# Link-Disjoint Paths for Reliable QoS Routing 

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#### Abstract

The problem of finding link/node-disjoint paths between a pair of nodes in a network has received much attention in the past. This problem is fairly well understood when the links in a network are only specified by a single link weight. However, in the context of Quality of Service routing, links are specified by multiple link weights and restricted by multiple constraints. Unfortunately, the problem of finding link/node disjoint paths in multiple dimensions faces different conceptual problems. This paper presents a first step to understanding these conceptual problems in link-disjoint Quality of Service routing and proposes a heuristic link-disjoint QoS algorithm that circumvents these problems.


## 1 Introduction

The problem of finding disjoint paths in a network has been given much attention in the literature due to its theoretical as well as practical significance to many applications, such as layout design of integrated circuits, survivable design of telecommunication networks and restorable/reliable routing. Paths between a given pair of source and destination nodes in a network are called link disjoint if they have no common (i.e. overlapping) links, and node disjoint if, besides the source and destination nodes, they have no common nodes. With the development of optical networks and the deployment of MPLS or GMPLS networks, the disjoint paths problem is receiving renewed interest as fast restoration after a network failure is crucial in such kind of networks. In robust communication networks, a connection usually consists of two link- or node-disjoint paths: one active path, and one backup path. A service flow will be redirected to the backup path if the active path fails. Load balancing, another important aspect for communication networks to avoid network congestion and optimize network throughput, also requires disjoint paths to distribute flows. Robustness and load balancing are, among others, both aspects of Quality of Service (QoS) routing.

In this paper we will focus on finding QoS-aware link-disjoint paths. In general a link-disjoint paths algorithm can be extended to a node-disjoint algorithm with the concept of node splitting, i.e. replacing one node with two nodes that are linked together by a link with zero weights [26]. Throughout this paper, we use the following notation. A network is denoted by a directed graph $G(V, E)$, where $V$ is the set of nodes and $E$ is the set of links. A directed link from node $u$ to node $v$ is represented as $u \rightarrow v, u, v \in V$. Each link is characterized by a link weight vector $\vec{w}$ consisting of $M$ link metrics $w_{m}(u \rightarrow v)$, for $m=1, \ldots, M$. We assume that only nonnegative link metrics are assigned to each link. However, in the process of computing disjoint paths, negative link

[^0]weights may be assigned to links. QoS metrics can be a) additive, e.g. delay, jitter, in which case the path-weight vector consists of summing the link-weight vectors of the links defining the path, b) multiplicative, e.g., one minus the packet loss probability, which can be considered as additive after taking the logarithm and c) min-max, e.g. bandwidth, and policy flags, in which case the minimum (or maximum) link weight defines the weight of a path. Min/max links that do not obey the constraints can be pruned from the topology, which is called topology filtering. Additive metrics cause more difficulties and therefore without loss of generality, we assume all metrics to be additive [29]. In the context of QoS routing or multi-constrained routing, a path is called feasible when its weight vector does not violate the constraints specified by the vector $\vec{L}$.

Since we mainly focus on finding link-disjoint paths, a path $P$, between a source $s$ and destination $t$ is considered to be a set of links that compose this path. With a slight abuse of notation, we choose $P$ to denote the path as well as its link set. If path $P_{1}$ is link-disjoint with $P_{2}$, there is no common link element in the link set representing each path and $P_{1} \cap P_{2}=\varnothing$, else $P_{1} \cap P_{2} \neq \varnothing$.

Definition of path length: Given a graph $G(V, E)$ with $M$ metrics per link, the non-linear length of a path $P$ from source node $s$ to destination node $t$ is defined as [5]:

$$
\begin{equation*}
l(P)=\max _{m=1, \ldots, M}\left(\frac{w_{m}(P)}{L_{m}}\right) \tag{1}
\end{equation*}
$$

where $w_{m}(P)=\sum_{(u \rightarrow v) \in P} w_{m}(u \rightarrow v)$.
The normalization in (1) by the constraints $\vec{L}$ ascertains that if $l(P)>1$, then one of the constraints has been violated. For $M=1$, the non-linear length of a path as defined in (1) reduces to a linear one, and the link weight vector $\vec{w}$ reduces to a scalar $w(u \rightarrow v)$. When no constraint is required, as in the LPP problem stated below, the linear length of a path is computed as $\sum_{(u \rightarrow v) \in P} w(u \rightarrow v)$, i.e. $L_{l}=1$. For simplicity of representation, the above notation of path length $l(P)$ is still used.

If path $P_{1}$ is link-disjoint with $P_{2}$, i.e. $P_{1} \cap P_{2}=\varnothing$, we have $l\left(P_{1} \cup P_{2}\right)=l\left(P_{1}\right)+l\left(P_{2}\right)$ for $M=1$. But for $M>$ 1, we have $l\left(P_{1} \cup P_{2}\right) \leq l\left(P_{1}\right)+l\left(P_{2}\right)$. Our target in this paper is to find a set of two link-disjoint paths that both obey multiple constants. We define the total length of two paths as

$$
\begin{equation*}
l\left(P_{1}\right)+l\left(P_{2}\right) \tag{2}
\end{equation*}
$$

for $M \geq 1$.

Link-disjoint Path Pair (LPP) Problem. Given a graph $G(V, E)$ with 1 metric per link ( $M=1$ ), for a sourcedestination pair ( $s, t$ ), find a set of two paths $P_{1}$ and $P_{2}$, such that $P_{1} \cap P_{2}=\varnothing$, and the total length $l\left(P_{1}\right)+l\left(P_{2}\right)$ is minimized.
The LPP problem can be solved in polynomial time [1][25][26].

Multiple Constrained Path (MCP) Problem. Given a graph $G(V, E)$ with $M>1$ metrics per link and a constraint vector $\vec{L}$, for a source-destination pair $(s, t)$, find a path that obeys the constraint vector $\vec{L}$,

$$
w_{m}(P) \leq L_{m}, \text { for } m=1, \ldots, M,
$$

where $w_{m}(P)=\sum_{u \rightarrow v \in E} w_{m}(u \rightarrow v)$, for $m=1, \ldots, M$.
The MCP problem is NP-complete [8][30].

Multiple Constrained Link-disjoint Path Pair (MCLPP) Problem. Given a graph $G(V, E)$ with $M>1$ metrics per link and a constraint vector $\vec{L}$, for a source-destination pair $(s, t)$, find a pair of link-disjoint paths $P_{1}$ and $P_{2}$, such that $P_{1} \cap P_{2}=\varnothing$, and both paths obey the constraint vector $\vec{L}$.

Theorem 1. MCLPP is NP-complete.


