph is included with probability p , independently from every other arc. Weights are added uniformly from the interval $[0,1]$. Four different sizes of graphs (512, 1024, 2048 and 4096 vertices) and five different probabilities (0.1, 0.3, 0.5, 0.7, 0.9) were selected as input. For each instance of the input five different graphs were created, all together 100 different graphs were created using this model.

Secondly, a different variant was used, called uniform Erdős-Rényi model. Input values for creating graphs were number of vertices and number of arcs, denoted by m . Out of all $n * (n - 1)$ possible arcs in a directed graph a random permutation was made to select the desired number of arcs. Weights are added uniformly from the interval $[0,1]$. Again the sizes of graphs were 512, 1024, 2048 and 4096 vertices and inputs for m were $5 * n$, $10 * n$, $20 * n$, $50 * n$, $100 * n$. All together 100 different graphs were created using this model.

Finally, using Albert-Barabási model the input values for creating random graphs were number of vertices of final graph, number of vertices of initial graph, denoted by c , and number of arcs added in each round, denoted by m . Initially a clique of size c is created. At each step, one new vertice is added, with m new arcs to the vertices already in the graph. With preferential attachment the vertices with higher degree have a higher probability to receive new arcs. The graph constructed after $n - c$ steps is undirected. To determine the direction of each arc, a function similar to a coin flip is employed. This function randomizes the orientation of the arcs, ensuring an

¹In Albert-Barabási model some seeds did not produce strongly connected graph and therefore more than 80 seeds were used.

equal distribution where half of the arcs lie above the main diagonal in the adjacency matrix, and the other half lie below it. For this evaluation the m value was fixed at 15 and C value was fixed at 30. Again the sizes of graphs were 512, 1024, 2048 and 4096 vertices and for each instance 20 different graphs were generated (80 together). Another number is chosen uniformly from interval $(0,1]$ for each arc weight.

4.3 Evaluation

As discussed Tree and Hourglass algorithms were compared with Floyd-Warshall algorithm. The results on graphs generated by binomial Erdős-Rényi model are shown in figure 1, by uniform Erdős-Rényi model are shown in figure 2 and by Albert-Barabási model are shown in figure 3. To better visualize the results, natural logarithm of arcs is used on the x axis. Both modifications performed much better than Floyd-Warshall algorithm, especially as the number of vertices gets higher. It is worth noting that, between the Tree and Hourglass algorithms, the Tree algorithm demonstrated slightly better results in the performance comparison.

