

# THE TRAVELING SALESMAN PROBLEM ON A CHAINED DIGRAPH

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**The traveling salesman problem (TSP) on a chained digraph is studied. An algorithm based on the ideas of dynamic programming is presented. A previously known reduction to a kind of the shortest path problem is extended to a special multi-TSP.**

## 1. INTRODUCTION

Suppose that a salesman, starting from his home city, is to visit exactly once each city on a given list of cities and then to return home. It is reasonable for him to select the order in which he visits the cities so that the total of the distances travelled in his tour is as small as possible. This problem is called the *traveling salesman problem* (TSP) (cf., e.g. monograph [13] and a recent expository article [12]).

Finding the traveling salesman's shortest route to pass  $n$  cities in such a way that each city is visited exactly once represents the traditional formulation of TSP. It is assumed that non-negative distances  $c(i, j) = c_{ij}$  between the cities  $i, j$  ( $1 \leq i < j \leq n$ ) are given and also that traveling salesman starts his trip from an arbitrary city. If the traveling salesman does not return to the starting city then the minimal traversed route is called an *open route* or a *path*.

We mention some graph theoretical interpretations. For standard graph theoretical notions see, e.g., [11], [3].

Let  $G$  be a *weighted* digraph on  $n$  vertices. To each arc  $(i, j)$  of  $G$  a *length* (*weight* or *cost*)  $c_{ij}$  is assigned. In particular, in the TSP  $c_{ij}$  is the distance between the cities  $i$  and  $j$ . We define  $c_{ij} = +\infty$  if the arc  $(i, j)$  does not exist in  $G$ . The

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matrix  $C = \| c_{ij} \|_1^n$  is called the *distance (weight or cost) matrix* of  $G$ . The *length (weight or cost)  $l(H)$*  of a subgraph  $H$  of  $G$  is defined to be the sum of lengths of arcs of  $H$ . In particular, the length of a path is the sum of lengths of arcs from which it consists. A path (cycle) consisting from  $k$  arcs is called a  *$k$ -path ( $k$ -cycle)*.

A cycle (path) passing through each vertex of  $G$  exactly once is called a *Hamiltonian cycle (path)*. Of course, a Hamiltonian cycle is an  $n$ -cycle and a Hamiltonian path is an  $(n - 1)$ -path. A salesman's tour (or *route*) is now a Hamiltonian cycle, or path. TSP can be reformulated in the following way:

*In a weighted digraph (graph) find a Hamiltonian path (cycle) of a minimal length.*

We recall general ideas for finding the best solution of the TSP and many other problems of combinatorial optimization. The first general algorithmic approach is *branch and bound*. The second one is the *polyhedral* approach. In the recent years, the polyhedral approach combined with branch and bound has proved to be an indispensable tool in computational solution of many hard large-scale real-world combinatorial problems.

Solutions to the TSP, which are not optimal but are "good" in some sense are called suboptimal solutions. "Good" suboptimal solutions are found by "clever" heuristics. There are many heuristics for the TSP described in the literature. One can also formulate the problem of finding the  $k$ -th best ( $k$ -best) solution(s) of a combinatorial optimization problem (see, e.g., [8]).

## 2. A CHAINED TSP

The TSP on a digraph with a chained structure is studied in this paper. Section 2.1 contains some basic notions while Section 2.2 provides an algorithm for solving TSP which is based on the dynamic programming (the DP-algorithm). In Section 2.3 the problem is reduced to the problem of finding a shortest path in an auxiliary weighted digraph (the SP-algorithm). A part of these results have been announced in [7].

### 2.1. Preliminaries

Let  $G$  be a digraph whose adjacency matrix has the form

$$A = \left\| \begin{array}{cccccc} A_1 & B_1 & O & \dots & O & \\ O & A_2 & B_2 & & O & \\ \vdots & & & & & \\ O & O & O & & B_{m-1} & \\ O & O & O & & A_m & \end{array} \right\|,$$

where  $A_1, A_2, \dots, A_m$  are square blocks of order  $n$ . Subgraphs of  $G$  with adjacency matrices  $A_1, A_2, \dots, A_m$  are denoted by  $G_1, G_2, \dots, G_m$  respectively. Digraph  $G$  is



vertices of  $X_m$ , where the length of a path is the sum of weights of all vertices and arcs from the path. Some efficient procedures for finding such a path and  $k$ -best solutions are given in [14], [15] as well (see Section 2.3).

