

On Symmetric Traveling Salesman Polytopes and

Separation

T. S. Arthanari
Department of Engineering Science
University of Auckland
New Zealand
t.arthanari@auckland.ac.nz

Abstract

The Symmetric Traveling Salesman Problem is posed in Arthanari [1], as a Multistage problem. The 0-1 program in question is the same as given in the previous paper, but the variables are those that arise in the relaxation of the variables in the previous paper. We consider the relaxation of the variables in the previous paper, and show that the relaxation is a linear programming problem. It is shown that a sequence of variables in the previous paper can be verified as a corresponding feasible relaxation. The relaxation is a linear programming problem, and the number of constraints is finite. Any violated inequality can be found otherwise.

1 Introduction

The Symmetric Traveling Salesman Problem (TSP) is a minimum cost tour that starts at a home city and visits every city in a given set of cities and returns to the home city. Being of the type of a combinatorial optimization problem (NP) of the 'hard kind', it is extensively studied (Jungert, [2]). Dantzig, [3], formulated the asymmetric Traveling Salesman Problem (ATSP) as a linear program on a graph (V, E) . The formulation of the symmetric Traveling Salesman Problem (STSP) as a multistage problem and given in [1] program in question is the same, but in the previous paper variables are those that arise in the relaxation of the variables in the previous paper. The relaxation of the variables in the previous paper is the same as given in the previous paper. The relaxation is a linear programming problem, and the number of constraints is finite. Any violated inequality can be found otherwise.

relaxation by shrinking E_n . He also uses this relaxation to separate in polynomial time over classes STSP inequalities known that the cycle in relaxation to a maximum length cycle.

In this paper we address several other problems in graph theory and in particular we address the problem of finding a maximum length cycle through a network, minimum length cycle through a network, and minimum length cycle through a network. In section 2 we give a formal definition of the problem and in section 3 we give a formal definition of the problem and in section 4 we give a formal definition of the problem. In section 5, we examine the consequences of the main finding and indicate directions for further research regarding complexity of this suggested approach.

2 Notations & Preliminaries

Let n be an integer, $3 \leq n$. Let $V_n = \{1, \dots, n\}$ be the set of vertices. Let $E_n = \{i, j\} \subset V_n, i < j$ be the set of edges W also call them pairs. Let $K_n = (V_n, E_n)$ denote the complete graph on n vertices. W denote the elements of E_n by where $= (i, j)$. W also set notation for (i, j) : For a subset $E \subseteq E_n$ we write the characteristic vector by $x_E \in \{0, 1\}^n$ where

$$x_E(e) = \begin{cases} 1 & \text{if } e \in E \\ 0 & \text{otherwise} \end{cases}$$

At times, we denote the characteristic vector by x_E . If $E \subseteq E_n, r \in \mathbb{R}$, then $r \cdot x_E$ denotes the characteristic vector of E by augmenting E with r edges. E_n be a subgraph of K_n , that is $V \subseteq V_n$ and $E \subseteq E_n$. Let $G = (V, E)$ be a graph, $V \subseteq V_n$ and $E \subseteq E_n$. For a graph $G = (V, E)$, form subsets $V \subseteq V_n$ we write

$$E(S) = \{i, j\} \subset E, i, j \in S$$

Let F be a subset of vertices of V , $F \subseteq V$, denote the set of edges with one ending in F and the other in $V \setminus F$. Let $d_F(v) = \deg(v) - |E(F, v)|$ be the degree of v in F . A subset F of V is called a Hamiltonian cycle, in case it is the edge set of a simple cycle, of length $|V|$, or $G = (V, H)$ is a subgraph with its connected and vertex degree $|V|$. We also call a Hamiltonian cycle \mathcal{H} . \mathcal{H} is a cycle of length $|V|$. Given $n \in \mathbb{N}$, E_n we define, $\mathcal{H}(E_n) = \{e \in E_n : d_{E_n}(e) = 2\}$.

Let $\mathcal{H}(E_n)$ denote the symmetric difference of all Hamiltonian cycles of length n . Let $\mathcal{H}(E_n) = \text{conv}(\mathcal{H}(E_n))$ where $\mathcal{H}(E_n)$ denotes the set of all Hamiltonian cycles of length n .

2.1 MI Formulation

Let $p_k = k(k-1)/2$, for $k \geq 2$. Let $\alpha_n = \sum_{k=4}^n p_k$.

