



# A restorable MPLS-based hose-model VPN network

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## Abstract

Design of a restorable MPLS-based Layer-3 VPN network with QoS guarantee is a new and important subject that has not been widely studied before. The main challenge arises from the fact that the Service Level Agreements (SLAs) of a L3-VPN usually only specify the maximum ingress and egress traffic rate, and provide no point-to-point traffic matrix information (i.e., a hose-model VPN). Conventional restoration and traffic engineering techniques do not apply to this type of traffic model. In this paper, we present a restoration network architecture and present two algorithms for solving the routing problem of this type of restoration networks. We demonstrate the effectiveness of our proposed restoration architecture by comparing the throughput performance with other approaches.

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## 1. Introduction

### 1.1. MPLS-based L3-VPN

MPLS-based VPN services can help an enterprise to converge existing disparate networks onto a consolidated, end-to-end infrastructure that can support combined data, voice, and video services. The underlying concept for VPN implementation is MPLS's *label stacking*. We can use a two-layer label stack: an external label and an internal label. The external label is the routing label and is used by the core router to route a packet to its destination

edge router. The internal label is the VPN label and it is used by the edge router to separate traffic from different VPNs (Figs. 1 and 2).

Depending on the protocol levels involved in constructing a VPN, we can divide VPNs into two types: Layer-2 (L2) and Layer-3 (L3) VPNs. A L2 VPN provides a secure point-to-point transport service. But if an enterprise needs connectivity among  $n$  points, full connectivity among them requires  $n(n-1)$  L2 VPN links. It is obviously not economical unless  $n$  is small. A L3 VPN (MPLS-based), on the other hand, is a set of sites of which the connectivity is provided by a provider's MPLS routing network and there are no virtual links set up directly among these sites. L3 VPNs provide three key benefits to enterprises: any-to-any connectivity through the use of forwarding tables, the ability to retain

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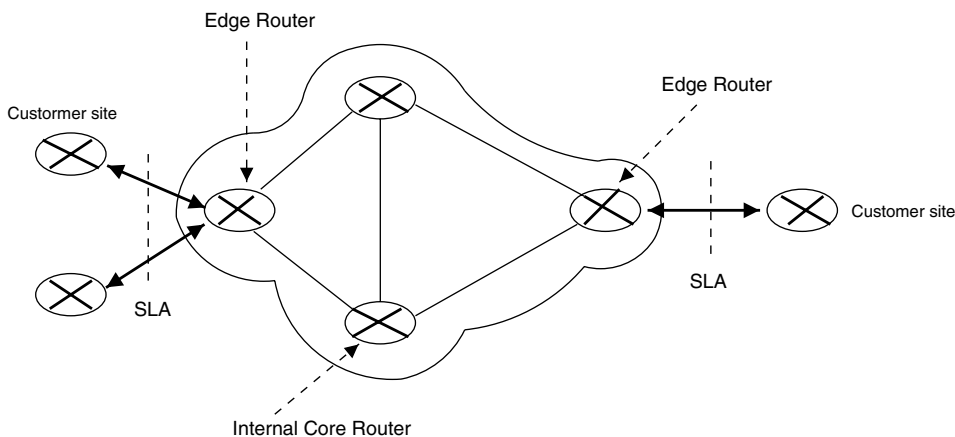


Fig. 1. An MPLS backbone network.

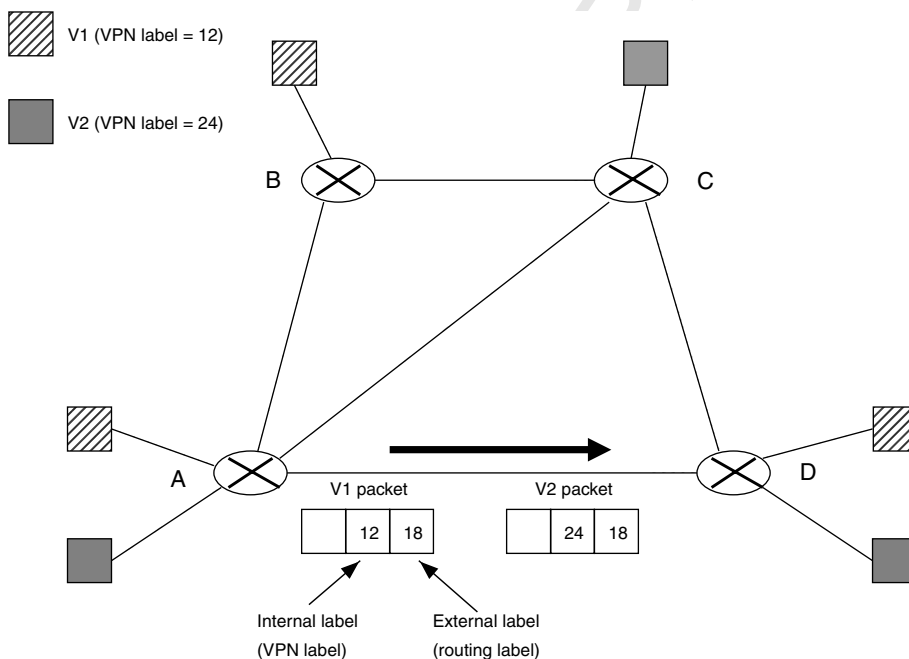


Fig. 2. There are two labels in a VPN packet. The internal label is the VPN label, and the external label is the routing label.

47 existing IP addressing plan by supporting overlapping  
48 IP addresses, and greater scalability at the  
49 site-to-site and data center levels [1].

