Shortest Paths in Nearly Conservative Digraphs

Zoltán Király *

September 26, 2014

Abstract

We introduce the following notion: a digraph D = (V, A) with arc weights $c : A \to \mathbb{R}$ is called nearly conservative if every negative cycle consists of two arcs. Computing shortest paths in nearly conservative digraphs is NP-hard, and even deciding whether a digraph is nearly conservative is coNP-complete.

We show that the "All Pairs Shortest Path" problem is fixed parameter tractable with various parameters for nearly conservative digraphs. The results also apply for the special case of conservative mixed graphs.

1 Introduction

We are given a digraph D = (V, A), a weight (or a length) function $c : A \to \mathbb{R}$ is called conservative (on D) if no directed cycle with negative total weight ("negative cycle" for short) exists, and c is called λ -nearly conservative if every negative cycle consists of at most λ arcs.

The APSP (All Pairs Shortest Paths) problem we are going to solve has two parts, first we must decide whether c is λ -nearly conservative, next, if the answer for the previous question is YES, then for all (ordered) pairs $s \neq t$ of vertices the task is to determine the length of the shortest (directed and simple) path from s to t.

In this paper we concentrate on the case $\lambda = 2$, a 2-nearly conservative weight function c is simply called *nearly conservative* in this paper. A mixed graph G = (V, E, A) on vertex set V has the set Eof undirected edges and the set A of directed edges (i.e., arcs). A weight function $c : E \cup A \to \mathbb{R}$ is called conservative if no cycle with negative total weight exists. For a mixed graph we can associate a digraph by replacing each undirected edge e having endvertices u and v by two arcs uv and vu with weights c(uv) = c(vu) = c(e). It is an easy observation that the resulting c is nearly conservative on the resulting digraph if and only if the original weight function was conservative on the original mixed graph, and in this case the solution of the APSP problem remains the same.

Arkin and Papadimitriou proved in [1] that the problems of detecting negative cycles and finding the shortest path in the absence of negative cycles are both NP-hard in mixed graphs. Consequently, checking whether c is nearly conservative on D is coNP-complete, and solving the APSP problem in the case c is nearly conservative on D is NP-hard. In this paper we give FPT algorithms for this problem related to various parameters.

Though it was a surprise to the author, he could not find any algorithm for dealing with these problems (despite the fact that many paper are written about the Chinese Postman problem on mixed graphs). We only found two more papers that are somehow related to this topic. In [4] for the special case of skew-symmetric graphs shortest "regular" paths are found in polynomial time if no negative "regular" cycle exist. In [2] for the similar special case of bidirected graphs minimum mean edge-simple cycles are found in polynomial time, this is essentially the same as finding minimum mean "regular" cycles in skew-symmetric graphs. The class of nearly conservative graphs seems to be not studied (and defined) in the literature, as well as we could not find any FPT result about APSP.

^{*}Department of Computer Science and Egerváry Research Group (MTA-ELTE), Eötvös University, Pázmány Péter sétány 1/C, Budapest, Hungary. Research was supported by grants (no. CNK 77780 and no. K 109240) from the National Development Agency of Hungary, based on a source from the Research and Technology Innovation Fund. E-mail: kiraly@cs.elte.hu

For defining the parameters we are going to use, we first define the notion of negative trees. Given D and c, we associate an undirected graph F = (V, E) as follows. Edge-set E consists of pairs $u \neq v$ of vertices for which both uv and vu are arcs in A, and c(uv) + c(vu) < 0. We can construct F in time O(|A|), and can also check whether it is a forest. We claim that if F is not a forest, then c is not nearly conservative on D, so our algorithm can stop with this decision. If F contains a cycle, then it corresponds to two oppositely directed cycles of D, and the sum of the total weights of these two cycles are negative, proving that c is not nearly conservative.

From now on we will suppose that F is a forest, and we call its nontrivial components (that have at least one edge) the **negative trees**.

Our first parameter k_0 is the number of negative trees, and we give an $O(2^{k_0} \cdot n^4)$ algorithm for the APSP problem (where n = |V|). Later we refine this algorithm for parameter k_1 , which is the maximum number of negative trees in any strongly connected component of D, and finally for parameter k_2 , which is the maximum number of negative trees in any weakly 2-connected block of any strongly connected component of D (for the definitions see the next section). Our final algorithm also runs in time $O(2^{k_2} \cdot n^4)$. Consequently, if there is a constant γ such that every weakly 2connected block of any strongly connected component of D has at most γ negative trees, then we have a polynomial algorithm.

