

The Ring Star Problem: Polyhedral Analysis and Exact Algorithm

Martine Labbé*

Gilbert Laporte †

Inmaculada Rodríguez Martín‡

Juan José Salazar González‡

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Abstract

In the Ring Star Problem the aim is to locate a simple cycle through a subset of vertices of a graph with the objective of minimizing the sum of two costs: a routing cost proportional to the length of the cycle, and an assignment cost from the vertices not in the cycle to their closest vertex on the cycle. The problem has several applications in telecommunications network design and in rapid transit systems planning. It is an extension of the classical location-allocation problem introduced in the early sixties, and closely related versions have been recently studied by several authors. This article formulates the problem as a mixed integer linear program, and strengthens it with the introduction of several families of valid inequalities. These inequalities are shown to be facet defining and are used to develop a branch-and-cut algorithm. Computational results show that instances involving up to 300 vertices can be solved optimally using the proposed methodology.

*ISRO and SMG, Université Libre de Bruxelles, C.P. 210/01, B-1050, Bruxelles, Belgium.

†Canada Research Chair in Distribution Management, École des Hautes Études Commerciales, 3000 chemin de la Côte-Sainte-Catherine, Montréal, Canada H3T 2A7.

‡DEIOC, Facultad de Matemáticas, Universidad de La Laguna, 38271 La Laguna, Tenerife, Spain.

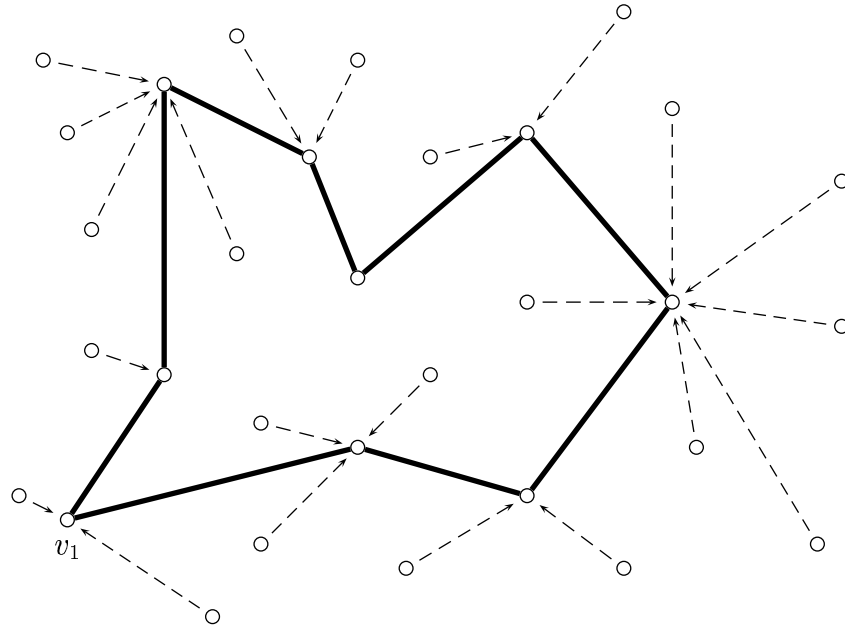


Figure 1: Ring-star solution

1 Introduction

In a typical telecommunications network, the traffic is gathered from many sources, progressively combined in order to fill links of increasing capacity and finally forwarded to its destination. Hence, most telecommunications networks are naturally structured in a multi-layer hierarchical structure.

Although real telecommunications (voice, video or data, nationwide or international networks) are usually structured into many such levels, most design problems considered by telecommunications companies concern only a part of the overall network. A generic telecommunications network consists of access networks which connect the terminals (user nodes) to concentrators (switches or multiplexers) and a backbone network which interconnects these concentrators or connects them to a central unit (root). Telecommunications networks differ in the designs of these access networks and the backbone network, see, e.g., Gourdin, Labbé and Yaman [7] or Klincewicz [10].

A solution to the design of some telecommunications networks is to connect terminals to concentrators by point-to-point links, which results in a star topology, and to interconnect the concentrators through a ring structure. This is the case in Digital Data Service (DDS) design, as mentioned in Xu, Chiu and Glover [17], for example. Roughly speaking, the problem then consists of selecting a subset of user nodes where concen-

trators will be installed, interconnect them by a ring network and assign the other user nodes to those concentrators. The total cost of all connections must be minimized. Figure 1 illustrates a potential solution, where the solid lines represent the *internet* (ring) and the dashed lines represent the *intranet* (assignments). This problem was introduced in Moreno Pérez, Moreno Vega and Rodríguez Martín [13] and solved heuristically.

Similar location-allocation problems arise in a number of planning contexts (see Cooper [3]). They consist of locating a number of facilities among the vertices of a graph so that the allocation costs (i.e., the sum of distances from the remaining vertices to their closest facility) is minimized. In some applications, as in rapid transit systems planning, the facilities must be routed by a vehicle. Thus, in addition to the allocation cost, the objective includes the cost of the shortest Hamiltonian cycle visiting all the facilities. See Labbé, Laporte and Rodríguez Martín [11] for a review of applications and a more general class of problems consisting of locating structures in a graph.

In this article we consider a problem common to both contexts, telecommunications and location-allocation, named *Ring Star Problem* (RSP). The problem can be formally stated as follows. Consider a mixed graph $G = (V, E \cup A)$ where $V = \{v_1, v_2, \dots, v_n\}$ is the vertex set, $E = \{[v_i, v_j] : v_i, v_j \in V, i < j\}$ is the edge set, and $A = \{(v_i, v_j) : v_i, v_j \in V\}$ is the arc set (loops (v_i, v_i) are included in A). We assume that $n \geq 5$ to eliminate degenerate and trivial instances. Vertex v_1 is referred to as the *root* (or *depot*). There is a non-negative *routing cost* c_{ij} associated with each edge $[v_i, v_j]$, and a non-negative *assignment cost* d_{ij} associated with each arc (v_i, v_j) . A solution to the RSP is a simple cycle through a subset V' of V including v_1 and at least two other vertices. The routing cost of a solution is the sum of the routing costs of all edges on the cycle. The assignment cost of a solution is defined as

$$\sum_{v_i \in V \setminus V'} \min_{v_j \in V'} d_{ij}. \quad (1)$$

The RSP consists of determining a solution for which the sum of routing and assignment costs is minimized. The problem is \mathcal{NP} -hard since the special case in which the assignment costs are very large compared to the routing cost is the classical *Traveling Salesman Problem* (TSP).

Lee, Chiu and Sanchez [12] defined a very closely related problem (referred to as *Steiner Ring Star Problem*) by considering an additional set of vertices W (representing *terminals* in telecommunications and *customers* in location-allocation) and setting the assignment cost to

$$\sum_{v_i \in W} \min_{v_j \in V'} d_{ij}. \quad (2)$$

They developed a branch-and-cut that solved instances with $|V| \leq 50$, $|W| \leq 90$ and $|V| + |W| \leq 100$. Xu, Chiu and Glover [17] proposed a tabu search algorithm for this problem, and tested it on instances with $|V| \leq 300$ and $|W| \leq 300$. In addition, Current

and Schilling [4] and Gendreau, Laporte and Semet [6] presented algorithms for variants of the RSP in which the routing cost is minimized subject to an upper bound on the cost of the largest assignment.

Our aim is to develop a polyhedral based exact algorithm for the RSP. The remainder of this article is organized as follows. In Section 2, the RSP is formulated as a mixed integer linear program to which several classes of valid inequalities are incorporated. In Section 3 we derive dimension and facet defining results for RSP. A branch-and-cut algorithm is described in Section 4. Section 5 presents extensive computational results on three classes of instances, one of them generated as to be able to compare our results with the ones in Lee, Chiu and Sanchez [12]. Conclusions follow in Section 6.

2 Mathematical Model

The RSP can be formulated as a mixed integer linear program as follows. For each edge $[v_i, v_j] \in E$, let x_{ij} be a binary variable equal to 1 if and only if edge $[v_i, v_j]$ appears on the cycle. For each arc $(v_i, v_j) \in A$, let y_{ij} be a binary variable equal to 1 if and only if vertex v_i is assigned to vertex v_j on the cycle. If a vertex v_i is on the cycle, it is then assigned to itself, i.e., $y_{ii} = 1$. In addition, for $S \subset V$ define $E(S) := \{[v_i, v_j] \in E : v_i, v_j \in S\}$ and $\delta(S) := \{[v_i, v_j] \in E : v_i \in S, v_j \notin S\}$. If $S = \{v_i\}$, we simply write $\delta(i)$ instead of $\delta(\{v_i\})$. For $E' \subseteq E$, define $x(E') := \sum_{[v_i, v_j] \in E'} x_{ij}$.