50 To support the QoS of a L3 VPN, customers are  
51 required to sign a Service Level Agreement (SLA)  
52 with the provider (Fig. 1). The SLA of a L3-VPN  
53 usually only specifies the ingress and egress rate,  
54 but provides no destination information. This type  
55 of VPNs are usually called hose-model VPNs [2].  
56 The other type of rate specification is the pipe-  
57 model that requires the customer to specify the traf-

fic rate between each source-destination pair. Hose-  
58 model VPNs obviously are much easier to use for  
59 customers.  
60

61 This paper will focus on hose-model L3 VPNs.  
62 Conventional MPLS traffic engineering tools usu-  
63 ally assume that traffic matrix  $T = \{d_{ij}\}$  of the  
64 VPN to be set up is given, where  $d_{ij}$  represents the  
65 average traffic intensity from node  $i$  to node  $j$ . How-  
66 ever, as pointed above a hose-model L3-VPN does  
67 not always provide that information. Instead, only  
68 the row sums  $\sum_j d_{ij} = \alpha_i$ , where  $\alpha_i$  (the ingress band-

width constraint) is the maximum rate of traffic that node  $i$  can send into the network, and *column* sums  $\sum_i d_{ij} = \beta_j$ , where  $\beta_j$  (the *egress* bandwidth constraint) is the maximum rate of traffic that node  $j$  can receive from the network, are given. To guarantee the QoS of a hose-model L3-VPN is a challenging task because for a given set of traffic constraints  $\alpha_i$  and  $\beta_i$ , there are many traffic matrices that can satisfy the constraints. Therefore, we must provide enough bandwidth for any traffic matrix that meets the ingress and egress constraints. The uncertainty inherited in a hose-model traffic pattern makes finding efficient routing and capacity planning algorithms a difficult task.

Many algorithms have been proposed for hose-model VPN provisioning [3–6]. But there is little work on hose-model VPN restoration. This is the focus of the paper. When a link or node fails, traversing traffic needs to be rerouted through alternate paths. There are two restoration approaches: link restoration and path restoration. In *link restoration* (also referred to as local restoration or fast restoration), each link of the network is protected by a set of pre-determined detour paths that connect the two endpoints of that link. Upon a link failure, traffic of the link is switched to the detour protecting paths. In *path restoration*, each path carrying working traffic is protected by a diverse backup path. Link restoration can be activated immediately when link failures are detected. In contrast, path restoration can be activated only after the failure information propagates to the source node. Link restoration is the preferred method for providing fast restoration in MPLS and optical networks [7,8] and we will focus on link restoration in the paper.

## 1.2. Prior works and our contributions

There are many papers on restoration, but few are related to hose-model traffic patterns. Ref. [9] presented the linear programming based approaches to solve the optimal capacity and flow assignment problem with link and path restoration strategies. Refs. [10,11] tackled a similar problem in mesh-based WDM optical networks. In all these papers, traffic matrix  $T$  is assumed given. Consequently they do not apply to a hose-model problem. For hose-model VPN protection, Ref. [12] proposed a restoration algorithm for a single hose-model VPN construction. It constructs a tree with each link protected by a detour path. In case of a link failure, traffic will be rerouted and the overall topology is

still a tree. This is obviously not an efficient protection scheme as we need to do it for every VPN. A recent paper [13] proposed to create a fully-connected virtual topology on top of the physical topology. It then routes packets across two-hop logical links in the fully-connected virtual topology. Some obvious drawbacks of the approach include longer paths, and thus longer end-to-end delays, and low efficiency. These problems become more severe for networks with sparse links as creating a fully-connected topology out of these networks will be more expensive than networks with dense links.

In this paper, we propose a restorable MPLS VPN network architecture with QoS guarantees. Our contributions include the following:

1. We present a restorable MPLS-based network architecture for supporting L3 VPNs with QoS guarantees. The protection is done for all VPNs, not just for one VPN.
2. We present an efficient decomposition algorithm to compute the optimal routes for this restoration network. This algorithm can include a hop-count limit on the restoration paths.
3. We compare the throughput performance of the new restoration architecture with that of conventional architectures. The results show that our proposed architecture can achieve a better performance for hose-model VPNs under different bandwidth requirements.

The rest of the paper is organized as follows. Section 2 presents the new restorable network architecture and the linear programming formulation for designing the working and restoration paths for the network. Section 3 presents an efficient decomposition algorithm. Section 4 presents performance evaluation of the proposed schemes. We conclude our discussion in Section 5.

## 2. Restorable L3-VPN network with QoS guarantee

We consider single-link failures [10]. Single-node failures can be analyzed similarly. The analysis in [5] showed that splitting traffic among multiple paths will have a much better performance than a single path approach for supporting hose-model VPNs. We will use the multi-path approach in this paper. Multiple paths for a given source-destination pair will be set up and load-balancing among them is done according to a set of pre-determined splitting ratios derived from the routing computation.

168 Load-balancing can be done such that no out-of-  
169 sequence transmissions occur for packets belonging  
170 to the same flow [14].

### 171 2.1. Network architecture

172 Our architecture is based on the non-blocking  
173 network approach outlined in [15]. The advantages  
174 of this approach for restoration will be discussed  
175 later. Let  $(\theta\tilde{\alpha}_i, \theta\tilde{\beta}_i)$  represent the maximum amount  
176 of total ingress and egress VPN traffic allowed to  
177 enter and leave the network at the edge router  $i$ ,  
178 where  $\tilde{\alpha}_i$  and  $\tilde{\beta}_i$  are given constants, describing the  
179 degree of unevenness of amount of traffic at each  
180 node. For example, suppose a network specifies  
181  $(\tilde{\alpha}_1 = 4, \tilde{\beta}_1 = 4)$  and  $(\tilde{\alpha}_2 = 12, \tilde{\beta}_2 = 12)$ , then it  
182 means that the amount of traffic allowed at edge  
183 router 2 (both ingress and egress) is three times that  
184 of router 1. Note that only the relative – not absolute –  
185 magnitudes of  $\tilde{\alpha}_i$  and  $\tilde{\beta}_i$  have significance as the  
186 real amount of admissible traffic is determined  
187 by  $\theta$ . We need to design the network such that as  
188 long as the ingress and egress traffic of node  $i$  is  
189 below  $(\theta\tilde{\alpha}_i, \theta\tilde{\beta}_i)$ , traffic routed through a link is  
190 always below its link capacity. A network with this  
191 property is called non-blocking in [15]. If the net-  
192 work is non-blocking, we only need to check if there  
193 is enough bandwidth left at the edge routers to  
194 which endpoints of the VPN are connected. There  
195 is no need to check the internal paths' available  
196 bandwidths. The decision of admitting a VPN is  
197 greatly simplified.

