# Compatible Weights and Valid Cycles in Non-spanning OSPF Routing Patterns 

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#### Abstract

Many IP (Internet Protocol) networks use OSPF (Open Shortest Path First) for determining the routing of traffic. OSPF routers compute routing paths using link weights set by the network administrator, and the routers send traffic on all shortest paths to the destination. An interesting question is whether or not a set of prespecified routing patterns can be realized in an OSPF network. If not, we seek structural properties that explain why no such weights exist. Mathematical models for finding weights and for combining routing patterns are presented. We show that two possibly non-spanning routing patterns forming a "valid cycle" cannot simultaneously be obtained in an OSPF network. Two new methods for finding valid cycles are presented, illustrated by numerical examples, and shown to be faster than those previously known.


Key words: Internet Protocol, OSPF, routing, compatible weights, valid cycle.

## 1. Introduction

The Internet consists of a huge number of routing domains, also called autonomous systems, and each domain is supervised by an administrator. The administrator has several responsibilities, and one of the most important is to determine how traffic is routed through the domain. The routing is determined by choosing a routing protocol, and by setting appropriate values on routing parameters. There are several different routing protocols available for autonomous systems, for example OSPF (Open Shortest Path First), RIP (Routing Internet Protocol) and IS-IS (Intermediate-System to Intermediate-System), see [19], [18], and [12].

We will here study networks where OSPF is used as routing protocol. In such networks, each link is assigned a positive integer link weight by the network administrator, and the routers send traffic on the paths that have a minimal sum of weights to each destination (called the shortest paths). If several paths have the same minimal sum of weights, the router splits the traffic addressed to the destination evenly on all outgoing links that belong to a shortest path. A "routing pattern" contains all the paths that are used for routing.

It is an easy task to compute the routing paths if the weights are known, and this is in fact exactly what the routers do when the routing paths are computed. However, from an administrator's point of view, it is not always as easy to find link weights that give a certain
set of desired routing paths. In fact it depends on the structure of the desired routing paths if such weight exists. The administrator's problem is studied in this paper, i.e. a set of desired routing paths is given, and the task is to find weights such that all desired paths are shortest paths with respect to the weights, while any other path between the same pairs of routers should have a larger sum of weights. If there are no such weights, we wish to explain which parts of the routing paths that are in conflict. Since this problem consists of finding a set of weights with certain properties, it will be called the Weight Finding Problem (WFP).
[15], [3] and [9] present mathematical models for different versions of the WFP, and the differences originate mainly from the structure of the prespecified routing paths. Some models require that at most one path is specified between a pair of nodes, while other models allow more than one. Load balancing can only be used if several paths are allowed between two nodes. Another variation is that the desired routing paths sometimes are directed, and sometimes undirected. The case with undirected routing paths yields symmetric routing, and this case is treated in [15] and [3]. The case with spanning and directed routing patterns is studied in [9]. Directed routing paths allow non-symmetric routing since it is possible to specify different paths in the different directions between two nodes. Symmetric routing can also be obtained using directed routing paths, simply by specifying pairs of oppositely directed paths, see [2].

In previous research, the OSPF protocol has also been modeled as side constraints in network design models and flow allocation models, see e.g. [4], [17], [23] and [14]. A common approach is to introduce integer variables for the link weights, and let the side constraints use the weights for ensuring that traffic is routed only on shortest paths to each destination. There are however disadvantages with this approach, one of which is a constraint coefficient that must be larger than the sum of weights on any path. This makes the LP-relaxation very weak, which is a disadvantage when solution methods are considered.

These difficulties may be avoided if the design model is not based on the weights. An alternate approach is therefore to develop network design models based on routing patterns which are obtainable from OSPF weights, but not explicitly including the weights. The OSPF protocol is then modeled by ensuring that the routing patterns satisfies necessary conditions for the existence of compatible weights. Such conditions have been developed by investigating infeasible instances of the WFP. Necessary conditions for undirected routing paths are presented in [3], and necessary conditions for directed and spanning routing patterns are presented in [11]. (A spanning routing pattern contains paths to/from all nodes, while in a non-spanning routing pattern, all nodes are not included.)

In this paper we consider the case with directed routing patterns and load balancing, and discuss several models for the WFP. Compared to the model in [11], it is not required that the routing patterns are spanning. We show that routing patterns with certain properties can be combined into a single routing pattern, and describe the advantages of doing so. We prove that two possibly non-spanning routing patterns can not be obtained simultaneously if they contain a structure called "valid cycle". In [11] this was shown only for spanning routing patterns. Two new methods for finding valid cycles are proposed, and computational results indicate that these new methods are faster than the one previously known, [6], [11].

The next section contains a detailed description of the problem and mathematical formulations of the weight finding problem defined for sets of routing paths. Section 3. investigates how non-spanning routing patterns can be combined into so called SP-graphs. Valid cycles are discussed in Section 4., and methods for finding valid cycles are presented in Section 5.. The case with symmetric routing paths is discussed in Section 6.. Section 7. contains numerical examples and Section 8.
contains results from computational tests. The last section contains conclusions.

## 2. Problem formulation

We consider the directed graph $G=(N, A)$ with the set of nodes $N$ and the set of $\operatorname{arcs} A$. A number of node pairs, $\left(o_{k}, d_{k}\right), k=1, \ldots, K$, is given, and for each $k$, a set $S_{k}$ of paths from node $o_{k}$ to node $d_{k}$ is given. $S_{k}$ contains the desired shortest paths between these two nodes. $S_{k}$ can either contain one single path or several paths (corresponding to splitting of the traffic), and each path in $S_{k}$ is represented by the included arcs. The case when each $S_{k}$ contains one path is called "the single path case". We will also use the set of all arcs in any of the paths in $S_{k}$, i.e. $R_{k}=\bigcup_{p \in S_{k}} p=\{(i, j): \exists p \in$ $\left.S_{k}:(i, j) \in p\right\}$. Each arc is included at most once in $R_{k}$, even if it is present in more than one of the paths, so $\left|R_{k}\right| \leq|A|$. We assume $S_{k}$ contains all paths given by the arcs in $R_{k}$.

In an OSPF-network, integral weights $w_{i j} \geq 1$ are associated with each link $(i, j) \in A$. The paths used by the routers are those with minimal sum of weights $w$, i.e. the routers find the shortest paths to the destinations. We will use the notation $W(p)=\sum_{(i, j) \in p} w_{i j}$, i.e. $W(p)$ denotes the total weight of path $p$. If $W(p) \leq W(q)$ for all paths $q$ between the same pair of nodes, $p$ is called a minimal weight path. If $W(p)<W(q)$ for all paths $q \neq p$ between the same pair of nodes, $p$ is called a unique minimal weight path. This notation is also extended to sets of paths as follows. Let $P$ be a set of paths between a pair of nodes. If $W(p)=W\left(p^{\prime}\right)$ for all paths $p \in P, p^{\prime} \in P$ and $W(p)<W(q)$ for all paths $p \in P$ and for all paths $q \notin P$ between the same pair of nodes, $P$ is called a unique minimal weight path set.

The goal is to find weights $w$ such that $S_{k}$ becomes a unique minimal weight path set from node $o_{k}$ to node $d_{k}$, for all $k$. Such a set of weights is said to be compatible with the given paths, so we use the term compatible weights. Obviously, a difficulty is that the same set of weights are used for all the sets $S_{k}$. The difference between two different sets of compatible weights is unimportant, so the question of whether or not a set of compatible weights exist is more important than which to choose. If compatible weights do not exist, we wish to find a small set of paths or links that prohibits the existence of compatible weights. Our practical goal then is to extract information about how to modify the paths, in order to enable the existence of compatible weights.

Lemma 1 If a desired routing path contains a directed cycle, no compatible weights exist.
Proof: Since a cycle leads to an already visited node, it can only have minimal weight if its total weight is zero or less. However, since no weight may be less than one, the total weight of a cycle is positive.

Let $p(s, t)$ denote a path $p$ (represented by the included arcs) from node $s$ to node $t, N(p)$ the set of nodes included in path $p$, and $N\left(S_{k}\right)$ the set of nodes included in any path in $S_{k}$.
Definition 1 If $p$ and $q$ are two paths and if $q \subset p$, then $q$ is called a subpath of $p$.
Lemma 2 If $p$ is a minimum weight path with respect to $w$ and if $q(s, t)$ is a subpath of $p$, then $q(s, t)$ is a minimum weight path from $s$ to $t$.
Proof: If not, there exists another path $r(s, t)$ with $W(r(s, t))<W(q(s, t))$. Then we can replace the subpath $q(s, t)$ by $r(s, t)$ in $p$, and get a path with less weight, which contradicts the assumption that $p$ is a minimum weight path.

Thus any subpath of any path in $S_{k}$ should be a minimal weight path.

There cannot be two different unique shortest paths between two nodes, so if two desired paths pass the same pair of nodes, the paths between these two nodes must be identical. This property has previously been called suboptimality in for example [3]. In [3], it is shown that in the undirected single path case, a necessary condition for the existence of compatible weights is that all pairs of desired paths are suboptimal.