The formulation is then:

$$\text{minimize} \quad \sum_{[v_i, v_j] \in E} c_{ij} x_{ij} + \sum_{(v_i, v_j) \in A} d_{ij} y_{ij} \quad (3)$$

subject to:

$$x(\delta(i)) = 2y_{ii} \quad \text{for all } v_i \in V, \quad (4)$$

$$\sum_{v_j \in V} y_{ij} = 1 \quad \text{for all } v_i \in V \setminus \{v_1\}, \quad (5)$$

$$x(\delta(S)) \geq 2 \sum_{v_j \in S} y_{ij} \quad \text{for all } S \subset V : v_1 \notin S, \quad (6)$$

$$x_{ij} \in \{0, 1\} \quad \text{for all } [v_i, v_j] \in E, \quad (7)$$

$$y_{ij} \geq 0 \quad \text{for all } (v_i, v_j) \in A, \quad (8)$$

$$y_{11} = 1 \quad (9)$$

$$y_{1j} = 0 \quad \text{for all } v_j \in V \setminus \{v_1\}, \quad (10)$$

$$y_{jj} \text{ integer} \quad \text{for all } v_j \in V \setminus \{v_1\}. \quad (11)$$

In this formulation, constraints (4) are degree constraints. They ensure that the degree of a vertex v_i is 2 if and only if it belongs to the cycle (i.e., $y_{ii} = 1$). Constraints (5) mean that either v_i is a vertex on the cycle (i.e., $y_{ii} = 1$), or v_i is assigned to a vertex

v_j on the cycle (i.e., $y_{ii} = 0$ and $y_{ij} = 1$). Constraints (6) are connectivity constraints since they state that S must be connected to its complement by at least two edges of the cycle whenever at least one vertex v_i is assigned to $v_j \in S$. The combination of (4), (7), (9) and (10) guarantees that the solution will contain at least one cycle including the root. Constraints (6) limit the number of cycles to one, and the combination of (4), (5), (6), and (8) means that every vertex not belonging to the cycle is assigned to a vertex on the cycle. Integrality conditions on the y_{ij} variables ($v_i \neq v_j$) are unnecessary since, for a given integer x , the objective is minimized when vertices not on the cycle are assigned to the closest vertices on the cycle. Integer solutions are trivially determined in case of ties. Implicitly the above model imposes that the cycle visits at least three vertices, including the root. Such a constraint guarantees that communication between vertices remains possible even if one vertex of the backbone network (e.g., the root) fails. Further, if reliability is unimportant, other potential solutions consisting of a degenerate cycle (i.e., only the root or only the root and another vertex) can easily be enumerated.

The linear relaxation of the model (3)-(11) can be strengthened through the introduction of additional valid inequalities. First observe that connectivity constraints (6) are very rich and reduce to the constraints

$$x_{ij} + y_{ij} \leq y_{jj} \quad (12)$$

whenever $S = \{v_i, v_j\} \subseteq V \setminus \{v_1\}$. Constraints (12) themselves dominate the classical inequalities:

$$x_{ij} \leq y_{ii} \text{ and } x_{ij} \leq y_{jj} \quad \text{for all } v_i, v_j \in V \setminus \{v_1\}.$$

Furthermore, new useful constraints similar to (12) are

$$x_{1j} \leq y_{jj} \text{ and } x_{i1} + y_{i1} \leq 1. \quad (13)$$

Inequalities valid for the cycle polytope (Bauer [2]) are also clearly valid for the RSP. We restrict our attention to the following *2-matching inequalities*:

$$x(E(H)) + x(T) \leq \sum_{v_i \in H} y_{ii} + \frac{|T| - 1}{2} \quad (14)$$

for all $H \subset V$ and $T \subset \delta(H)$ satisfying:

- (i) $\{v_i, v_j\} \cap \{v_k, v_\ell\} = \emptyset$ for $[v_i, v_j], [v_k, v_\ell] \in T$ and $[v_i, v_j] \neq [v_k, v_\ell]$,
- (ii) $|T| \geq 3$ and odd.

An additional family of valid inequalities follows from the fact that some assignments are incompatible. More specifically, variables y_{ij} and y_{ik} are incompatible when $j \neq k$, thus the *Stable Set Problem* (SSP) associated to those incompatibility relations between the assignment variables is a relaxation of RSP. A similar observation has been made by

Avella and Sassano [1] for the p-Median Problem. This property leads (for example) to the following odd-hole (or clique) inequalities:

$$y_{ij} + y_{jk} + y_{ki} \leq 1 \quad (15)$$

for all three different vertices $v_i, v_j, v_k \in V \setminus \{v_1\}$. See Padberg [14] for a partial polyhedral description of the stable set polytope.

Finally, the combination of the two mentioned relaxations (i.e., the cycle and the stable set problems) produces new valid inequalities for RSP, such as:

$$x(\delta(S)) \geq 2(y_{ij} + y_{jk} + y_{ki}) \quad (16)$$

for all three different vertices $v_i, v_j, v_k \in V \setminus \{v_1\}$ and all $S \subseteq V \setminus \{v_1\}$ such that $v_i, v_j, v_k \in S$.

In next section we study conditions under which these inequalities are facet-defining for the RSP polytope.

3 Polyhedral Analysis

We now derive some polyhedral results for model (3)–(11). Denote by P the polytope defined by the convex hull of feasible solutions to RSP and by Q the associated TSP polytope, i.e.,

$$P := \text{conv} \left\{ (x, y) \in \mathbb{R}^{|E|+|A|} : (x, y) \text{ satisfies (4)–(11)} \right\},$$

$$Q := \{(x, y) \in P : y_{ii} = 1 \text{ for all } v_i \in V \setminus \{v_1\}\}.$$

Both polytopes are linked by the intermediate polytopes:

$$P(F) := \{(x, y) \in P : y_{ii} = 1 \text{ for all } v_i \notin F \cup \{v_1\}\},$$

for all $F \subseteq V \setminus \{v_1\}$, in the sense that $P(\emptyset) = Q$ and $P(V \setminus \{v_1\}) = P$. This connection allows us to extend known results from Q to P . For any $\alpha \in \mathbb{R}^{|E|}, \beta \in \mathbb{R}^{|A|}$ and $\gamma \in \mathbb{R}$, define the hyperplane

$$\mathcal{H}(\alpha, \beta, \gamma) := \{(x, y) \in \mathbb{R}^{|E|+|A|} : \alpha x + \beta y = \gamma\}.$$

Lemma 3.1 *Let $v_{i_1}, v_{i_2}, \dots, v_{i_{n-1}}$ be an ordering of the vertices of $V \setminus \{v_1\}$ and let $F_k = \{v_{i_1}, \dots, v_{i_k}\}$ for all $k = 1, \dots, n-1$. If for each $k = 1, \dots, n-1$ and each $v_{i_l} \in V \setminus \{v_{i_k}\}$ there exists a feasible solution (x, y) to the RSP such that*

(i) $y_{i_l i_l} = 1$ for all $l > k$ (i.e., all vertices of $V \setminus F_k$ belong to the cycle),

(ii) $y_{i_{i_l}} + y_{i_{l+1}} = 1$ for all $l < k$ (i.e., all vertices of $F_k \setminus \{v_{i_k}\}$ are either on the cycle or assigned to the root v_1),

(iii) $y_{i_k i_t} = 1$ (i.e., vertex v_{i_k} does not belong to the cycle and is assigned to vertex v_{i_t}),
and

(iv) $\alpha x + \beta y = \gamma$,

then $\dim(\mathcal{H}(\alpha, \beta, \gamma) \cap P) \geq \dim(\mathcal{H}(\alpha, \beta, \gamma) \cap Q) + (|V| - 1)^2$.