### 198 2.2. Optimal routing and link restoration

199 Our goal is to compute the working and link-rest-  
200 oration paths to maximize the admissible amount  
201 of traffic (i.e.,  $\theta$ ). Recall that  $\tilde{\alpha}_i$  and  $\tilde{\beta}_i$  are given con-  
202 stants and their values are determined from past  
203 traffic demands (if there is no prior information  
204 about the network, we can simply assume the  
205 ingress–egress capacity at an edge node is propor-  
206 tional to the total capacity of network links incident  
207 at that node). The network is described as a directed  
208 graph  $G(V, E)$ , where  $V$  is the set of vertices (nodes)  
209 and  $E$  is the set of links. Let  $Q \subseteq V$  be the set of edge  
210 routers through which traffic is admitted into the  
211 network. We first introduce the following nota-  
212 tions:  $c_e$  The capacity for link  $e \in E$   
213  $\chi_{ij}^e$  The routing variable, representing the por-  
214 tion of working traffic from node  $i \in Q$  to  
215  $j \in Q$  routed through link  $e$ .

$A(e)$  the capacity on link  $e$  reserved for working  
traffic. When link  $e$  fails,  $A(e)$  amount of  
traffic has to be rerouted along a set of res-  
toration (detour) paths that connect the  
two endpoints of this link.

$y_f^e$  The amount of restoration traffic that will  
be routed through link  $e$  in case link  $f$  fails.

Given the ingress and egress traffic constraints at  
all the edge nodes, a traffic matrix is called *valid* if it  
satisfies the specified traffic constraints. Let  $H =$   
 $[(\tilde{\alpha}_1, \tilde{\beta}_1), \dots, (\tilde{\alpha}_n, \tilde{\beta}_n)]$  and  $\tilde{H} = [(\theta\tilde{\alpha}_1, \theta\tilde{\beta}_1), \dots, (\theta\tilde{\alpha}_n,$   
 $\theta\tilde{\beta}_n)]$ . Let  $M$  be the set of valid traffic matrices  
 $T = \{d_{ij}\}$  constrained by  $H$ . If  $T \in M$ , then  $\theta T$  will  
be a valid traffic matrix for the constraint  $\tilde{H}$ . In  
the following discussion, we consider  $T \in M$  and  
use the form  $\theta T$  to indicate a valid traffic matrix  
constrained by  $\tilde{H}$ .

Our problem is to determine the working flow  $\chi_{ij}^e$   
and the restoration flow  $y_f^e$  that can maximize  $\theta$ . We  
formulate it as the following:

$$\max \theta \quad (1a)$$

$$\text{s.t. } \sum_{e \in \Gamma^+(v)} \chi_{ij}^e - \sum_{e \in \Gamma^-(v)} \chi_{ij}^e = 0, \quad i, j \in Q, \quad v \in V, \quad v \neq i, j \quad (1b)$$

$$\sum_{e \in \Gamma^+(v)} \chi_{ij}^e - \sum_{e \in \Gamma^-(v)} \chi_{ij}^e = 1, \quad i, j \in Q, \quad v \in V, \quad v = i \quad (1c)$$

$$\sum_{e \in \Gamma^+(v)} \chi_{ij}^e - \sum_{e \in \Gamma^-(v)} \chi_{ij}^e = -1, \quad i, j \in Q, \quad v \in V, \quad v = j \quad (1d)$$

$$\sum_{i, j \in Q} \chi_{ij}^e \cdot (\theta d_{ij}) \leq A(e), \quad e \in E, \quad T \in M \quad (1e)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = 0, \quad f = (o, t) \in E, \quad v \neq o, t \quad (1f)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = A(f), \quad f = (o, t) \in E, \quad v = o \quad (1g)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = -A(f), \quad f = (o, t) \in E, \quad v = t \quad (1h)$$

$$A(e) + y_f^e \leq c_e, \quad e, f \in E, \quad e \neq f \quad (1i)$$

$$y_e^e = 0, \quad e \in E \quad (1j)$$

$$\chi, y, \theta, A \geq 0 \quad (1k) \quad 239$$

where  $\Gamma^+(v)$  and  $\Gamma^-(v)$  are the set of outgoing and  
incoming links of node  $v$ , and  $o$  and  $t$  represent  
the originating and terminating nodes of link  $f$ . 242

Constraints (1b)–(1d) represent the flow conservation constraints for working traffic at intermediate, source, and destination nodes and constraints (1f)–(1h) represent the flow conservation constraints for restoration traffic. Constraint (1e) ensures that the total amount of working traffic on any link does not exceed the working capacity  $A(e)$ . Constraint (1i) ensures that the sum of working traffic and the restoration traffic that appears on a link due to failure of any other link does not exceed the link capacity. Constraints (1j,1k) provide the ranges for the variables.