Let $x_k(e)$ be defined $2 \leq k \leq n$ and $8k \geq 2 \leq V_n$.

Let x denote $(x_{12}, \dots, x_{n-1, n})^T \in \mathbb{R}^{\alpha_n}$.

Here $x_k = (x_{12}, \dots, x_{k-1, k})^T \in \mathbb{R}^{2 \leq k \leq n}$.

Let $C_{1j} = c_{1k} + c_{jk}$! C_{1j} where $1 \leq k < j \leq n$, the cost vector,

using C_{1j} denote the vector C_{1j} for $1 < j < k \leq n$.

Let $2 \leq R \subseteq \mathbb{R}^n$ be defined for $e \in 2 \leq R_n$.

We have the following formulation [?], which will be called the multistage

-insertion formulation (MI formulation) (short):

$$(1) \quad \text{minimize } CX$$

subject to

$$(2) \quad x_k \in \mathbb{R}^{k-1}, \quad 1 \leq k \leq n, \quad V_3$$

$$(3) \quad x_k(e) + u_e = 1, \quad 8e \in 2 \leq E_3$$

$$(4) \quad x_j(-1) \setminus E_{j-1} + \sum_{k=j+1}^n x_k(e) + u(e) = 0, \quad 8e = (1, j) \in 2 \leq E_{n-1} \cap E_3$$

$$(5) \quad x_n(-1) \setminus E_{n-1} + u(e) = 0, \quad 8e = (1, n) \in 2 \leq (V_n \setminus 1)$$

$$(6) \quad X \in \mathbb{R}^{2 \leq B(n)} \quad \& \quad u \in \mathbb{R}^+_{2 \leq n}$$

Remark: 1

MI formulation model is the following multistage insertion at the start

we have the unique tour given by $f(1, 2), \dots, (2, 3), \dots, (k-1),$ tour resulting

from the earlier decisions. We have an $n!$ tour at the stage the cost

of these decisions are equal by the objective function which gives the total

insertion at the start. Each insertion is exact one

pair.

(?) ensures that each of the pairs $(2, 3)$ and $(2, 3)$ are used for insertion

by utmost one:

(?) ensures that each of the pairs $(4, 2)$ to $(n-1, 2)$ is used for

(?) define a tour of the pairs $(n-1, n)$ to $(1, n)$, resulting from all

the insertion pairs through 3:

relaxing the insertion constraint (MI?) we get the MI Polyn (\mathbb{R}^n) given

by the above constraints and problem.

It has been shown that this formulation is equivalent to the FJ formulation. The sense that $\mathcal{H}(e) \in E_n$ feasible here relaxations feasible that E_{p_n} and given a feasible FJ relaxation we have an (X, u) feasible $M1$ formulation. Notice that the variables are the same as in the FJ formulation and minimizing z , as can be seen from the definition of CX .

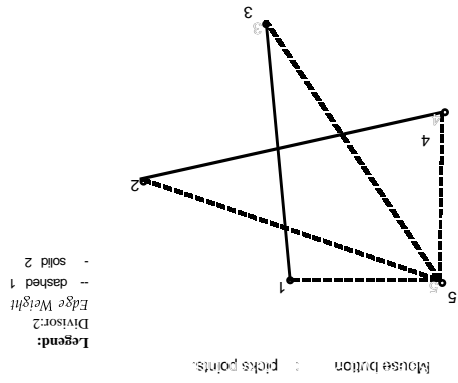
In general let $A^{[n]}$ denote the matrix corresponding to $(?)$. Let $A^{(n)}$ denote the submatrix of $A^{[n]}$ corresponding to x_n . Let r denote the vector r_0 . Then we can write equivalently

$$E^{[n]} = \begin{pmatrix} 0 & \dots & 0 \\ 1_{p_3} & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 0 & \dots & 0 \\ 0 & \dots & 1_{p_{n-2}} \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 1 & \dots & 1_{p_{n-1}} \end{pmatrix} = \begin{pmatrix} 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 0 & \dots & 1_{p_{n-1}} \end{pmatrix} A^{(n)} = \begin{pmatrix} 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 0 & \dots & 1_{p_{n-1}} \end{pmatrix} A^{[n]}$$

and

$$A^{[n]} = \begin{pmatrix} 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 1_{p_3} & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 0 & \dots & 0 \\ 0 & \dots & 1_{p_{n-2}} \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 1 & \dots & 1_{p_{n-1}} \end{pmatrix} = \begin{pmatrix} 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ \vdots & \vdots & \vdots \\ 0 & \dots & 0 \\ 0 & \dots & 0 \\ 0 & \dots & 1_{p_{n-1}} \end{pmatrix} A^{[n]}$$

where M_i is the p_i node-arc incidence matrix.