Figure 1. The assignment of link weights to the links in the chain topology between nodes $i$ and $i+1$
Proof: Given a chain topology with $n+1$ nodes and $2 n$ links, each with a two-component weight vector as depicted in Figure 1 and a set of numbers $a_{i} \in A, 0 \leq a_{i} \leq S$, for $i=1, \ldots, n$, where $S=\sum_{i=1}^{n} a_{i}$. The constraints are chosen as follows: $L_{1}=n S$-( $S / 2$ ), and $L_{2}=(S / 2)$.

To solve the MCLPP problem, we need to find two paths P and $\mathrm{P}^{\prime}$ from node 1 to node $n+1$ that obey the constraints. Since, for all link weight vectors, the sum of the components equals $S$, we have that $w_{1}(P)+w_{2}(P)=n S$ and $w_{l}\left(P^{\prime}\right)+w_{2}\left(P^{\prime}\right)=n S$. Accordingly, a solution satisfying the constraints is only found if $w_{l}\left(P\right.$ and $\left.P^{\prime}\right)=n S-(S / 2)$ and $w_{2}\left(P\right.$ and $\left.P^{\prime}\right)=(S / 2)$. The problem has now become an instance of the well-known NP-complete partition problem [8] and can only be solved by finding the set $A^{\prime} \subseteq A$, for which $\sum_{a_{i} \in A^{\prime}} a_{i}=(\mathrm{S} / 2)$. A feasible path $P$ exists if the set $A^{\prime}$ exists. A feasible path $P$ consists of the lower link if $a_{i} \in A^{\prime}$ and the upper link if $a_{i} \notin A^{\prime}$. The path $P^{\prime}$ then follows the remaining links.

In this paper we focus on solving the MCLPP problem. Related work on finding disjoint paths in one dimension between a source and a destination will be reviewed in Section 2 and a simple link-disjoint algorithm LBA will be explained in Section 3. In Section 4 an extension of LBA to multiple dimensions is discussed and shown to be difficult. Therefore, a heuristic algorithm DIMCRA for solving the MCLPP problem is proposed in Section 5. We conclude this article in Section 6.

## 2 Related work

### 2.1 Link-disjoint Paths Routing in One Dimension

An intuitive method to determine two shortest link-disjoint paths between a pair of source and destination nodes consists of two steps. The first step retrieves the shortest path between a given pair of nodes in a graph. The second step is to remove all the links of that path from the graph, and to find the shortest path in the pruned graph. We will refer to this method as the Remove-Find (RF) method. Although the RF method is direct and simple, it has at least two disadvantages due to the removal of links belonging to the first shortest path: (a) provided that two link-disjoint paths exist, there is no guarantee that they will be found as illustrated in Section 3.1 and (b) the second link-disjoint shortest path may have a significantly larger length.

To surmount the disadvantages of the RF method, other methods have been devised to find a pair of shortest link-disjoint paths with minimal total length [1][3][4][19][24][25][26][31]. In [25], Suurballe proposes an algorithm, referred to as Suurballe's algorithm, to find $K$ node-disjoint paths with minimal total length using the path augmentation method. The path augmentation method is originally used to increase the size of a matching with an augmenting path [6] and to find a maximum flow or a minimum cost flow in a network [7][21]. The problem to find link/node disjoint paths can be viewed as a special case of the minimum cost flow problem as demonstrated in [1][25][26]. The basic idea of Suurballe's algorithm is to construct a solution set of two disjoint paths based on the shortest path and a shortest augmenting path. $K$ disjoint paths can be obtained by augmenting the $K-1$ optimal disjoint paths with this algorithm. In 1984, Suurballe and Tarjan [26] improved Suurballe's algorithm such that pairs of link-disjoint paths from one source node to $n$ destination nodes could be efficiently obtained in a single Dijkstra-like computation. This algorithm is referred to as the S-T algorithm. To find $n$ pairs of disjoint paths, the S-T algorithm requires $O\left(E \log _{(1+E / n)} n\right)$ time and Suurballe's algorithm $O\left(n^{2} \log n\right)$, where $n$ is the number of destination nodes and $E$ is the number of links. Kar et al. [12] and Kodialam and Lakshman [13][14] incorporated the S-T algorithm into their algorithms to find a pair of link-disjoint paths serving as active and backup paths for routing bandwidth guaranteed connections. Liang [17] extended the S-T algorithm to find two link-disjoint paths between a pair of nodes with optimization in both network load and routing cost.

In 1994, Bhandari [1] proposed an algorithm to find a pair of span-disjoint paths between two nodes in optical-fiber networks. The disjoint paths algorithm used by Bhandari is a modified version of Suurballe's algorithm [25] that requires a special link weight transformation to facilitate the use of Dijkstra's. Bhandari made a simplification to Suurballe's algorithm by directly setting all the link weights on the first shortest path negative. Shaikh [23] made an extension to Bhandari's algorithm [1] to solve the span-disjoint paths problem in more complicated structured optical networks.

It is proved in [16][22] that the LPP problem will be NP-complete if it is required that the maximal length of the two disjoint paths, i.e. $\max \left(l\left(P_{1}\right), l\left(P_{2}\right)\right)$, is minimized. In addition, Ho and Mouftah [10] proposed another optimal object function $\alpha \cdot l\left(P_{1}\right)+l\left(P_{2}\right)$, where $P_{1}$ and $P_{2}$ are the active path and the backup path, respectively. The parameter $\alpha$ can be set large for a shared protection scheme ( $1: N$ or $M: N$ ) and could be as small as unity for a dedicated protection scheme (1:1). When $\alpha=1$, it reduces to the object function used in [1][25][26].

Heuristic algorithms based on matrix calculation [28] or recursive matrix-calculation [20] to solve the $K$ shortest link-disjoint paths problem with a bounded hopcount have been proposed as well. There are also some algorithms for finding $K$-best paths, i.e. $K$ disjoint or maximally disjoint paths with minimum total length between a pair of nodes, in a trellis graph [3][31]. An optimal algorithm for finding $K$-best paths without hopcount limitation between a pair of nodes is given by Lee and Wu in [15], where they transfer the $K$-best paths problem into a maximum network flow and minimum cost network flow algorithm via some modifications to the original graph. Distributed algorithms for the link/node-disjoint paths algorithms can be found in [4][19][24].

### 2.2 Disjoint Paths Routing in Multiple Dimensions

To the best of our knowledge there is no literature on the MCLPP problem. Recently some papers on disjoint paths in QoS routing have emerged. However, they only considered bandwidth and/or delay as their QoS metrics [2][14][11][9][18]. The maximally disjoint shortest and widest paths (MADSWIP) algorithm from Taft-Plotkin, et al. [27], involves a modified version of the S-T algorithm to find a pair of disjoint paths. MADSWIP can produce a pair of widest or shortest maximally link-disjoint paths from a source node to all other nodes. Moreover it tries to find two paths simultaneously to satisfy the maximally link-disjointness to each other in a QoS routing context. However the link metrics used in their algorithm are bandwidth and delay, where only the latter metric is additive.