We will now generalize the concept of suboptimality for the case with directed paths and load balancing, under the name "subpath consistency" (since it is a requirement that certain subpaths shall be consistent). We say that two sets of desired paths are subpath inconsistent if both sets contain directed paths from one node to another, and if these sets are not identical. Two sets of desired paths which are not subpath inconsistent with respect to any pair of nodes are called subpath consistent. If we let $S_{k}^{S}(s, t)$ be the set of all subpaths from node $s$ to node $t$ of paths in $S_{k}$, we can make a precise definition of subpath consistency.
Definition $2 S_{k}$ and $S_{l}$ are subpath consistent if $S_{k}^{S}(s, t)=S_{l}^{S}(s, t)$ for all $s \in N$ and $t \in N$ such that $S_{k}^{S}(s, t) \neq \emptyset$ and $S_{l}^{S}(s, t) \neq \emptyset$.
Thus, if $S_{k}$ and $S_{l}$ are subpath inconsistent, $S_{k}^{S}(s, t) \neq$ $\emptyset, S_{l}^{S}(s, t) \neq \emptyset$, and $\left(S_{k}^{S}(s, t) \backslash S_{l}^{S}(s, t)\right) \cup\left(S_{l}^{S}(s, t) \backslash\right.$ $\left.S_{k}^{S}(s, t)\right) \neq \emptyset$ for some $s, t \in N$. Clearly, if $\mid N\left(S_{k}\right) \cap$ $N\left(S_{l}\right) \mid<2$, there does not exist two such nodes, so two sets of paths containing at most one node in common are
always subpath consistent. As the definition of subpath consistency really concerns subpaths, the following is not surprising.
Lemma 3 If the paths $p_{1}$ and $p_{2}$ are subpath consistent, then any two subpaths $q_{1} \subset p_{1}$ and $q_{2} \subset p_{2}$ are subpath consistent.
Corollary 1 If $S_{k}$ and $S_{l}$ are subpath consistent, then for any subpath $p(s, t)$ of any path in $S_{k}$, either $s$ and $t$ does not both belong to $N\left(S_{l}\right)$, or $p(s, t)$ is a subpath of some path in $S_{l}$.
We will now show that subpath consistency is a necessary condition for the existence of compatible weights.
Lemma 4 If two sets of desired paths are subpath inconsistent, there exist no compatible weights.
Proof: If $S_{k}$ and $S_{l}$ are subpath inconsistent, then there exists a subpath $p(s, t) \in\left(S_{k}^{S}(s, t) \backslash S_{l}^{S}(s, t)\right) \cup$ $\left(S_{l}^{S}(s, t) \backslash S_{k}^{S}(s, t)\right)$ for some $s \in N$ and $t \in N$ such that $S_{k}^{S}(s, t) \neq \emptyset$ and $S_{l}^{S}(s, t) \neq \emptyset$. If $p(s, t) \in S_{k}^{S}(s, t) \backslash S_{l}^{S}(s, t)$, then $p(s, t)$ should be a minimal weight path according to $S_{k}$ but not according to $S_{l}$. If $p(s, t) \in S_{l}^{S}(s, t) \backslash S_{k}^{S}(s, t)$, then $p(s, t)$ should be a minimal weight path according to $S_{l}$ but not according to $S_{k}$. Since none of these cases can be true, there exist no compatible weights.

Let us now set up a mathematical model for finding compatible weights, if they exist. The model should give weights $w$ such that all paths in $S_{k}$ should be the shortest paths from the node $o_{k}$ to the node $d_{k}$. All other paths should be more expensive. Let us denote the set of all paths in the graph from node $o_{k}$ to node $d_{k}$ by $P_{k}$. We require that all paths in $P_{k} \backslash S_{k}$ should have a larger sum of weights than the paths in $S_{k}$. Since the weights will be integral, the least difference will be one, so we get constraints (1) below. Furthermore, all shortest paths between a pair of nodes must obviously have the same cost, which yields constraints (2). (In the single path case, constraints (2) are not present, as there is only one path in each $S_{k}$.)

P1:

$$
\begin{align*}
& \sum_{(i, j) \in q} w_{i j}-\sum_{(i, j) \in p} w_{i j} \geq 1 \forall p \in S_{k} \\
& \forall q \in P_{k} \backslash S_{k}, \forall k  \tag{1}\\
& \sum_{(i, j) \in r} w_{i j}-\sum_{(i, j) \in p} w_{i j}=0 \forall p \in S_{k}: p \neq r \\
& \forall r \in S_{k}, \forall k  \tag{2}\\
& w_{i j} \geq 1, \text { integer } \forall(i, j) \in A \tag{3}
\end{align*}
$$

By construction it is obvious that any feasible solution
to P 1 is a set of compatible weights. On the other hand, any set of compatible weights must satisfy (1), (2) and (3), and must therefore be a feasible solution to P1. Thus we have the following.
Lemma 5 Pl has a feasible solution if and only if there exists a compatible set of weights.
A feasible solution to P1 is not unique. For example, multiplying a feasible set of weights with a positive integer, yields another feasible solution. We consider all feasible sets of weights as equivalent, although in practice, unnecessarily large weights might be avoided. Lemmas 4 and 5 yield the following.
Lemma 6 If two sets of paths are not subpath consistent, then PI has no feasible solution.
A disadvantage of the model P1 is the large number of constraints, which comes from the large number of possible paths in a network. Each path in $S_{k}$ is compared to all other paths from $o_{k}$ to $d_{k}$, either in constraint(s) (4) or in (5).

Letting $C_{k}$ denote the (unknown) minimal weight of a path from $o_{k}$ to $d_{k}$, P1 can be rewritten as follows, where $C_{k}$ is variable.

P2:

$$
\begin{align*}
\sum_{(i, j) \in q} w_{i j} & \geq C_{k}+1 \forall q \in P_{k} \backslash S_{k}, \forall k  \tag{4}\\
\sum_{(i, j) \in p} w_{i j} & =C_{k} \forall p \in S_{k}, \forall k  \tag{5}\\
w_{i j} & \geq 1, \text { integer } \forall(i, j) \in A \tag{6}
\end{align*}
$$

The number of constraints is less in P2 than in P1 (unless $\left|S_{k}\right|=1 \forall k$ ). By simply using (5) one can easily verify that any feasible solution of P1 is feasible in P2, and that any feasible solution of P2 is feasible in P1. Thus P2 has a feasible solution if and only if P1 has a feasible solution.

According to [16], there exists a node potential, $\pi_{i} \forall i \in N$, satisfying $w_{i j}+\pi_{i}-\pi_{j} \geq 0 \forall(i, j) \in A$ if all directed cycles in a graph have a non-negative sum of weights. Furthermore, the node potential $\pi$ can be chosen integer-valued if $w$ is integer-valued. It is also shown that $p$ is a minimum weight path if and only if there is a potential satisfying $w_{i j}=\pi_{j}-\pi_{i} \forall(i, j) \in p$. If we sum up these equalities for a minimal weight path $p$ starting in node $s$ and ending in node $t$, we get $W(p(s, t))=\pi_{t}-\pi_{s}$, since all other node potentials cancel out. A path $q$ from $s$ to $t$ which is not a minimum weight path must therefore satisfy $W(q(s, t))>\pi_{t}-\pi_{s}$.

In the weight finding problem, we need different node potentials for each set $S_{k}$, so we introduce $\pi_{i}^{k} \forall i \in$ $N, \forall k=1, \ldots, K$. We have

$$
W(p)=\pi_{d_{k}}^{k}-\pi_{o_{k}}^{k} \forall p \in S_{k}, \forall k
$$

and according to Lemma 2 , this applies to any subpath of $p$, i.e.

$$
\begin{gathered}
W(p(s, t))=\pi_{t}^{k}-\pi_{s}^{k} \text { for all } \\
p(s, t) \in S_{k}^{S}(s, t) \quad \forall s \in N\left(S_{k}\right), t \in N\left(S_{k}\right) \forall k
\end{gathered}
$$

Especially, this applies to subpaths consisting of single links.

$$
w_{i j}=\pi_{i}^{k}-\pi_{j}^{k} \forall(i, j) \in p, \forall p \in S_{k}, \forall k
$$

Here, $(i, j) \in p, p \in S_{k}$ is more efficiently written as $(i, j) \in R_{k}$, since this avoids representing links more than once. Furthermore we should have

$$
W(q) \geq \pi_{d_{k}}^{k}-\pi_{o_{k}}^{k}+1 \forall q \in P_{k} \backslash S_{k}, \forall k
$$

(since the weights and $\pi_{i}^{k}$ are integer-valued). We thus get the following model.

P3:

$$
\begin{align*}
& w_{i j}+\pi_{i}^{k}-\pi_{j}^{k}=0 \forall(i, j) \in R_{k}, \forall k  \tag{7}\\
& \sum_{(i, j) \in q} w_{i j}+\pi_{o_{k}}^{k}-\pi_{d_{k}}^{k} \geq 1 \forall q \in P_{k} \backslash S_{k}, \forall k  \tag{8}\\
& w_{i j} \geq 1, \text { integer } \forall(i, j) \in A \tag{9}
\end{align*}
$$

Theorem 1 P3 has a feasible solution if and only if P1 has a feasible solution.
Proof: Assume that we have a feasible solution to P1. Now let $\pi_{i}^{k}$ be equal to the minimum sum of weights from $o_{k}$ to $i$. We then get $w_{i j}=\pi_{j}^{k}-\pi_{i}^{k} \forall(i, j) \in$ $R_{k}, \forall k$, so (7) is satisfied. The paths in $P_{k} \backslash S_{k}$ are not minimum weight paths, so all constraints in (8) are satisfied. Constraints (9 are identical to (3), so we conclude that P3 has a feasible solution.

Assume now that we have a feasible solution to P3. Summing up constraints (7) over any path $p$ in $S_{k}$ yields $W(p)=\pi_{d_{k}}^{k}-\pi_{o_{k}}^{k}$. This ensures that constraints (2) are satisfied. Inserting $W(p)=\pi_{d_{k}}^{k}-\pi_{o_{k}}^{k}$ into (8) immediately yields constraints (1). Thus a feasible solution to P 3 is also feasible in P1.

We conclude that P3 has a feasible solution if and only if there exists a set of compatible weights.

## 3. Combining paths into SP-graphs

P3 may have a large number of constraints in (8), due to the large number of paths in the sets $P_{k} \backslash S_{k}$. Each path in $P_{k} \backslash S_{k}$ contains at least one subpath (possibly a single arc) that starts and ends at nodes spanned by $R_{k}$, and does not included any arc in $R_{k}$. These subpaths are not included in any path $p \in S_{k}$, and if we ensure that these subpaths are not part of a minimal weight path from node $o_{k}$ to node $d_{k}$, we have ensured that the paths in $P_{k} \backslash S_{k}$ have a larger sum of weights than the paths in $S_{k}$. Since many paths in $P_{k} \backslash S_{k}$ include the same subpath, this can be used for reducing the size of the model.

We may also decrease the size of the model by combining sets of minimum weight paths into shortest path graphs, SP-graphs. We will use the term "origin" ("destination") to denote any node in an SP-graph without predecessors (successors). An SP-graph is defined to be a set of arcs which contains at least one path from each origin to each destination, and does not contain any directed cycle. (Consequently each SP-graph must contain at least one origin and at least one destination.)
$R_{k}$ is an example of an SP-graph with a single origin and a single destination. Since an SP-graph no longer is associated with a single node pair $\left(o_{k}, d_{k}\right)$, we introduce the index $l$ for SP-graphs and let $A_{l}$ denote SP-graph $l$.

A set of weights is compatible with an SP-graph if all paths in the SP-graph are minimum weight paths and any other path from an origin to a destination is not a minimum weight path. A set of weights is called compatible if it is compatible with each SP-graph. Let us also extend the meaning of subpath consistency to SP-graphs. Let $A_{l}^{S}(s, t) \subseteq A_{l}$ denote the set of arcs included in any path from $s$ to $t$ in $A_{l}$.