Proof. We will prove by induction on $|F_k|$ that $\dim(\mathcal{H}(\alpha, \beta, \gamma) \cap P(F_k)) \geq \dim(\mathcal{H}(\alpha, \beta, \gamma) \cap Q) + |F_k|(|V| - 1)$. If $k = 0$, i.e., if $F_k = \emptyset$, then this inequality holds since $P(F_k) = Q$. Now, by the induction hypothesis, there exist $\dim(\mathcal{H}(\alpha, \beta, \gamma) \cap Q) + |F_{k-1}|(|V| - 1) + 1$ affinely independent points such that $y_{j_k j_k} = 1$ in $\mathcal{H}(\alpha, \beta, \gamma) \cap P(F_{k-1})$ and thus in $\mathcal{H}(\alpha, \beta, \gamma) \cap P(F_k)$. An additional set of $|V| - 1$ affinely independent points such that $y_{j_k j_k} = 0$ in $\mathcal{H}(\alpha, \beta, \gamma) \cap P(F_k)$ are obtained as in the lemma's hypothesis. \square

Proposition 3.1 $\dim(P) = |E| + |V|^2 - 3|V| + 1$.

Proof. Recall that we are assuming $|V| \geq 5$. Then

$$\begin{aligned} \dim(P) &= \dim(\mathcal{H}(0, 0, 0) \cap P) \\ &\geq \dim(\mathcal{H}(0, 0, 0) \cap Q) + (|V| - 1)^2 \quad (\text{by Lemma 3.1 using any sequence}) \\ &= \dim(Q) + (|V| - 1)^2 \\ &= |E| - |V| + (|V| - 1)^2 \quad (\text{see Grötschel and Padberg [8]}). \end{aligned}$$

Now, $\dim(P) \leq |E| + |V|^2 - 3|V| + 1$ since the RSP can be described with $|E| + |V|^2$ variables and $|V|$ type (4), $|V| - 1$ type (5), and $|V|$ type (9)–(10) linearly independent equality constraints. \square

Another important consequence of Lemma 3.1 is that any facet-defining inequality $\alpha x + \beta y \leq \gamma$ for Q satisfying (i)–(iv) is also facet-defining for P . We will use this result for the polyhedral analysis of P .

Proposition 3.2 *The inequality $x_{ij} \geq 0$ defines a facet of P for each $[v_i, v_j] \in E$.*

Proof. The hyperplane defined by $x_{ij} = 0$ satisfies the assumptions of Lemma 3.1 for any sequence. Indeed for any vertex $v_k \in V \setminus \{v_1\}$, a simple cycle spanning $V \setminus \{v_k\}$ and not containing edge $[v_i, v_j]$, together with the assignment of v_k to any vertex of the cycle, yields an (x, y) vector of P such that $x_{ij} = 0$. Hence, $\dim(\{(x, y) \in P : x_{ij} = 0\}) \geq \dim(\{(x, y) \in Q : x_{ij} = 0\}) + (|V| - 1)^2 = (|E| - |V| - 1) + (|V| - 1)^2$, since $x_{ij} \geq 0$ induces a facet of Q (see Grötschel and Padberg [8]). The thesis then follows from Proposition 3.2. \square

Note at this stage that the inequality $x_{ij} \leq 1$ is not facet inducing since it is dominated by the connectivity constraint (12).

Proposition 3.3 *The inequality $y_{ij} \geq 0$ defines a facet of P for each $(v_i, v_j) \in A$, $i \neq j$ and $i \neq 1$.*

Proof. The proof is obtained along the lines of the proof of Lemma 3.1 except that F is now initialized as $F = \{v_i\}$. Since in this case $Q \subset P(F)$, there exist $\dim(Q) + 1 = |E| - |V| + 1$ affinely independent vectors (Hamiltonian cycles) such that $y_{ij} = 0$. An additional $|V| - 2$ affinely independent vectors are obtained by taking a simple cycle spanning $V \setminus \{v_i\}$ and assigning v_i to each vertex of the cycle different from v_j (so $y_{ij} = 0$), which proves that $y_{ij} \geq 0$ defines a facet of $P(F)$ for $F = \{v_i\}$. Moreover, for all $v_k \in V \setminus \{v_1, v_i\}$, any simple cycle spanning $V \setminus \{v_k\}$ and the successive assignment of v_k to the vertices of the cycle yield (x, y) vectors of P satisfying conditions (i)–(iv) of Lemma 3.1. \square

If $i = j$, the inequality $y_{ij} \geq 0$ is not facet inducing since it is implied by equalities (4) and $x_{ij} \geq 0$ for $(v_i, v_j) \in F$. If $i \neq j$ and $i = 1$, then $y_{ij} = 0$ since $y_{11} = 1$, and therefore $y_{ij} \geq 0$ is not facet inducing. The inequality $y_{ij} \leq 1$ is not facet inducing since it is implied by constraints (5) and (8).

Proposition 3.4 *The valid inequality $x(\delta(S)) \geq 2 \sum_{v_j \in S} y_{ij}$ defines a facet of P for each $S \subseteq V \setminus \{v_1\}$, $2 \leq |S| \leq |V| - 3$, $v_i \in S$.*

Proof. The hyperplane defined by $x(\delta(S)) = 2 \sum_{v_j \in S} y_{ij}$ satisfies the assumptions of Lemma 3.1 for any sequence containing v_i in last position. Indeed, for any vertex $v_k \in V \setminus \{v_1, v_i\}$ there exists a simple cycle spanning $V \setminus \{v_k\}$ and with two edges in $\delta(S)$ such that, together with the assignment of v_k to each vertex of the cycle, yields (x, y) vectors of P satisfying $x(\delta(S)) = 2 \sum_{v_j \in S} y_{ij} (= 2)$. Furthermore, for v_i , the last vertex of the sequence, a simple cycle spanning $(S \cup \{v_1\}) \setminus \{v_i\}$ with two edges in $\delta(S)$ can be constructed. The assignment of v_i to each vertex of $S \setminus \{v_i\}$ yields $|S| - 1$ solutions (x, y) satisfying the hypothesis (i) to (iv) of Lemma 3.1. An additional $|V \setminus S|$ such solutions (x, y) are obtained by constructing a simple cycle spanning $V \setminus S$ (which contains at least three vertices), and successively assigning v_i to each vertex of $V \setminus S$. The thesis follows by noting that in Q , the inequality $x(\delta(S)) \geq 2 \sum_{v_j \in S} y_{ij}$ reduces to $x(\delta(S)) \geq 2$, which is facet defining (Grötschel and Padberg [8]). \square

Proposition 3.4 does not hold if $|S| = 1$ or $|S| = |V| - 2$ since constraints (6) are then dominated by (4) or $x(\delta(S)) \geq 2$, respectively. The validity of the latter inequality comes from the fact that the cycle must contain at least one vertex from S whenever $S \subset V \setminus \{v_1\}$ and $|S| \geq |V| - 2$. Proposition 3.4 is also invalid if $v_i \notin S$ since then

$$x(\delta(S \cup \{v_i\})) = x(\delta(S)) + x(\delta(i)) - 2 \sum_{v_j \in S} x_{ij} \geq 2 \sum_{v_j \in S \cup \{v_i\}} y_{ij} = 2(y_{ii} + \sum_{v_j \in S} y_{ij}).$$

This implies that $x(\delta(S)) \geq 2 \sum_{v_j \in S} (y_{ij} + x_{ij})$, which dominates the correspondent constraint (6).

Proposition 3.5 *The inequality $x_{1j} \leq y_{jj}$ defines a facet of P for $j \neq 1$.*

Proof. The hyperplane defined by $x_{1j} = y_{jj}$ satisfies the assumptions of Lemma 3.1. This can be seen by taking a simple cycle spanning $V \setminus \{v_k\}$ and successively assigning v_k to each vertex of the cycle. Furthermore, this cycle must contain edge $[v_1, v_j]$ whenever $v_k \neq v_j$. The thesis follows from the fact that $x_{1j} \leq y_{jj} = 1$ induces a facet of Q (Grötschel and Padberg [8]). \square

The constraints $x_{i1} + y_{i1} \leq 1$ for all $i \neq 1$ do not induce facets of P since these are dominated by the constraints $x_{1i} \leq y_{ii}$. Indeed, $x_{1i} \leq y_{ii} = 1 - \sum_{v_j \in \delta(H) \setminus \{v_i\}} y_{ij} \leq 1 - y_{i1}$.

Proposition 3.6 *The inequality $x(E(H)) + x(T) \leq \sum_{v_i \in H} y_{ii} + (|T| - 1)/2$ for each $H \subset V$ and $T \subset \delta(H)$ satisfying:*

- (i) $\{v_i, v_j\} \cap \{v_k, v_\ell\} = \emptyset$ for $[v_i, v_j], [v_k, v_\ell] \in T$ and $[v_i, v_j] \neq [v_k, v_\ell]$,
- (ii) $|T| \geq 3$ and odd,

defines a facet of P when $|V| \geq 6$.