## 3 Path Augmentation for Solving LPP

In this section we will present a simplified variant of Bhandari's Algorithm [1], referred to as LBA (Link-disjoint version of Bhandari's Algorithm), which can produce an optimal solution for the LPP problem. The basic steps of LBA are given in Section 3.1. The fundamental concepts of this algorithm are discussed in Section 3.2. The optimality is proved in Section 3.3 and in Section 3.4, LBA is shown to be loop-free.

### 3.1 The steps of LBA

Bhandari's algorithm [1] was designed to find a pair of span-disjoint paths in an optical network. We modify Bhandari's algorithm into a link-disjoint path pair algorithm LBA by omitting the node-splitting operation that ensures the node-disjointness and the graph transformations that ensure span-disjointness.

Before explaining the operation of LBA we first introduce some notations that will be used further. If we reverse the direction and the sign of the link weights of each link on the path $P_{1}$ between $s$ and $t$, i.e. $w(v \rightarrow u)=$ $-w(u \rightarrow v), \forall(u \rightarrow v) \in P_{1}$, then we will have a path directed from $t$ to $s$, denoted by $-P_{1}$, which consists of the reversed $P_{1}$ links. We define ${ }^{1} l\left(-P_{1}\right)=-l\left(P_{1}\right)$. A set, which consists of the $P_{l}$ links whose reversed links appear on $P_{2}$ and vice versa, is denoted as $P_{1} \widetilde{\cap} P_{2}=\left\{(u \rightarrow v)\right.$ and $(v \rightarrow u) \mid(u \rightarrow v) \in P_{1}$ and $\left.(v \rightarrow u) \in P_{2}\right\}$. In all the figures, bold lines represent links on the shortest path(s) in a graph or its corresponding modified graph, dashed lines represent reversed links which do not exist in the original graph and bold dashed lines represent such reversed links that appear on the shortest path. The steps of the LBA algorithm are as follows:

[^1]Given a directed graph $G(V, E)$, for a source-destination pair $(s, t)$,
Step 1. Find the shortest ${ }^{2}$ path $P_{1}$ from node $s$ to node $t$;
Step 2. Replace $P_{1}$ with $-P_{1}$, a modified graph $G\left(V, E^{\prime}\right)$ is created;
Step 3. Find a shortest path $P_{2}$ from node $s$ to node $t$ in the modified graph $G\left(V, E^{\prime}\right)$; if $P_{2}$ does not exist, then stop;
Step 4. Take the union of $P_{1}$ and $P_{2}$, remove from the union the link set which consists of the $P_{1}$ links whose reversed links appear in $P_{2}$, and vice versa, then group the remaining links into two paths $P_{1}^{\prime}$ and $P_{2}^{\prime}$, i.e. $P_{1}^{\prime} \cup P_{2}^{\prime}=\left(P_{1} \cup P_{2}\right) \backslash\left(P_{1} \tilde{\cap} P_{2}\right)$.


Figure 2. Example of the operation of LBA
We will explain the steps of LBA with an example in Figure 2. Suppose that we are required to find a set of two shortest disjoint paths between $a$ and $b$. In Step 1, the shortest path from $a$ to $b$ is found as $P_{1}=a c d b$, with minimum length 4. In Step 2, a modified graph $G\left(V, E^{\prime}\right)$ is created by reversing the direction and the sign of the weight of each link on $P_{1}$. For instance, the link $c \rightarrow d$ with weight 1 is replaced by the link $d \rightarrow c$ with weight -1 . In Step 3, the shortest path in the modified graph $P_{2}=a d c b$ has length 6. In Step 4, $P_{1} \tilde{\cap} P_{2}=\{c \rightarrow d, d \rightarrow c\}$ is removed from the union $P_{1} \cup P_{2}$. The solution set of disjoint paths $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}=\{a c b, a d b\}$ is obtained. The total length of this path set equals $5+5=10$, which is exactly the minimal total length of two link-disjoint paths in this graph.

[^2]

Figure 3. Example 1 of the operation of RF


Figure 4. Example 2 of the operation of RF
For comparison, in Figure 3, we apply the RF method on the same topology with the same requirements. In step 1 the shortest path $a c d b$ is retrieved. In step 2, a modified graph is created by removing all the links on $a c d b$ The shortest path in the modified graph is $a e b$ with length 11 . Thus the set $\{a c d b, a e b\}$ has a total length $4+11=$ 15 , which is longer than 10 as found with LBA. This example illustrates that the RF method cannot guarantee to find the optimal solution. More important, in the graph shown in Figure 4(a), although there exist two linkdisjoint paths between $a$ and $b$, RF cannot find the second path in step 2 as shown in Figure 4(b). LBA, on the other hand, still returns the optimal set in this case.

### 3.2 LBA is Based on the Shortest Path

In this subsection, we will clarify why the optimal solution set of LBA, as well as other path augmentation algorithms $[1,18,19]$, is based on the shortest path. Although the theory presented here is based on (or can be derived from) the theory of min-cost flow [7][21], it is instructive to give an outline.

We will first show that the optimal set for the LPP problem is based on the shortest path. Secondly, we will show that the optimal set of two link-disjoint paths has the smallest difference in length from the shortest path among all the possible sets of link-disjoint paths. Finally, we will show that the logical difference set (defined below) can be viewed as a path.

Given a digraph $G(V, E)$ and a pair of source-destination nodes $(s, t)$, the relation between a set of two linkdisjoint paths $\left\{P_{d 1}, P_{d 2}\right\}$ and the shortest path $P_{1}$ belongs to one of the following types:

1. $\quad P_{1}$ itself is $P_{d 1}$ or $P_{d 2}$, i.e. $P_{1}=P_{d 1}$ or $P_{1}=P_{d 2}$;
2. $\quad P_{1}$ overlaps with both paths $P_{d 1}$ and $P_{d 2}$, i.e. $P_{1} \cap P_{d 1} \neq \varnothing, P_{1} \neq P_{d 1}$ and $P_{1} \cap P_{d 2} \neq \varnothing, P_{1} \neq P_{d 2}$;
3. $P_{1}$ only overlaps with one path in the set $\left\{P_{d 1}, P_{d 2}\right\}$, but not with the other one, i.e. $P_{1} \cap P_{d 1} \neq \varnothing, P_{1} \neq$ $P_{d 1}$ and $P_{1} \cap P_{d 2}=\varnothing\left(\right.$ or $P_{1} \cap P_{d 2} \neq \varnothing, P_{1} \neq P_{d 2}$ and $P_{1} \cap P_{d 1}=\varnothing$ );
4. $\quad P_{1}$ is link-disjoint with both paths in $\left\{P_{d 1}, P_{d 2}\right\}$, i.e. $P_{1} \cap\left(P_{d 1} \cup P_{d 2}\right)=\varnothing$.