Definition $3 A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ are subpath consistent if $A_{l^{\prime}}^{S}(s, t)=A_{l^{\prime \prime}}^{S}(s, t)$ for all $s \in N$ and $t \in N$ such that $A_{l^{\prime}}^{S}(s, t) \neq \emptyset$ and $A_{l^{\prime \prime}}^{S}(s, t) \neq \emptyset$.

We will now give a model for finding compatible weights to a set of SP-graphs (possibly with multiple origins and destinations). Let $V_{l}$ be the set of node pairs spanned by each $A_{l}$, i.e. let $V_{l}=\left\{(s, t): s \in N\left(A_{l}\right)\right.$ and $\left.t \in N\left(A_{l}\right)\right\}$. For each $(s, t) \in V_{l}$, let $T_{s t}^{l}$ be the set of paths from node $s$ to node $t$ completely outside $A_{l}$. This means that each path in $T_{s t}^{l}$ starts at node $s$, ends at node $t$, and does not pass any other node spanned by
$A_{l}$. We can now formulate the following model.
P4:

$$
\begin{gather*}
w_{i j}+\pi_{i}^{l}-\pi_{j}^{l}=0 \forall(i, j) \in A_{l}, \forall l \\
\sum_{(i, j) \in q} w_{i j}+\pi_{s}^{l}-\pi_{t}^{l} \geq 1 \forall q \in T_{s t}^{l}, \forall(s, t) \in V_{l}, \forall l \tag{11}
\end{gather*}
$$

$$
\begin{equation*}
w_{i j} \geq 1, \text { integer } \forall(i, j) \in A \tag{12}
\end{equation*}
$$

The first set of constraints is in principle identical to (7), and ensures that all paths in $A_{l}$ are minimum weight paths. The second set of constraints ensures that each subpath that starts and ends in $A_{l}$ and does not pass any other node spanned by $A_{l}$ is not a part of a minimum weight path from an origin of $A_{l}$ to a destination of $A_{l}$. Theorem 2 P4 has a feasible solution if and only if compatible weights exist.
P 4 is in principle a modification of P3. If each SPgraph is equal to an arc set $R_{k}$, the result follows from Lemma 5 and Theorem 1. Below we show how SPgraphs can be combined into larger and fewer SP-graphs without changing the feasibility of P4. As an alternative, a general proof of Theorem 2 is found in [10].

The number of constraints (11) could be large if there are many nodes not spanned by each SP-graph. However, if each SP-graph spans all nodes in the node set, set (11) consists of only $\sum_{l}\left(|A|-\left|A_{l}\right|\right)$ constraints, so here we see an incentive to make the SP-graphs as large as possible. There are however restrictions on which sets of paths can be combined. Starting from the case when each SP-graph has one origin and one destination, in which case P4 has a feasible solution if and only if P3 has one, we will increase the sizes of the SP-graphs, while maintaining the property that P 4 has a feasible solution if and only if compatible weights exist.

Assume that a number of SP-graphs are given. We now seek the answer to the following question. Does combining two of them, $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$, into one SP-graph, i.e. letting $A_{C}=A_{l^{\prime}} \cup A_{l^{\prime \prime}}$, yield the same result in P4 as using $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ separately? Let P4s denote the case when $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ are treated separately, and P 4 c the case when they are combined. We allow this combination if either compatible weights exist for both P 4 s and P 4 c , or compatible weights do not exist for any of P 4 s and P 4 c .

We will consider the case when $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ has one and the same origin, denoted by $o$. Let $\delta^{-}(\bar{N})$ denote all arcs leaving node set $\bar{N}, \delta^{+}(\bar{N})$ all arcs entering node set $\bar{N}$, and $\gamma(\bar{N})$ all arcs with both endpoints in


Fig. 1. A partition allowing combination.
node set $\bar{N} . N(\bar{A})$ denotes the set of nodes spanned by the arc set $\bar{A}$.

We find that two SP-graphs can be combined if they have the same (single) origin, are identical in a subgraph containing the origin, do not contain any arcs entering the identical subgraph, and are completely separate in the rest of the graph. More specifically, let $N_{0} \cup N_{1} \cup$ $N_{2}=N\left(A_{l^{\prime}} \cup A_{l^{\prime \prime}}\right)$, i.e. the set of nodes spanned by $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ is partitioned into the sets $N_{0}, N_{1}$ and $N_{2}$. Now assume that $o \in N_{0}$ and $\gamma\left(N_{0}\right) \cap A_{l^{\prime}}=\gamma\left(N_{0}\right) \cap A_{l^{\prime \prime}}$, i.e. that $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ are identical in $N_{0}$. Also assume that $N\left(A_{l^{\prime}}\right) \cap N_{2}=\emptyset$ and $N\left(A_{l^{\prime \prime}}\right) \cap N_{1}=\emptyset$, i.e. that no part of $A_{l^{\prime}}$ is in $N_{2}$ and no part of $A_{l^{\prime \prime}}$ is in $N_{1}$. Furthermore, we require that $A_{l^{\prime}} \cap \delta^{+}\left(N_{0}\right)=\emptyset$ and $A_{l^{\prime \prime}} \cap \delta^{+}\left(N_{0}\right)=\emptyset$, i.e. that no arc in $A_{l^{\prime}}$ or $A_{l^{\prime \prime}}$ enters the node set $N_{0}$. See Figure 1 for an example of such a partition. Here $A_{l^{\prime}}$ is indicated with dashed lines and $A_{l^{\prime \prime}}$ by solid lines. Both have node 1 as origin. The partition is given by $N_{0}=\{1,2,3,4\}, N_{1}=\{5,6,7\}$ and $N_{2}=\{8,9,10,11\}$. An important consequence of the assumptions is that $A_{l^{\prime}}$ consists of arcs in $\gamma\left(N_{0}\right)$, arcs in $\gamma\left(N_{1}\right)$ and arcs directed from nodes in $N_{0}$ to nodes in $N_{1}$. So if a path in $A_{l^{\prime}}$ leaves the node set $N_{0}$, the path enters and stays inside $N_{1}$ until the destination is reached. Similarly, $A_{l^{\prime \prime}}$ consists of arcs in $\gamma\left(N_{0}\right)$, arcs in $\gamma\left(N_{2}\right)$ and arcs directed from nodes in $N_{0}$ to nodes in $N_{2}$. The paths in $A_{l^{\prime \prime}}$ first visit a sequence of nodes in $N_{0}$ and then a sequence of nodes in $N_{2}$.
Lemma 7 If there exists a partitioning $N_{0} \cup N_{1} \cup N_{2}=$ $N\left(A_{l^{\prime}} \cup A_{l^{\prime \prime}}\right)$, such that $o \in N_{0}$ is the only origin in $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}, \gamma\left(N_{0}\right) \cap A_{l^{\prime}}=\gamma\left(N_{0}\right) \cap A_{l^{\prime \prime}}, N\left(A_{l^{\prime}}\right) \cap$ $N_{2}=\emptyset, N\left(A_{l^{\prime \prime}}\right) \cap N_{1}=\emptyset, A_{l^{\prime}} \cap \delta^{+}\left(N_{0}\right)=\emptyset$ and $A_{l^{\prime \prime}} \cap \delta^{+}\left(N_{0}\right)=\emptyset$, then P4c has a feasible solution if and only if P4s has one. Therefore the SP-graphs may
be combined.
Proof: We study how P4 changes as a result of combining $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$. The $w$-variables are the same in P 4 s and P 4 c , but the two sets of node prices $\pi^{1}$ and $\pi^{2}$ in P 4 s will be combined into one set, $\pi^{C}$, in P 4 c .

Assume first that P 4 s has a feasible solution, $\left(\bar{w}, \bar{\pi}^{1}, \bar{\pi}^{2}, \ldots, \bar{\pi}^{l}\right)$. The solution remains feasible if we increase/decrease all components in $\bar{\pi}^{2}$ with a constant. Letting $\hat{\pi}_{i}^{2}=\bar{\pi}_{i}^{2}+\left(\bar{\pi}_{o}^{1}-\bar{\pi}_{o}^{2}\right) \forall i \in N$ yields $\bar{\pi}_{o}^{1}=\hat{\pi}_{o}^{2}$, which implies $\bar{\pi}_{i}^{1}=\hat{\pi}_{i}^{2} \forall i \in N_{0}$, since $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ are identical in $N_{0}$ and since (10) yields $\bar{w}_{i j}=\bar{\pi}_{j}^{1}-\bar{\pi}_{i}^{1} \forall(i, j) \in \gamma\left(N_{0}\right) \cap A_{l^{\prime}}$ and $\bar{w}_{i j}=\hat{\pi}_{j}^{2}-\hat{\pi}_{i}^{2} \forall(i, j) \in \gamma\left(N_{0}\right) \cap A_{l^{\prime \prime}}$.

P 4 s and P 4 c have one constraint in set (10) for each arc in $A_{C}$. For arcs in $\gamma\left(N_{0}\right)$, the constraints are identical, so the same $\pi$-solution can be used. We set $\pi_{i}^{C}=$ $\bar{\pi}_{i}^{1} \forall i \in N_{0}$. Next we note that $\bar{\pi}^{1}$ together with $\bar{w}$ will ensure that (10) is satisfied for arcs in $A_{l^{\prime}} \backslash \gamma\left(N_{0}\right)$, so we set $\pi_{i}^{C}=\bar{\pi}_{i}^{1} \forall i \in N_{1}$. Similarly, $\hat{\pi}^{2}$ together with $\bar{w}$ satisfies (10) for arcs in $A_{l^{\prime \prime}} \backslash \gamma\left(N_{0}\right)$, so we set $\pi_{i}^{C}=\hat{\pi}_{i}^{2} \forall i \in N_{2}$.

What remains is constraint set (11), stating that the paths completely outside $A_{C}$ should not be minimum weight paths. We have $\pi_{i}^{C}=\bar{\pi}_{i}^{1} \forall i \in N_{0} \cup N_{1}$, so P4c satisfies all constraints from set (11) concerning paths starting and ending in $N_{0} \cup N_{1}$ without passing nodes in $N_{2}$. Similarly, we have $\pi_{i}^{C}=\bar{\pi}_{i}^{2} \forall i \in N_{0} \cup N_{2}$, so P4c satisfies all constraints from set (11) concerning paths starting and ending in $N_{0} \cup N_{2}$ without passing nodes in $N_{1}$.