The preliminary version of this paper appeared in [6] for the special case of mixed graphs. In that paper we also gave a strongly polynomial algorithm for finding shortest exact walk (a walk with given number of edges) in any non-conservative mixed graph.

2 Definitions

For our input digraph D we may assume it is simple. An arc from u to v is called a *loose arc* if there is another arc from u to v with a smaller weight. In a shortest path between s and t (if $s \neq t$) neither loops nor loose arcs can appear. Consequently, as a preprocessing, we can safely delete these (and also keep only one copy from multiple arcs having the same weight).

However for our purposes multiple arcs will be useful, so we will use them for describing the algorithm. We use the convention that the notation uv always refers to the shortest arc from u to v.

We call an arc uv of D special if vu is also an arc, and moreover c(uv) + c(vu) < 0. Other arcs are called *ordinary*. For a special arc uv the special arc vu is called its *opposite*. As a part of the preprocessing, we add some loose arcs to D. For every special arc uv we add an arc a from v to u with weight c(a) = -c(uv). By the definition of special arcs, these are really loose arcs, as -c(uv) > c(vu). We call these arcs added ordinary arcs, or shortly loose arcs. We call the improved digraph also D, and its arc set is called A. Arc set A is decomposed into $A = A_s \cup A_o$, where A_s is the set of special arcs, and A_o is the set of ordinary (original or added) arcs. (The main purpose of this procedure is the following. We will sometimes work in the ordinary subdigraph $D_o = (V, A_o)$, and we need to maintain the same reachability: if there is a path from s to t in D, then there is also a path from sto t in D_o .) Our main property remained true: if c is nearly conservative on D, then every negative cycle consists of two oppositely directed special arcs. Remark: special arcs may have positive length, so loose arcs may have negative length. We call a path ordinary if all its arcs are ordinary. Note that by the assumptions $|A| \leq 2n^2$, where n = |V|.

Given D and c, we associate an undirected graph F = (V, E) as follows. Edge-set E consists of unordered pairs $u \neq v$ of vertices for which uv is a special arc in A_s . As we detailed in the Introduction, if F is not a forest, then c is not nearly conservative on D. We consider this process as the last phase of the preprocessing: we determine F, and if it is not a forest, then we stop with the answer "NOT NEARLY CONSERVATIVE".

From now on we suppose that F is a forest, and we call its nontrivial components (that have at least one edge) the **negative trees**. If T is a negative tree, then $\mathbf{V}(\mathbf{T})$ denotes its vertex set, and $\mathbf{A}(\mathbf{T})$ denotes the set of special arcs that correspond to its edges. If $s, t \in V(T)$ are two vertices of T, then $\mathbf{d}^{\mathbf{T}}(\mathbf{s}, \mathbf{t})$ denotes the length of unique path from s to t in A(T).

A walk from v_0 to v_ℓ (or a v_0v_ℓ -walk) is a sequence

$$W = v_0, a_1, v_1, a_2, v_2, \dots, v_{\ell-1}, a_\ell, v_\ell$$

where $v_i \in V$ for all *i*, and a_j is an arc from v_{j-1} to v_j for all *j*. A walk is closed if $v_0 = v_\ell$. A closed walk is also called here a v_0v_0 -walk. A number ℓ of arcs used by a walk *W* is denoted by

|W|. The length (or weight) c(W) of a walk W is defined as $\sum_{j=1}^{\ell} c(a_j)$. If W_1 is a s_1v -walk and W_2 is a vt_2 -walk, then their concatenation is denoted by $W_1 + W_2$. For a walk W we use the notation $W[v_i, v_j]$ for the corresponding part $v_i, a_{i+1}, \ldots, a_j, v_j$ if i < j.

A walk W is **special-simple** if no special arc is contained twice in it, moreover, if W contains special arc uv, then it does not contain its opposite vu. A walk is a *path* if all the vertices v_0, \ldots, v_ℓ are distinct. A closed walk is a *cycle* if all the vertices v_0, \ldots, v_ℓ are distinct, with the exception of $v_0 = v_\ell$. If $|W| = \ell = 0$, then we call the walk also an empty path (its length is 0), and in this paper unconventionally the empty path will also be considered as an empty cycle. The distance $d_D(s,t) = d(s,t)$ of t from s is the length of the shortest path from s to t (where $s, t \in V$).