Constraint (1e) is not a linear constraint. But we can introduce a new routing variable  $x_{ij}^e = \lambda_{ij}^e \cdot \theta$  and rewrite Eq. (1) as the following:

$$\max \theta \quad (2a)$$

$$\text{s.t. } \sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = 0, \quad i, j \in Q, \quad v \in V, \quad v \neq i, j \quad (2b)$$

$$\sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = \theta, \quad i, j \in Q, \quad v \in V, \quad v = i \quad (2c)$$

$$\sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = -\theta, \quad i, j \in Q, \quad v \in V, \quad v = j \quad (2d)$$

$$\sum_{i, j \in Q} x_{ij}^e d_{ij} \leq A(e), \quad e \in E, \quad T \in M \quad (2e)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = 0, \quad f = (o, t) \in E, \quad v \neq o, t \quad (2f)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = A(f), \quad f = (o, t) \in E, \quad v = o \quad (2g)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = -A(f), \quad f = (o, t) \in E, \quad v = t \quad (2h)$$

$$A(e) + y_f^e \leq c_e, \quad e, f \in E, \quad e \neq f \quad (2i)$$

$$y_e^e = 0, \quad e \in E \quad (2j)$$

$$x, y, \theta, A \geq 0 \quad (2k)$$

Although Eq. (2) is a linear programming formulation, it cannot be solved directly because constraint (2e) lists every valid  $T$  in  $M$  and there are too many of them. The problem is solved with the following property. Different forms of this property have been given in [16,17].

**Property 1.** Given  $H = [(\tilde{\alpha}_1, \tilde{\beta}_1), \dots, (\tilde{\alpha}_n, \tilde{\beta}_n)]$ , routing  $x_{ij}^e$  and working capacity reservation  $A(e)$  can satisfy constraint (2e) for all traffic matrices in  $M$  if and only if there exist non-negative weights  $\pi_e(i)$  and  $\lambda_e(i)$  for each  $e \in E$  and  $i \in Q$  such that

$$(i) \sum_{i \in Q} \tilde{\alpha}_i \pi_e(i) + \sum_{i \in Q} \tilde{\beta}_i \lambda_e(i) \leq A(e) \quad \text{for each } e \in E. \quad (272)$$

$$(ii) x_{ij}^e \leq \pi_e(i) + \lambda_e(j) \quad \text{for each } e \in E \text{ and every } i, j \in Q. \quad (274)$$

**Proof.** The proof is provided in Appendix A.  $\square$  (278)

Property 1 allows us to replace constraint (2e) in Eq. (2) with requirements (i)–(ii) in Property 1 and transform the formulation into the following:

$$\max \theta \quad (3a)$$

$$\text{s.t. } \sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = 0, \quad i, j \in Q, \quad v \in V, \quad v \neq i, j \quad (3b)$$

$$\sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = \theta, \quad i, j \in Q, \quad v \in V, \quad v = i \quad (3c)$$

$$\sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = -\theta, \quad i, j \in Q, \quad v \in V, \quad v = j \quad (3d)$$

$$\sum_{i \in Q} \tilde{\alpha}_i \cdot \pi_e(i) + \sum_{i \in Q} \tilde{\beta}_i \cdot \lambda_e(i) \leq A(e), \quad e \in E \quad (3e)$$

$$x_{ij}^e \leq \pi_e(i) + \lambda_e(j), \quad i, j \in Q, \quad e \in E \quad (3f)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = 0 \quad f = (o, t) \in E, \quad v \neq o, t \quad (3g)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = A(f), \quad f = (o, t) \in E, \quad v = o \quad (3h)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = -A(f), \quad f = (o, t) \in E, \quad v = t \quad (3i)$$

$$A(e) + y_f^e \leq c_e, \quad e, f \in E, \quad e \neq f \quad (3j)$$

$$y_e^e = 0, \quad e \in E \quad (3k)$$

$$x, y, \pi, \lambda, \theta, A \geq 0 \quad (3l) \quad (284)$$

The above linear programming (LP) problem can be solved by standard LP solvers like Cplex [18]. Then we can derive the set of working paths and link restoration paths from the flow variables  $x_{ij}^e$  and  $y_f^e$ . (285–288)

### 3. Adding hop-count limit to restoration paths (289)

If the goal is only to maximize network throughput, some of the computed restoration paths may (290–291)

be long and this makes the restoration latency unacceptable. In the following we present a decomposition scheme that will lead to a path formulation for the design of restoration paths. The path formulation allows us to add a hop count limit on the restoration paths. Another benefit of the approach is that the new approach is faster than the one in Section 2.

### 3.1. Two-stage decomposition algorithm

The computation of the working flow and the restoration flow in Eq. (3) can be partitioned into two separate stages. At the first stage, corresponding to constraints (3b)–(3f), we assume the working capacity vector  $A$  is given, where

$$A = [A(1), A(2), \dots, A(m)] \quad (4)$$

and  $m$  is the number of links (assuming the set of links is labeled from 1 to  $m$ ). We can compute the optimal routing. The process also generates a new working capacity vector  $\tilde{A}$ . A working capacity vector is called *feasible* if it satisfies constraints (3g)–(3k), meaning that the network has enough capacity left to protect it. At the second stage, we will test if the newly generated  $\tilde{A}$  from stage 1 is feasible or not. If not, we will modify  $\tilde{A}$  in the 2nd stage to make it feasible and pass the result back to stage 1 for another round of iteration.

*Stage 1:* Assume  $A$  (i.e., all  $A(e)$ ) is given. We determine routing and maximum  $\theta$  by solving the following linear programming problem.

$$\max \theta \quad (5a)$$

$$\text{s.t. } \sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = 0, \quad i, j \in Q, \quad v \in V, \quad v \neq i, j \quad (5b)$$

$$\sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = \theta, \quad i, j \in Q, \quad v \in V, \quad v = i \quad (5c)$$

$$\sum_{e \in \Gamma^+(v)} x_{ij}^e - \sum_{e \in \Gamma^-(v)} x_{ij}^e = -\theta, \quad i, j \in Q, \quad v \in V, \quad v = j \quad (5d)$$

$$\sum_{i \in Q} \tilde{\alpha}_i \cdot \pi_e(i) + \sum_{i \in Q} \tilde{\beta}_i \cdot \lambda_e(i) \leq A(e), \quad e \in E \quad (5e)$$

$$x_{ij}^e \leq \pi_e(i) + \lambda_e(j), \quad i, j \in Q, \quad e \in E \quad (5f)$$

$$x, \pi, \lambda, \theta \geq 0 \quad (5g)$$

We use  $\theta(A)$  to denote the optimal value of  $\theta$  in Eq. (5) since it is a function of the working capacity vector  $A$ . Let  $R$  denote the set of feasible  $A$ . Property 2 in Appendix B shows that  $\theta(A)$  is a concave function on  $R$ . This allows us to use the subgradient scheme [19] to compute a new working capacity vector  $\tilde{A}$  to

improve  $\theta$ .  $\tilde{A}$  can be computed as  $\tilde{A} \leftarrow A + \tau\gamma$ , where  $\gamma$  is a subgradient vector at point  $A$  and  $\tau$  is the step size.  $\gamma$  is called a *subgradient vector* of  $\theta(A)$  at the point  $A$  if

$$\theta(A) - \theta(\bar{A}) \leq \gamma \cdot (A - \bar{A}), \quad A \in R \quad (6)$$

holds. We show how to compute  $\gamma$  below.