## 2.2. A dynamic programming algorithm

Following [6] we describe an algorithm for solving TSP on a digraph  $G$  having a chained structure. The algorithm is based on the ideas of dynamic programming and therefore we call it a DP-*algorithm*.

For a vertex  $x$  in  $G_i$  ( $i = 1, 2, \dots, m$ ) let  $d(x)$  be the length of a shortest path starting from  $x$  and passing through all vertices of  $G_i, G_{i+1}, \dots, G_m$ . Such a path contains a Hamiltonian path in each of digraphs  $G_i, G_{i+1}, \dots, G_m$ . Let  $h(x)$  be the length of a shortest Hamiltonian path in  $G_i$  which starts at the vertex  $x$ . Finally, let  $s(x)$  be the length of a shortest path starting at the  $x$  and passing through all the vertices of the digraphs  $G_{i+1}, \dots, G_m$ .

We determine the quantity  $d(x)$  first for vertices  $x$  from  $G_m$ , then for those from  $G_{m-1}$  and so on, back to the digraph  $G_1$ . The minimal value of  $d(x)$  for vertices  $x$  from  $G_1$  yields the solution of the TSP on  $G$ . Minimizing  $d(x)$  also provides an optimal path as will be seen soon.

For vertices  $x$  from  $G_m$  we have  $d(x) = h(x)$ . The quantity  $h(x)$  and the corresponding Hamiltonian path in  $G_m$  can be determined by any standard algorithm for the TSP.

Suppose we have determined  $d(x)$  for vertices  $x$  from  $G_{i+1}$  and we proceed to determining  $d(x)$  for vertices from  $G_i$ .

First we determine  $s(x)$  for vertices  $x$  from  $G_i$ . Let  $c(x, y)$  be the length of the arc joining vertex  $x$  with vertex  $y$ . Obviously, we have

$$s(x) = \min_{y \in G_{i+1}} (c(x, y) + d(y)).$$

Further, we form an auxiliary digraph  $H_i$  by extending digraph  $G_i$  with a new vertex  $z_i$ . From each of vertices  $x$  of  $G_i$  an arc goes toward  $z_i$ . The length of this arc is defined to be  $c(x, z_i) = s(x)$ . It is obvious now that  $d(x)$  is equal to the length of a shortest Hamiltonian path in  $H_i$  starting at  $x$  and terminating in  $z_i$ .

## 2.3. Reduction to a kind of the shortest path problem

Following the paper [14] by M. MILOSAVLJEVIĆ and M. OBRADOVIĆ, we describe how the TSP on a chained digraph can be reduced to the problem of finding a shortest path in an auxiliary weighted digraph. The corresponding algorithm is called an SP-algorithm.

The length  $l(H_G)$  of any (global) Hamiltonian path  $H_G$  in  $G$  is the sum of lengths  $l(H_{G_i})$  of (partial) Hamiltonian paths  $H_{G_i}$  in  $G_i$ ,  $i = 1, 2, \dots, m$ , plus the sum of weights of the interconnecting arcs. Accordingly, the criterion function for the minimal (global) Hamiltonian path can be expressed in the form

$$(1) \quad J(H_G) = \min l(H_G) = \min \left\{ \sum_{i=1}^m l(H_{G_i}) + \sum_{i=1}^{m-1} e_{pq}^{(i)} \right\},$$

where  $e_{pq}^{(i)}$  is an entry of  $E_i$ , with  $p$  the ending vertex of  $H_G$ , and  $q$  the starting vertex of  $H_{G_{i+1}}$ .