Lemma 1. Given a directed graph $G(V, E)$ and a source-destination pair $(s, t)$, if the optimal set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ of $L P P$ exists, $P_{1}^{\prime} \cup P_{2}^{\prime}$ must contain either the first shortest path $P_{1}$ itself or some $P_{1}$ links on each of its two paths.

Proof: If $P_{1}^{\prime} \cup P_{2}^{\prime}$ is of type (4), then each path in $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ is link-disjoint with $P_{1}$. As $P_{1}$ is the shortest path, both $\left\{P_{1}, P_{1}^{\prime}\right\}$ and $\left\{P_{1}, P_{2}^{\prime}\right\}$ have a total length shorter than $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$. Hence the optimal set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ cannot be of type (4) and $P_{1}^{\prime} \cup P_{2}^{\prime}$ must contain some or all $P_{1}$ links to be the optimal set.

If $P_{1}^{\prime} \cup P_{2}^{\prime}$ is of type (3), only one path in $P_{1}^{\prime} \cup P_{2}^{\prime}$ contains some $P_{1}$ links, without loss of generality, suppose $P_{1}^{\prime}$ contains some $P_{1}$ links, and the other path $P_{2}^{\prime}$ is link-disjoint with $P_{1}$, then $\left\{P_{1}, P_{2}^{\prime}\right\}$ is a set which is shorter than $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$, Hence the optimal set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ cannot be of type (3).

Therefore, if the optimal set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ exists, $P_{1}^{\prime} \cup P_{2}^{\prime}$ must be either of type (1) or (2).

Property 1. The optimal set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ has the smallest difference in length

$$
\begin{equation*}
Y=l\left(P_{1}^{\prime}\right)+l\left(P_{2}^{\prime}\right)-l\left(P_{1}\right) \geq 0 \tag{3}
\end{equation*}
$$

from the shortest path $P_{1}$, among all the possible sets of link-disjoint path pairs.

In the set $P_{1}^{\prime} \cup P_{2}^{\prime} \cup\left(-P_{1}\right)$, the $P_{1}$ links contained in the set $P_{1}^{\prime} \cup P_{2}^{\prime}$ will form loops with the $-P_{1}$ links. For example, if a $P_{1} \operatorname{link} u \rightarrow v$ is contained in the set $P_{1}^{\prime} \cup P_{2}^{\prime}$, then it will create a loop with the link $v \rightarrow u$ on $-P_{1}$ between the nodes $u$ and $v$. The length of this loop is zero because $w(v \rightarrow u)=-w(u \rightarrow v)$. Let us denote $O_{l}$ $=\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right) \widetilde{\cap}\left(-P_{1}\right)$, which means that the set $O_{l}$ consists of each $P_{1}$ link in the union of $P_{2} \cup P_{1}$ and its corresponding $-P_{1}$ link. We define the logical difference set ${ }^{3}$ between $P_{1}^{\prime} \cup P_{2}^{\prime}$ and $P_{1}$ as $\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right)-P_{1}=P_{1}^{\prime} \cup P_{2}^{\prime} \cup\left(-P_{1}\right) \backslash O_{l}$. In fact, $l\left(O_{l}\right)=0$ because the set $O_{l}$ consists of loops with zero length, each consisting of a pair of opposite $P_{1}$ and $-P_{1}$ links. With $l\left(-P_{1}\right)=-l\left(P_{1}\right)$, we have

$$
l\left(\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right)-P_{1}\right)=l\left(\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right) \cup\left(-P_{1}\right)\right)-l\left(O_{l}\right)=l\left(P_{1}^{\prime}\right)+l\left(P_{2}^{\prime}\right)+l\left(-P_{1}\right)=l\left(P_{1}^{\prime}\right)+l\left(P_{2}^{\prime}\right)-l\left(P_{1}\right),
$$

[^3]which is exactly $Y$ in (3).
Lemma 2 shows that the logical difference set forms the shortest path in the modified graph where $P_{1}$ is replaced with $-P_{1}$.

Lemma 2. Given a directed graph $G(V, E)$ and pair $(s, t)$ and let $P_{1}$ be the shortest path in this graph. We define $G\left(V, E^{\prime}\right)$ as the graph $G(V, E)$ for which the path $P_{1}$ is replaced with $-P_{1}$. The logical difference set $P_{1}^{\prime} \cup P_{2}^{\prime}-P_{1}$ between the optimal set of two link-disjoint paths $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ and the shortest path $P_{1}$ forms the shortest path $P_{2}$ from node s to node $t$ in $G\left(V, E^{\prime}\right)$.

Proof: We will first prove that $P_{2}=P_{1}^{\prime} \cup P_{2}^{\prime}-P_{1}$ is a complete path from $s$ to $t$ in $G\left(V, E^{\prime}\right)$, then we will prove that $P_{2}$ is the shortest path in $G\left(V, E^{\prime}\right)$.
Part A. From Lemma 1, the optimal set of two link-disjoint paths $P_{1}^{\prime} \cup P_{2}^{\prime}$ must contain either the first shortest path $P_{1}$ itself or some $P_{1}$ links on each of its two paths.

If $\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right) \supset P_{1}$, without loss of generality, suppose $P_{1}^{\prime}=P_{1}$, then $O_{l}=P_{1} \cup\left(-P_{1}\right)$. With the definition of logical difference set, we have $P_{2}=\left(\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right) \cup\left(-P_{1}\right)\right) \backslash O_{l}=\left(P_{1} \cup P_{2}^{\prime} \cup\left(-P_{1}\right)\right) \backslash\left(P_{1} \cup\left(-P_{1}\right)\right)=P_{2}^{\prime}$. Hence $P_{2}$ must be a complete path from $s$ to $t$.
If $P_{1}^{\prime} \cup P_{2}^{\prime}$ contains some $P_{1}$ links on each of its two paths, as $-P_{1}$ is the path from $t$ to $s$ in $G\left(V, E^{\prime}\right)$, and neither $P_{1}^{\prime}$ nor $P_{2}^{\prime}$ contains any $-P_{1}$ links, then the union $P_{1}^{\prime} \cup P_{2}^{\prime} \cup\left(-P_{1}\right)$ contains two cycles: one cycle consists of $P_{1}^{\prime}$ and $-P_{1}$, the other consists of $P_{2}^{\prime}$ and $-P_{1}$. When the set $O_{l}$ is removed from the union set, the remaining links compose the logical difference set $P_{2}$. Hence $P_{2}$ must be a complete path from $s$ to $t$.