**Property 3.** Suppose  $\bar{A}$  is a working capacity vector as defined by Eq. (4). Let  $\bar{A} \in R$  and  $\gamma$  be a subgradient of  $\theta(A)$  at  $\bar{A}$ . Then

$$\gamma = [\bar{\omega}_1, \bar{\omega}_2, \dots, \bar{\omega}_m]. \quad (7)$$

where  $\bar{\omega}$  are the corresponding optimal dual variables for constraint (5e).

**Proof.** From the definition of subgradient,  $\gamma$  at  $\bar{A}$  can be computed as follows. For  $\theta(A)$ , let  $\omega$  be the corresponding optimal dual variables for constraint (5e). Then from linear programming theory,

$$\begin{aligned} \theta(A) - \theta(\bar{A}) &= \sum_e \omega_e A(e) - \sum_e \bar{\omega}_e \bar{A}(e) \\ &\leq \sum_e \bar{\omega}_e A(e) - \sum_e \bar{\omega}_e \bar{A}(e) \\ &= \sum_e \bar{\omega}_e [A(e) - \bar{A}(e)] \end{aligned} \quad (8)$$

We can rewrite Eq. (8) as  $\theta(A) - \theta(\bar{A}) \leq [\bar{\omega}_1, \bar{\omega}_2, \dots, \bar{\omega}_m] \cdot (A - \bar{A})$ . From the definition of subgradient, we thus have  $\gamma = [\bar{\omega}_1, \bar{\omega}_2, \dots, \bar{\omega}_m]$ .  $\square$

*Stage 2:* At stage 2, we check if the new working capacity vector  $\tilde{A} = [\tilde{A}(1), \dots, \tilde{A}(m)]$  produced by stage 1 is feasible or not. If not, we will modify it and make it feasible. This is done with the following LP formulation.

$$\max r \quad (9a)$$

$$\text{s.t. } \sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = 0, \quad f = (o, t) \in E, \quad v \neq o, t \quad (9b)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = A(f), \quad f = (o, t) \in E, \quad v = o \quad (9c)$$

$$\sum_{e \in \Gamma^+(v)} y_f^e - \sum_{e \in \Gamma^-(v)} y_f^e = -A(f), \quad f = (o, t) \in E, \quad v = t \quad (9d)$$

$$A(e) + y_f^e \leq c_e, \quad e, f \in E, \quad e \neq f \quad (9e)$$

$$y_e^e = 0, \quad e \in E \quad (9f)$$

$$A(e) \geq \tilde{A}(e) \cdot r, \quad e \in E \quad (9g)$$

$$y, r, A \geq 0 \quad (9h)$$

Constraints (9b)–(9e) ensure that the computed working capacity is feasible. Constraint (9g) implies that  $r = \min_e \{A(e)/\tilde{A}(e)\}$ . Thus, if  $r \geq 1$ ,  $\tilde{A}$  is obviously feasible and it will be put back into Eq. (5) for the next iteration. If  $r < 1$  ( $\tilde{A}$  is not feasible), then  $(r\tilde{A})$  will be feasible and we will pass this vector back to stage 1 for further iterations.

### 3.2. Path formulation for restoration paths

The formulation in stage 2 is link based. But we now transform it into a *path-flow* formulation so that we can impose a hop-count limit on the restoration paths. Also, a faster computation algorithm is available for the new form. Let  $P_e$  denote the set of all paths, except  $e$ , from the originating node to the terminating node of link  $e$ . Let  $y(p)$  denote the restoration traffic on path  $p$  if its protected link fails. To restore the traffic for any failed link  $e$ , we must have  $A(e) = \sum_{p \in P_e} y(p)$  for all  $e \in E$ . The path-flow formulation of Eq. (9) is as follows:

$$\max r \quad (10a)$$

$$\text{s.t. } \sum_{p \in P_e} y(p) \geq r \cdot \tilde{A}(e) \quad e \in E \quad (10b)$$

$$\sum_{p \in P_e} y(p) + \sum_{p: p \in P_f, e \in p} y(p) \leq c_e \quad f \neq e, e, f \in E \quad (10c)$$

$$y, r \geq 0 \quad (10d)$$

The above path-flow formulation can be solved efficiently with a primal-dual approach adapted from the technique developed for the maximum concurrent flow problem in [20]. In addition, the hop-count limit can be easily included in the algorithm. The dual formulation of Eq. (10) is to associate a variable  $\sigma_e$ , for each link  $e$ , corresponding to constraint (10b) and a non-negative variable  $w(e,f)$ , for each pair  $e,f \in E$ ,  $e \neq f$ , corresponding to constraint (10c). The dual formulation can be written as

$$\min \sum_{e \in E} c_e \sum_{f \in E, f \neq e} w(e,f) \quad (11a)$$

$$\text{s.t. } \sum_{e' \in p} w(e', e) + \sum_{f \in E, f \neq e} w(e, f) \geq \sigma_e p \in P_e, \quad e \in E \quad (11b)$$

$$\sum_e \tilde{A}(e) \sigma_e \geq 1 \quad (11c)$$

If we set  $z_e$  to the minimum value of the left-hand-side (LHS) of constraint (11b), then  $w(e,f)$  will be a dual feasible solution that satisfies constraint

(11b). In addition, constraint (11c) can be easily satisfied if we divide all weights  $w(e,f)$  by  $\sum_e \tilde{A}(e) z_e$ .