As explained in Section 2.1, minimization of (1) leads to a convenient layered structure representation (Fig. 1), where this task is accomplished by selecting a minimal length path through the digraph  $F$  starting from  $X_1$  and ending at  $X_m$ .

Let us introduce notation (see Fig. 1) as follows: vertex  $a_{ij}$  corresponds to the  $j$ -th Hamiltonian path in  $G_i$ ,  $i = 1, 2, \dots, m$ ;  $j = 1, \dots, Q$ ,  $Q = n!$ , where the numeration of Hamiltonian paths in each  $X_i$  is arbitrary. The weight  $c(a_{ij})$  of vertex  $a_{ij} \in X_i$  is equal to the length of the  $j$ -th Hamiltonian path in  $G_i$ . The weight of the arc  $(a_{il}, a_{i+1,k})$ , denoted by  $c(a_{il}, a_{i+1,k})$ , is equal to the corresponding entry of matrix  $E_i$  as mentioned earlier. Without loss of generality we suppose  $c(a_{ij}) > 0$ ,  $i = 1, 2, \dots, m$ ,  $j = 1, \dots, Q$ ,  $c(a_{il}, a_{i+1,k}) > 0$ ,  $i = 1, 2, \dots, m-1$ ,  $l, k = 1, \dots, Q$ . If the set of vertices of  $F$  is denoted by  $X$  and if  $U$  is the set of arcs, we shall write  $F = F(X, U)$ .

We consider the minimization of the criterion function

$$J(j_1, j_2, \dots, j_m) = \sum_{i=1}^m \left( c(a_{ij_i}) + c(a_{ij_i}, a_{i+1, j_{i+1}}) \right),$$

(with  $j_1, j_2, \dots, j_m$  a path of length  $m-1$  through digraph  $F$  and  $c(a_{mj_m}, a_{m+1, j_{m+1}}) = 0$ ), which is an equivalent of criterion (1).

In order to construct an efficient algorithm for minimization of criterion (1), we transform the weighted digraph  $F$  into another weighted digraph  $F'$  in the following way. Consider a set of mappings  $j \rightarrow n(i, j)$ ,  $i = 1, \dots, m$ ;  $j = 1, \dots, Q$ , actually  $m$  permutations on  $Q$  elements, under condition that

$$\begin{aligned} c(b_{ij}) &= c(a_{in(i,j)}), \quad i = 1, \dots, m, \quad j = 1, \dots, Q; \\ c(b_{ij}, b_{i+1,l}) &= c(a_{in(i,j)}, a_{i+1, n(i+1,l)}), \quad i = 1, \dots, m, \quad l = 1, \dots, Q; \\ c(b_{ij}) &\leq c(b_{ik}), \quad j < k, \quad i = 1, \dots, m, \quad j, k = 1, \dots, Q; \end{aligned}$$

where  $\{b_{ij}\}$  is the set of vertices of  $F'$ . In other words, we have rearranged each subset in such a way that  $\{c(b_{ij})\}_{j=1}^Q$  forms a non-decreasing sequence, keeping length of all arcs the same as in  $F$ .

Of course, if  $J(j_1^*, j_2^*, \dots, j_m^*) = \min J(j_1, j_2, \dots, j_m)$  on the digraph  $F$ , and  $J^*(j_1^{**}, j_2^{**}, \dots, j_m^{**}) = \min J(j_1, j_2, \dots, j_m)$  on the digraph  $F'$ , it is obvious that  $J^*(j_1^{**}, j_2^{**}, \dots, j_m^{**}) = J(j_1^*, j_2^*, \dots, j_m^*)$ ,  $j_i^{**} = n(i, j_i^*)$ .

Observe that the shortest path could be expected to pass the reordered digraph  $F'$  through its "upper" part. Hence, the "lower" part of  $F'$  is irrelevant for a procedure of finding the shortest path and consequently need not be memorized and processed. The subgraph which is sufficient for processing is given by the following theorem from [14].