Part B. Assume that the shortest path in $G\left(V, E^{\prime}\right)$ is $P_{3} \neq P_{2}$, then we must have $l\left(P_{3}\right)<l\left(P_{2}\right)$. As $l\left(P_{2}\right)=l\left(P_{1}^{\prime}\right)+l\left(P_{2}^{\prime}\right)-l\left(P_{1}\right)$ we have $l\left(P_{3}\right)+l\left(P_{1}\right)<l\left(P_{1}^{\prime}\right)+l\left(P_{2}^{\prime}\right)$, which contradicts the assumption that $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ is the optimal set.

### 3.3 LBA Is Loop-free

Many routing algorithms assume non-negative link weights to avoid a loop of negative length appearing on a path. However, negative link weights introduced to a graph in LBA will not cause loops in the routing process.

(a) The shortest path $\boldsymbol{P}_{\mathbf{1}}(s, t)$

(b) A loop containing some $-P_{1}$ link

Figure 5. A loop contains some negative link

Theorem 2: Given a digraph $G(V, E)$ and source-destination pair $(s, t)$ and let $P_{1}$ be the shortest path in this graph. The modified graph $G\left(V, E^{\prime}\right)$ is defined as the graph $G(V, E)$ for which $P_{1}$ is replaced with $-P_{1}$. A loop containing some negative link(s) in $G\left(V, E^{\prime}\right)$ will not have a negative length.
Proof: Assume $s v_{1} \ldots v_{i} v_{i+1} \ldots v_{n} t$ is the shortest path $P_{1}$ from node $s$ to node $t$ in $G(V, E)$, as shown in Figure 5(a). The corresponding path $-P_{1}$ in $G\left(V, E^{\prime}\right)$ (Figure 5(b)) has a link $\left(v_{i+1} \rightarrow v_{i}\right)$ which appears on loop $P_{l}=u_{i} v_{i+1} v_{i} u_{i}$. Suppose the loop $P_{l}$ has a negative length $l\left(P_{l}\right)=w\left(u_{i} \rightarrow v_{i+1}\right)+w\left(v_{i+1} \rightarrow v_{i}\right)+w\left(v_{i} \rightarrow u_{i}\right)<0$. Because $w\left(v_{i+1} \rightarrow v_{i}\right)$ $=-w\left(v_{i} \rightarrow \mathrm{v}_{i+1}\right)$, we must have $w\left(v_{i} \rightarrow u_{i}\right)+w\left(u_{i} \rightarrow v_{i+1}\right)<w\left(v_{i} \rightarrow v_{i+1}\right)$. Hence the sub-path $s v_{1} \ldots v_{i} u_{i} v_{i+1}$ is shorter than the sub-path $s v_{1} \ldots v_{i} v_{i+1}$. This contradicts the assumption that $s v_{1} \ldots v_{i} v_{i+1} \ldots v_{n} t$ is the shortest path. $\square$

### 3.4 Optimality of the solution produced with LBA

Theorem 3. Given a directed graph $G(V, E)$ and source-destination pair ( $s, t$ ), the algorithm LBA returns the optimal set for the LPP problem.

Proof: Let $P_{1}$ be the shortest path in the original graph $G(V, E)$ found in step 1 of LBA and $P_{2}$ be the shortest path in the modified graph $G\left(V, E^{\prime}\right)$, found in step 3 of LBA. $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ is the solution set generated by LBA. The proof consists of three parts.
Part A. (Proof of Link-disjointness) By construction of the solution set, we must have $P_{1}^{\prime} \cap P_{2}^{\prime}=\varnothing$.
Part B. (Proof of Minimal Total Length) Suppose the optimal set of link-disjoint paths is $\left\{P_{1}^{\prime \prime}, P_{2}^{\prime \prime}\right\}$ instead of $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$. According to Lemma 2, the logical difference set of $\left\{P_{1}^{\prime \prime}, P_{2}^{\prime \prime}\right\}$ with $P_{1}$ is the shortest path in the modified graph $G\left(V, E^{\prime}\right)$. This contradicts that $P_{2}$ is the shortest path in modified graph $G\left(V, E^{\prime}\right)$.

Part C. (Proof of Loop-freeness) On Theorem 2, LBA is loop-free. Thus the solution set returned by LBA must be the optimal set.

## 4 Extending LBA to Multiple Dimensions

The extension of LBA to multiple dimensions using SAMCRA [29] is called MLBA (Multiple-constrained LBA). A brief description of SAMCRA, which serves as the multiple-constrained shortest path routing algorithm in MLBA, is given in Section 4.1. The basic steps of MLBA (Multiple-constrained LBA) are presented in Section 4.2. The problems appearing in multiple dimensions are addressed in Section 4.3.

### 4.1 Brief Introduction of SAMCRA

SAMCRA [29] is an exact multiple-constrained routing algorithm based on three concepts: (a) non-linear path length, (b) k-shortest path routing, and (c) non-dominance. The non-linear length function defined in (1) is necessary for exactness and implies that a sub-path of a shortest path is not necessarily a shortest itself. It is therefore necessary to keep track of multiple sub-paths at each intermediate node on the path between a pair of nodes. A (sub)-path $P_{1}$ is dominated by a (sub)-path $P_{2}$ if $w_{m}\left(P_{1}\right) \geq w_{m}\left(P_{2}\right)$, for $m=1, \ldots, M$, with an inequality
sign for at least one $m$. This operation reduces the search space and removes loops from a route when nonnegative link weights are used.

### 4.2 Operations of MLBA

The basic steps of MLBA are the same as those for LBA except that the shortest path routing algorithm is replaced with SAMCRA in MLBA. We will illustrate the operation of MLBA with the example topology shown in Figure 6(a). For the sake of simplicity, we have assigned each link a two-dimensional weight vector, but it is also possible to use an $M$-dimensional weight vector ( $M>1$ ). The complexity of solving the MCLPP problem will increase with $M$, but as shown in [29], the complexity may decrease (and even become polynomial) if $M$ tends to infinity. To solve the MCLPP problem, we are required to find two link-disjoint paths from source node $A$ to destination node $B$ that both obey the constraints vector $\vec{L}=(20,20)$. Among the solutions to MCLPP we prefer the one with the minimum total length. The shortest multiple-constrained path from node $A$ to node $B$ is the path $a c d b$. Its path weight vector is $(4,5)$. The optimal set of two shortest link-disjoint paths (according to (2)) in this topology is $\{a c b, a d b\}$, with path vectors $(5,6)$ and $(5,5)$ respectively and minimum total length $0.3+0.25=$

(a) Step 1

(c) Step 3

(b) Step 2

(d) Step 4

Figure 6. Example of the operation of MLBA
0.55 .