Example: 2.1 Consider $n = 5$, then $p_1 = 1$; and $p_n = 9$. Let X be given by,

$$X_0 = \begin{pmatrix} 2 & 0 & 1 & 2 & 1 \\ 2 & 0 & 0 & 0 & 1 \\ 2 & 0 & 0 & 1 & 2 \\ 2 & 0 & 0 & 1 & 2 \end{pmatrix}$$

where X_0 denotes the matrix

X satisfies the equality constraint of the $M1$ relaxation also the corresponding obtained from these with the equation $(?)$ is, 0: We have,

$$u_0 = \begin{pmatrix} 0 \\ \vdots \\ \vdots \\ \vdots \\ 0 \end{pmatrix} \quad A^{[5]} X = \begin{pmatrix} 0 \\ 1 \\ 2 \\ 1 \\ 0 \\ 1 \\ 2 \\ 1 \\ 2 \end{pmatrix}$$

3 Main Results

Let k denote the set of all Hamiltonian eigenvalues $(V^k; E^k)$, for $k \in \mathbb{N}$. We call them, as mentioned earlier, k -tours: Every irreducible M I formulation is a Hamiltonian eigenvalue and vice versa. Recall that \mathbb{H}^n denotes the set of $(X; u)$ feasible for the relaxation (X=k) be the restricted \mathbb{H}^n upto the k -th stage that is $(X; u) = (x^k; \dots; x^k)$ for $k \in \mathbb{N}$. For a given $(X; u)$ feasible for the relaxation (X=k) denote the consequence of the decision made in $(X; u)$ as $(X; u)$ relaxed from them I relaxation $\mathbb{H}^n = k$ by substituting in the equation and solving for the variables x^k in \mathbb{H}^k , for $k \in \mathbb{N}$. We denote by $D(T)$, the corresponding \mathbb{H}^k such that $x^k(e) = 1g$, where X is the irreducible solution corresponding to $D(T)$ is called because of that: This is $(X; T) \in \mathbb{H}^k$.

Given $(X; u) \in \mathbb{H}^n$, $k \in \mathbb{N}$, and $T \in \mathbb{H}^k$; we define

$$m^k(X; T) = \min_{k \in \mathbb{N}} \{k \cdot f(x^k(e)) : e \in D(T)\} \quad (1)$$

Let $m^k(X; S) = \min_{k \in \mathbb{N}} \{k \cdot f(x^k(e)) : e \in S\}$ for some $T \in \mathbb{H}^k$; S denotes these two pairs over S :

Remark 2

To solve the problem $(X; T)$ over the set of \mathbb{H}^k we attach the weight k to each $(X; T) \in \mathbb{H}^k$ by $(X; T) \in \mathbb{H}^k$ is a convex combination of some k -tours and T is one among them, then $(X; T)$ is positive, otherwise it is 0.

Let X correspond to an n -tour then X is an integer vector and so $(X; k) \in \mathbb{H}^k$ for all $k \in \mathbb{N}$. So it is that k -tour then $(X; T) = 1$: Also, in this case, if $T \in \mathbb{H}^k$, then $m^k(X; S) = 1$ as well. This is so because, $(X; T) = 0$ if $T \in \mathbb{H}^k$ and $T \in \mathbb{H}^k$.

Lemma 1 Given $(X; u) \in \mathbb{H}^n$, consider X and the corresponding \mathbb{H}^n . We have

$$U(T) = [T_1, T_2, T_3, T_4]$$

where $D(T_1) = e_1 = f_1, 2g; D(T_2) = e_2 = f_1, 3g$, and $D(T_3) = e_3 = f_2, 3g$.

Recall that a standard orthogonality vector T and x_k denote the vector (x_1, x_2, \dots, x_k) .

This lemma shows that for every feasible $(X; u)$, we have, the consequence $U(T)$ is (\mathbb{H}^k) . The next lemma gives the result on the star-optimality.

