## Discrete Optimization

# Solving the shortest path tour problem 

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#### Abstract

In this paper, we study the shortest path tour problem in which a shortest path from a given origin node to a given destination node must be found in a directed graph with non-negative arc lengths. Such path needs to cross a sequence of node subsets that are given in a fixed order. The subsets are disjoint and may be different-sized. A polynomial-time reduction of the problem to a classical shortest path problem over a modified digraph is described and two solution methods based on the above reduction and dynamic programming, respectively, are proposed and compared with the state-of-the-art solving procedure. The proposed methods are tested on existing datasets for this problem and on a large class of new benchmark instances. The computational experience shows that both the proposed methods exhibit a consistent improved performance in terms of computational time with respect to the existing solution method.


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## 1. Introduction

The shortest path tour problem $(\mathcal{S P T P})$ is a variant of the shortest path problem $(\mathcal{S P P})$ and appeared for the first time in the scientific literature in Bertsekas's dynamic programming and optimal control book [2].

The $\mathcal{S P I P}$ has been recently studied by Festa in [13]. The paper of Festa is the first systematic contribution for solving the $\mathcal{S P I P}$. The author proved that the problem belongs to the complexity class $\mathbf{P}$. The polynomial Karp-reduction of the $\mathcal{S P T P}$ to the sin-gle-source single-destination $\mathcal{S P P}$ involves the construction of an expanded graph in which different algorithms for the $\mathcal{S P P}$ were tested and compared on pseudo-randomly generated instances. The results presented in [13] showed that Dijkstra's algorithm outperforms all the competitor algorithms.

Applications of the $\mathcal{S P T P}$ arise for example in the context of the manufacture workpieces, where a robot has to perform at least one operation selected from a set of $S$ types of operations. In such case, the problem may be modeled as a $\mathcal{S P} T \mathcal{P}$ in which operations are associated with nodes of a directed graph and the time needed for a tool change is represented by the distance between two nodes (see [13]).

The main scientific contribution of this paper consists in analyzing some basic theoretical properties of the $\mathcal{S P T P}$, in designing a dynamic programming-based algorithm $(\mathcal{D P} \mathcal{A})$ for solving it, and showing how an ad hoc algorithm for acyclic graphs may be used to solve the $\mathcal{S P I P}$ after efficiently reducing it to a classical $\mathcal{S P P}$

[^0]through the method referred to as modified graph algorithm $(\mathcal{M G \mathcal { G }})$.

The remainder of the paper is organized as follows. The problem is formally described in Section 2. The state-of-the-art algorithm to address the $\mathcal{S P I P}$ is presented in Section 3. Some properties concerning the reducibility of the problem to a classical $\mathcal{S P P}$ and the relevant consequences in terms of solvability are described in Section 4. A dynamic programming algorithm is illustrated in Section 5. Computational results and the analysis of the performance of the proposed algorithms are presented in Section 6. The paper ends with some concluding remarks stated in Section 7.

## 2. Problem description

Consider a directed graph $G=(N, A)$ defined by a set of nodes $N:=\{1, \ldots, n\}$ and a set of arcs $A:=\{(i, j) \in N \times N: i \neq j\}$, where $|A|=m$. A non-negative length $c_{i j}$ is assigned to each $\operatorname{arc}(i, j) \in A$. Let $F(i):=\{j \in N:(i, j) \in A\}$ and $B(i):=\{j \in N:(j, i) \in A\}$ be the forward star and backward star associated with each node $i \in N$, respectively. Moreover, let $S$ denote a certain number of node subsets $T_{1}, \ldots, T_{S}$ such that $T_{h} \cap T_{k}=\emptyset, h, k=1, \ldots, S, h \neq k$.

Given two nodes $i_{1}, i_{\pi} \in N, i_{1} \neq i_{\pi}$, the path $P_{i_{1}, i_{\pi}}$ from $i_{1}$ to $i_{\pi}$ is defined as a sequence of nodes $P_{i_{1}, i_{\pi}}=\left\{i_{1}, \ldots, i_{\pi}\right\}$ such that $\left(i_{j}, i_{j+1}\right)$ $\in A, j=1, \ldots, \pi-1$. Observe that $i_{j}, j=1, \ldots, \pi$, represents the node index occurring in position $j$ in path $P_{i_{1}, i_{\pi}}$. A path $P_{i_{1}, i_{\pi}}$ is said to be elementary whether $i_{l} \neq i_{j}, l, j=1, \ldots, \pi$ and $l \neq j$. We refer to the length of path $P_{i_{1}, i_{\pi}}$ as $l\left(P_{i_{1}, i_{\pi}}\right)$ representing the sum of the lengths of the arcs connecting consecutive nodes in $P_{i_{1}, i_{n}}$, i.e., $l\left(P_{i_{1}, i_{\pi}}\right)=\sum_{j=1}^{\pi-1} c_{j, j+1}$.