Now we study paths starting in $N_{2}$ and ending in $N_{1}$. Suppose that one constraint in set (11) is not satisfied in P 4 c , for example the constraint defined for the path $q(s, t)$ from $s \in N_{2}$ to $t \in N_{1}$. Since $\pi$ can be chosen integer-valued if $w$ is integer-valued, see [16], we then have $\sum_{(i, j) \in q(s, t)} \bar{w}_{i j}+\pi_{s}^{C}-\pi_{t}^{C} \leq 0$. Now consider a path $q=\{q(o, s), q(s, t)\}$, where $q(o, s)$ is an arbitrary path from $o$ to $s$ completely inside $A_{C}$. We then get

$$
\begin{aligned}
0 & =\sum_{(i, j) \in q(o, s)}\left(\bar{w}_{i j}+\pi_{i}^{C}-\pi_{j}^{C}\right)=\sum_{(i, j) \in q(o, s)} \bar{w}_{i j}+ \\
\pi_{o}^{C}- & \pi_{s}^{C} \geq \sum_{(i, j) \in q(o, s)} \bar{w}_{i j}+\pi_{o}^{C}-\pi_{s}^{C} \\
& +\sum_{(i, j) \in q(s, t)} \bar{w}_{i j}+\pi_{s}^{C}-\pi_{t}^{C}=\sum_{(i, j) \in q} \bar{w}_{i j}+\pi_{o}^{C}- \\
\pi_{t}^{C} & =\sum_{(i, j) \in q} \bar{w}_{i j}+\bar{\pi}_{o}^{1}-\bar{\pi}_{t}^{1} .
\end{aligned}
$$

The first equality holds since (10) is satisfied for all $(i, j) \in q(o, s)$. The node prices cancel out in the second and the fourth equality, and the last equality holds since
$\pi_{i}^{C}=\bar{\pi}_{i}^{1} \forall i \in N_{0} \cup N_{1}$. The inequality follows from the assumption that (11) is not satisfied for $q(s, t)$.

We thus get $\sum_{(i, j) \in q} \bar{w}_{i j}+\bar{\pi}_{o}^{1}-\bar{\pi}_{t}^{1} \leq 0$, i.e. $W(q) \leq$ $\bar{\pi}_{t}^{1}-\bar{\pi}_{o}^{1}$, so the path $q$ does not have a larger sum of weights than the path from $o$ to $t$ in $A_{l^{\prime}}$. The path $q$ is not completely inside $A_{l^{\prime}}$ since $s \in N_{2}$, so this contradicts that the solution is feasible in P 4 s . Thus, assuming that (11) is not satisfied yields a contradiction, so we conclude that (11) is satisfied for all paths starting in $N_{2}$ and ending in $N_{1}$. This reasoning can be repeated for paths starting in $N_{1}$ and ending in $N_{2}$, so we conclude that P 4 c has a feasible solution if P 4 s has a feasible solution.

Finally we note that if P4c has a feasible solution $\left(\bar{w}, \bar{\pi}^{C}, \bar{\pi}^{3}, \ldots \bar{\pi}^{l}\right)$, we immediately get a feasible solution to P 4 s by using $\pi^{1}=\bar{\pi}^{C}$ and $\pi^{2}=\bar{\pi}^{C}$.

The result also holds if the two SP-graphs are identical in a subgraph containing the common destination, and completely separate in the rest of the graph. This means that two SP-graphs can be combined if there exists a partitioning $N_{0} \cup N_{1} \cup N_{2}=N\left(A_{l^{\prime}} \cup A_{l^{\prime \prime}}\right)$ such that $d \in N_{0}$ is the only destination of $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}, \gamma\left(N_{0}\right) \cap A_{l^{\prime}}=\gamma\left(N_{0}\right) \cap A_{l^{\prime \prime}}, N\left(A_{l^{\prime}}\right) \cap N_{2}=\emptyset$, $N\left(A_{l^{\prime \prime}}\right) \cap N_{1}=\emptyset, A_{l^{\prime}} \cap \delta^{-}\left(N_{0}\right)=\emptyset$ and $A_{l^{\prime \prime}} \cap$ $\delta^{-}\left(N_{0}\right)=\emptyset$. This can be shown by simply reversing the roles of origins and destinations in the previous proof.
Corollary 2 If two SP-graphs are subpath consistent and have the same node as the only origin, they may be combined.
Proof: Letting $N_{0}=N\left(A_{l^{\prime}}\right) \cap N\left(A_{l^{\prime \prime}}\right)$, it is easy to see that the assumptions of Lemma 7 are satisfied.

If $N_{1}=\emptyset$ (i.e. if all destinations of $A_{l^{\prime}}$ are in $N_{0}$ ) in the partitioning in Lemma 7 then $A_{l^{\prime}}$ is a subset of $A_{l^{\prime \prime}}$. In that case all sets of weights that are compatible with $A_{l^{\prime \prime}}$ will also be compatible with $A_{l^{\prime}}$, so $A_{l^{\prime}}$ does not add anything to P 4 . $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ may be combined, but the combination of them will be identical to $A_{l^{\prime \prime}}$. Therefore we can simply delete $A_{l^{\prime}}$. The same applies to $A_{l^{\prime \prime}}$ if $N_{2}=\emptyset$.

An SP-graph can sometimes be combined with a part of another SP-graph. This could be useful when two SPgraphs are similar, but do not have the same origin or the same destination. Recall that $A_{l}^{S}(s, t)$ is the set of arcs included in any path from node $s$ to node $t$ in $A_{l}$. Lemma 8 P4 remains feasible/infeasible if $A_{l}^{S}(s, t)$ is defined as a separate SP-graph and added to the set of $S P$-graphs, for any pair of nodes $s$ and $t$.
Proof: The lemma is only interesting if $A_{l}^{S}(s, t) \neq \emptyset$. If P4 has a feasible solution, it follows from Theorem 2 that some set of weights is compatible with each origi-
nal SP-graph. $A_{l}^{S}(s, t)$ contains exactly the same paths from $s$ to $t$ as $A_{l}$, so the weights are also compatible with $A_{l}^{S}(s, t)$. Thus P4 remains feasible if the SPgraph $A_{l}^{S}(s, t)$ is added. P 4 is infeasible when no set of weights is compatible with each of the original SPgraphs. Adding an SP-graph cannot change this fact. $\square$

Lemma 7 can be used repeatedly, since several destinations are allowed. If we start from $A_{l^{\prime}}=R_{1}, A_{l^{\prime \prime}}=$ $R_{2}$, etc., and repeatedly combine SP-graphs with the same origin, the resulting SP-graphs will contain an arborescence. If all resulting SP-graphs are spanning, each path in $T_{s t}^{l}$ consists of a single arc, so P4 can be simplified to the following problem.

P5:

$$
\begin{array}{r}
w_{i j}+\pi_{i}^{l}-\pi_{j}^{l}=0 \forall(i, j) \in A_{l}, \forall l \\
w_{i j}+\pi_{i}^{l}-\pi_{j}^{l} \geq 1 \forall(i, j) \notin A_{l}, \forall l \\
w_{i j} \geq 1, \text { integer } \forall(i, j) \in A \tag{15}
\end{array}
$$

Letting $m$ denote the number of SP-graphs, we find that P5 has only $|A|(m+1)$ constraints since each arc either is included in (13) or (14). In [11] we prove that, if all SP-graphs are spanning, P5 has a feasible solution if and only if there exists a set of compatible weights.

Let us also mention a case where SP-graphs may not be combined. Assume that $o_{1} \in N\left(A_{l^{\prime}}\right) \backslash N\left(A_{l^{\prime \prime}}\right)$, $d_{1} \in N\left(A_{l^{\prime}}\right) \backslash N\left(A_{l^{\prime \prime}}\right), o_{2} \in N\left(A_{l^{\prime \prime}}\right) \backslash N\left(A_{l^{\prime}}\right), d_{2} \in$ $N\left(A_{l^{\prime \prime}}\right) \backslash N\left(A_{l^{\prime}}\right)$, and $N\left(A_{l^{\prime}}\right) \cap N\left(A_{l^{\prime \prime}}\right) \neq \emptyset$. If we combine $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$, this SP-graph will not only contain the paths from $o_{1}$ to $d_{1}$ and from $o_{2}$ to $d_{2}$, but also from $o_{1}$ to $d_{2}$ and from $o_{2}$ to $d_{1}$. A feasible solution of P 4 s ensures that the paths from $o_{1}$ to $d_{1}$ and from $o_{2}$ to $d_{2}$ are minimal weight paths, but not that the paths from $o_{1}$ to $d_{2}$ and from $o_{2}$ to $d_{1}$ are of minimal weight. Therefore, P4s being feasible does not imply that P4c is feasible. Our conclusion is the following.
Lemma 9 Two SP-graphs cannot be combined if they create a new origin-destination pair.
Thus, if the two sets of paths have at least one node in common, but different origins and different destinations, they may not be combined.

## 4. Valid cycles

Our main goal is to analyze instances that lack compatible weights. In [7] this is done by finding unbounded solutions to the LP-dual of the weight finding problem. These unbounded solutions are represented by cycles, in which dual variables can be changed infinitely. Here
we will not use duality, but verify the effect of such cycles in another way.

Recall that each $A_{l}$ contains at least one path from each origin to each destination, and no $A_{l}$ contains a directed cycle. Now consider two SP-graphs, $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$, and a cycle $C \subseteq A, C=F \cup B$, where $F$ is the set of arcs considered to be used forward and $B$ the set of arcs considered to used backward, in relation to SP-graph $A_{l^{\prime}}$.
Definition 4 A cycle $C=F \cup B$ is called feasible if $B \subseteq A_{l^{\prime}}$ and $F \subseteq A_{l^{\prime \prime}}$.
Definition 5 The arc $(i, j)$ is called eligible if $(i, j) \in$ $\left(F \backslash A_{l^{\prime}}\right) \cup\left(B \backslash A_{l^{\prime \prime}}\right)$.
In words, an eligible arc lies in $F$ but not in $A_{l^{\prime}}$ or in $B$ but not in $A_{l^{\prime \prime}}$.
Definition 6 The cycle $C=F \cup B$ is called improving if it contains at least one eligible arc.
Definition 7 A cycle $C=F \cup B$ is called valid if there exist two indices $l^{\prime}$ and $l^{\prime \prime}$ such that the cycle is feasible and improving.
Theorem 3 If there exists a valid cycle, then there exists no compatible set of weights.
The proof for the spanning case (i.e. when all SP-graphs span all nodes) is given in [11]. Here we give a more general proof, covering both the spanning and nonspanning cases.