$n = 5$.

Lemma 2 Given $(X; u) \in \mathbb{H}^5$, if X satisfies $k = 4$,

$$m^k(X; T) = \frac{e^{2T}}{X} \cdot x^{k+1}(T) = \text{for all } T \in \mathbb{H}^k$$

then $2 \leq Q_5$.

This lemma forms the basis for the main theorem. A straightforward proof attempts to exploit the AT problem relation = 5: The inequality $(x; T) \cdot x^{k+1}(T)$; it is established that the inequality $T \cdot H^q$ is not in this sense. It can be seen from the main theorem, we require an inequality that is satisfied by every subset H^k of V^{n-1} . For ensuring \mathcal{Q}^n :

Proof of Lemma ?? : Consider $(x; u) \in F_n$ & x satisfying

$$(10) \quad x^q(D; T) \stackrel{\text{def}}{=} m^q(x; T) \cdot \sum_{x \in T} x_5(e) \cdot 8 \cdot T \cdot 2 \cdot H^q \quad e \in T$$

Look at the AT problem as defined below:

Find f_{min} such that,

$$x \quad f_{\text{min}} = x^q(D; T) = x^q(e) \stackrel{\text{def}}{=} e_i; i \in 2 \cdot 0 = f_{1; 2; 3; g}$$

(f; 1; 2; A)

$$x \quad f_{\text{min}} = x_5(e) \stackrel{\text{def}}{=} d_i; i \in 2 \cdot D = f_{1; 1; \dots; 6; g}$$

where $A = f(i; 1); i \in 2 \cdot 0$ and $F(1) = f(1; 1); (1; 6); F(2) = f(2; 2); (1; 5);$

and $F(3) = f(3; 3); (1; 4); g$: Assuming zero capacity for the arcs F , and using arc $i; j$ for supply and demand at i and j , we have the total available quantity to total demand and supply. This solves the problem easily (see equation (1)).

Consider the network obtained by adding a super source, with an arc capacity c_{ij} connecting i and j and similar arcs $i; j$ added, with an arc from i and j , (with capacity c_{ij}). The maximal flow through this network is ()

the AT problem is feasible.

Claim: The maximal flow through the network is indeed.

Hence we have a feasible flow for the AT problem under consideration.

consider the (1) result in x^q , assigned by lemma ??.

Let M denote the matrix corresponding to the T_n ; $i \in 2 \cdot 0$ as column numbers that order we have $M \cdot x^q = U(1)$: From the feasibility of the AT problem we have $f_{\text{min}} > 0$ and add up to: Restrict the amount of $f_i; i \in 1$ such that $f_{\text{min}} > 0$:

We have,

$$T_n \cdot A \stackrel{H}{(5)} = T_n \cdot f_{\text{min}}$$

(11)

where T_n is the $1 \times n$ row vector obtained from T_n by inserting f_{min} in T_n : And $A \stackrel{H}{(5)}$ is defined as x^q ; refer to the column corresponding to (e) : From feasibility of the AT problem $(x; u) \in F_n$; $n = 5$,

$$M \cdot x^q \cdot A \stackrel{H}{(5)} \cdot x_5 = u$$

(12)

Replacing x_5 and x^q in this expression by the expression f_{min} , we obtain,

$$M \cdot \begin{matrix} @BB \\ p \\ \vdots \\ C \\ A \\ @BB \\ p \\ \vdots \\ C \\ A \end{matrix} \cdot \begin{matrix} f_{3n} \\ f_{1n} \\ 1 \\ 0 \\ p \\ 1 \\ f_{1n} \\ 1 \end{matrix} \cdot \begin{matrix} A \\ \vdots \\ C \\ A \end{matrix} \stackrel{H}{(5)} = u$$

(13)

Theorem 3.1 (Main Theorem) Given $(X; u) \in F_n$, it satisfies

$$m_k(X; S) \cdot x_{k+1}(S) \quad (14)$$

Proof: $8 \leq k \leq n$ and $S \in H_k$, then $2 \leq n$.

Proof:

The proof is by induction on number of cities.

Base for induction: the result is true for $n = 5$, from lemma 2.2.

Assume the result is true for $n-1$. We shall show that the result is true for

n .

Consider $(X; u) \in F_n$.