The algorithm proceeds iteratively. At each iteration, for each link  $e$ , the *shortest* path  $p \in P_e$  that minimizes the LHS of constraint (11b) is computed, flow is sent on the path, and the primal and dual variables are updated accordingly. Note that we can impose a hop-count limit in this step when we compute the shortest paths (e.g., we can use the Bellman–Ford algorithm [21]). This may reduce the working capacity a little bit (see Section 4), but the restoration latency can be restricted by adding this constraint.

## 4. Performance evaluation

In this section we compare the performance for different schemes. The primary performance measure is the maximum admissible bandwidth of traffic the network can sustain. In the following experiments, we assume the preference parameters  $\tilde{\alpha}_i$  and  $\tilde{\beta}_i$  at edge node  $i$  is set proportional to the total capacity of network links incident at node  $i$ . This is a logical assumption because if there is more traffic demand from a node, more links will be added to that node. As we mentioned in the introduction, most existing restoration algorithms assume the traffic matrix is given, and they can not be applied to problems with hose-model traffic patterns. In the following, we only compare our scheme with those that can be applied to hose-model traffic patterns. The schemes we compare include the following:

- Linear programming, no link protection (LP\_NP)*: the optimal scheme by solving Eq. (5) by setting the working capacity = link capacity (i.e.,  $A(e) = c_e$  for all  $e \in E$ ).
- Linear programming, with link protection (LP\_P)*: the optimal scheme by solving Eq. (3) with standard LP solvers.
- Decomposition and Iterative scheme (DI)*: the working and restoration flows are computed by the decomposition algorithm discussed in Section 3.
- Non-iterative scheme (NI)*: the working capacity is computed by the link partition scheme that will be described below.
- Shortest path routing and restoration (SPRR)*: Use the shortest paths for the working traffic. If multiple shortest paths exist, traffic will be evenly split among them. SPRR also uses the

shortest restoration paths to protect a link. Once working and restoration paths are given, we can easily compute the maximum  $\theta$ .

Two non-iterative schemes of (d) are described below. Without considering what type of traffic patterns to be supported in the network, we just compute the amount of working capacity under the condition that it can be protected.

*Non-Iterative Scheme 1 (NI1):*

$$\max \sum_{e \in E} \sum_{p \in P_e} y(p) \quad (12a)$$

$$\text{s.t.} \sum_{p \in P_e} y(p) + \sum_{p: p \in P_f, e \in p} y(p) \leq c_e, \quad f \neq e, \quad e, f \in E \quad (12b)$$

$$y \geq 0 \quad (12c)$$

The objective function (12a) is to maximize the sum of all links' working capacity. Once we have  $y(p)$ , we can derive  $A(e)$ . We then put  $A(e)$  back into Eq. (5) to find the maximum  $\theta$  for the hose-model pattern. This is done in one iteration.

*Non-Iterative Scheme 2 (NI2):*

Similar to the previous non-iterative scheme, we compute the working capacity first. But we change the objective function in (12a) to  $\bar{r} = \min_{e \in E} \{r_e\}$ , where  $r_e$  is the fraction of the capacity of link  $e$  reserved for the working traffic.

$$\max \bar{r} \quad (13a)$$

$$\text{s.t.} \sum_{p \in P_e} y(p) \geq \bar{r} \cdot c_e \quad e \in E \quad (13b)$$

$$\sum_{p \in P_e} y(p) + \sum_{p: p \in P_f, e \in p} y(p) \leq c_e, \quad f \neq e, \quad e, f \in E \quad (13c)$$

$$y, \bar{r} \geq 0 \quad (13d)$$

#### 4.1. Speed of convergence of the decomposition scheme

We first evaluate the effectiveness of the decomposition algorithm with the Sprint IP backbone topology shown in Fig. 3 [22]. We assume all nodes are edge nodes and all links have the same capacity of 1000 U. Although the theoretical rate of convergence for basic subgradient algorithm is linear [23], its convergence speed is much better in practice [24]. Fig. 4 shows how many iterative phases the decomposition algorithm needs to perform before getting a near-optimal solution. The straight line in Fig. 4

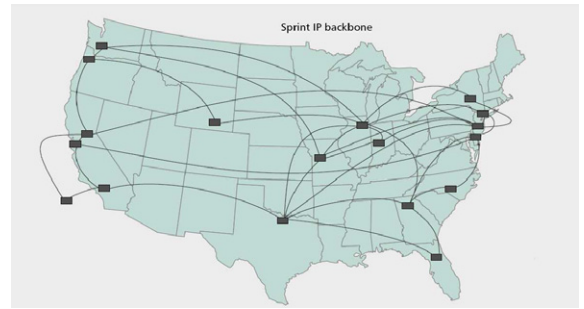


Fig. 3. The Sprint US backbone topology used for performance evaluation.

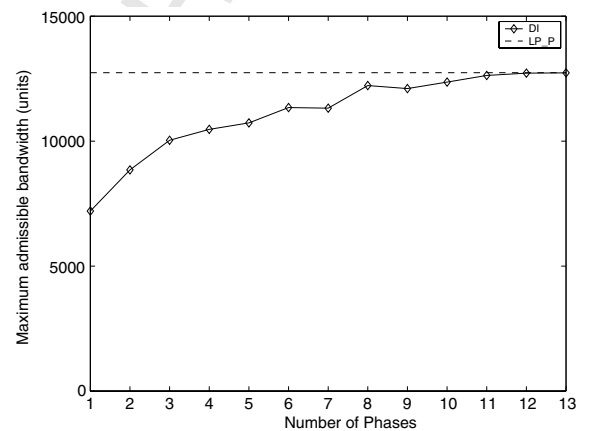


Fig. 4. Maximum admissible bandwidth for the decomposition scheme in the Sprint topology.

indicates the maximum admissible bandwidth computed by the LP\_P scheme. As we can see, after 10 phases, the maximum admissible bandwidth computed by the decomposition scheme is very close to the optimal value. For larger networks presented later, we normally can get a near-optimal solution in less than a hundred phases for the decomposition scheme.