Proof. The hyperplane defined by $x(E(H)) + x(T) \leq \sum_{v_i \in H} y_{ii} + (|T| - 1)/2$ satisfies the assumptions of Lemma 3.1. Indeed, for each vertex $v_k \in V \setminus \{v_1\}$ it is trivial to design a simple cycle spanning $V \setminus \{v_k\}$ and satisfying hypothesis (i)–(iv) of Lemma 3.1, and v_k can be assigned to each vertex of the cycle. The thesis follows from the fact that $x(E(H)) + x(T) \leq \sum_{v_i \in H} y_{ii} + (|T| - 1)/2$ defines a facet of Q (Grötschel and Padberg [8]). \square

We now turn to the two inequalities (15) and (16) related to the SSP relaxation.

Proposition 3.7 *The inequality $y_{ij} + y_{jk} + y_{ki} \leq 1$ defines a facet of P for each three different vertices $v_i, v_j, v_k \in V \setminus \{v_1\}$.*

This proof is obtained using induction on the number $n := |V|$ of vertices. By using the software PORTA [15] we were able to enumerate all facet defining inequalities for $n = 5$ (see <http://webpages.u11.es/users/jjsalaza/ring>). This allows us to state that the result holds true for $n = 5$.

Assume now that (15) is facet defining for polytope P associated to a mixed graph $G = (V, E \cup A)$ with $V = \{v_1, \dots, v_n\}$, $n \geq 5$. Let $v_{n+1} \notin V$ and define P' as the polytope associated to $G' = (V', E' \cup A')$ where $V' = V \cup \{v_{n+1}\}$, $E' = E \cup \{[v_{n+1}, v_i] : i = 1, \dots, n\}$ and $A' = A \cup \{(v_{n+1}, v_i), (v_i, v_{n+1}) : i = 1, \dots, n\}$. First, consider $\dim(P)$ affinely independent vectors of P satisfying (15) at equality. These vectors can be transformed into vectors of P' by adding arc (v_{n+1}, v_1) to the corresponding RSP solution graphs. Further, those $\dim(P)$ new vectors are still affinely independent and tight for (15). To complete the proof, we need to exhibit $\dim(P') - \dim(P) = 3n - 2$ affinely independent points of P' , tight for (15) and such that $y_{n+1,1} = 0$. Each mixed graph of Figure 2

points	$y_{n+1,1}$	$y_{n+1,n+1}$	$y_{n+1,l} : l \neq 1, n+1$	$y_{l,n+1} : l \neq n+1$	$x_e : e \in \delta(n+1)$
P	1 \vdots 1	0	0	0	0
$(a), (b)$	0 \vdots 0	1 \vdots 1	0	0	$\begin{matrix} 1 & 1 & 0 & 0 & \dots & 0 \\ \vdots & 0 & 1 & 0 & \dots & 0 \\ \vdots & 0 & 0 & 1 & & \vdots \\ \vdots & \vdots & & & \ddots & 0 \\ 1 & 0 & 0 & \dots & 0 & 1 \end{matrix}$
(g)	0	1	0	0	0 1 1 0 \dots 0 0
$(c), (d)$	0 \vdots 0	1 \vdots 1	0	$\begin{matrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \dots & 0 & 1 \end{matrix}$	\dots
$(e), (f)$	0 \vdots 0	0 \vdots 0	$\begin{matrix} 1 & 0 & \dots & 0 \\ 0 & 1 & & \vdots \\ \vdots & & \ddots & 0 \\ 0 & \dots & 0 & 1 \end{matrix}$	0	0

Table 1: Affinely independence certificate for Proposition 3.7

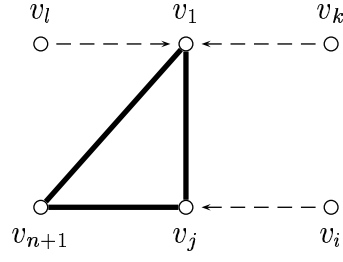
(a), (c) and (e) provides three such points (obtained by permuting v_i , v_j and v_k). This gives nine points. Next, each mixed graph of Figures 2 (b), (d) and (f) provides $n - 4$ such points (one for each $l \neq 1, i, j, k$). Finally, Figure 2 (g) gives the last point. To see that all these $3n - 2$ points are affinely independent, it suffices to check that the new components of their characteristic vectors, i.e. corresponding to the addition of v_{n+1} to the graph, form a non-singular matrix (see Table 1). \square

If one vertex, say v_k , coincides with the root v_1 , Proposition 3.7 does not hold since then (15) is dominated by (12). Indeed, in this case

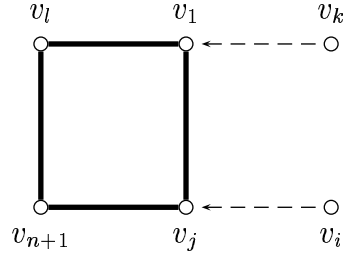
$$y_{ij} + y_{j1} + y_{1i} = y_{ij} + y_{j1} \leq y_{jj} + y_{j1} \leq \sum_{v_l \in V} y_{jl} = 1.$$

Proposition 3.8 *The inequality $x(\delta(S)) \geq 2(y_{ij} + y_{jk} + y_{ki})$ defines a facet of P when $|V| \geq 7$ for each three different vertices $v_i, v_j, v_k \in V \setminus \{v_1\}$ and each $S \subseteq V \setminus \{v_1\}$ such that $v_i, v_j, v_k \in S$ and $4 \leq |S| \leq |V| - 3$.*

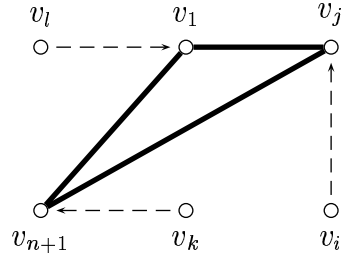
Proof. The proof is similar to that of Proposition 3.7.



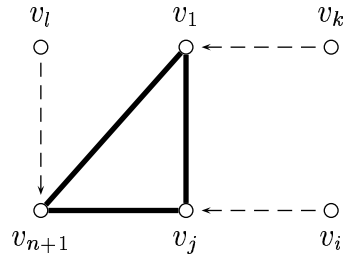
(a)



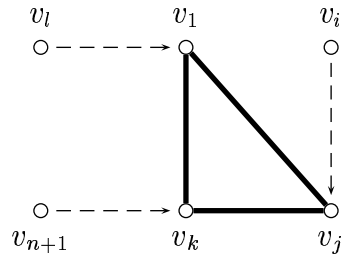
(b)



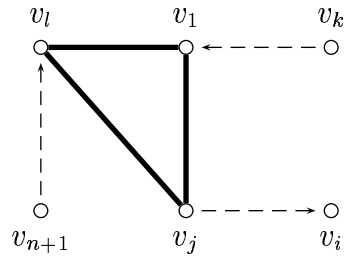
(c)



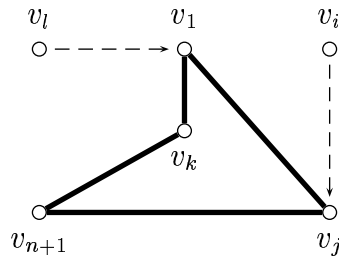
(d)



(e)



(f)



(g)

Figure 2: RSP solutions for the proof of Proposition 3.7

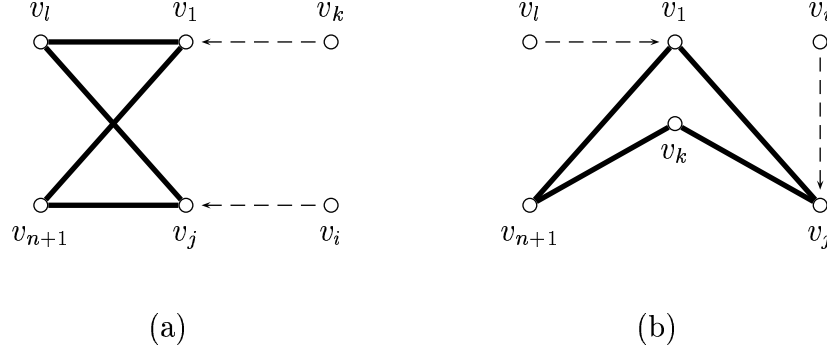


Figure 3: RSP solutions for the proof of Proposition 3.8

When $n = 7$, we can verify that (16) is facet defining by using PORTA [15] (see <http://webpages.uill.es/users/jjsalaza/ring>). Further the $\dim(P)$ affinely independent vectors of P can be extended exactly in the same way.