**Theorem 1.** Let  $j_1^*, j_2^*, \dots, j_m^*$  be the shortest path in the subgraph  $F'(X_S, U_S)$  of the reordered digraph  $F'(X, U)$  defined by

$$X_S = \{b_{ij} \mid i = 1, \dots, m; j = 1, \dots, j_i^{**}\},$$

where  $U_S$  is the corresponding set of arcs. The indices  $j_i^{**}$ ,  $i = 1, \dots, m$  are obtained as minimal indices  $j$  for which

$$c(b_{ij}) \geq c(b_{ij_i^*}) + c(b_{ij_i^*, b_{i+1, j_{i+1}^*}}) - \min_{l, k \in \{1, \dots, Q\}} \{c(b_{il}, b_{i+1, k})\}.$$

Then, the problem of finding an  $(m-1)$ -path of minimal length in  $F'(X, U)$  has the same solution as the same problem in subgraph  $F'_S(X_S, U_S)$ .

In [14] an algorithm for finding  $k$ -best solution of the TSP on a chained digraph (i.e. of the shortest path problem on  $F$ ) is described.

### 3. A SPECIAL MULTI-TSP

In this section we shall consider the following multi-traveling salesman problem.

*Given a weighted digraph  $G$  on  $n = km$  ( $k, m$  integers) vertices, determine  $m$  disjoint  $(k-1)$ -paths with a minimal total length.*

Note that this problem is quite different from the standard multi-TSP in which the number of cities in tours of salesmen are not prescribed. The standard problem is usually solved by a transformation which reduces it to the ordinary TSP [2]. This transformation cannot be applied successfully to our problem.

A problem similar to our problem is the so called *clover leaf problem* [9], [1]. In the clover leaf problem each salesman starts from a specified city and after visiting a fixed number of cities returns to the starting city. Note that our problem can be reduced to the clover leaf problem by introducing an additional city which is equidistant to all other cities.

Some branch and bound and dynamic programming algorithms for our problem have been described in reports [4], [5], [6]. Note that in [10] the problem has been formulated in terms of integer programming.

Let  $G$  be a chained digraph, consisting of digraphs  $G_1, G_2, \dots, G_k$ . We shall consider  $m$ -TSP on digraph  $G$  in the presence of some additional constraints [6].

Each of digraphs  $G_1, G_2, \dots, G_k$  has  $n$  vertices, where  $n$  is divisible by  $m$ . Feasible collections of  $m$  disjoint  $(\frac{k}{m}-1)$ -paths have the additional property of inducing a collection of  $m$  disjoint  $(n-1)$ -paths on each of digraphs  $G_1, G_2, \dots, G_k$ . A collection of  $m$  disjoint paths of equal lengths will be called an  *$m$ -collection*.

In order to get a feasible  $m$ -collection for  $G$  we start with an  $m$ -collection in  $G_1$ , continue it with an  $m$ -collection in  $G_2$ , and so on up to  $G_k$ . Terminal vertices ( $m$  in number) of paths from an  $m$ -collection in  $G_i$  are joined by arcs with  $m$  starting vertices of paths from an  $m$ -collection in  $G_{i+1}$ . There are  $m!$  ways of joining the two  $m$ -collections by  $m$  arcs. In order to get an optimal join we have to solve an assignment problem of dimension  $m$ .

Now we can generalize the SP-algorithm described in Section 2.3, to the multi-TSP. We form the auxiliary digraph  $F$  in such a way that vertices of  $F$  correspond now to  $m$ -collections (instead of Hamiltonian paths) from digraphs  $G_i$ . The weight of a vertex in  $F$  is equal to the length of the corresponding  $m$ -collection from a  $G_i$ .

The weight of an arc in  $F$  is the sum of the weights of the arcs joining  $m$ -collections corresponding to the starting and the terminal vertex of the arc, i.e. to the length of an optimal assignment solved when joining two  $m$ -collections by arcs. The algorithms of M. MILOSAVLJEVIĆ and M. OBRADOVIĆ can be now applied, without difficulties, for finding both an optimal solution and best suboptimal solutions.

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