The number of iterative phases needed for achieving convergence given in Fig. 4 does not depend on the type of CPUs we use, but the real computation time for each phase will be machine dependent. Fig. 5 compares the running-times (in seconds) of the two approaches on randomly generated topologies measured on a 3-GHz Pentium-4 PC with 2 GB of memory. The results clearly show that the DI scheme is much faster than the LP\_P scheme. The running-time of the LP\_P scheme grows quickly with the size of the network. In con-



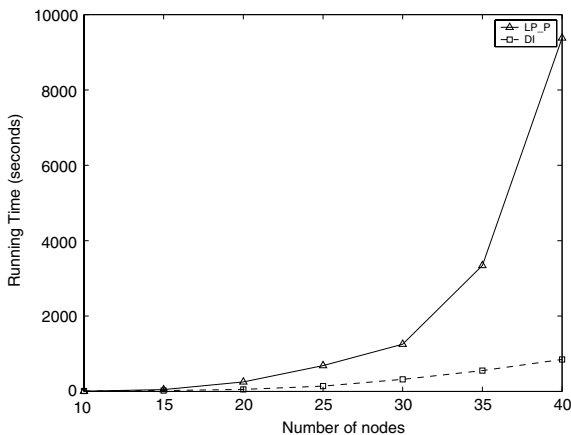


Fig. 5. The running time of the LP\_P and DI schemes in various randomly generated topologies with different number of nodes.

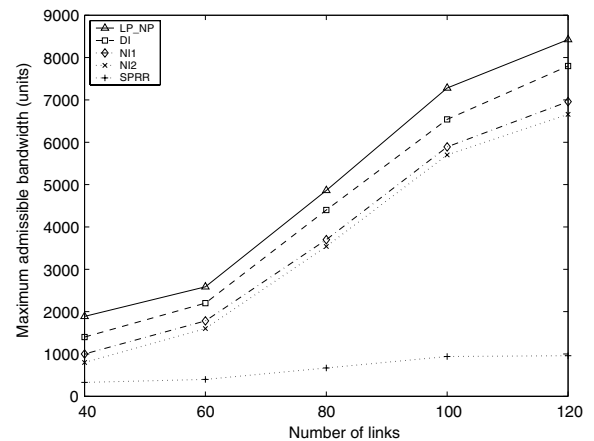


Fig. 6. Maximum admissible bandwidth vs. the number of links in Experiment 1.

trast, the running-time of the DI scheme grows in a much slower pace.

#### 4.2. Throughput comparison

We evaluate the throughput of LP\_NP, DI, NI1, NI2, and SPRR, based on randomly generated topologies with the following varying parameters: (1) number of links in the network, (2) number of nodes in the network, and (3) number of edge nodes.

- *Experiment 1:* 20 node topologies with 40–120 bidirectional links. The number of edge nodes is set to 10.
- *Experiment 2:* 10–50 node topologies. The number of links is twice the number of nodes in the topology. The number of edge nodes is set to 10.
- *Experiment 3:* 40 node topologies with 80 bidirectional links. The number of edge nodes is varied from 6 to 20.

The link capacity is 100 U. Figs. 6–8 are the average results of ten independent runs. LP\_NP is presented only to show how much traffic we need to sacrifice to ensure restoration. Comparing LP\_NP and DI, we find that DI reduces the admissible traffic by 7.4–25.8% in Experiment 1, 11–35.1% in Experiment 2, and 11.8–19.9% in Experiment 3.

DI performs much better than NI1 and NI2 in all the experiments. The performance gap between DI and NI1 (which achieves the secondary high performance among the restoration schemes) ranges from 10.8% to 40% in Experiment 1, from 23.5% to 32.8% in Experiment 2, and from 19.8% to 33.3%

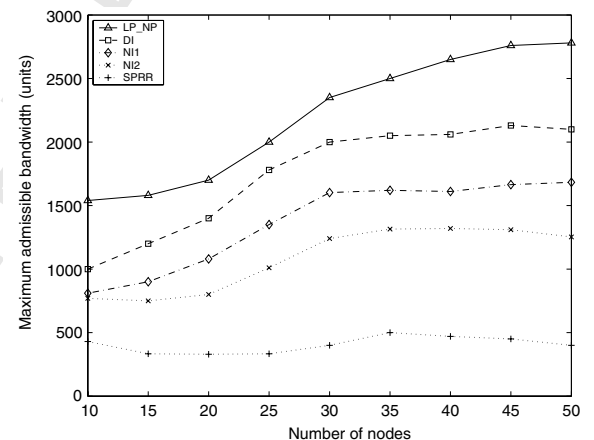


Fig. 7. Maximum admissible bandwidth vs. the number of nodes in Experiment 2.

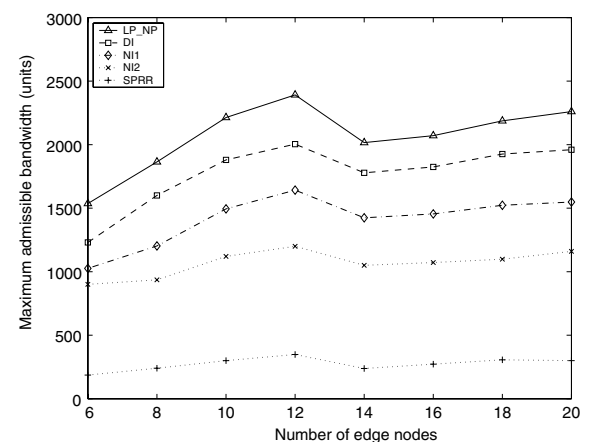


Fig. 8. Maximum admissible bandwidth vs. the number of edge nodes in Experiment 3.

in Experiment 3. The reason is because the two NI schemes partition the link capacity without taking in account the hose-model traffic pattern. This results in the reduction of admissible capacity. Among the two, NI1, which maximizes the working capacity for the working traffic, performs better than NI2. We also observe that SPRR performs much worse than the other schemes. This is because SPRR uses shortest paths for working and restoration traffic. Load-balancing can not be done as efficiently as the other schemes.