When $v_{n+1} \in S$, the $3n - 2$ additional affinely independent vectors illustrated in Figure 2 are also tight for (16). When $v_{n+1} \notin S$, Figure 2 (b) yields $n - |S| - 1$ vectors tight for (16), i.e. only if $v_l \notin S$ and $v_l \neq v_1, v_{n+1}$. Figure 3 (a) provides $|S| - 3$ additional tight vectors, i.e. for $v_l \in S \setminus \{v_i, v_j, v_k\}$. Further, Figure 2 (g) must be replaced by Figure 3 (b). \square

Observe that inequalities (16) are not facet defining when $|S| = 3$ since in that case they are the sum of inequalities (12) for arcs (v_i, v_j) , (v_j, v_k) and (v_k, v_i) . Also, (16) cannot be facet defining if $|V \setminus S| \leq 2$ since then $x(\delta(S))$ must be equal to at least two.

4 Branch-and-cut algorithm

We now outline the main ingredients of our branch-and-cut algorithm for finding an optimal solution cycle \overline{C} of the RSP. Let $V(\overline{C})$ be the set of vertices in the solution.

Step 1 (Initialization). Set the iteration count $t := 1$. Compute an upper bound \bar{z} on the optimal RSP as follows. Starting with $V(\overline{C}) := \{v_1\}$, successively insert a vertex $v_i \notin V(\overline{C})$ to minimize

$$L(i, \lambda) := \lambda \text{inc}(\overline{C}, i) + (1 - \lambda)(-\text{dec}(\overline{C}, i)),$$

where $\text{inc}(\overline{C}, i)$ is the minimum increment produced in the routing cost of cycle \overline{C} when inserting v_i , and $\text{dec}(\overline{C}, i)$ is the decrement produced in the accessibility cost. This operation is repeated as long as $L(i, \lambda)$ decreases with a new insertion. This is done for each $\lambda \in \{0.1, 0.3, 0.5, 0.7, 0.9\}$ and the best solution is selected. The initial linear

subproblem is then defined as

$$\text{minimize} \quad \sum_{[v_i, v_j] \in E} c_{ij} x_{ij} + \sum_{(v_i, v_j) \in A} d_{ij} y_{ij}$$

subject to:

$$\begin{aligned} x(\delta(i)) &= 2y_{ii} && \text{for all } v_i \in V \\ \sum_{v_j \in V} y_{ij} &= 1 && \text{for all } v_i \in V \setminus \{v_1\} \\ y_{11} &= 1 \\ x_{1i} &\leq y_{ii} && \text{for all } v_i \in V \setminus \{v_1\}, \end{aligned}$$

and the subproblem is inserted in a list \mathcal{L} .

Step 2 (Termination check and subproblem selection). If the list \mathcal{L} is empty, stop. Otherwise select a subproblem from the list according to a best-first policy.

Step 3 (Subproblem solution). Set $t := t + 1$. Solve the subproblem and let z be the solution objective value. If $z \geq \bar{z}$, go to Step 2. Otherwise, if the solution is feasible for the RSP, set $\bar{z} := z$, update \bar{C} , and go to Step 2.

Step 4 (LP-based heuristic). If the solution is fractional and t is a multiple of 5, apply the following heuristic. Given the fractional solution (x^*, y^*) , sort the x_{ij}^* variables in decreasing order of their values and take in turn their associated edges until a cycle C^* is obtained. If v_1 is not in C^* , introduce it in the best position. Finally, apply the classical 2-optimality procedure to further improve the routing cost of C^* . Let z^* be the objective function value of solution C^* . If $z^* \leq \bar{z}$, set $\bar{z} := z^*$ and update \bar{C} with C^* .

Step 5 (Constraint separation and generation). Introduce up to 300 violated connectivity constraints (6) and 2-matching constraints (14). If no constraints can be generated, go to Step 6. Otherwise, go to Step 3.

Step 6 (Branching). Create two subproblems by branching on a fractional y_{ii} or x_{ij} variable. The first branching strategy consists of finding a y_{ii} variable. To this end, we applied the “strong branching” rule (see Jünger and Thienel [9] for details) within the five y_{ii} variables with fractional value closest to 0.5. If all these variables are integer, select a x_{ij} variable using the same criterion. Insert both subproblems in \mathcal{L} and go to Step 2.

We have developed the following separation procedures for Step 5. Denote by $G^* = (V^*, E^*)$ the *support graph* associated with a given (fractional) solution (x^*, y^*) , i.e.,

$V^* := \{v_i \in V : 0 < y_{ii}^* < 1\}$ and $E^* := \{[v_i, v_j] \in E : 0 < x_{ij}^* < 1\}$. We also define $A^* := \{(v_i, v_j) \in A : i \neq j, 0 < y_{ij}^* < 1\}$.

Separation of connectivity constraints (6).

The inequalities (6) involving sets S such that $|S| = 2$ reduce to (12), and any of them can be violated only if the corresponding edge $[v_i, v_j]$ belongs to E^* or the corresponding arc (v_i, v_j) belongs to A^* . So, they are separated by examining only edges in E^* and arcs in A^* . Next, if the support graph G^* is not connected, then each subset $S \subset V$ corresponding to a connected component and such that $v_1 \notin S$ generates a violated connectivity constraint for each $v_i \notin S$. If G^* is connected, finding a most violated connectivity constraint $x(\delta(S)) \geq 2 \sum_{v_j \in S} y_{ij}$ is equivalent to finding the largest violation of $x(\delta(S)) + 2 \sum_{v_j \notin S} y_{ij} \geq 2$. For a given vertex $v_i \in V \setminus \{v_1\}$ this reduces to a maximum flow problem defined as follows. Let $G^{**} = (V^{**}, E^{**})$ be a graph such that $V^{**} = V^* \cup \{v_{n+1}\}$ and $E^{**} = E^* \cup \{[v_{n+1}, v_j] : v_j \in V^*\}$. The capacity of each edge $e \in E^*$ is equal to x_e^* and that of the new edges $[v_{n+1}, v_j]$ is equal to $2y_{ij}^*$. Let $S' \subset V^{**}$ be such that $v_{n+1} \in S'$, $v_1 \notin S'$ and the capacity Δ of the cut $\delta(S')$ in G^{**} is minimum. If $\Delta \geq 2$, there is no connectivity constraint involving v_i violated by the current solution (x^*, y^*) . Otherwise, $S = S' \setminus \{v_{n+1}\}$ defines a most violated connectivity constraint (6) involving v_i .

Separation of 2-matching constraints (14)

To separate the 2-matching inequalities, we use the heuristic procedure proposed in Fischetti, Salazar and Toth [5]. This algorithm can identify several violated 2-matching constraints for the current fractional solution.

Separation of constraints (15) and (16)

Clearly, inequalities (15) can be separated through a complete enumeration of v_i, v_j, v_k such that $y_{ij}^* > 0$, $y_{jk}^* > 0$, and $y_{ki}^* > 0$. Indeed, when (say) $y_{ij} = 0$ and (12) holds, then (15) also holds. In a similar way, for each v_i, v_j, v_k such that $y_{ij}^* > 0$, $y_{jk}^* > 0$, and $y_{ki}^* > 0$, a min-cut separating v_1 from $\{v_i, v_j, v_k\}$ in G^* gives the most-violated constraint (16), if any. Again, when (say) $y_{ij} = 0$ and (6) holds, then (16) also holds.

5 Computational results

The branch-and-cut algorithm described in the previous sections was implemented in the C++ programming language. ABACUS 2.3 linked with CPLEX 6.0 was used as a branch-and-cut framework. See Jünger and Thienel [9] for details on this software. The

performance of the algorithm was tested on three different classes of test instances. The root was always chosen as the first vertex.

Class I is based on TSP instances from TSPLIB 2.1 (Reinelt [16]) involving between 50 and 200 vertices (problems *eil51* to *kroB200*). Denote by l_{ij} the distance between vertices v_i and v_j given in the TSP files. To obtain optimal solutions visiting approximately 100%, 75%, 50% and 25 % of the total number of vertices in the instances, we set $c_{ij} = \lceil \alpha l_{ij} \rceil$, $d_{ij} = \lceil (10 - \alpha) l_{ij} \rceil$ for $\alpha \in \{3, 5, 7, 9\}$, and $d_{ii} = 0$ for all $v_i \in V$.