Proof: Theorem 3 will be proved by summing the weights around a valid cycle, once in view of SP-graph $A_{l^{\prime}}$ and once in view of SP-graph $A_{l^{\prime \prime}}$. Obviously these two sums should be equal, but we will get a contradiction, verifying that P 4 cannot have a feasible solution.

Let $\Gamma_{l}(C)$ denote the sum of weights around the valid cycle $C=F \cup B$ in view of SP-graph $A_{l}$. Now we split $C$ into subpaths, depending on whether an arc is in $F$ or $B$ and also whether it is in $A_{l^{\prime}}$ or not. Let $P_{l^{\prime}}^{F}$ be the set of subpaths in $F \cap A_{l^{\prime}}, P_{l^{\prime}}^{B}$ the set of subpaths in $B \cap A_{l^{\prime}}, Q_{l^{\prime}}^{F}$ be the set of subpaths in $F \backslash A_{l^{\prime}}$, and $Q_{l^{\prime}}^{B}$ be the set of subpaths in $B \backslash A_{l^{\prime}}$. Furthermore, let $p_{r l^{\prime}}^{F}$ denote subpath $r$ in the set $P_{l^{\prime}}^{F}$, etc.

We require that each subpath starts and ends in nodes spanned by $A_{l^{\prime}}$, and does not pass any other node spanned by $A_{l^{\prime}}$. This means that each subpath in $P_{l^{\prime}}^{F} \cup P_{l^{\prime}}^{B}$ consists of a single arc, while the subpaths in $Q_{l^{\prime}}^{F} \cup Q_{l^{\prime}}^{B}$ may consist of more than one arc. The cycle $C$ consists of all these subpaths, and we have the following.

$$
F \cap A_{l^{\prime}}=\bigcup_{r \in P_{l^{\prime}}^{F}} p_{r l^{\prime}}^{F}, \quad B \cap A_{l^{\prime}}=\bigcup_{r \in P_{l^{\prime}}^{B}} p_{r l^{\prime}}^{B}, \quad F \backslash A_{l^{\prime}}=
$$

$\bigcup_{r \in Q_{l^{\prime}}^{F}} q_{r l^{\prime}}^{F}, \quad B \backslash A_{l^{\prime}}=\bigcup_{r \in Q_{l^{\prime}}^{B}} q_{r l^{\prime}}^{B}$.
We now get the following sum of weights around the cycle.

$$
\Gamma_{l^{\prime}}(C)=\sum_{(i, j) \in F} w_{i j}-\sum_{(i, j) \in B} w_{i j}=\sum_{(i, j) \in F \cap A_{l^{\prime}}} w_{i j}+
$$

$$
\sum_{(i, j) \in F \backslash A_{l^{\prime}}} w_{i j}-\sum_{(i, j) \in B \cap A_{l^{\prime}}} w_{i j}-\sum_{(i, j) \in B \backslash A_{l^{\prime}}} w_{i j}=
$$

$$
\sum_{r \in P_{l^{\prime}}^{F}} \sum_{(i, j) \in p_{r l^{\prime}}^{F}} w_{i j}+\sum_{r \in Q_{l^{\prime}}^{F}} \sum_{(i, j) \in q_{r l^{\prime}}^{F}} w_{i j}-\sum_{r \in P_{l^{\prime}}^{B}} \sum_{(i, j) \in p_{r l^{\prime}}^{B}} w_{i j}-
$$

$$
\sum_{r \in Q_{l^{\prime}}^{B}} \sum_{(i, j) \in q_{r l^{\prime}}^{B}} w_{i j}
$$

Now introduce $s$ by letting $w_{i j}=\pi_{j}^{l}-\pi_{i}^{l}+$ $s_{i j}^{l} \forall(i, j) \in A$. We have $s_{i j}^{l}=0 \forall(i, j) \in A_{l}$, due to constraint 4.1. We also note that for any path $p(s, t)$ for any $s$ and $t$, we have

$$
\begin{aligned}
& W(p(s, t))=\sum_{(i, j) \in p(s, t)}\left(\pi_{j}^{l}-\pi_{i}^{l}+s_{i j}^{l}\right)=\pi_{t}^{l}- \\
& \pi_{s}^{l}+\sum_{(i, j) \in p(s, t)} s_{i j}^{l}
\end{aligned}
$$

For $\Gamma_{l^{\prime}}(C)$, the $\pi$ 's cancel if we make this substitution, so we get

$$
\begin{aligned}
& \Gamma_{l^{\prime}}(C)=\sum_{r \in P_{l^{\prime}}^{F}} \sum_{(i, j) \in p_{r l^{\prime}}^{F}} s_{i j}^{l^{\prime}}+\sum_{r \in Q_{l^{\prime}}^{F}} \sum_{(i, j) \in q_{r l^{\prime}}^{F}} s_{i j}^{l^{\prime}}- \\
& \sum_{r \in P_{l^{\prime}}^{B}} \sum_{(i, j) \in p_{r l^{\prime}}^{B}} s_{i j}^{l^{\prime}}-\sum_{r \in Q_{l^{\prime}}^{B}} \sum_{(i, j) \in q_{r l^{\prime}}^{B}} s_{i j}^{l^{\prime}} .
\end{aligned}
$$

Now we note that $Q_{l^{\prime}}^{B}=\emptyset$, since $B \subseteq A_{l^{\prime}}$, so the last term is equal to zero. Furthermore $s_{i j}^{l}=0$ for each arc in $A_{l^{\prime}}$, i.e. for each arc in $p_{r l^{\prime}}^{F}$ and in $p_{r l^{\prime}}^{B}$. That makes terms one and three equal to zero, so only term two remains. Now we note that if $F \subseteq A_{l^{\prime}}$, we have $Q_{l^{\prime}}^{F}=\emptyset$, which means that $\Gamma_{l^{\prime}}(C)=0$. The other possibility is $F \nsubseteq A_{l^{\prime}}$, which implies $Q_{l^{\prime}}^{F} \neq \emptyset$. Each path $r \in Q_{l^{\prime}}^{F}$ starts and ends at nodes in $N\left(A_{l^{\prime}}\right)$ and does not pass any other node spanned by $A_{l^{\prime}}$, so there is one constraint in 4.2 for each $r \in Q_{l^{\prime}}^{F}$. Therefore, we have $\sum_{(i, j) \in q_{r l^{\prime}}^{F}} s_{i j}^{l^{\prime}} \geq 1 \forall r \in Q_{l^{\prime}}^{F}$, i.e. $\Gamma_{l^{\prime}}(C)>0$. There are thus two possibilities, namely $Q_{l^{\prime}}^{F}=\emptyset$, which yields $\Gamma_{l^{\prime}}(C)=0$, and $Q_{l^{\prime}}^{F} \neq \emptyset$, which yields $\Gamma_{l^{\prime}}(C)>0$.

Let us now calculate $\Gamma_{l^{\prime \prime}}(C)$, which is the sum of weights around the same cycle, but taking $A_{l^{\prime \prime}}$ into account. Doing the same kind of partitioning of the arcs in the cycle, we get

$$
\Gamma_{l^{\prime \prime}}(C)=\sum_{r \in P_{l^{\prime \prime}}^{F}} \sum_{(i, j) \in p_{r l^{\prime \prime}}^{F}} s_{i j}^{l^{\prime \prime}}+\sum_{r \in Q_{l^{\prime \prime}}^{F}} \sum_{(i, j) \in q_{r l^{\prime \prime}}^{F}} s_{i j}^{l^{\prime \prime}}-
$$

$$
\sum_{r \in P_{l^{\prime \prime}}^{B}} \sum_{(i, j) \in p_{r l^{\prime \prime}}^{B}} s_{i j}^{l^{\prime \prime}}-\sum_{r \in Q_{l^{\prime \prime}}^{B}} \sum_{(i, j) \in q_{r l^{\prime \prime}}^{B}} s_{i j}^{l^{\prime \prime}} .
$$

Now we know that $F \subseteq A_{l^{\prime \prime}}$, so $Q_{l^{\prime \prime}}^{F}=\emptyset$. Furthermore, $s_{i j}^{l^{\prime \prime}}=0$ for all $(i, j) \in A_{l^{\prime \prime}}$, i.e. for all $(i, j)$ in $p_{r l^{\prime \prime}}^{F}$ or in $p_{r l^{\prime \prime}}^{B}$. This makes the first three terms above equal to zero, so only the last term remains. As above we can show that the this term is zero if $Q_{l^{\prime \prime}}^{B}=\emptyset$, and negative if $Q_{l^{\prime \prime}}^{B} \neq \emptyset$. So again there are two possibilities, namely $Q_{l^{\prime \prime}}^{B}=\emptyset$, which yields $\Gamma_{l^{\prime \prime}}(C)=0$, and $Q_{l^{\prime \prime}}^{B} \neq \emptyset$, which yields $\Gamma_{l^{\prime \prime}}(C)<0$.

Thus $\Gamma_{l^{\prime}}(C)=\Gamma_{l^{\prime \prime}}(C)=0$ if and only if $Q_{l^{\prime}}^{F}=$ $\emptyset$ and $Q_{l^{\prime \prime}}^{B}=\emptyset$. A valid cycle has $F \nsubseteq A_{l^{\prime}}$ and/or $B \nsubseteq A_{l^{\prime \prime}}$, while $Q_{l^{\prime}}^{F}=\emptyset$ if $F \subseteq A_{l^{\prime}}$ and $Q_{l^{\prime \prime}}^{B}=\emptyset$ if $B \subseteq A_{l^{\prime \prime}}$, so for a valid cycle, at least one of the sets $Q_{l^{\prime}}^{F}$ and $Q_{l^{\prime \prime}}^{B}$ is non-empty. If $\Gamma_{l^{\prime}}(C)=0$ then $Q_{l^{\prime}}^{F}=\emptyset$, which implies that $Q_{l^{\prime \prime}}^{B} \neq \emptyset$ and $\Gamma_{l^{\prime \prime}}(C)<0$. On the other hand, if $\Gamma_{l^{\prime \prime}}(C)=0$ then $Q_{l^{\prime \prime}}^{B}=\emptyset$, which implies that $Q_{l^{\prime}}^{F} \neq \emptyset$ and $\Gamma_{l^{\prime}}(C)>0$. A third possibility is that $Q_{l^{\prime}}^{F} \neq \emptyset$ and $Q_{l^{\prime \prime}}^{B} \neq \emptyset$, which yields $\Gamma_{l^{\prime}}(C)>0$ and $\Gamma_{l^{\prime \prime}}(C)<0$.

