

## MINIMUM COST PATH IN A HUGE GRAPH

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**ABSTRACT.** Suppose we have a weighted graph  $G = (V, E, c)$ , where  $V$  is the set of vertices,  $E$  is the set of arcs, and  $c : E \rightarrow \mathbf{R}_+$  is the cost function. Determining a minimum cost path between two given nodes of this graph can take  $\mathcal{O}(m \log n)$  time, where  $n = |V|$  and  $m = |E|$ . If this graph is huge, say  $n \approx 700000$  and  $m \approx 2000000$ , determining a minimum cost path can be a serious time consuming task. So we must develop an algorithm that quickly determines a path having the cost near the optimum one.

**Keywords:** minimum cost path, huge graph, strongly connected component

### 1. INTRODUCTION

If we develop a route planning application, it is very important to use efficient algorithms that determine a path between two distinct nodes. But what if the application manages a huge graph? This is the case of a complete roads map of a medium country, like Romania. In this case we simply ask to find a path that has the cost near the optimum one, but this path must be found very quickly. A fast algorithm is very important if the application is running on a server, and must satisfy the requests that come from many users by Internet.

To develop the algorithm proposed in this paper, we need some remarks, such as:

- (i) each link (arc) can be either a main road or a secondary road;
- (ii) the number of main roads (class  $MR$ ) is very small as compared to the secondary ones (class  $SR$ ); suppose the cardinality of  $MR$  is 8-10% of the cardinality of  $MR \cup SR$ ;
- (iii) the number of vertices that belong to a main road (class  $MV$ ) is very small as compared to the number of vertices that belongs to a secondary road (class  $SV$ ); suppose the cardinality of  $MV$  is 8-10% of the cardinality of  $MV \cup SV$ ;
- (iv) the main roads are uniformly scattered among the secondary ones.

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2000 *Mathematics Subject Classification.* 05C40.

1998 *CR Categories and Descriptors.* G.2.2 [**Mathematics of Computing**]: Discrete Mathematics – *Graph Theory*.

We need to exploit the following idea (see figure 1). Given two distinct vertices  $s$  and  $t$ , each of them being of  $SV$  type, we first determine a minimum cost path from  $s$  to the nearest vertex  $s_1$  that is of  $MV$  type. We also determine a minimum cost path from  $t$  to the nearest vertex  $t_1$  (reversing!) that is of  $MV$  type. Next we determine a minimum cost path from  $s_1$  to  $t_1$ , using only the main roads. The solution of the problem is the union of these three paths. This algorithm is very fast because:

- the paths from  $s$  to  $s_1$  and from  $t_1$  to  $t$  can be quickly determined: see remark (iv);
- the path from  $s_1$  to  $t_1$  can also be quickly determined: see remarks (ii) and (iii).

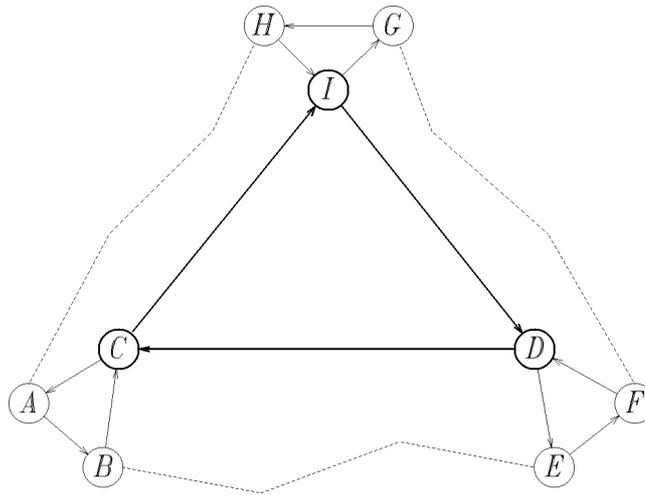


FIGURE 1. Examples of paths determined using the *NearOptimumPath* algorithm. From  $A$  to  $C$ :  $A - B - C$ , from  $C$  to  $D$ :  $C - I - D$ , from  $B$  to  $F$ :  $B - C - I - D - E - F$

In order to use the algorithm sketched above, we have to prepare two supplementary structures.

We scan the original graph to find all the vertices of  $MV$  type. Using these nodes we build a partial subgraph that has only nodes of  $MV$  type and arcs of  $MR$  type. Let this partial subgraph be  $G_s = (V_s, E_s)$ . We also build the graph  $G_i = (V, E_i)$  - the inverse graph of  $G$ , where the set  $E_i$  is defined as follows:

if  $(x, y) \in E$  then  $(y, x) \in E_i$ , id est, for each arc  $(x, y)$  from  $E$  we insert the inverse arc  $(y, x)$  to  $E_i$ .

We describe below the proposed algorithm.

**Algorithm** *NearOptimumPath*;

*Input.* The original graph  $G$ , the inverse graph  $G_i$ , the partial subgraph  $G_s$ ; two distinct nodes  $s$  and  $t$ .

*Output.* Near optimum path from  $s$  to  $t$ .

**begin**

\* **if** ( $s$  is not of  $MV$  type) **then**

determine, in graph  $G$ , using the algorithm of Dijkstra and selection trees, a minimum cost path  $D_1$  from  $s$  to the nearest node  $s_1$  of  $MV$  type;

**if** ( $t$  is detected before reaching a node of  $MV$  type) **then**

**stop:** we found the searched path;

**else** ( $s$  is of  $MV$  type)

**let**  $s_1 := s$ ;  $D_1 := \emptyset$ ;

\* **if** ( $t$  is not of  $MV$  type) **then**

determine, in graph  $G_i$ , using the algorithm of Dijkstra and selection trees, a minimum cost path  $D_2$  from  $t$  to the nearest node  $t_1$  of  $MV$  type;

**else** ( $t$  is of  $MV$  type)

**let**  $t_1 := t$ ;  $D_2 := \emptyset$ ;

\* determine, in graph  $G_s$ , using the algorithm of Dijkstra and selection trees, a minimum cost path  $D_3$  from  $s_1$  to  $t_1$ ;

\* report the union of these three paths:  $D := D_1 \cup \text{reverse}(D_2) \cup D_3$ ;

**end** (algorithm).

When can we use this algorithm? The following theorem below answer this question.