Class II was generated in the following way. We generated vertices with coordinates in $[0, 1000] \times [0, 1000]$ and computed l_{ij} as the Euclidean distance between v_i and v_j rounded up to the nearest integer. Costs c_{ij} , d_{ij} and d_{ii} are defined as in Class I, using the same values of α . For each pair $(|V|, \alpha)$ we generated ten instances.

Class III consists of instances generated as in Lee, Chiu and Sanchez [12]. We first generated vertices with coordinates in $[0, 1000] \times [0, 1000]$ and then computed l_{ij} as the Euclidean distance between v_i and v_j rounded up to the nearest integer. Costs $c_{ij} := d_{ij} := l_{ij}$, and costs d_{ii} were randomly generated in the interval $[0, 1000]$ for all $v_i \in V$. For each value $|V|$ we generated ten instances.

Tables 2 to 5 summarize the computational behavior of our branch-and-cut code on the three classes of instances. The column headings are defined as follows.

Name: problem name (for Class I);

|V|: number of vertices (for Classes II and III);

α : scale factor described above (for Classes I and II);

succ: number of instances solved to optimality over 10 trials (for Classes II and III);

p^* : percentage of vertices visited by the optimal cycle;

opt: optimal objective value (for Class I);

%-UB0: percentage ratio UB/opt, where UB is the objective function value of the initial heuristic solution.

h-time: initial heuristic time.

%-LB: percentage ratio LB/opt, where LB is the objective value of the LP-relaxation computed at the root node of the branch-decision tree;

%-UB: percentage ratio UB/opt, where UB is the objective value of the best solution at the root node of the branch-decision tree;

pair: number of Constraints (12) generated;

sec: number of Constraints (6) with $|S| > 2$ generated;

2mat: number of Constraints (14) generated;

nodes: total number of nodes examined (1 means that the problem required no branching);

time: total computing time.

The computing times reported are expressed in the format *h:mm:ss*, and refer to CPU time on Pentium III personal computer running at 866 MHz. We imposed a time limit of two hours for each run. For the instances exceeding the time limit, we report '>2:00:00' in the *time* column, and compute the corresponding results by considering the best available solution as optimal. Hence, for the time-limit instances the column %-LB gives an overestimation of the percentage approximation error. Tables 2 and 3 refer to Class I (TSPLIB based instances). Tables 4 and 5 refer to instances in Class II and Class III respectively, and contain average results over ten trials.

Preliminary experiments revealed that inequalities (15) and (16) are not useful in the sense that they were not violated after all others had been separated. Moreover, even when violations were detected at the root node, introducing these constraints did not help to reduce the gap between the lower and the upper bounds, and did not reduce the computing time. Hence we decided not to use these inequalities in our tests.

Results presented in Tables 2 to 4 indicate that the proposed branch-and-cut algorithm can solve instances involving up to 200 vertices within modest computing times. For a given problem size, the most difficult instances tend to be those where relatively few vertices belong to the optimal cycle. This can be explained by the fact that the location component of the problem is then important. At the other extreme, instances in which all vertices belong to the cycle reduce to a TSP which is a relatively easy problem for the values of $|V|$ considered. The %-LB column indicates that the lower bound developed at the root of the branch-and-cut tree is very tight, typically within 0.5% of the optimum. The heuristic is also very powerful and usually yields solution within 5% of the optimum. It seems to perform better when not all vertices belong to the cycle. All valid inequalities, apart from (15) and (16), are frequently generated within the search tree. The number of nodes in the tree is relatively low, and several instances (in particular the TSPLIB based instances) are solved without any branching. Overall we were able to solve to optimality 124 of the 132 TSPLIB based instances in Class I. For the random instances in Class II, the corresponding statistic is 267 out of 280, and the largest instances involved 200 vertices.

Table 5 shows that for random instances in Class III, i.e., those generated as in [17, 12], we were able to solve to optimality 122 out of 130 instances, involving up to 300 vertices. The largest instances we were able to solve contain three times as many vertices as those solved by Lee, Chiu and Sanchez [12], who considered a closely related problem.

6 Conclusions

We have presented what we believe to be the first exact algorithm for the *Ring Star Problem*, a problem arising in several network design contexts. An mixed integer linear programming formulation including several classes of facet defining inequalities was proposed, together with a branch-and-cut algorithm. The proposed approach was tested on three classes of instances. The largest solved involved 300 vertices. Recall that two of the classes are the benchmark instances in Moreno Pérez, Moreno Vega and Rodríguez Martín [13], while the third set of instances were generated as those used in Lee, Chiu and Sanchez [12] and in Xu, Chiu and Glover [17].

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name	α	%-UB0	h-time	p^*	opt	%-LB	%-UB	pair	sec	2mat	nodes	time
eil51	3	107.51	0:00:00	100.00	1278	100.00	107.51	4	155	91	3	0:00:02
eil51	5	102.76	0:00:00	74.51	1995	100.00	100.00	61	42	0	1	0:00:01
eil51	7	101.47	0:00:00	33.33	2113	100.00	100.00	238	1251	0	1	0:00:10
eil51	9	101.29	0:00:00	11.76	1244	100.00	100.00	529	883	0	1	0:00:10
berlin52	3	105.06	0:00:00	100.00	22626	100.00	100.00	12	46	0	1	0:00:01
berlin52	5	102.78	0:00:00	78.85	36115	100.00	100.00	85	957	36	1	0:00:04
berlin52	7	100.94	0:00:00	46.15	37376	100.00	100.00	221	655	1	1	0:00:06
berlin52	9	100.78	0:00:00	9.62	20361	100.00	100.00	539	941	0	1	0:00:12
brazil58	3	102.16	0:00:01	100.00	76185	100.00	100.00	18	105	3	1	0:00:02
brazil58	5	101.46	0:00:01	68.97	115045	100.00	100.00	104	2994	285	1	0:00:14
brazil58	7	100.24	0:00:00	48.28	126807	100.00	100.00	237	1543	2	1	0:00:12
brazil58	9	100.00	0:00:00	15.52	83690	100.00	100.00	577	1502	0	1	0:00:28
st70	3	103.41	0:00:01	100.00	2025	99.80	103.41	32	660	110	11	0:00:13
st70	5	102.41	0:00:01	78.57	3110	99.84	102.41	110	2837	243	3	0:00:24
st70	7	101.41	0:00:01	42.86	3402	100.00	100.00	283	1920	0	1	0:00:20
st70	9	100.15	0:00:00	25.71	2610	100.00	100.00	806	1935	0	1	0:01:10
eil76	3	107.25	0:00:02	100.00	1614	100.00	100.00	6	139	43	1	0:00:05
eil76	5	100.61	0:00:02	73.68	2460	99.92	100.61	113	1858	56	5	0:00:21
eil76	7	102.76	0:00:01	42.11	2504	100.00	100.00	343	2228	0	1	0:00:37
eil76	9	101.35	0:00:00	15.79	1710	100.00	100.00	930	2440	0	1	0:02:28
pr76	3	101.55	0:00:02	100.00	324477	98.59	101.55	128	15270	6043	727	0:17:39
pr76	5	106.21	0:00:02	80.26	500395	99.66	106.21	171	8847	1489	71	0:02:27
pr76	7	103.89	0:00:01	53.95	555858	99.80	103.89	350	6744	254	13	0:01:54
pr76	9	100.72	0:00:00	17.11	424359	100.00	100.00	874	3011	0	1	0:03:04
gr96	3	102.75	0:00:15	100.00	163935	99.30	102.75	38	1608	415	31	0:01:43
gr96	5	104.31	0:00:13	79.17	252850	100.00	100.00	144	4802	292	1	0:01:06
gr96	7	105.97	0:00:07	50.00	273599	100.00	100.00	409	3254	2	1	0:01:08
gr96	9	105.99	0:00:02	27.08	232823	100.00	100.00	1236	3766	0	1	0:06:53
rat99	3	107.35	0:00:06	100.00	3633	99.89	107.35	12	1178	234	13	0:00:38
rat99	5	103.65	0:00:05	89.90	5885	100.00	100.00	150	2437	300	1	0:00:35
rat99	7	104.26	0:00:02	42.42	6436	100.00	100.00	480	3197	0	1	0:01:20
rat99	9	105.01	0:00:01	21.21	5150	100.00	100.00	1285	4541	0	1	0:07:19
kroA100	3	100.86	0:00:06	100.00	63846	99.80	100.86	24	886	211	11	0:00:37
kroA100	5	100.21	0:00:05	80.00	100785	99.78	100.21	149	6285	246	3	0:01:26
kroA100	7	106.18	0:00:03	55.00	115388	100.00	100.00	428	3368	4	1	0:01:16
kroA100	9	103.15	0:00:00	21.00	94265	100.00	100.00	1250	3150	0	1	0:05:57
kroB100	3	105.42	0:00:06	100.00	66423	99.50	105.42	30	2051	310	39	0:01:38
kroB100	5	104.35	0:00:05	77.00	104550	100.00	100.00	147	4319	271	1	0:00:57
kroB100	7	103.95	0:00:02	46.00	118111	100.00	100.00	435	3322	0	1	0:01:17
kroB100	9	100.99	0:00:00	16.00	93938	100.00	100.00	1303	4238	0	1	0:07:34
kroC100	3	102.42	0:00:06	100.00	62247	99.81	102.42	48	1074	215	9	0:00:33
kroC100	5	103.43	0:00:05	81.00	99065	100.00	100.00	141	3283	280	1	0:00:45
kroC100	7	102.40	0:00:03	50.00	113533	100.00	100.00	438	3200	0	1	0:01:07
kroC100	9	102.07	0:00:01	23.00	92894	100.00	100.00	1319	4056	0	1	0:07:05
kroD100	3	105.58	0:00:06	100.00	63882	99.88	105.58	40	948	186	5	0:00:32
kroD100	5	105.25	0:00:05	82.00	101645	100.00	100.00	154	2651	119	1	0:00:38
kroD100	7	101.07	0:00:02	47.00	116849	99.94	101.07	440	4213	7	3	0:01:50
kroD100	9	102.82	0:00:00	23.00	92102	100.00	100.00	1251	4123	0	1	0:06:58
kroE100	3	104.55	0:00:06	100.00	66204	99.28	104.55	48	821	374	41	0:01:44
kroE100	5	103.67	0:00:05	77.00	104915	99.94	103.67	170	6678	447	11	0:01:54
kroE100	7	105.60	0:00:02	51.00	116471	100.00	100.00	441	3010	1	1	0:01:06
kroE100	9	100.96	0:00:01	20.00	96116	100.00	100.00	1228	5068	0	1	0:08:25
rd100	3	101.67	0:00:06	100.00	23730	100.00	100.00	28	524	122	1	0:00:16
rd100	5	102.34	0:00:05	76.00	37975	99.87	102.34	165	6766	282	5	0:01:40
rd100	7	102.61	0:00:02	44.00	40915	100.00	100.00	433	5848	26	1	0:01:53
rd100	9	102.59	0:00:01	21.00	31776	100.00	100.00	1287	4368	0	1	0:07:01
eil101	3	106.84	0:00:06	100.00	1887	99.84	106.84	16	824	227	19	0:00:46
eil101	5	103.61	0:00:05	72.28	2905	100.00	100.00	301	1810	257	1	0:00:31
eil101	7	102.29	0:00:02	38.61	2926	99.93	102.29	1680	7918	18	7	0:04:00
eil101	9	101.48	0:00:00	17.82	1955	100.00	100.00	3291	4970	0	1	0:08:18
lin105	3	103.43	0:00:07	100.00	43137	100.00	100.00	56	808	145	1	0:00:21
lin105	5	100.81	0:00:06	80.95	69365	100.00	100.00	233	2588	341	1	0:00:43
lin105	7	107.97	0:00:04	53.33	83597	100.00	100.00	1149	4047	6	1	0:01:38
lin105	9	102.37	0:00:02	31.43	69920	100.00	100.00	3995	4998	0	1	0:07:30