Now let us run MLBA on this topology. In Step 1, the shortest path $P_{1}=a c d b$ is found. In Step 2, the original graph is modified by replacing all the $P_{1}$ links with $-P_{1}$ links. In this case, each component of link weight vector of a $-P_{1}$ link is set negative. For instance, the link $c \rightarrow d$ with weight vector $(1,1)$ is replaced with the link $d \rightarrow c$ with corresponding weight vector $(-1,-1)$. In Step 3, the shortest path in the modified graph found with

SAMCRA is $P_{2}=a d c b$, with path weight vector $(3,3)+(-1,-1)+(4,5)=(6,6)$. In Step 4, the set $O_{l}$ consisting of a pair of opposite $P_{1}$ and $P_{2}$ links $(c \rightarrow d)$ and $(d \rightarrow c)$ are removed from the union of $P_{1}$ and $P_{2}$. Then the optimal solution set $\{a c b, a d b\}$ is returned.

### 4.3 Problems due to the Non-linear Length in Multiple Dimensions

### 4.3.1 Loops caused by Negative Link Weights

For $M=1$, SAMCRA acts just like Dijkstra's algorithm, therefore MLBA reduces to LBA and negative link weights along $-P_{1}$ will not cause a loop in the routing process of MLBA. For $M>1$, Theorem 2 still holds and a loop containing some $-P_{1} \operatorname{link}(\mathrm{~s})$ still has a non-negative length. However, some of the components of the loop weight vector may be negative, causing MLBA to pass this loop a finite number of times. We will explain this looping through Figure 7, where each link possesses two link metrics. Suppose that the shortest path $P_{1}$ is sadf, depicted with bold lines in Figure 7(a). The link weights vector ( $x_{1}, x_{2}$ ) of link $c \rightarrow d$ must be chosen to ensure that the path sacdf is longer than sadf, i.e.

$$
\max \left[\begin{array}{l}
w_{1}(s \rightarrow a)+w_{1}(a \rightarrow c)+x_{1}+w_{1}(d \rightarrow f) \\
w_{2}(s \rightarrow a)+w_{2}(a \rightarrow c)+x_{2}+w_{2}(d \rightarrow f)
\end{array}\right]>\max \left[\begin{array}{l}
w_{1}(s \rightarrow a)+w_{1}(a \rightarrow d)+w_{1}(d \rightarrow f) \\
w_{2}(s \rightarrow a)+w_{2}(a \rightarrow d)+w_{2}(d \rightarrow f)
\end{array}\right] .
$$

Numerically,

$$
\max \left[\begin{array}{l}
x_{1}  \tag{4}\\
x_{2}
\end{array}\right]>\max \left[\begin{array}{l}
w_{1}(a \rightarrow d)-w_{1}(a \rightarrow c) \\
w_{2}(a \rightarrow d)-w_{2}(a \rightarrow c)
\end{array}\right]=\left[\begin{array}{l}
5-2 \\
4-1
\end{array}\right]=\left[\begin{array}{l}
3 \\
3
\end{array}\right]
$$

After Step 2 of MLBA is executed, there appears a loop $P_{l}=d a c d$ shown with double lines in Figure 7(b), containing the link $d \rightarrow a$ with negative link weights $(-5,-4)$.

(a) The shortest path is sadf.

(b) The loop dacd contains a negative link $d \rightarrow a$.

Figure 7. Non-dominance may fail to remove a loop in the case of $\boldsymbol{M} \boldsymbol{> 1}$.
If equation (4) holds and each component of vector $\left(x_{1}, x_{2}\right)$ is greater than 3 , then the sub-path sdacd will be dominated by the direct link $s \rightarrow d$ with weight vector $(5,8)$ and will be removed by the non-dominance check in SAMCRA. However, if equation (4) holds but one component of ( $x_{1}, x_{2}$ ) is not greater than 3 , say $x_{1}<3, x_{2}>3$, then

$$
\left[\begin{array}{l}
w_{1}\left(P_{l}\right)=w_{1}(d \rightarrow a)+w_{1}(a \rightarrow c)+w_{1}(c \rightarrow d)=(-5)+2+x_{1}<0 \\
w_{2}\left(P_{l}\right)=w_{2}(d \rightarrow a)+w_{2}(a \rightarrow c)+w_{2}(c \rightarrow d)=(-4)+1+x_{2}>0
\end{array}\right],
$$

where $\left(w_{1}\left(P_{l}\right), w_{2}\left(P_{l}\right)\right)$ is the path vector of the loop $P_{l}$. In this case, the sub-path sdacd is not dominated by the link $s \rightarrow d$, although $l\left(P_{l}\right)>0$. Hence, loops can occur in MLBA that continue until one of the constraints is violated. Unfortunately, checking all paths to assure that they are loop-free is computationally too expensive.

As mentioned in Section 2, in Suurballe's algorithm [25] and the S-T algorithm [26], a transformation of link weights $w^{\prime}(u \rightarrow v)=w(u \rightarrow v)+d(u)-d(v)$ is applied to each link, where $d(u)$ is the distance from source node $s$ to node $u$ on the shortest path tree. This transformation is made to guarantee that the links on the shortest path tree have zero link weights and those links not on the tree have link weights greater than zero in the modified graph. However, an artifact of a non-linear length is that subsections of shortest paths are not necessarily shortest paths [5][29]. Consequently, for $M>1$, Suurballe's transformation cannot ensure non-negative link weights and loops may emerge.

### 4.3.2 Total Length of the Solutions Produced with MLBA

We assume for the moment that the constraints are large enough such that all paths are feasible. If $M=1$, it is proved in Section 3.4 that the solution set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ produced with MLBA has the minimum total length. With the total length defined in (2), Lemma 1 in section 3 still holds for $M>1$. The optimal solution set of two linkdisjoint multiple-constrained paths with minimum total length either contains the first shortest path $P_{1}$ itself or some $P_{1}$ links on each of its two paths. Also, the optimal set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ still obeys Property 1. Unfortunately, the logical difference set $\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right)-P_{1}$ is not necessarily the shortest path $P_{2}$ in the modified graph, since $l\left(\left(P_{1}^{\prime} \cup P_{2}^{\prime}\right)-P_{1}\right)=l\left(P_{1}^{\prime}\right)+l\left(P_{2}^{\prime}\right)-l\left(P_{1}\right)$ does not necessarily hold for $M>1$. Hence, Lemma 2 may not hold for $M$ $>1$ and the solution set constructed based on $P_{1}$ and $P_{2}$ is not necessarily the optimal set with minimum total length. Moreover, the solution set may also violate the constraints or a feasible solution may not be found.