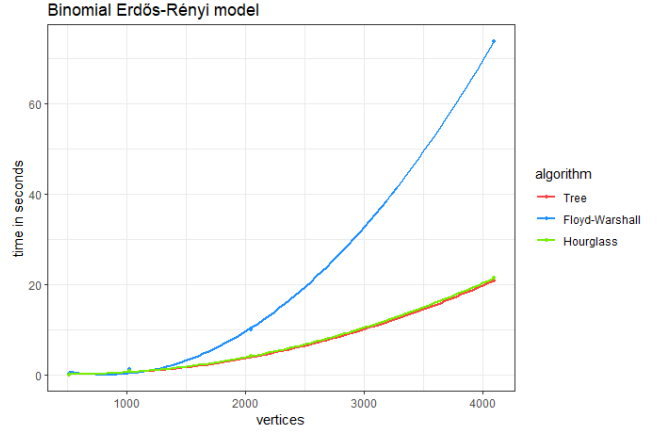


Figure 1: Binomial Erdős-Rényi model results

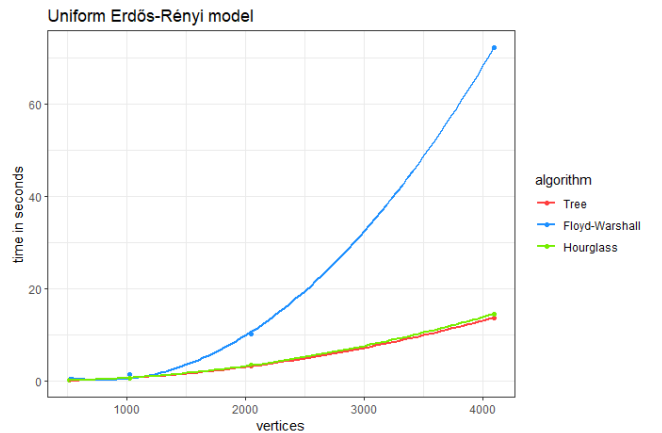


Figure 2: Uniform Erdős-Rényi model results

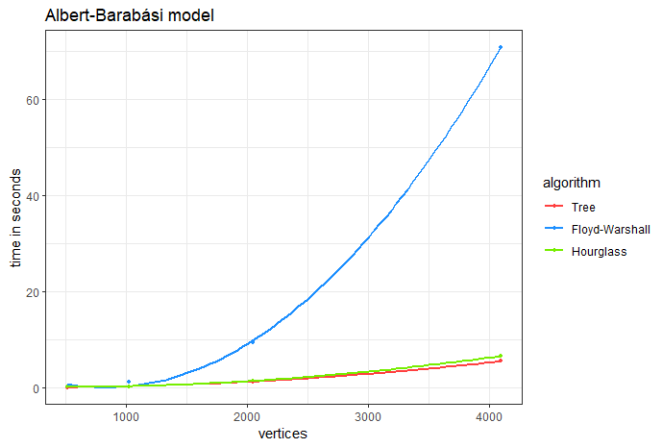


Figure 3: Albert-Barabási model results

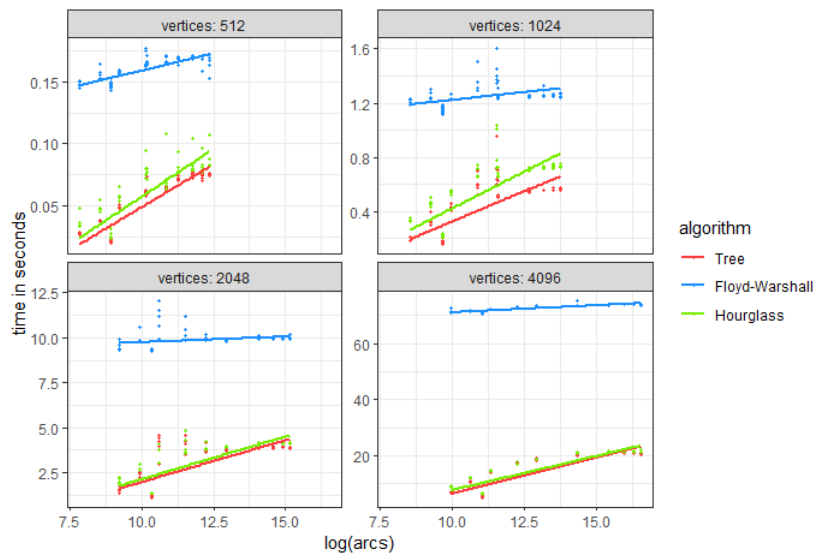


Figure 4: Combined results

In figure 4 results of all three variants of graphs are shown. As expected both modifications were slower as the graph got more dense and the number of vertices stayed the same. Moreover also Floyd-Warshall algorithm was a little bit slower as graph got denser which is unexpected.

5 CONCLUSIONS

In this paper, a straightforward evaluation of three algorithms for finding shortest paths in a network was presented. Random networks were generated using the Erdős-Rényi and Albert-Barabási models. The paper’s results showed that both modifications of the algorithms performed better than the Floyd-Warshall algorithm, especially when the size of the graphs (vertices) was larger.

To further enhance the comprehensiveness of the evaluation, the Tree and the Hourglass algorithms will be further assessed

on graphs that are not strongly connected. By incorporating non-strongly connected graphs into the assessment, deeper insights into the behavior and robustness of the algorithms in real-world scenarios, are expected to be gained.

REFERENCES

- [1] R. K. Ahuja, T. L. Magnanti, and J. B. Orlin. 1993. *Network Flows: Theory, Algorithms, and Applications*. Prentice-Hall, Inc., USA.
- [2] A. Brodnik, M. Grgurovič, and R. Požar. 2021. Modifications of the Floyd-Warshall Algorithm with Nearly Quadratic Expected-Time. *ARS MATHEMATICA CONTEMPORANEA* (2021). <https://amc-journal.eu/index.php/amc/article/view/2467>
- [3] R. W. Floyd. 1962. Algorithm 97: Shortest path. *Commun. ACM* 5 (1962), 345. <https://doi.org/10.1145/367766.368168>