Look at $(X; u) \in F_{n-1}$, obtained from $(X; u) \in F_n$ by

Now $(X; u) \in F_{n-1}$ from feasibility of $(X; u) \in F_n$ by

induction hypothesis.

9.1.1. Let $(X; u) \in F_n$ such that $u_{n-1} = 1$ with $u_n > 0$,

$u_n = 1$ and $u_{n-1} = 1$.

Now consider the problem with:

0 ! ! sources : $i = 1, \dots, p$; where $p = j, j$
 $D ! !$ sinks : $i = 1, \dots, q$; where $q = j, j$
 $a_i ! !$ supply : $a_i = u_i$ at i ;
 $b_j ! !$ demand : $b_j = x_{n-1}(u_j)$ at j ;
 ≥ 1 if $i = j$ is not forbidden
 $c_{ij} = 1$ capacity: $c_{ij} = 1$ otherwise.

Find $f_{ij} \geq 0, i, j \in D$, feasible

From requirements (?)

9.1.2. Let $(X; u) \in F_n$ such that $u_{n-1} = 1$ and $u_n > 0$, where

such that $(L \setminus O; D \setminus W)$ contains a $(f; u)$ not forbidden $(L \setminus O; D \setminus W)$, 1:
 But $b(D) = 1, b(D \setminus W) \leq a(O \setminus L) \cdot 1$: Therefore, requirements (?) for
 such L are trivially satisfied.

So, let us assume that $(L \setminus O; W \setminus D)$, then consists only of forbidden arcs.

Therefore $(L \setminus O; W \setminus D) = 0$: We require $(D \setminus L), a(O \setminus L)$: Now $b(D \setminus O) = 1$ and $c(L; W) = 1$: But here $c(L; W) = c(L \setminus O; D \setminus W)$. It is
 such that $(L \setminus O; D \setminus W)$ contains a $(f; u)$ not forbidden $(L \setminus O; D \setminus W)$, 1:
 But $b(D) = 1, b(D \setminus W) \leq a(O \setminus L) \cdot 1$: Therefore, requirements (?) for
 such L are trivially satisfied.

Thus, $X \cdot X \cdot m_{n-1}(X; T) \cdot x_n(u) = x_n(u)$
 Hence requirements (?) are met. Thus we have a feasible flow for the prob-
 lem. This means it is nonnegative vector adding to one as maximum flow in
 the network = 1: The result is proved, as will do the other lemma 2.2.

Hence the theorem.

Notice that it is not difficult to see that the AT problem is defined for k

is satisfied for $n \geq 2$.

4 Conclusions

This paper identifies a new class of linear inequalities for the separation of a set S from a convex set C . It is shown that the separation problem can be reformulated as a linear programming problem. The new inequalities are shown to be stronger than the existing ones. The paper also discusses the computational complexity of the new inequalities.

Acknowledgments

Prof. David Ryan, Mr. Jim Green and Ms. Sarah Ewald are thanked for their assistance in providing the resources and secretarial assistance for the project. The author also acknowledges the support of the National Science Foundation.

References

[1] C. Aggarwal, R. K. Ahuja, J. B. Orlin, and J. M. Mitchell. Linear-time algorithms for the h -median problem. *SIAM J. Comput.*, 28(3):638–652, 1999.

[2] J. S. Arthanarajah. The traveling salesman problem. *SIAM Review*, 34(1):1–27, 1992.

[3] J. S. Arthanarajah, Usha, Equivalences of h -median and h -center problems. *SIAM J. Comput.*, 28(3):653–663, 1999.

[4] J. S. Arthanarajah, Usha, On the equivalence of the multi-stage insertion and cycle shift in formulation of the traveling salesman problem. *SIAM J. Comput.*, 28(3):664–674, 1999.

[5] B. Carr. Separating hyperplanes for integer programming. *SIAM J. Comput.*, 28(3):675–689, 1999.

[6] G. B. Dantzig, Fulkerson, and Johnson. Linear inequalities and traveling salesman problems. *SIAM J. Comput.*, 28(3):690–709, 1999.

[7] L. R. Ford and D. R. Fulkerson. *Flows and Networks*. Prentice-Hall, 1962.

[8] M. Jünger, Reinelt, and Rinaldi. The traveling salesman problem. *SIAM Review*, 34(1):1–27, 1992.

[9] M. Usha. New inequalities for the h -median problem. *SIAM J. Comput.*, 28(3):710–720, 1999.