**Theorem 1.** *The algorithm NearOptimumPath can find a path between any two distinct vertices of the graph  $G$  if and only if  $G$  and  $G_s$  are both strongly connected.*

*Proof.* These two conditions are obviously sufficient, and the graph  $G$  must also be strongly connected. We have to prove that the graph  $G_s$  must also be strongly connected. Indeed, suppose that the graph  $G_s$  is not strongly connected. It is possible that the algorithm wrongly reports that there is no path between two given nodes, even if such a path exists - the graph  $G$  is strongly connected. Let examine the figure 2:

- one can find two distinct nodes  $s$  and  $t$ , each of them being of  $MV$  type, but there is no path from  $s$  to  $t$  having only arcs of  $MR$  type;
- one can find two distinct nodes  $s$  and  $t$ , at least one being of  $SV$  type, and the algorithm find two nodes  $s_1$  and  $t_1$ , but there is no path from  $s_1$  to  $t_1$  having only arcs of  $MR$  type.  $\square$

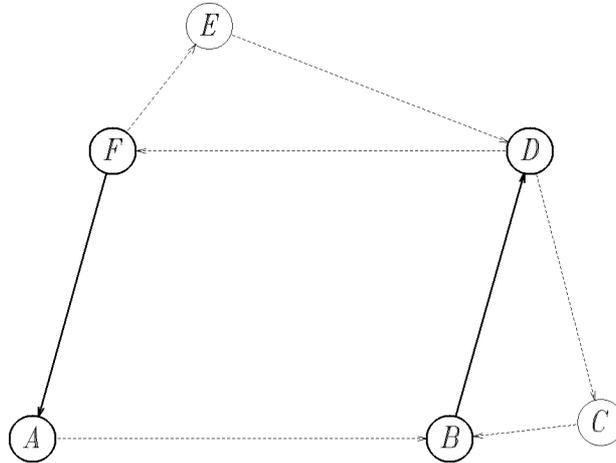


FIGURE 2. The algorithm can not find any path from  $A$  to  $D$  or from  $A$  to  $E$ , because the partial subgraph engendered by the arcs  $(F, A)$  and  $(B, D)$  is not strongly connected

How quickly can we determine a path using this algorithm? The running time of an implementation that uses this algorithm, as compared to the original Dijkstra's algorithm, is proportional to the percentage of the number of the main roads and main vertices.

We saw that this method needs to arrange the information into an organized structure to accelerate searching. This preprocessing phase is necessary because queries will be performed repeatedly on the same graph; these are so called repetitive-mode queries. How much time is needed to arrange the data for searching? To answer this question we present below an algorithm that determines the strongly connected components of a directed graph. This presentation is a review of the algorithm presented in [3].

**Algorithm** *StronglyConnectedComponents*;

*Input.* A directed graph  $G = (V, E)$ .

*Output.* An array  $C$  : the strongly connected components, each vertex being marked with the component number that contains it.

**begin**

**for** (each vertex  $x \in V$ ) **do**

$Mk[x] := False$ ;

$Md := 0$ ;

**for** (each vertex  $x \in V$ ) **do**

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if ( $Mk[x] = False$ ) then
     $ScanMark(x)$ ;
Sort  $D$  on decreasing order, storing for each mark the associated vertex in  $X$ ;
for (each vertex  $x \in V$ ) do
     $C[x] := False$ ;
Build the inverse graph  $G'$  corresponding to the graph  $G$ ;
 $nc := 0$ ;
Warning! The procedure  $ScanCnx$  manages the graph  $G'$ ;
    for (each vertex  $x \in X$ ) do
        if ( $C[x] = 0$ ) then begin
             $nc := nc + 1$ ;  $ScanCnx(x)$ ;
        end
end (algorithm).
Procedure  $ScanMark$ (vertex  $x$ );
begin
     $Mk[x] := True$ ;
    for (each vertex  $y$ , successor of  $x$ ) do
        if ( $Mk[y] = 0$ ) then     $ScanMark(y)$ ;
     $Md := Md + 1$ ;  $D[x] := Md$ ;
end (procedure).
Procedure  $ParcCnx$ (vertex  $x$ );
begin
     $C[x] := nc$ ;
    for (each vertex  $y$ , successor of  $x$ ) do
        if ( $C[y] = 0$ ) then
             $ScanCnx(y)$ ;
end (procedure).

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The determination of the strongly connected components of a directed graph needs  $\mathbf{O}(n \log n + m)$  time, and building the inverse graph  $G_i$  and the partial subgraph  $G_s$  takes  $\mathbf{O}(m)$  time. So we have

**Theorem 2.** *The supplementary structures used by the algorithm  $NearOptimumPath$  are determined in  $\mathbf{O}(n \log n + m)$  preprocessing time.*

We can use the following technique for a parallel architecture. The searching process starts two execution threads: the first thread uses the algorithm  $NearOptimumPath$ , and the second thread uses the original Dijkstra's algorithm. If one of these two threads finds a path (which may be optimal or not), it must stop the other thread. In this case we don't need to impose a very restrictive condition: it is not necessary that the partial subgraph  $G_s$  be strongly connected.

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