The relation: there is a path in D from s to t and also from t to s, is obviously an equivalence relation, its classes are called the strongly connected components of D. (Notice that a negative tree always resides in one strongly connected component.) A weakly 2-connected block of a digraph is a 2-connected block of the underlying undirected graph (where arcs are replaced with undirected edges).

An algorithm is FPT for a problem with input size n and parameter k if there is an absolute constant γ , and a function f such that the running time is $f(k) \cdot O(n^{\gamma})$. (Originally FPT stands for "fixed parameter tractable", and it is an attribute of the problem, however in the literature usually the corresponding algorithms are also called FPT.) In this paper we give FPT algorithms for the APSP problem for nearly conservative digraphs.

In the simplest version we assume that there is just one negative tree and it is spanning V. Next we give an algorithm for the case where we still have only one negative tree, but it is not spanning V. These algorithms are polynomial and simple.

Then we use various parameters: k_0 is the number of negative trees in D, k_1 is the maximum number of negative trees in any strongly connected component of D, and k_2 is the maximum number of negative trees in any weakly 2-connected block of any strongly connected component of D. (Clearly $k_0 \ge k_1 \ge k_2$.) The main goal of this paper to give an $O(2^{k_2} \cdot n^4)$ algorithm for the APSP problem for the case $\lambda = 2$, i.e., for deciding whether c is nearly conservative on D, and if it is, then for calculating the distances $d_D(s, t)$ for each (ordered) pair of vertices $s, t \in V$.

In the next section we show some lemmas. In Section 4 we give some polynomial algorithms for the case of one negative tree. In Section 5 we give an FPT algorithm where the parameter k_0 is the total number of negative trees in D. Next, in Section 6 we extend it to the case where k_2 only bounds the number of negative trees in any weakly 2-connected block of any strongly connected component.

Our main goal is only giving the length of the shortest paths, in Section 7 we detail how the actual shortest paths themselves can be found.

Finally in Section 8 we conclude the results, show their consequences to mixed graphs, and pose some open problems.

3 Lemmas

In this section we formulate some lemmas. Though each of them can be easily proved using the newly introduced notions and the statements of the preceding lemmas, we could not find these statements in the literature (neither in an implicit form).

We premise some unusual aspects of nearly conservative weight functions. Usually shortest path algorithms use the following two facts about conservative weight functions. If P is a shortest sx-path and Q is a shortest xt-path, then P + Q contains an st-path not longer than c(P) + c(Q). If P is a shortest st-path containing vertices u and v (in this order), then P[u, v] is a shortest uv-path. These two statements are NOT true for nearly conservative weight functions.

Remember that $D = (V, A_s \cup A_o)$ is the improved digraph with loose arcs, and the associated graph F is a forest.

Lemma 1. Weight function c is nearly conservative on D if and only if there is no negative specialsimple closed walk.

Proof. If C is a negative cycle consisting of at least three arcs, then it is also a negative special-simple closed walk. On the other hand, suppose that C is a negative special-simple closed walk with a minimum number of arcs, and assume that C is not a cycle, that is there are $0 < i < j \leq \ell$ such

that $v_i = v_j$. Now C decomposes into two special-simple closed walks with less arcs, clearly at least one of them has negative length, a contradiction.

Lemma 2. If c is nearly conservative on D, and $s, t \in V$, and Q is a special-simple st-walk, then we also have an st-path P with $c(P) \leq c(Q)$, and P contains only arcs of Q.

Proof. Let Q be a shortest special-simple st-walk (which exists by the previous lemma and as c is nearly conservative) having the minimum number of arcs.

By the previous lemma, if s = t, then the empty path serves well as P. So we may assume that $s \neq t$ and Q is not a path, i.e., there are $0 \leq i < j \leq \ell$ such that $v_i = v_j$. Now Q decomposes to a special-simple sv_i -walk Q_1 , a special-simple closed walk C through v_i and an special-simple v_jt -walk Q_2 . By the previous lemma C is nonnegative, so $c(Q_1 + Q_2) \leq c(Q)$, consequently $Q_1 + Q_2$ is a not longer special-simple st-walk with less number of arcs, a contradiction.