The $\mathcal{S P T P}$ aims at finding a shortest path $\mathcal{P}_{s, d}$ from origin node $s \in V$ to destination node $d \in V$ in the directed graph $G$, such that it
visits successively and sequentially the following subsets $T_{k}$, $k=0, \ldots, S+1$, such that $T_{0}=\{s\}$ and $T_{S+1}=\{d\}$. Note that sets $T_{k}$, $k=1, \ldots, S$, must be visited in exactly the same order in which they are defined.

Consequently, a path $P_{i_{1}, i_{\pi}}$ is said to be a feasible solution for the $\mathcal{S P T P}$ if:
$\exists g_{0}, g_{1} \ldots g_{S+1} \in[1, \pi]: g_{0}<g_{1}<\cdots<g_{S+1}$,
$i_{g_{0}} \in P_{i_{1}, i_{\pi}} \cap T_{0}, i_{g_{1}} \in P_{i_{1}, i_{\pi}} \cap T_{1}, \ldots, i_{g_{S+1}} \in P_{i_{1}, i_{\pi}} \cap T_{S+1}$.
Conditions (1) mean that an increasing sequence of natural numbers exists such that the corresponding nodes of the path $P_{i_{1}, i_{\pi}}$ belong to the ordered sequence of subsets $T_{0}, T_{1}, \ldots, T_{S+1}$. A small instance of the $\mathcal{S P T P}$ is depicted in Fig. 1, where $N=\{s=1$, $2,3,4,5,6, d=7\}, \quad S=2, T_{0}=\{s=1\}, \quad T_{1}=\{3\}, \quad T_{2}=\{2,4\}$, $T_{3}=\{d=7\}$. The shortest path from node 1 to node 7 is $\mathcal{P}_{1,7}=\{1,3,7\}$ with length 5 , while the shortest path tour between the same origin and destination nodes is $P_{1,7}=\{1,3,2,3,7\}$ with length 11. Such path is not elementary, since it passes twice through node 3.

## 3. The state-of-the-art

The state-of-the-art consists of the expanded graph method proposed by Festa in [13], and referred to as $\mathcal{E G \mathcal { A }}$ in the sequel. A brief description of how the $\mathcal{E G} \mathcal{A}$ works is given in the following.

The $\mathcal{E G A}$ relies on a polynomial-time reduction algorithm that transforms any $\mathcal{S P T P}$ instance defined on a single-stage graph $G$ into a single-source single-destination $\mathcal{S P P}$ instance defined on a multi-stage graph $G^{\prime}=\left(V^{\prime}, A^{\prime}\right)$ with $S+2$ stages, each replicating $G$, and such that $V^{\prime}=\{1, \ldots,(S+2) n\}$ and $\left|A^{\prime}\right|=(S+1) m$. More precisely, the reduction algorithm performs the following operations:
i. $V^{\prime}:=\{1, \ldots,(S+2) n\} ; A^{\prime}:=\emptyset$;
ii. at each iteration, an $\operatorname{arc}(a, b)$ is added to $A^{\prime}$. In particular, for each stage $k \in\{0, \ldots, S\}$, for each node $v \in\{1, \ldots, n\}$, and for each adjacent node $w \in F S(v),(a, b)=(v+k n, w+(k+1) n)$ with length $c_{v w}$, if $w \in T_{k+1} ;(a, b)=(v+k n, w+k n)$ with length $c_{v w}$, otherwise.

Since $\left|A^{\prime}\right|=(S+1) m$, the computational complexity of the reduction algorithm is $O(S m) .{ }^{1}$ Once the multi-stage graph $G^{\prime}$ is obtained, to solve the resulting $\mathcal{S P P}$ any shortest path algorithm can be applied. By applying Dijkstra's algorithm that uses a binary heap for storing temporary node labels, the overall worst case computational complexity of $\mathcal{E G \mathcal { A }}$ is $O\left(\left|A^{\prime}\right| \log \left|V^{\prime}\right|+\left|V^{\prime}\right| \log \left|V^{\prime}\right|\right)$, which is dominated by $O\left(\left|A^{\prime}\right| \log \left|V^{\prime}\right|\right)$, that is $O(S m \log n)$.

## 4. A modified graph method

In this section, we will focus on some basic properties related to the reducibility of any $\mathcal{S P T P}$ instance into a single-source singledestination $\mathcal{S P P}$ instance.

### 4.1. On the reduction of the $\mathcal{S P T P}$ to the $\mathcal{S P P}$

Given an instance of the $\mathcal{S P T P}$ on a directed graph $G=(N, A)$ the following definition is applied.