## 5 DIMCRA

In the previous section, we have shown that it is not trivial to extend LBA to multiple dimensions. Due to the problems existing in MLBA, we propose a heuristic algorithm DIMCRA (link-Disjoint Multiple Constraints Routing Algorithm) for the MCLPP problem.

### 5.1 Operations of DIMCRA

DIMCRA $(G, s, t)$ : Given a directed graph $G(V, E)$, a constraint vector $\vec{L}$ and a source-destination pair $(s, t)$,
Step 1. Find the shortest path $P_{1}$ obeying $\vec{L}$ with SAMCRA; if $P_{l}$ does not exist, then stop;
Step 2. Reverse the direction of all the links on the shortest path $P_{1}$, and set the sign of their link weights zero, $w_{m}(v \rightarrow u)=0, \forall(u \rightarrow v) \in P_{1}$ and $m=1, \ldots, M$. A modified graph $G^{\prime}$ is created;

Step 3. Find the shortest path $P_{2}$ constrained by $2 \vec{L}$ in the modified graph $G^{\prime}$ with SAMCRA; if $P_{2}$ does not exist, then stop;
Step 4. Make the union of $P_{1}$ and $P_{2}$, remove from the union the $P_{1}$ links whose reversed links appear on $P_{2}$, and vice versa, then group the remaining links into a set of two paths $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$, i.e. $P_{1}^{\prime} \cup P_{2}^{\prime}=\left(P_{1} \cup P_{2}\right) \backslash\left(P_{1} \widetilde{\cap} P_{2}\right)$.

Step 5. Check the length of each path in the set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$. If the path $P_{i}^{\prime}(1 \leq i \leq 2)$ violates the constraints, then update the modified graph $G^{\prime}$ by removing the link set $P_{i}^{\prime} \backslash\left(P_{i}^{\prime} \cap P_{1}\right)$ from it, and go to Step 3 . Otherwise stop and return the current solution set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$.

Compared to MLBA, DIMCRA uses a different transformation to create the modified graph. In Step 2 of DIMCRA, the shortest path links are still reversed in direction but the corresponding direction-reversed links are assigned with zero link weight vectors instead of negative ones. Therefore the loop problem caused by negative link weights that mainly destroys the efficiency of MLBA is bypassed. In MLBA, $P_{2}$ is required to obey the constraints, which may cause some feasible sets to be ignored by MLBA. In fact, when $\vec{w}\left(P_{2}\right)>\vec{L}$, if $P_{2}$ contains no reversed $P_{1} \operatorname{link}(\mathrm{~s})$, then $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ is actually $\left\{P_{1}, P_{2}\right\}$ and cannot be a feasible set. But if $P_{2}$ contains some reversed $P_{1} \operatorname{link}(\mathrm{~s})$, it is possible that $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ is a feasible set, for instance, $l\left(P_{1}\right)=0.6, l\left(P_{1}^{\prime}\right)=0.8, l\left(P_{2}^{\prime}\right)=0.9$, and $l\left(P_{2}\right)=1.1$. However, if $\vec{w}\left(P_{2}\right)>2 \vec{L}$, then we must have $\vec{w}\left(P_{1}^{\prime}+P_{2}^{\prime}\right)=\vec{w}\left(P_{2}+P_{1}-P_{1 r}\right)>\vec{w}\left(P_{2}\right) \geq 2 \vec{L}$, where $P_{1 r}$ denotes the set of $P_{1}$ links whose reversed links appear on $P_{2}$, and $P_{1 r}$ must be a proper subset of $P_{1}$. Therefore, in Step 3 of DIMCRA, the constraint check on path $P_{2}$ in SAMCRA is performed with $2 \vec{L}$ as the constraints vector, otherwise a feasible solution set may not be found. We have also added an extra step, Step 5 of DIMCRA, to check that the constraints are obeyed. If only with $\vec{w}\left(P_{1}^{\prime}+P_{2}^{\prime}\right) \leq 2 \vec{L}$ DIMCRA does not always ensure both paths within constraints, i.e. $\vec{w}\left(P_{1}^{\prime}\right) \leq \vec{L}$ and $\vec{w}\left(P_{2}^{\prime}\right) \leq \vec{L}$. Hence Step 5 of DIMCRA checks both paths in the solution set returned at Step 4. If each of them obeys the constraints, DIMCRA will return the solution set and stop. On the other hand, if either of them does not obey the constraints, DIMCRA is redirected to Step 3 to continue the search for a feasible set. In Step 3, if no $P_{2}$ exists, DIMCRA will stop with no solution. We will illustrate the operation of DIMCRA with the following examples.

Example 1: Consider the example graph in Figure 8(a). We are required to find a set of two link-disjoint paths between $a$ and $b$, each within the constraints $\vec{L}=(20,20)$ and preferably with the minimum total length. In Step 1, the shortest path $P_{1}=a c d b$ is found. In Step 2, each $P_{1}$ link is reversed and is assigned with zero link weights. In Step 3, the shortest path in the modified graph $G$ ' is found as $P_{2}=a d c b$, with path vector $(3,3)+(0,0)$ $+(4,5)=(7,8)$, as shown with bold lines in Figure 8(c). In Step 4, only for the $P_{1}$ link $c \rightarrow d$, its reversed link $d \rightarrow$ $c$ appears on $P_{2}$ and vice versa. Thus these two links are removed from the union of $P_{1}$ and $P_{2}$, and the remaining


## Figure 8. Example 1 of the operation of DIMCRA

links are grouped into a set of two paths $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}=\{a c b, a d b\}$, shown with bold lines. In Step 5, the constraints check is executed on both paths. As each of them obeys the constraints, DIMCRA stops. In this case, the optimal solution set of $\{a c b, a d b\}$ is returned by DIMCRA. The solution set that would have been returned by RF, is not optimal.

Example 2: Consider the graph in Figure 9(a), which is the same as in the previous example except that the


Figure 9. Example 2 of the operation of DIMCRA
link $e \rightarrow b$ is assigned a different vector $(2,1)$. The constraints remain the same. In this example the optimal set of

two link-disjoint multiple-constrained paths is still the set $\{a d b, a c b\}$ with path vectors $(5,5)$ and $(5,6)$ respectively, and the minimum total length $5 / 20+6 / 20=0.55$. In Step 3, the shortest path in the modified graph is found as $P_{2}=a e b$ with path vector (7,7), shown in Figure 9(c). In Step 4, as for each $P_{1}$ link, its reversed link does not appear on $P_{2}$, or vice versa, the solution set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}$ is constructed as $\{a c d b$, aeb $\}$, exactly $P_{1}$ and $P_{2}$ themselves. The total length of $\{a c d b, a e b\}$ is $5 / 20+7 / 20=0.6$. In this example, DIMCRA failed to return the optimal set, but DIMCRA's solution set is close to the optimal one and both paths obey the constraints. RF would have returned the same solution.