Suppose T is a negative tree, $u, v \in V(T)$, and P is a *uv*-path in D' = D - A(T). If $c(P) < -d^{T}(v, u)$, then c is not nearly conservative on D because otherwise $P + P_{vu}^{T}$ would be a negative special-simple closed walk, where P_{vu}^{T} is the *vu*-path in A(T). Otherwise, if $c(P) \geq -d^{T}(v, u)$, then we have a *uv*-path P' in D' consisting of loose arcs such that $c(P') \leq c(P)$. Using this train of thought we get the following lemmas that play the central role in our algorithms.

Lemma 3. Let T be a negative tree, and assume that c is nearly conservative on D. If P is a shortest st-path using some vertex of V(T), then let u be the first vertex of P in V(T), and let v be the last vertex of P in V(T). Then P[u, v] uses only special arcs from A(T). Consequently, if $s, t \in V(T)$, then $d(s, t) = d^{T}(s, t)$.

Proof. Remember that a uv-path in A(T) may have positive length. Fortunately, by the definition of u and v, there are no vertices of P preceding u or following v inside V(T), and this fact can be used successfully.

Suppose P is a shortest st-path. By the observation made before the lemma, for any $u', v' \in V(T)$, any subpath of form P[u', v'] that uses no arcs from A(T) can be replaced by loose arcs without increasing the length. After we made all these replacements, we replaced P[u, v] by a special-simple uv-walk Q' such that Q' contains only arcs in A(T) and loose arcs, and $c(Q') \leq c(P[u, v])$. By Lemma 2, Q' contains a uv-path P' with $c(P') \leq c(Q')$. We got P' by eliminating cycles, if any cycle had positive length, then we get c(P') < c(Q'). Suppose now that all eliminated cycles had zero length, meaning that each one had the form x, a, y, yx, x, where a is the loose arc from x to y and yx is the special arc from y to x. If after deleting all these cycles P' still has a loose arc a from x to y then it can be replaced safely with the special arc xy yielding again a path strictly shorter than Q'. Thus the only possibility where we can only get a P' with the same length (as Q') is that the special-simple uv-walk Q' consisted of the uv-path P_{uv}^T inside A(T) and additionally some zero length cycle described above, and moreover $P' = P_{uv}^T$. Now we claim that in this case the path P[u, v] used only arcs from A(T), i.e., it was also P_{uv}^T . Otherwise there are vertices $x, y \in V(T)$ such that x is on P_{uv}^T , y is not on it, and Q' contains one loose arc and one special arc between x and y. However in this case vertex x had to be included twice in path P, a contradiction.

To finish the proof observe that P[s, u] + P' + P[v, t] is an *st*-path, and in the case $P[u, v] \neq P_{uv}^T$ it would be shorter than the shortest path *P*.

Lemma 4. Let T be a negative tree, and assume that c is nearly conservative on digraph D' = D - A(T) defining distance function d'. Then c is nearly conservative on D if and only if for any pair of vertices $u, v \in V(T)$ we have $d'(u, v) \ge -d^T(v, u)$.

Proof. We showed that the condition is necessary. Suppose that C is a negative cycle in D having at least three arcs. If it has at most one vertex in V(T), then it is also a negative cycle in D'. We claim that we can construct a special-simple negative closed walk C' which uses only loose arcs and arcs in A(T). To achieve this goal, repeatedly take any subpath C[u, v], where $u, v \in V(T)$, but inner vertices of C[u, v] are in V - V(T). By the condition $c(C[u, v]) \geq -d^T(v, u)$, which means that changing C[u, v] to the uv-path consisting of loose arcs does not increase the length of C. We arrived at a contradiction, as the special-simple closed walk C' contains a negative cycle which is impossible by the definition of loose arcs.