It is in neither of these cases possible that $\Gamma_{l^{\prime}}(C)=$ $\Gamma_{l^{\prime \prime}}(C)$. Thus there is a contradiction, which proves the theorem

This verifies that the existence of a valid cycle implies that a compatible set of weights does not exist, regardless of whether or not the SP-graphs span all nodes. However, the fact that no valid cycle exists is not enough to guarantee that compatible weights exist. Since valid cycles are obtained by comparing only two SP-graphs, it is important to include as much information as possible in each SP-graph. This is done by combining as many sets of desired paths as possible into each SP-graph.

There are instances that do not have compatible weights or valid cycles. In such cases, there are more complicated structures that can be used to explain the lack of compatible weights, see [8]. However, such structures cannot be found as efficiently as valid cycles. By combining sets of paths, one might transform such a complicated structure into a valid cycle, thereby enabling the usage of the more efficient method in Section 5 ..
Lemma 10 A valid cycle must contain at least three nodes and three arcs.
A proof of Lemma 10 is given in [11]. A stronger result is true if subpath consistency holds.
Lemma 11 If the SP-graphs are subpath consistent, a valid cycle must contain at least four nodes and four arcs.
Proof: Since an SP-graph may not contain a directed
cycle, a valid cycle with three arcs must contain two arcs in $F$ and one in $B$ (or two arcs in $B$ and one in $F$ ). Then the two arcs of the same type, $F$ (or $B$ ), must be adjacent, and thus form a path from one node, $i$, to another, $j$. Furthermore, the remaining arc, in $B$ (or $F$ ), must also lead from $i$ to $j$, in order to complete the cycle. So there are two different desired paths from $i$ to $j$. Now either both of these paths are contained in both SP-graphs, in which case no arc in this cycle can be eligible, or both of these paths are not contained in both SP-graphs, in which case the SP-graphs are not subpath consistent.

The same reasoning can be used to show that, under subpath consistency, a valid cycle with four arcs and nodes cannot contain adjacent arcs in $F$ (or $B$ ). However, as shown in Section 7., a valid cycle can contain four arcs and nodes if it is alternating between arcs in $F$ and in $B$. One might want to use this in the method presented in Section 5., but unfortunately there is probably no way of checking subpath consistency that is significantly faster than the methods in Section 5.. If this is true, it is best to use our methods for simultaneously checking for valid cycles and subpath consistency.
Lemma 12 If two SP-graphs are subpath inconsistent, there exists a valid cycle.
Proof: If $A_{k}$ and $A_{l}$ are subpath inconsistent there exists some $s \in N\left(A_{k}\right) \cap N\left(A_{l}\right)$ and $t \in N\left(A_{k}\right) \cap N\left(A_{l}\right)$ such that $A_{k}^{S}(s, t) \neq \emptyset, A_{l}^{S}(s, t) \neq \emptyset$ and $\left(A_{k}^{S}(s, t) \backslash\right.$ $\left.A_{l}^{S}(s, t)\right) \cup\left(A_{l}^{S}(s, t) \backslash A_{k}^{S}(s, t)\right) \neq \emptyset$. A valid cycle is now obtained by letting $B$ be a path from $s$ to $t$ in $A_{k}^{S}(s, t)$ and $F$ a path from $s$ to $t$ in $A_{l}^{S}(s, t)$, where at least one arc in $B$ lies in $A_{k}^{S}(s, t) \backslash A_{l}^{S}(s, t)$ or at least one arc in $F$ lies in $A_{l}^{S}(s, t) \backslash A_{k}^{S}(s, t)$.

Furthermore, there are examples (see Section 7. and [11]) with SP-graphs that are subpath consistent, but still have valid cycles. Therefore we can draw the following conclusion.
Lemma 13 The absence of a valid cycle is a stronger necessary condition for the existence of compatible weights than subpath consistency.

## 5. Methods for finding valid cycles

A practical method for finding valid cycles must consider each pair of SP-graphs, $l^{\prime}=1, \ldots, m-1$ and $l^{\prime \prime}=l^{\prime}+1, \ldots, m$. The arcs in one of the SP-graphs will be used backwards, if they are included in the cycle, and are therefore labeled with B . The arcs in the other SP-graph are used forwards, if they are included in the cycle, and are therefore labeled with F. Arcs with
one label, B or F , are eligible, while arcs labeled with both are not, and arcs with no label are not used at all. This can be seen as constructing a graph $\bar{G}$ which contains the arcs in one of the SP-graphs and the reversed arcs in the other SP-graph.

After this, one can either remove parts of the graph that cannot be a part of a cycle, or start to search for a valid cycle. The first approach is used in VC0, the method given in [7], and in VC1, the first of our new methods, while the second approach is used in VC2, the second of our new methods.

After removing arcs and nodes that cannot be a part of a feasible cycle, VC0 proceeds as follows. An eligible arc is chosen, and we try to find a feasible cycle containing this arc. If there exists no such cycle, the eligible arc is removed. If an isolated subgraph (a subgraph with in-degree or out-degree zero containing no eligible arc) is found in the process, this subgraph is also eliminated. This kind of graph reduction is repeated until the whole graph is eliminated, or a valid cycle is found.

The new variations of the method are based on strongly connected components of $\bar{G}$ (i.e. subgraphs containing a directed path between each pair of nodes). Clearly each feasible cycle corresponds to a directed cycle in $\bar{G}$, and all nodes in a directed cycle must belong to the same strongly connected component of $\bar{G}$. Arcs between two strongly connected components cannot be part of a feasible cycle.
Lemma 14 A valid cycle lies within one strongly connected component of $\bar{G}$.
A valid cycle must contain an eligible arc, so if a strongly connected component contains no eligible arc, it contains no valid cycle. On the other hand, if it contains an eligible arc, this arc is included in at least one directed cycle of $\bar{G}$ (i.e. in at least one feasible cycle), so a valid cycle exists. We have thus proved the following.
Lemma 15 A valid cycle exists if and only if a strongly connected component of $\bar{G}$ contains an eligible arc.
Lemma 10 states that a valid cycle contains at least three nodes, so strongly connected components that consist of less than three nodes cannot contain valid cycles. One may also note that all arcs in $\bar{G}$ are reversed if the order of the SP-graphs are changed. The strongly connected components are the same in the reversed graph as in the original graph, so it suffices to investigate each pair of SP-graphs once.

We now specify the two new methods for finding valid cycles. The first method, VC1, finds strongly connected components using Tarjan's method, [22], and is
similar to VC0 in that it uses reductions in order to reduce the size of the graph.

The following parts of $\bar{G}$ cannot be part of a valid cycle, and can therefore be removed.

- All arcs with endpoints in two different strongly connected components.
- All strongly connected components with less than three nodes.
- All strongly connected components that does not contain any eligible arc.

Comparing to VC0, we find the following. The main difference is that we do not need to return to the reduction phase. Eliminating a whole strongly connected component does not affect the other components. Furthermore, VC0 needs to use a shortest path method in order to try to find a valid cycle, starting from one endpoint of a certain eligible arc and searching for a path to the other endpoint of the eligible arc. If no path exists between these two nodes, the arc is removed, and the original reduction phase is reentered. In the new methods, however, we know that a valid cycle exists and it can easily be found by a simple depth first search (DFS) between the endpoints of the eligible arc. Just as in VC0, we have the following result.
Lemma 16 If the graph is completely eliminated by the reduction phase, there exists no valid cycle with the two SP-graphs considered.
Let us now present this method in an algorithmic form. First we give Tarjan's method. $S$ is a stack.

## Tarjan's method

(1) Set $k=1$. Set $v(i)=1 \forall i \in N$. Set $S=\emptyset$.
(2) For each $i \in N$, if $v(i)=1$ then do $\operatorname{SEARCH}(i)$.

## Procedure SEARCH $(k)$

(1) Set $v(i)=0, n(i)=k$ and $k=k+1$.
(2) Set $l(i)=n(i)$, and push $i$ onto $S$.
(3) For each node $j$ such that $(i, j) \in A$ do:

If $v(j)=1$ then
(a) $\operatorname{SEARCH}(j)$.
(b) Set $l(i)=\min (l(i), l(j))$.
else if $n(j)<n(i)$ and $j$ is on $S$ then $l(i)=$ $\min (n(j), l(i))$.
(4) If $l(i)=n(i)$ then repeat pop $x$ from $S$ until $x=i$. The main algorithm will be as follows.

## Algorithm VC1

(1) Choice of SP-graphs: If all pairs of SP-graphs have been compared, go to 6 . Otherwise choose two SP-graphs $A_{l^{\prime}}$ and $A_{l^{\prime \prime}}$ not previously compared.
(2) Graph construction: Construct a graph $\bar{G}$ by adding the arcs in $A_{l^{\prime}}$ to the reversed arcs of $A_{l^{\prime \prime}}$.

Arcs in $\left(A_{l^{\prime}} \backslash A_{l^{\prime \prime}}\right) \cup\left(A_{l^{\prime \prime}} \backslash A_{l^{\prime}}\right)$ are marked eligible.
(3) Reduction phase: Find the strongly connected components in $\bar{G}$ with Tarjan's method.
Remove all arcs between two different strongly connected components.
Remove all strongly connected components with less then three nodes.
Remove all strongly connected components that do not contain an eligible arc.
Remove nodes with only one adjacent arc.
(4) Cycle check: If all nodes are eliminated: No valid cycle found. Go to 1 . Otherwise, there exists a valid cycle. (Optionally: Terminate the method.)
(5) Find valid cycle: Form a valid cycle by finding an eligible $\operatorname{arc}(i, j) \in A_{l^{\prime \prime}}$ and a path from node $j$ to node $i$, or an eligible $\operatorname{arc}(i, j) \in A_{l^{\prime}}$ and a path from node $i$ to node $j$. Terminate the method. No compatible weights exist.
(6) No valid cycle found: No valid cycle exists. Terminate the method. (Compatible weights may exist.)
If the optional stopping criterion in step 4 is used, we will know that there exists a valid cycle, but we will not find it. In step 5, the path is found by a simple depthfirst search.