Example 3. We again consider Example 2 except with different constraints ( 6,6 ). Running DIMCRA, we obtain the same results as in Example 2 (for Step 1 to Step 4). But in Step 5, when the constraints check is made on each path in the solution set $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}=\{a c d b, a e b\}$, the longer path $P_{2}^{\prime}=a e b$ with path vector $(7,7)$ does not obey the constraints. This means that the currently built solution set is not feasible. The links that only appear on $P_{2}^{\prime}=a e b$, i.e. link $a \rightarrow e$ and $e \rightarrow b$, are removed from the modified graph shown in Figure 9(b). The updated modified graph is shown in Figure 10(a) and DIMCRA is redirected to Step 3. In Step 3, a shortest path in the updated modified graph is found as $P_{2}=a d c b$, depicted in Figure $10(b)$. In Step 4, the solution set is $\{a c b, a d b\}$, as shown in Figure 10(c). At last, in Step 5, each path in the current solution set obeys the constraints. The optimal set $\{a c b, a d b\}$ is returned and DIMCRA stops. RF would have failed to return a solution.

In Step 3 the constraints are set to $2 \vec{L}$. With these modified constraints, if the shortest path $P_{2}$ in the modified graph violates the constraints $\vec{L}$ but obeys $2 \vec{L}$, it can be returned by SAMCRA in Step 3, hence a feasible set

(a) Step 1
(c) Step 3


(b) Step 2

(d) Step 4

(e) Step 5

Figure 11. Example of the operation of DIMCRA with constraints $(\mathbf{1 0 , 1 0})$
related to such kind of $P_{2}$ will not be ignored, as illustrated in Example 3. Moreover, if a path $P_{2}$ does not exist in the updated modified graph, DIMCRA will stop. Thus DIMCRA will not bounce back and forth between Step 3 and 5.

With the constraints check on each path in the solution set, Step 5 guarantees that DIMCRA returns a feasible set of two link-disjoint multiple-constrained paths, as illustrated in the above examples.

However it may occur, as illustrated in Figure 11, that DIMCRA cannot return a feasible set even if there exists one. The RF method also fails to return the feasible set in this case.

### 5.2 Properties of DIMCRA

As proved in Section 3, the way to construct a solution set by reversing the shortest path $P_{1}$, finding a shortest path $P_{2}$ in the modified graph and constructing the solution set based on these two shortest paths $P_{1}$ and $P_{2}$ guarantees the disjointness of the two paths in the solution set. Setting the direction-reversed $P_{1}$ links with zero link weights guarantees the loop-freeness of DIMCRA. For, if no negative link weights are used in a graph, a loop can be avoided by the non-dominance check in SAMCRA. Comparing with the operation of setting direction-reversed $P_{1}$ links negative, the operation of setting such reversed $P_{1}$ links with zero link weights still encourages the choice of such reversed $P_{1}$ links on a path but with less intensity.
Unfortunately DIMCRA does not always find the set of feasible link-disjoint paths. Hence, it may be possible to further optimize DIMCRA, such that it can guarantee to always find a set of feasible link-disjoint paths, if they exist. However, DIMCRA in its current state is better than the RF method (as was indicated in the examples). Both methods return the same solution when $\left\{P_{1}^{\prime}, P_{2}^{\prime}\right\}=\left\{P_{1}, P_{2}\right\}$ and $P_{1} \cap P_{2}=\varnothing$. In all other cases DIMCRA
either returns a more optimal solution than RF or RF does not find a solution where DIMCRA does. Since, to our knowledge, no other algorithms for solving MCLPP exist, the performance of DIMCRA is difficult to assess.

## 6 Conclusions

The Link-disjoint path problem occurs in network design where aspects as survivability, load balancing and network resource utilization are strived for. This problem has barely been investigated in the QoS routing context where a path is characterized by multiple metrics. A simple algorithm for solving the LPP problem for is presented in this paper. The problems surrounding the extension of this simple algorithm to multiple dimensions are discussed. A heuristic algorithm DIMCRA is proposed to find link-disjoint multiple-constrained paths between a pair of source and destination nodes. If DIMCRA returns a link-disjoint pair of paths they always obeys the constraints. However, DIMCRA's solution is not necessarily optimal in terms of minimizing the total length of the returned paths or guaranteeing to always find the feasible set. Its performance however is better than the simple Remove-Find method.

Some open issues remain, namely: making DIMCRA exact whilst still efficient, allowing maximally disjoint paths or bridges and simulating the performance.

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[^1]:    ${ }^{1}$ With the definition of length in (1), we have $l\left(-P_{1}\right)=-l\left(P_{1}\right)$ only for $M=1$.

[^2]:    ${ }^{2}$ If there exist more than one shortest path in the original graph or in the modified graph, either one of them can be chosen. Choosing different shortest paths may lead to different solution sets. However, these solution sets will have the same minimum total length.

[^3]:    ${ }^{3}$ The logical difference set $P_{2}-P_{1}$ also can be computed as $P_{2}-P_{1}=\left\{(u \rightarrow v) \mid(u \rightarrow v) \in P_{2} \backslash\left(P_{2} \cap P_{1}\right)\right\} \cup\{(v \rightarrow u)$ $\left.\mid(u \rightarrow v) \in P_{1} \backslash\left(P_{2} \cap P_{1}\right)\right\}$, which means that if a link $u \rightarrow v$ of $P_{2}$ does not appear on $P_{1}$, then this link belongs to the difference set $P_{2}-P_{1}$, and if a link $u \rightarrow v$ of $P_{1}$ does not appear on $P_{2}$, then its direction reversed link $v \rightarrow u$ belongs to the difference set $P_{2}-P_{1}$, with a link weight $w(v \rightarrow u)=-w(u \rightarrow v)$. In set theory, the difference operation is defined as $P_{2}-P_{1}=P_{2} \backslash\left(P_{1} \cap P_{2}\right)$, and the symmetric difference operation is defined as $P_{2}-P_{1}=$ $\left(P_{2} \cup P_{1}\right) \backslash\left(P_{1} \cap P_{2}\right)$. The concept of logical difference set in this paper resembles the symmetric difference set but it is not the same.