4 Polynomial algorithms for the case $k_0 = 1$

First we give an $O(n^2)$ algorithm for the very restricted case, where we have only one negative tree T, and moreover it spans V. We claim first that c is nearly conservative on D if and only if for each ordinary arc uv we have $c(uv) \geq -d^T(v, u)$. If $c(uv) < -d^T(v, u)$, then we have a negative special-simple closed walk, so c is not nearly conservative by Lemma 1. Suppose now that $c(uv) \geq -d^T(v, u)$ holds for each ordinary arc uv, and C is a negative cycle in D with at least three arcs. As in the proof of Lemma 4, replace each ordinary arc uv of C by a uv-path consisting of loose arcs, this does not increase the length. We arrive at special-simple closed walk using only special and loose arcs that is negative. However this contradicts to the definition of loose arcs. We also got that in this case for any pair $s, t \in V$ the length of the shortest path is $d^T(s, t)$ by Lemma 3. Consequently it is enough to give an $O(n^2)$ algorithm for calculating distances $d^T(s, t)$. We suppose that $V = \{1, \ldots, n\}$ and initialize a length-n all-zero array D_u for each vertex u. Then we fill up these arrays in a top-down fashion starting from the root vertex 1. Let \mathcal{P} denote the subset of vertices already processed, initially it is $\{1\}$. If a parent u of an unprocessed vertex v is already processed, we process v: for each processed vertex x we set $D_v(x) = c(vu) + D_u(x)$, and set $D_x(v) = D_x(u) + c(uv)$, and put v into \mathcal{P} .

Next we give an $O(n^4)$ algorithm for the case where we have only one negative tree T, but we do not assume it to span V.

In digraph $D' = D - A(T) = D_o = (V, A_o)$ using the Floyd-Warshall algorithm (see in any lecture notes, e.g., in [3]), it is easy to check whether c is conservative on D' in time $O(n^3)$. If it is not conservative, then we return with output "NOT NEARLY CONSERVATIVE" (as in this case c clearly cannot be nearly conservative on D), and if it is conservative, then this algorithm also calculates the length d'(s,t) of all shortest paths in D' (for $s, t \in V$). If vertex t is not reachable from vertex s, then it gives $d'(s,t) = +\infty$ (remember that reachability is the same in D' as in D).

Then we calculate the distances $d^T(u, v)$ in time $O(n^2)$ as in the previous section. By Lemma 4, c is nearly conservative on D if and only if for all pairs $u, v \in V_T$ we have $d'(u, v) \ge -d^T(u, v)$, this can be checked in time $O(n^2)$. It remains to calculate the pairwise distances. If P is a shortest *st*-path, then it is either a ordinary path (having length d'(s, t)), or it has a first arc $uu' \in A(T)$ and a last arc $v'v \in A(T)$. The part P[u, v] must reside inside A(T) by Lemma 3.

Lemma 5. If c is nearly conservative on D, and T is the only negative tree, then the distance d(s,t) is

$$d(s,t) = \min\left(d'(s,t), \min_{u,v \in V_T} [d'(s,u) + d^T(u,v) + d'(v,t)]\right).$$

Proof. This is a consequence of Lemma 3. The trick used here is that a shortest *su*-path and a shortest *vt*-path in D' need not be arc-disjoint, this is the main purpose for which we introduced the notion of special-simple, so for the relation LHS \leq RHS we have to use Lemma 2.

These values can be easily calculated for all pairs in total time $O(n^4)$, so we are done.

5 FPT algorithm for parameter k_0

In this section we suppose that there are at most k_0 negative trees in D. Let T_1, \ldots, T_{k_0} be the negative trees, remember that we defined $A(T_i)$ as the set of special arcs that correspond to the edges of T_i . We denote by V_T the vertex set $\bigcup_i V(T_i)$.

First we compute distances d^{T_i} for all $1 \le i \le k_0$ in total time $\sum O(|V(T_i|^2) = O(n^2)$. Next we compute distances d' in digraph $D' = D - \bigcup_i A(T_i) = D_o$ in time $O(n^3)$, or stop if c is not nearly conservative on D'.

We use dynamic programming for the calculation remained. For all $J \subseteq \{1, \ldots, k_0\}$ we define the *J*-subproblem as follows. Solve the APSP problem in digraph $D_J = D - \bigcup_{i \in \{1, \ldots, k_0\} = J} A(T_i)$, and let

 d_J denote the corresponding distance function if c is nearly conservative on D_J (otherwise, if c is not nearly conservative on D_J for any J, we stop). We already solved the \emptyset -subproblem, $d_{\emptyset} \equiv d'$.