The second method, VC2, uses Kosaraju's method, [20], for finding strongly connected components. The method is based on two depth-first searches. The first DFS starts at an arbitrary node and number the nodes in postorder (i.e. the nodes are numbered in the order the DFS backs up from the nodes). The graph $\bar{G}$ is then reversed, and the next DFS starts at the node with highest number. A strongly connected component is found each time the second DFS terminates, and the nodes reached belong to the same component. As long as the second DFS have not reached all nodes, the DFS is restarted at the unreached node with highest number. When all strongly connected components are found, we investigate if some component contains an eligible arc. If this is the case, a valid cycle can be found by a DFS. The other possible result is that no strongly connected component contains an eligible arc, and Lemma 15 then tells us that no valid cycle exists.

## Kosaraju's method

(1) Perform a DFS of $\bar{G}$ and number the nodes in order of completion of the recursive calls.
(2) Reverse the directions of every arc in $\bar{G}$.
(3) Perform a DFS on the reversed graph, starting the search from the highest numbered node according
to the numbering assigned in step 1 . If the DFS does not reach all nodes, start the next DFS from the highest numbered remaining node.
(4) Each tree in the resulting spanning forest is a strong component of $\bar{G}$

## Algorithm VC2

Replace steps 3 and 4 in VC1 by the following:
3. Find strongly connected components: Find the strongly connected components of $\bar{G}$ with Kosaraju's method.
4. Cycle check: If no strongly connected component contains an eligible arc: No valid cycle found. Go to 1 . Otherwise, there exists a valid cycle. (Optionally: Terminate the method.)
Theorem 4 After a finite number of steps, algorithms VC1 and VC2 will terminate, either with a valid cycle or a proof that no valid cycle exists.
The complexity of finding strongly connected components in a graph with $n$ nodes and $a$ arcs is $O(n+a)$ both with Tarjan's method and with Kosaraju's method, see [1] and [13]. These methods are used on graphs with $|N|$ nodes and at most $2\left|A_{L}\right|$ arcs, where $\left|A_{L}\right|$ denotes largest number of arcs in an SP-graph. An eligible arc can be found in $O\left(\left|A_{L}\right|\right)$, so if $m$ is the number of SPgraphs, the complexity of VC1 and VC2 is $O\left(m^{2}(|N|+\right.$ $\left.\left|A_{L}\right|\right)$ ). For complete SP -graphs (i.e. extremely much splitting) we would have $O\left(\left|A_{L}\right|\right)=O\left(|N|^{2}\right)$, while for simple spanning trees we get $O\left(\left|A_{L}\right|\right)=O(|N|)$, and for non-spanning SP-graphs $\left|A_{L}\right|$ might be even less.

If a valid cycle is found, modification of the SPgraphs can be made as described in [7], in order to produce compatible weights. In short, one can either make a feasible cycle infeasible, by removing an arc from one SP-graph, or make an improving cycle non-improving, by adding an arc to one SP-graph (and thereby making an eligible arc non-eligible). Moving an arc within an SP-graph may sometimes yield both effects at the same time. In [7], we only considered spanning SP-graphs, which often made removing arcs from SP-graphs impossible. Allowing SP-graphs that are not spanning, makes removing arcs from SP-graphs a possible alternative to adding or moving arcs.

## 6. The symmetric single path case

Some network operators require that the routing paths should be symmetric, which means that the same path should be used in both directions between two nodes. Some operators also require that the routing paths should be unique. We call this case the symmet-
ric single path case. Our models have been developed for the more general case, i.e. when the desired paths not necessarily have to be symmetric and unique, but we can deal with the symmetric single path case as well. For our models, the symmetric single path case is obtained by including a single path in each set $S_{k}$, and by ensuring that the reversed path is included in some other set $S_{k^{\prime}}$.

The symmetric single path case has previously been treated by using undirected graphs. The desired routing paths are represented by undirected paths, and we let $\phi_{l}$ denote the desired path between node $o_{l}$ and node $d_{l}$. Mathematical models for the symmetric single path case in undirected graphs are presented in [3]. A directed model also allows the same weight to be used, but may also give different weights for the different directions. Some operators require that the same weight should be used in both directions of a link, which is possible by using an undirected model (or a directed model with the constraints $w_{i j}=w_{j i} \forall(i, j) \in A$, see e.g. [5]).

Let us now describe how undirected paths can be converted to symmetric SP-graphs, so that our methods can be used. Two SP-graphs are introduced for each undirected path $\phi_{l}$, and $A_{l^{\prime}}$ consists of the arcs in the directed path from $o_{l}$ to $d_{l}$ visiting the same nodes in the same order as $\phi_{l}$, and $A_{l^{\prime \prime}}$ contains the reversed arcs in $A_{l^{\prime}}$. This way we will consider both "directions" of each undirected path, so all information from the given undirected paths is retained. Furthermore, the directed path in each SP-graph can be found in an undirected path, so no additional requirements have been introduced. The symmetric single path case has now been converted from undirected paths to directed SP-graphs, so P4 can be used for finding compatible weights and a valid cycle method can be used for explaining why no compatible weights exists. We may also use Lemma 7 and Lemma 8 in order to combine the SP-graphs as much as possible.

If this approach is used on undirected paths that are suboptimal, the constructed SP-graphs will be subpath consistent. It then follows from Corollary 2 that two SPgraphs with the same origin can be combined to an outtree with two branches. The combined SP-graph contains the same paths as the original two SP-graphs, so it is subpath consistent with any other SP-graph. We can therefore use Lemma 7 repeatedly and combine all SPgraphs with the same origin to an out-tree with several branches. (Due to Lemma 8, we may also use subpaths of the original paths.) The same holds for SP-graphs with the same destination, so we may also construct

SP-graphs formed as in-trees.
In [3] two necessary conditions for the existence of compatible weights, stronger than subpath consistency, are given for the undirected case, namely the cyclic compatibility condition and the generalized cyclic compatibility. In Section 7., examples from [3] are treated by the above approach. We find valid cycles in instances not satisfying the cyclic compatibility condition, in instances satisfying the cyclic compatibility condition and not satisfying the generalized cyclic compatibility condition, and in instances satisfying the generalized cyclic compatibility condition. In other words, these cyclic conditions are not stronger necessary conditions for the existence of compatible weights than the absence of valid cycles. We have not been able to prove that there exists a valid cycle for each instance without compatible weights where the cyclic compatibility condition or the generalized cyclic compatibility condition is not satisfied, but have not found any instance satisfying the cyclic compatibility condition and the generalized cyclic compatibility condition that has no valid cycle either.

In [3], no efficient method for checking if the cyclic compatibility condition or the generalized cyclic compatibility condition is satisfied is given. The straightforward way would be to enumerate for each edge all cycles containing the edge and checking all the desired paths with both endpoints on the cycle. Doing this, as described here, can obviously not be done in polynomial time. We believe that a more efficient approach is to convert the undirected routing paths to directed SPgraphs, and then search for valid cycles. This can be done in polynomial time.

## 7. Examples

Let us first study an example from [7]. Two subpath consistent SP-graphs are shown in Figures 2a-b. We construct the graph $\bar{G}$ by adding the arcs in $A_{2}$ and the reversed arcs in $A_{1}$, see Figure 2c. Note that arc $(1,6)$ is the only arc which is included in both SP-graphs, so all arcs except $(1,6)$ are eligible. The strongly connected components of $\bar{G}$ are $\{1,6\}$ and $\{2,3,4,5\}$, see Figure 2 d . The first strongly connected component has only two nodes, and in addition contains no eligible arc, so it is removed. The second component contains several eligible arcs, so we know that a valid cycle exists. We get the cycle 2-4-3-5-2, and conclude that there do not exist any compatible set of weights for these two SP-graphs.


Fig. 2. (a) The ${ }^{(\mathrm{d} \text { in-tree }} A_{1}$; (b) the out-tree $A_{2}$; (c) the graph $\bar{G}$; (d) the strongly connected components of $\bar{G}$.

The next example comes from [3]. Five undirected paths are given; 5-6, 1-4-5, 2-3-5, 3-1-6 and 4-2-6. The paths do not satisfy the cyclic compatibility conditions, so we know that there do not exist any compatible weights. The directed SP-graphs are constructed as described in Section 6.. Figures 3a-b show those that give a valid cycle. When a valid cycle method is applied, these SP-graphs are combined to the directed graph in Figure 3 c , and its strongly connected components are $\{1,2,3,4\}$ and $\{5,6\}$, see Figure 3d. The last one does not contain any eligible arc, but the first one does, so a


Fig. 3. (a) The out-tree from node 5; (b) the in-tree to node 6 ; (c) the graph $\bar{G}$; (d) the strongly connected components of $\bar{G}$.
valid cycle is found.
In [3] another, similar, example is given, which satisfies the cyclic compatibility condition, but not the generalized cyclic compatibility condition. However, exactly the same valid cycle as found in the previous example is present. So, at least in these two examples, the difference between the cyclic compatibility condition and the generalized cyclic compatibility condition seems to be unrelated to the possible existence of a valid cycle.

Our next example also comes from [3]. It is based on the undirected graph and the undirected routing paths in


Fig. 4. (a) Undirected graph and routing paths; (b) the in-tree to node 3; (c) the in-tree to node 5.

Figure 4 a . These paths satisfy all necessary conditions presented in [3], but there exists no compatible set of weights.

A set of SP-graphs is constructed as described in Section 6.. Now we apply a valid cycle method, and a valid cycle is found when the in-tree to node 3 and the in-tree to node 5 are investigated, see Figures 4b-c. The valid cycle method finds three strongly connected components, namely $\{1,2,4,6\},\{3\}$ and $\{5\}$. The first component consists of four eligible arcs, and a depth first search results in the valid cycle 1-6-4-2-1. If we analyze the valid cycle, we find that the eligible arcs originate from the routing paths 2-1-3, 3-4-6, 2-4-5 and $5-1-6$. It can be verified that discarding any of these routing paths eliminates this particular conflict.

Let us finally consider an example which shows the advantages of combining SP-graphs. There is no valid cycle for the SP -graphs $A_{1}=\{(1,3),(3,2)\}$, $A_{2}=\{(1,4),(4,5)\}, A_{3}=\{(6,4),(4,2)\}$ and $A_{4}=\{(6,3),(3,5)\}$. Now combine $A_{1}$ and $A_{2}$ into $A_{5}$, and $A_{3}$ and $A_{4}$ into $A_{6}$, which is possible according to Lemma 7. Applying VC1 or VC2 to $A_{5}$ and $A_{6}$ yields the strongly connected components $\{1\},\{6\}$ and $\{2,3,4,5\}$. The last component contains several eligible arcs, forming the valid cycle 2-3-5-4-2.