[^1]Definition 1. Let $G^{(a)}=\left(N^{(a)}, A^{(a)}, c^{(a)}\right)$ be a weighted directed graph obtained from $G$ in such a way that:

- $N^{(a)}=\bigcup_{k=0}^{S+1} T_{k} ;$
- $A^{(a)}=\bigcup_{k=0}^{S} A_{k}^{(a)}$,
where

$$
A_{k}^{(a)}:=\left\{(i, j) \in T_{k} \times T_{k+1}: i \in T_{k} \text { and } j \in T_{k+1}\right\} ;
$$

- $c^{(a)}: A^{(a)} \mapsto \mathbb{Z}^{+}$is a function that associates an integer non-negative number $c_{i j}^{(a)}$ to each $\operatorname{arc}(i, j) \in A^{(a)}$, where $c_{i j}^{(a)}:=l\left(P_{i, j}\right)$ is the length of a shortest path from node $i \in T_{k}$ to node $j \in T_{k+1}$ on graph $G$.

The following property holds for the arc set $A^{(a)}$.
Property 1. Let $G^{(a)}$ be the weighted directed graph associated with an instance of the $\mathcal{S P T P}$, then $\left|A^{(a)}\right| \leqslant n(n-2)$.

Solving a $\mathcal{S P T P}$ instance may be performed by finding a shortest path in $G^{(a)}$. Indeed, the following property holds.

Property 2. Every path $P_{s, d}^{(a)}$ from s to $d$ in $G^{(a)}$ defines a SPIP solution in $G$ with the same cost, and vice versa.

Such a property derives from the construction of graph $G^{(a)}$ given in Definition 1.

The optimal cost of any $\mathcal{S P} T \mathcal{P}$ instance is equal to the cost of the shortest path from node $s$ to node $d$ computed on $G^{(a)}$, as shown in the following property.

Property 3. The cost of a shortest path $\mathcal{P}_{s, d}^{(a)}$ in $G^{(a)}$ is equal to the cost of an optimal $\mathcal{S P T P}$ in $G$.

Proof. Such property can be proved by contradiction. From Property 2, it follows that the shortest path $\mathcal{P}_{s, d}^{(a)}$ in $G^{(a)}$ corresponds to a feasible path tour $P_{s, d}^{*}$ in G. Hence, suppose that a shortest path tour $P_{s, d}$ different from $P_{s, d}^{*}$ exists in $G$, such that $l\left(P_{s, d}\right)<l\left(P_{s, d}^{*}\right)$. This means that a feasible path $P_{s, d}^{(a)^{\prime}}$ exists in $G^{(a)}$, associated with $P_{s, d}$, whose cost is less than the cost of the shortest path $P_{s, d}^{(a)}$. This conclusion contradicts the initial assumption.

### 4.2. Solving the $\mathcal{S P T P}$ as $\mathcal{S P P}$

The framework of the proposed $\mathcal{M G \mathcal { A }}$ is sketched in Algorithm 1:

## Algorithm 1. Modified graph algorithm

```
Step 1 (Initialization)
    Compute a shortest path from each node \(i \in T_{k}\) to each node
        \(j \in T_{k+1}, k=0, \ldots, S\).
    Step 2 (Graph Construction)
    Build the digraph \(G^{(a)}=\left(N^{(a)}, A^{(a)}\right)\) as detailed in Definition 1.
    Step 3 (Topological enumeration of \(N^{(a)}\) )
    Step 4 (Shortest Path Computation)
    Let \(l\left(\mathcal{P}_{1, i}\right)\) denote the length of the shortest path from node
    \(s=1\) to node \(d=i\), and \(p(i)\) denote
    the predecessor node of \(i\) in \(\mathcal{P}_{1, i}\).
    Set \(l\left(\mathcal{P}_{1,1}\right):=0, p(1):=1\) and iteration \(_{\mathcal{M G \mathcal { A }}}:=0\).
    for all \(j=2, \ldots,\left|N^{(a)}\right|\) do
        \(l\left(\mathcal{P}_{1, j}\right)=\min _{i \in N^{(a)}:(i, j) \in A^{(a)}}\left\{l\left(\mathcal{P}_{1, i}\right)+c_{i j}\right\}\). Update \(p(j)\) with the
        value of \(i\) whereby the minimum of \(l\left(\mathcal{P}_{1, j}\right)\) occurs.
        Update the number of iterations as
        iteration \(_{\mathcal{M G \mathcal { A }}}=\) iteration \(_{\mathcal{M G \mathcal { A }}}+1\).
    end for
```


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[^1]:    ${ }^{1}$ For bounding the computational complexity of an algorithm, several asymptotic notations are used [6]. If $n$ is the input size, the computational complexity of an algorithm $f(n)$ is $O(g(n))$, if there exist positive constants $a$ and $n_{0}$ such that $0 \leqslant f(n) \leqslant a g(n)$ for all $n \geqslant n_{0}$.