Lemma 6. Suppose we solved the (J - i)-subproblem for every $i \in J$ and found that c is nearly conservative on D_{J-i} . By Lemma 4, we can check whether c is conservative on D_J using only

distance functions d^{T_i} and d_{J-i} for one element $i \in J$. If yes, then we have

$$d_{J}(s,t) = \min\left(d_{\emptyset}(s,t), \min_{i \in J}[\min_{u,v \in V(T_{i})}(d_{\emptyset}(s,u) + d^{T_{i}}(u,v) + d_{J-i}(v,t)]\right)$$

Proof. First we show that LHS≥RHS. Let P be a shortest path in D_J . Either P is disjoint from $\bigcup_{j \in J} V(T_j)$, in this case its length is $d_{\emptyset}(s,t)$ in graph D_J . The other possibility is that P has some first vertex u in $\bigcup_{j \in J} V(T_j)$, say $u \in V(T_i)$. Let v denote the last vertex of P in $V(T_i)$. That is, P[s,u] goes inside D_{\emptyset} and P[v,t] goes inside D_{J-i} , and, by Lemma 3, P[u,v] goes inside $A(T_i)$.

To show that LHS \leq RHS we only need to observe that if P_1 is an *su*-path in D_{\emptyset} , P_2 is a *uv*-path in $A(T_i)$, and P_3 is a *vt*-path in D_{J-i} , then $P_1 + P_2 + P_3$ is a special-simple *st*-walk.

Remember that 'solving the APSP problem' is defined in this paper as first checking nearly conservativeness, and if c is nearly conservative, then calculate all shortest paths. As solving one subproblem needs $O(n^4)$ steps, we proved the following

Theorem 1. If D has k_0 negative trees, then the dynamic programming algorithm given in this section correctly solves the APSP problem in time $O(2^{k_0} \cdot n^4)$.

The weak blocks of a digraph refer to the 2-connected blocks of the underlying undirected graph. It is well known that the block-tree of an undirected graph can be determined in time $O(n^2)$ by DFS. If we have this decomposition and we also calculated APSP inside every weak block, then we can also calculate APSP for the whole digraph in additional time $O(n^3)$. Consequently we have

Corollary 7. If every weak block of D contains at most k'_0 negative trees, then we can solve the APSP problem in time $O(2^{k'_0} \cdot n^4)$.

6 General FPT algorithm for parameters k_1 and k_2

Suppose every strongly connected component of D contains at most k_1 negative trees. By the previous section we can solve the APSP problem inside each strongly connected component in total time $O(2^{k_1} \cdot n^4)$. If for any of them we found that c is not nearly conservative, then we stop and report the fact that c is not nearly conservative on D. Henceforth in this section we assume that for every strongly connected component K of D, c is nearly conservative on K. (In this situation clearly c is nearly conservative on D.) The distance function restricted to component K is denoted by d_K . If $s, t \in V(K)$, then every st-path goes inside K, thus $d(s,t) = d_K(s,t)$. It remains to calculate APSP in D for pairs s, t, that are in different strongly connected components.

We construct a new acyclic digraph D^* by first substituting every strongly connected component K by acyclic digraph D_K^* as follows. Suppose $V(K) = \{x_1^K, x_2^K, \ldots, x_r^K\}$, the vertex set of D_K^* will consist of 2r vertices, $\{a_1^K, a_2^K, \ldots, a_r^K, b_1^K, b_2^K, \ldots, b_r^K\}$. For each $1 \le i, j \le r$ the digraph D_K^* contains arc $a_i^K b_j^K$ with length $d_K(x_i^K, x_j^K)$.

In order to finish the construction of D^* , for every arc $x_i^K x_j^L$ of D connecting two different strongly connected components $K \neq L$, digraph D^* contains the arc $b_i^K a_j^L$ with length $c(x_i^K x_j^L)$. It is easy to see that D^* is truly acyclic and has 2n vertices. As D^* is a simple digraph, paths can be given by only listing the sequence of its vertices. We can calculate APSP in D^* in time $O(n^3)$ by the method of Morávek [7] (see also in [3]) if we run this famous algorithm from all possible sources s. It gives distance function d_{D^*} (where if t is not reachable from s, then we write $d_{D^*}(s,t) = +\infty$). The total running time is still $O(2^{k_1} \cdot n^4)$. We remark that if every strongly connected component has a spanning negative tree, then the running time is $O(n^3)$.