## 8. Implementation and computational tests

Computational tests are performed on random networks with 10 to 90 nodes (RAND) and on real life networks from the survivable network design data library [21]; COST266, DI-YUAN, FRANCE, GERMANY50, and PIORO40. We use five different groups of test instances; RC, TR, RND, RM1 and RM5. The computer used is a $2 \times 750 \mathrm{MHz}$ Sun-Blade-1000 with 3 Gb physical memory.

The test instances have been generated by computing spanning shortest path graphs to each node in the network with respect to a set of weights, and then changing one or several SP-graphs such that it is not certain that compatible weights exist. The instances in group RC are generated by adding one arc to a random SP-graph. (The added arc belongs to a shortest path if the weight is decreased by one.) A random arc is added to a random SP-graph in TR, and a random number (between 1 and $|N|$ ) of random arcs are added to random SP-graphs in RND. The instances in group RM1 are generated by removing one arc from a random SP-graph, and the instances in group RM5 are generated by repeating this five times. We ensure that each SP-graph is connected. Furthermore, an instance is discarded if a modified SPgraph contains a directed cycle, and this is why groups TR and RND sometimes contain fewer instances.

We have also generated test instances with nonspanning SP-graphs. A spanning SP-graph contains routing paths from one or several origins to one destination, and we generate non-spanning SP-graphs by computing the sets $A_{l}^{S}\left(o_{l}, d_{l}\right)$ for each origin and destination of each SP-graph. The non-spanning SP-graphs can be combined back to the spanning SP-graphs using Lemmas 7 and 8 , so a non-spanning instance has compatible weights if the spanning instance it was generated from has compatible weights. This also means that the spanning and non-spanning versions contain the same information, so they constitute two different ways of solving the same basic problem.

Computational results for RAND networks and instances with spanning SP-graphs are summarized in Table 1 . We use instances from groups RC and TR. $N$ denotes the number of nodes in the group, $I$ denotes the number of instances in the group, and $N_{A}$ denotes the average number of arcs in the SP-graphs. Column $N_{V}$ shows the number of instances with valid cycles, and the remaining two columns show the solution times for VC 1 and VC 2 . These columns show the total solution time (in seconds) for all instances in the different
groups, so by dividing these numbers with the numbers of instances, the average time for one instance can be obtained.

Table 1

|  | $N$ | $I$ | $N_{A}$ | $N_{V}$ | VC1 | VC2 |
| :---: | :---: | ---: | ---: | ---: | ---: | ---: |
|  | 10 | 40 | 9.8 | 18 | 0.02 | 0.08 |
|  | 15 | 40 | 15.1 | 34 | 0.04 | 0.06 |
|  | 20 | 40 | 20.6 | 33 | 0.09 | 0.10 |
| RAND | 30 | 40 | 32.6 | 33 | 0.45 | 0.32 |
|  | 40 | 40 | 42.8 | 39 | 1.10 | 0.78 |
|  | 50 | 40 | 53.3 | 39 | 1.97 | 1.39 |
|  | 90 | 40 | 101.4 | 38 | 11.30 | 8.48 |

Computational results for RAND networks and spanning SP-graphs

If we compare the solution times for VC 1 and VC 2 , we find that VC 1 is faster for smaller instances and that VC2 is faster for larger instances. VC0 was applied to the same random instances (with 10-30 nodes) in [6]. Comparing the solution times for VC 0 with the ones obtained for VC1 and VC2, we find that the newer methods are up to 500 times faster then the original method. VC1 and VC2 has time complexity $O\left(m^{2}\left(\left|A_{L}\right|+|N|\right)\right)$ and VC0 has time complexity $O\left(m^{2}|N|^{2}\left|A_{L}\right|\right)$ (reduced to $O\left(m^{2}|N|\right)$ for the single path case), where $m$ is the number of SP-graphs and $A_{L}$ is the SP-graph with the largest number of arcs. However, VC0 was implemented in $\mathrm{Tcl} / \mathrm{Tk}$ (www.tcl.tk) within the framework of the graphical package VINEOPT (www.vineopt.com). $\mathrm{Tcl} / \mathrm{Tk}$ is a scripting language and much slower than an implementation in C (VC1) and $\mathrm{C}++(\mathrm{VC} 2)$.

Table 2 shows computational results obtained for spanning SP-graphs and real-life networks. The solution times for the two methods are very similar. The average solution times per instance are smaller than $0.04 \mathrm{sec}-$ onds for all groups, so we conclude that both methods work well for these types of instances.

In [6], the LP-problem P5 was solved for the same instances, with the code LPSOLVE (lpsolve.sourceforge.net). We find, for example, that the COST266 instances took approximately 40 seconds per instance on average, the PIORO40 instances approximately 200 seconds and the GERMANY50 instances approximately 500 seconds. We thus conclude that it is much quicker to use VC1 or VC2 than to try to solve P5. We also recall that P5 cannot be used for non-spanning instances; instead we would have to use P3, which has more constraints and will take even longer time to solve.

Table 3 shows computational results obtained for the RAND networks and the non-spanning SP-graphs. Column $N_{S}$ shows the average number of SP-graphs in each group. As for the previous groups of test instances, we find that VC1 is faster for smaller instances and that VC2 is faster for larger instances.

We also find that test instances with non-spanning SPgraphs require much more solution time than the spanning ones. (Recall that these non-spanning instances contain the same paths as the spanning, and that they are separated as much as possible.) For an instance with 90 nodes, a valid cycle method investigates approximately 4000 pairs of spanning SP-graphs or 5600000 pairs of non-spanning SP-graphs when no valid cycle can be found. However, investigating a pair of spanning SP-graphs require only $13 \%$ more solution time than a pair of the non-spanning ones. Our conclusion is clearly to combine SP-graphs as much as possible.

## 9. Conclusions

The problem of finding weights that gives prespecified routing patterns in IP networks using the OSPF protocol is considered. We discuss models that yield compatible weights, and use them to analyze the situation when no compatible weights exist. We show that the absence of valid cycles is a necessary condition for the existence of compatible weights for possibly nonspanning SP-graphs. We present two new methods for finding valid cycles by investigating strongly connected components. Computational experiments indicate that the new methods are faster than the previously known method. We show that SP-graphs with certain properties can be combined into a single SP-graph. Computational experiments indicate that the methods are faster if SPgraphs are combined as much as possible. Furthermore, we show how to apply our methods to the symmetric single path case (instead of using undirected models).

An interesting direction for future research is to include a valid cycle method into a framework for finding the optimal design of an OSPF network, or into a framework for optimizing the performance of existing OSPF networks.

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Table 2

Computational results for real-life networks and spanning SP-graphs.

|  | COST266: 37 nodes, 114 arcs |  |  |  |  | DI-YUAN: 11 nodes, 84 arcs |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | I | $N_{A}$ | $N_{V}$ | VC1 | VC2 | $I$ | $N_{A}$ | $N_{V}$ | VC1 | VC2 |
| RC | 30 | 38.56 | 25 | 0.47 | 0.34 | 30 | 10.94 | 3 | 0.02 | 0.05 |
| TR | 29 | 38.57 | 27 | 0.35 | 0.20 | 30 | 10.94 | 7 | 0.02 | 0.04 |
| RND | 30 | 38.96 | 30 | 0.08 | 0.03 | 30 | 11.33 | 15 | 0.01 | 0.06 |
| RM1 | 30 | 38.51 | 26 | 0.36 | 0.26 | 30 | 10.76 | 2 | 0.03 | 0.05 |
| Rm5 | 30 | 38.40 | 30 | 0.03 | 0.03 | 30 | 10.45 | 7 | 0.02 | 0.03 |
|  | FRANCE: 25 nodes, 90 arcs |  |  |  |  | GERMANY50: 50 nodes, 176 arcs |  |  |  |  |
|  | I | $N_{A}$ | $N_{V}$ | VC1 | VC2 | I | $N_{A}$ | $N_{V}$ | VC1 | VC2 |
| RC | 30 | 26.28 | 28 | 0.11 | 0.10 | 30 | 54.26 | 27 | 0.99 | 0.75 |
| TR | 30 | 26.28 | 28 | 0.10 | 0.12 | 29 | 53.91 | 29 | 0.50 | 0.27 |
| RND | 29 | 26.57 | 29 | 0.03 | 0.04 | 27 | 54.35 | 27 | 0.06 | 0.06 |
| RM1 | 30 | 26.20 | 23 | 0.18 | 0.10 | 30 | 54.22 | 27 | 1.06 | 0.77 |
| Rm5 | 30 | 26.04 | 29 | 0.03 | 0.03 | 30 | 54.14 | 30 | 0.20 | 0.10 |
|  | PIORO40: 40 nodes, 178 arcs |  |  |  |  |  |  |  |  |  |
|  | I | $N_{A}$ | $N_{V}$ | VC1 | VC2 |  |  |  |  |  |
| RC | 30 | 42.18 | 27 | 0.39 | 0.32 |  |  |  |  |  |
| TR | 30 | 42.18 | 27 | 0.49 | 0.27 |  |  |  |  |  |
| RND | 30 | 42.64 | 30 | 0.05 | 0.04 |  |  |  |  |  |
| RM1 | 30 | 42.13 | 23 | 0.75 | 0.59 |  |  |  |  |  |
| Rm5 | 30 | 42.03 | 30 | 0.11 | 0.10 |  |  |  |  |  |

Table 3

Computational results for RAND networks and non-spanning SP-graphs.

|  | $N$ | $I$ | $N_{S}$ | $N_{A}$ | $N_{V}$ | VC1 | VC2 |
| :---: | :---: | :---: | ---: | :---: | ---: | ---: | ---: |
| RAND | 10 | 40 | 59.9 | 2.09 | 18 | 0.36 | 1.16 |
|  | 15 | 40 | 115.4 | 2.89 | 34 | 1.55 | 3.11 |
|  | 20 | 40 | 210.2 | 3.25 | 33 | 5.83 | 9.43 |
|  | 30 | 40 | 422.8 | 4.10 | 33 | 48.44 | 59.36 |
|  | 40 | 40 | 762.0 | 4.93 | 39 | 243.67 | 245.67 |
|  | 50 | 40 | 1099.7 | 5.84 | 39 | 610.71 | 549.67 |
|  | 90 | 40 | 3344.0 | 8.66 | 38 | 14422.50 | 10690.67 |

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