Table 2: Computational results for instances in Class I

name	α	%-UB0	h-time	p^*	opt	%-LB	%-UB	pair	sec	2mat	nodes	time
pr107	3	100.36	0:00:08	100.00	132909	100.00	100.00	0	1171	15	1	0:00:23
pr107	5	100.89	0:00:06	68.22	210465	100.00	100.00	327	3418	169	1	0:00:50
pr107	7	100.63	0:00:03	42.99	259571	100.00	100.00	1198	4979	1	1	0:01:52
pr107	9	101.08	0:00:01	26.17	264918	100.00	100.00	4218	5366	0	1	0:06:45
gr120	3	103.26	0:00:14	100.00	20826	99.92	103.26	40	1337	186	9	0:01:01
gr120	5	103.75	0:00:10	75.83	31480	100.00	100.00	420	8782	366	1	0:02:38
gr120	7	102.19	0:00:06	41.67	32301	100.00	100.00	1211	6133	0	1	0:03:19
gr120	9	103.19	0:00:01	22.50	24322	100.00	100.00	3799	6332	0	1	0:11:34
pr124	3	104.96	0:00:14	100.00	177090	98.82	104.96	251	10055	1594	195	0:12:32
pr124	5	103.45	0:00:13	90.32	286115	99.79	103.45	559	7777	782	17	0:03:58
pr124	7	102.48	0:00:10	66.94	358853	100.00	100.00	1263	7590	21	1	0:03:27
pr124	9	102.65	0:00:03	39.52	340153	100.00	100.00	5122	7999	1	1	0:16:05
bier127	3	104.70	0:00:15	100.00	354846	99.84	104.70	64	1883	341	29	0:02:27
bier127	5	108.23	0:00:13	76.38	539955	100.00	108.23	526	10150	638	3	0:03:32
bier127	7	101.35	0:00:06	44.09	567110	99.99	101.35	2140	11081	27	5	0:07:11
bier127	9	100.24	0:00:01	14.96	347845	100.00	100.00	5989	7333	0	1	0:36:52
ch130	3	107.09	0:00:17	100.00	18330	99.85	107.09	100	2049	384	21	0:02:07
ch130	5	106.60	0:00:15	83.08	28790	100.00	100.00	400	7044	608	1	0:02:38
ch130	7	103.87	0:00:08	47.69	32707	99.64	100.77	3183	19575	143	15	0:11:24
ch130	9	101.72	0:00:01	16.15	23639	100.00	100.00	5179	7112	0	1	0:21:56
pr136	3	104.15	0:00:20	100.00	290316	99.45	104.15	42	11092	5600	147	0:15:04
pr136	5	101.43	0:00:18	86.76	468520	100.00	100.00	401	6192	437	1	0:02:23
pr136	7	102.59	0:00:06	43.38	491981	100.00	100.00	1786	8073	0	1	0:06:01
pr136	9	104.14	0:00:02	25.74	387327	100.00	100.00	6152	10262	0	1	0:25:33
gr137	3	104.13	0:01:02	100.00	208929	99.76	104.13	46	1764	348	13	0:02:36
gr137	5	101.16	0:00:57	83.94	329465	99.74	101.16	603	20033	1097	15	0:08:59
gr137	7	102.44	0:00:36	54.74	366022	100.00	100.00	1965	11516	16	1	0:07:06
gr137	9	107.02	0:00:05	26.28	335009	100.00	100.00	6359	10498	0	1	0:31:09
pr144	3	103.07	0:00:25	100.00	175611	99.83	103.07	126	1830	232	21	0:02:35
pr144	5	104.87	0:00:24	94.44	290945	99.55	104.87	1108	45085	4218	323	0:51:01
pr144	7	103.83	0:00:18	63.89	383041	100.00	100.00	1607	15046	236	1	0:07:22
pr144	9	104.68	0:00:03	23.61	366833	100.00	100.00	7966	11221	0	1	0:25:44
ch150	3	104.44	0:00:29	100.00	19584	99.79	104.44	77	4270	547	35	0:05:07
ch150	5	104.64	0:00:25	77.33	31170	99.98	100.00	550	10366	456	3	0:05:15
ch150	7	104.02	0:00:13	48.00	34930	99.99	104.02	1891	10208	3	3	0:09:21
ch150	9	104.50	0:00:02	18.67	26371	100.00	100.00	7689	12746	0	1	0:54:12
kroA150	3	103.53	0:00:29	100.00	79572	99.49	103.53	84	11656	2161	115	0:17:05
kroA150	5	102.62	0:00:24	80.67	125435	100.00	100.00	399	7538	695	1	0:03:44
kroA150	7	103.97	0:00:13	48.67	140961	99.76	103.97	2106	10498	90	5	0:09:05
kroA150	9	102.79	0:00:03	19.33	113080	100.00	100.00	7254	10606	0	1	0:45:36
kroB150	3	107.31	0:00:29	100.00	78390	99.51	107.31	78	4667	1491	65	0:10:33
kroB150	5	104.52	0:00:25	79.33	122875	99.49	104.52	4171	181892	6524	221	1:52:12
kroB150	7	104.48	0:00:12	50.67	135382	100.00	100.00	1729	8595	1	1	0:07:01
kroB150	9	102.02	0:00:03	18.67	108885	100.00	100.00	7372	10483	0	1	0:43:22
pr152	3	103.41	0:00:31	100.00	221046	99.51	103.41	162	5210	2296	45	0:09:42
pr152	5	100.00	0:00:34	87.50	376155	96.05	100.00	1595	58290	6674	285	> 2:00:00
pr152	7	100.00	0:00:18	51.32	475052	96.66	100.00	5576	84188	2042	59	> 2:00:00
pr152	9	101.78	0:00:03	21.05	475440	100.00	100.00	10902	14875	0	1	0:52:15
u159	3	107.83	0:00:37	100.00	126240	99.80	107.83	58	1346	610	7	0:02:27
u159	5	104.35	0:00:33	84.28	204250	99.97	104.35	673	16748	897	7	0:09:08
u159	7	101.92	0:00:20	55.97	235221	100.00	100.00	2040	10668	3	1	0:09:24
u159	9	101.92	0:00:05	25.79	199552	100.00	100.00	8388	13736	0	1	0:59:50
rat195	3	108.44	0:01:23	100.00	6969	99.68	108.44	78	17482	2048	61	0:29:41
rat195	5	108.30	0:01:05	84.62	11320	99.92	108.30	562	10841	1033	11	0:12:21
rat195	7	102.53	0:00:28	40.00	12319	100.00	100.00	2826	21795	14	1	0:34:25
rat195	9	100.00	0:00:07	16.92	9395	94.21	100.00	8965	16115	0	1	> 2:00:00
d198	3	102.50	0:01:28	100.00	47340	99.74	102.50	60	10573	1102	65	0:25:54
d198	5	102.49	0:01:10	79.80	76945	99.94	102.49	2516	106574	7424	5	1:12:20
d198	7	102.52	0:00:38	51.01	94300	99.87	101.34	8298	79000	716	59	1:48:59
d198	9	100.00	0:00:09	19.19	97899	96.34	100.00	15536	26958	0	1	> 2:00:00
kroA200	3	100.00	0:02:05	100.00	93699	93.59	100.00	234	23765	2930	129	> 2:00:00
kroA200	5	107.78	0:01:12	74.00	138885	99.85	107.78	2410	149600	3181	39	> 2:00:00
kroA200	7	103.77	0:00:43	49.00	158227	99.83	102.29	7169	72865	919	37	1:31:05
kroA200	9	100.00	0:00:09	19.00	124678	97.15	100.00	10387	16377	0	1	> 2:00:00
kroB200	3	106.26	0:01:32	100.00	88311	99.81	106.26	106	4080	868	33	0:13:44
kroB200	5	104.97	0:01:18	77.00	138905	99.93	104.97	1025	29056	1023	7	0:25:44
kroB200	7	101.96	0:00:48	49.00	156638	100.00	100.00	2854	16926	4	1	0:25:19
kroB200	9	100.00	0:00:08	18.50	127800	95.13	100.00	10055	14849	0	1	> 2:00:00