Theorem 2. Suppose $s = x_{i_0}^{K_0} \in V(K_0)$ and $t = x_{j_r}^{K_r} \in V(K_r)$ where $K_0 \neq K_r$ are different strongly connected components of D. Then the shortest st-path in D has length exactly $d_{D^*}(a_{i_0}^{K_0}, b_{j_r}^{K_r})$.

Proof. Vertex t is not reachable from s in D if and only if $b_{j_r}^{K_r}$ is not reachable from $a_{i_0}^{K_0}$ in D^* . Otherwise, suppose that $a_{i_0}^{K_0}, b_{j_0}^{K_0}, a_{i_1}^{K_1}, b_{j_1}^{K_1}, \dots, a_{i_r}^{K_r}, b_{j_r}^{K_r}$ is a shortest path P in D^* . For $0 \le \ell \le r$ let path P_ℓ be a shortest path in D from $x_{i_\ell}^{K_\ell}$ to $x_{j_\ell}^{K_\ell}$, this path obviously goes inside K_ℓ . We can construct an st-path Q in D with the same length as P has in D^* : $Q = P_0 + x_{j_0}^{K_0} x_{i_1}^{K_1} + P_1 + x_{j_1}^{K_1} x_{i_2}^{K_2} + P_2 + \ldots + P_{r-1} + x_{j_{r-1}}^{K_{r-1}} x_{i_r}^{K_r} + P_r$.

For the other direction, suppose that there are strongly connected components K_0, K_1, \ldots, K_r , such that the shortest *st*-path Q in D meets these components in this order, and for all ℓ the path Qarrives into K_ℓ at vertex $x_{i\ell}^{K_\ell}$ and leaves K_ℓ at vertex $x_{j\ell}^{K_\ell}$. As Q is a shortest path it clearly contains a path of length $d_{K_\ell}(x_{i\ell}^{K_\ell}, x_{j\ell}^{K_\ell})$ inside K_ℓ for each ℓ , consequently the following path has the same length in D^* : $P = a_{i_0}^{K_0}, b_{j_0}^{K_0}, a_{i_1}^{K_1}, b_{j_1}^{K_1}, \ldots, a_{i_r}^{K_r}, b_{j_r}^{K_r}$.

Using Corollary 7 we easily get the following more general statements.

Corollary 8. If every weak block of any strongly connected component of D contains at most k_2 negative trees, then we can solve the APSP problem in time $O(2^{k_2} \cdot n^4)$.

Corollary 9. If there is an absolute constant γ , such that in any weak block of any strongly connected component of D there are at most γ negative trees, then there is a polynomial time algorithm for the APSP problem that runs in time $O_{\gamma}(n^4)$.

7 Finding the paths

In this section we assume that c is nearly conservative on D.

We usually are not only interested in the lengths of the shortest paths, but also some (implicit) representation of the paths themselves. The requirement for this representation is that for any given s and t, one shortest st-path P must be computable from it in time $O(\ell)$ if ℓ is the number of arcs in P.

It is well known (see e.g., in [3]) that both the algorithm of Floyd and Warshall and the algorithm of Morávek can compute predecessor matrices Π (by increasing the running time by a constant factor only), with the property that for each $s \neq t$ the entry $\Pi(s,t)$ points to the last-but-one vertex of a shortest *st*-path. This representation clearly satisfies the requirement described in the previous paragraph.

For a digraph H let Π_H denote the predecessor matrix of this type, and suppose that for each strongly connected component K we computed Π_K , and we also computed Π_{D^*} . Then Π_D is easily computable as follows. Suppose that $s = x_{i_0}^{K_0}$ and $t = x_{j_r}^{K_r}$, and $\Pi_{D^*}(a_{i_0}^{K_0}, b_{j_r}^{K_r}) = a_{i_r}^{K_r}$. If $i_r \neq j_r$, then define $\Pi_D(s,t) = \Pi_{K_r}(x_{i_r}^{K_r}, x_{j_r}^{K_r})$, otherwise let $b_{j_{r-1}}^{K_{r-1}} = \Pi_{D^*}(a_{i_0}^{K_0}, a_{i_r}^{K_r})$ and define $\Pi_D(s,t) = x_{j_{r-1}}^{K_{r-1}}$. It remained to compute the predecessor matrices Π_K in the case where K is a strongly connected the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case where K is a strongly connected to be the predecessor matrices Π_K in the case Π_K is the predecessor matrices Π_K is the predece

It remained to compute the predecessor matrices Π_K in the case where K is a strongly connected component of D. In accordance with Section 5 from now on we call K as D (and forget the other vertices of the digraph), and the matrix we are going to determine is simply Π .

If s and t are vertices of the same negative tree T_i , then the method given in the first paragraph of Section 4 easily calculates $\Pi(s,t) = \Pi_{A(T_i)}(s,t)$. Next we call the Floyd-Warshall algorithm on D', and it can give $\Pi_{D'}$, then during the dynamic programming algorithm we determine matrices Π_{D_J} for all J.

Given s and t, by Lemma 6 if the minimum is $d_{\emptyset}(s,t)$, then $\Pi_{D_J}(s,t) = \Pi_{D_{\emptyset}}(s,t)$, otherwise we find i, u, v giving the minimum value. If $v \neq t$, then $\Pi_{D_J}(s,t) = \Pi_{D_{J-i}}(s,t)$, otherwise if v = t but $u \neq v$, then $\Pi_{D_J}(s,t) = \Pi_{A(T_i)}(s,t)$, and finally if $v = t = u \neq s$ then $\Pi_{D_J}(s,t) = \Pi_{D_{\emptyset}}(s,t)$.

Extending this setup for weak blocks is obvious.

8 Conclusion and open problems

We gave FPT algorithms for the NP-hard APSP problem in nearly conservative graphs regarding with various parameters.

For mixed graphs we have the following consequence. As nonnegative undirected edges can be replaced by two opposite arcs, we may assume that every undirected edge has negative length. Here the negative trees are the nontrivial components made up by undirected edges, and APSP problem is to check whether c is conservative on a mixed graph G, and if YES, then calculate the pairwise distances.

Remember, that for mixed graphs the APSP problem contains checking conservativeness, and if c is conservative on the mixed graph, then all shortest paths should be calculated.

Corollary 10. If every weak block of any strongly connected component of a mixed graph contains at most k_2 negative trees, then we can solve the APSP problem in time $O(2^{k_2} \cdot n^4)$.

Finally we pose three open problems. A weight function is even-nearly conservative if every negative cycle consist of an even number of arcs.

Question 1. Is there an FPT algorithm for shortest paths if c is 3-nearly conservative? (The parameter should not contain the number of negative triangles.)

Question 2. Is there a polynomial or FPT algorithm for recognizing even-nearly conservative weights? This would be interesting even if we restrict the digraph to be symmetric (i.e., every arc has its opposite).

Question 3. Is there an FPT algorithm for shortest paths if c is λ -nearly conservative, using some parameter k of "inconvenient components" (should be defined accordingly) and also λ ?

Acknowledgment

The author is thankful to András Frank who asked a special case of this problem, and also to Dániel Marx who proposed the generalization to nearly conservative digraphs.

References

- E.M. ARKIN, C.H. PAPADIMITRIOU On negative cycles in mixed graphs, Operations Research Letters 4 (3) (1985), pp. 113–116.
- [2] M.A. BABENKO, A.V. KARZANOV Minimum mean cycle problem in bidirected and skewsymmetric graphs, *Discrete Optimization* 6 (2009), pp. 92–97.
- [3] T.H. CORMEN, C.E. LEISERSON, R.L. RIVEST, C. STEIN Introduction to Algorithms, MIT Press, Cambridge third edition, (2009)
- [4] A.V. GOLDBERG, A.V. KARZANOV Path problems in skew-symmetric graphs, Combinatorica 16 (3) (1996), pp. 353–382.
- [5] R.M. KARP A characterization of the minimum cycle mean in a digraph, *Discrete Mathematics* 23 (1978), pp. 309–311.
- [6] Z. KIRÁLY Shortest paths in mixed graphs, Egres Technical Report TR-2012-20, www.cs.elte.hu/egres/
- [7] J. MORÁVEK A note upon minimal path problem, Journal of Mathematical Analysis and Applications 30 (1970), pp. 702–717.