Table 3: Computational results for instances in Class I (continued)

V	α	succ	p^*	%-UB0	h-time	%-LB	%-UB	pair	sec	2mat	nodes	time
50	3	10	100.00	103.06	0:00:00	99.59	102.30	16.60	437.70	94.10	12.60	0:00:05
	5	10	78.60	100.94	0:00:00	99.91	100.11	70.70	941.30	69.10	2.00	0:00:04
	7	10	50.40	102.53	0:00:00	100.00	100.00	201.50	971.30	1.90	1.00	0:00:06
	9	10	12.40	100.95	0:00:00	100.00	100.00	543.50	977.90	0.00	1.00	0:00:15
75	3	10	100.00	104.30	0:00:02	99.83	102.70	22.20	548.70	148.40	9.20	0:00:15
	5	10	78.67	103.40	0:00:02	99.90	100.74	113.50	2777.90	207.60	7.00	0:00:32
	7	10	49.33	102.46	0:00:01	99.99	100.28	321.60	2364.80	5.00	1.40	0:00:33
	9	10	15.87	101.47	0:00:00	100.00	100.00	895.60	2279.60	0.00	1.00	0:02:02
100	3	10	100.00	104.40	0:00:06	99.74	104.08	35.30	1520.10	398.70	19.40	0:01:07
	5	10	77.70	103.96	0:00:05	99.87	102.08	159.90	8149.40	652.20	19.40	0:02:49
	7	10	46.00	102.35	0:00:02	99.78	100.89	514.60	31894.50	988.70	74.40	0:12:53
	9	10	19.40	102.31	0:00:01	100.00	100.00	1262.40	4051.60	0.00	1.00	0:07:36
125	3	10	100.00	106.17	0:00:14	99.77	105.31	51.00	2949.20	481.30	24.00	0:02:44
	5	10	80.08	104.92	0:00:12	99.92	103.48	520.40	13255.80	740.40	10.00	0:05:02
	7	10	45.44	103.09	0:00:06	99.87	100.51	1902.60	10757.70	140.50	6.40	0:06:23
	9	10	18.16	102.36	0:00:01	100.00	100.00	5409.70	7471.00	0.00	1.00	0:22:05
150	3	10	100.00	105.66	0:00:29	99.75	104.95	118.20	7356.40	1350.30	72.00	0:10:37
	5	10	80.20	105.07	0:00:25	99.91	102.41	775.90	32248.70	1494.70	21.00	0:16:30
	7	10	46.67	103.57	0:00:12	99.91	101.38	3718.50	23575.00	415.70	32.20	0:23:09
	9	10	18.47	102.09	0:00:02	100.00	100.00	7424.30	11163.00	0.00	1.00	0:54:17
175	3	10	100.00	106.69	0:00:54	99.84	105.96	94.90	4248.90	858.80	29.60	0:08:10
	5	8	79.71	105.12	0:00:46	99.88	103.20	1128.38	39025.38	1897.50	17.50	0:27:03
	7	9	46.16	103.53	0:00:23	99.89	101.90	4206.56	30322.00	478.78	19.00	0:34:58
	9	9	18.16	102.98	0:00:04	100.00	100.00	9630.22	15011.78	0.11	1.00	1:43:30
200	3	10	100.00	106.74	0:01:32	99.78	106.74	169.60	12025.50	1937.70	84.80	0:32:20
	5	9	79.56	104.57	0:01:19	99.93	103.53	1006.22	40507.22	1871.78	10.33	0:33:14
	7	8	47.75	104.17	0:00:38	99.92	102.45	4383.25	32413.25	369.13	14.00	0:48:01
	9	0										

Table 4: Computational results for instances in Class II

V	succ	p^*	%-UB0	h-time	%-LB	%-UB	pair	sec	2mat	nodes	time
10	10	36.00	101.12	0:00:00	100.00	100.00	12.20	12.50	0.00	1.00	0:00:00
20	10	25.00	101.74	0:00:00	100.00	100.00	61.90	92.70	0.00	1.00	0:00:00
30	10	25.33	103.29	0:00:00	100.00	100.00	104.20	133.50	0.50	1.00	0:00:00
40	10	23.00	101.13	0:00:00	100.00	100.00	154.00	258.10	1.70	1.00	0:00:01
50	10	20.80	103.74	0:00:00	100.00	100.00	225.90	338.10	1.50	1.00	0:00:02
75	10	20.53	101.87	0:00:00	99.99	100.03	401.20	1008.60	5.20	1.20	0:00:12
100	10	19.50	103.20	0:00:01	99.98	100.17	561.60	1777.40	14.30	1.60	0:00:35
125	10	18.08	102.55	0:00:01	100.00	100.18	1013.00	2882.60	22.10	1.40	0:01:16
150	10	17.20	102.37	0:00:03	99.98	100.44	1348.10	4785.40	25.10	2.00	0:02:53
175	10	17.01	103.29	0:00:05	99.96	100.65	2534.10	13865.80	98.00	6.20	0:12:33
200	10	16.50	102.32	0:00:08	99.96	100.79	3177.60	26594.00	165.40	11.00	0:28:22
250	8	15.80	103.39	0:00:17	99.98	101.32	3792.75	22781.75	133.63	4.25	0:32:54
300	4	14.75	102.62	0:00:30	99.99	100.25	9220.25	40934.25	191.25	1.50	1:08:41

Table 5: Computational results for instances in Class III