#### 4.3. Impact of hop-count limit on throughput

We use the Sprint backbone topology (Fig. 3) to study the throughput degradation due to adding a hop-count limit. Fig. 9 plots the maximum admissible bandwidth as the maximum allowable hop count of the restoration paths. As can be seen, the maximum admissible bandwidth does not change much beyond a hop-count of 6. The results allow us to make an intelligent tradeoff between the throughput and the restoration latency.

#### 4.4. Dynamic construction of hose-model VPNs

The performance of the proposed approach shown in the previous sections can be further improved as described in this section. We compare the performance of the proposed approach with that of the conventional approaches in a dynamic environment where VPNs come and go. The performance measure we use is the *rejection ratio* which

is defined as the percentage of the total VPN requests that is rejected.

In a dynamic VPN environment, previously proposed VPN provisioning algorithms, as pointed out in [15], have the drawback of computing working and link-restoration paths every time a new VPN is added. This is time consuming and can create a scalability problem if the frequency of adding and deleting VPNs is high. The non-blocking network approach does not have the same problem. For our approach, we will use one additional measure to further improve the performance presented in the previous sections. We use a server to record how much bandwidth of each link has been taken for existing VPNs. Recall that in the proposed approach, the paths are fixed. When a new VPN arrives, we use the formulation of Eq. (A.1) (given in Appendix A) to find the maximum amount of bandwidth required (i.e., the worst-case traffic pattern) along the paths for this new VPN. The computation required for this is much less than finding the optimal set of working and restoration paths of each new VPN. The throughput presented in the previous sections does not track this information and only uses the information of the ingress and egress nodes of the VPN to decide if the VPN can be admitted.

We conduct experiments on the Sprint backbone topology (Fig. 3). The VPN requests are generated following a Poisson process and the holding time is exponentially distributed. The VPN endpoints are randomly attached to the edge nodes, and the number of endpoints of each VPN is chosen randomly between 5 and 15. The ingress and egress

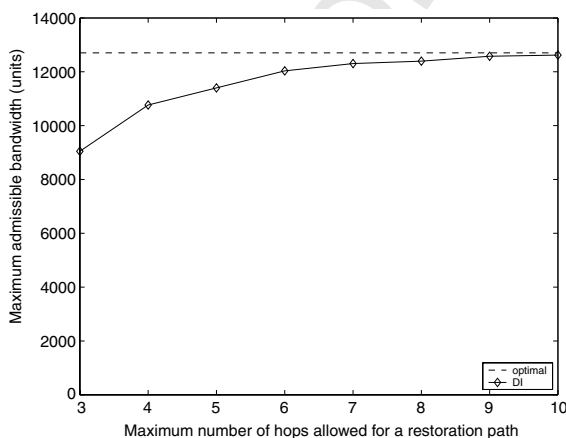


Fig. 9. Maximum admissible bandwidth vs. maximum hop count of a restoration path.

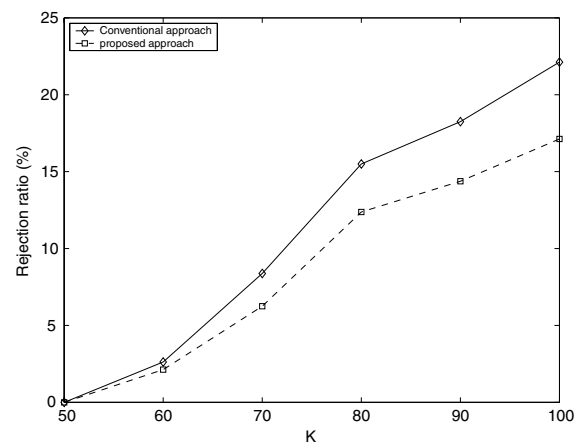


Fig. 10. The comparison of rejection ratio of the conventional and our proposed VPN construction approaches.

bandwidth requirement of a VPN endpoint are assumed to be the same, and its value is chosen randomly between 1 and  $K$ , where  $K$  is a selected parameter that indicates the degree of variability of VPN bandwidth requirements. The results are shown in Fig. 10. As can be seen, the rejection ratios of the two approaches are very close if  $K$  is small. However, as  $K$  becomes larger, which implies greater variability of the bandwidth requirements of each VPN, our proposed approach has a better performance than the conventional approach.

## 5. Conclusions

In this paper, we presented a new restoration architecture for L3-VPN networks. We also presented a linear programming formulation for computing the optimal routing and link restoration for this architecture. Furthermore, we showed an efficient decomposition algorithm that can compute a near-optimal solution with much less computational overhead. The proposed architecture has many advantages, including no need to set up an external routing table, no need to set up restoration paths for a new VPN, and high throughput performance. The proposed decomposition algorithm is computation efficient and allows us to include a hop-count limit to bound the restoration latency of the VPN in case restoration occurs. The techniques developed in this paper can also be applied to other restoration networks.

## Appendix A. Proof of Property 1

**Property 1.** Given  $H = [(\tilde{\alpha}_1, \tilde{\beta}_1), \dots, (\tilde{\alpha}_n, \tilde{\beta}_n)]$ , routing  $x_{ij}^e$  and working capacity reservation  $A(e)$  can satisfy constraint (2e) for all traffic matrices in  $M$  if and only if there exist non-negative weights  $\pi_e(i)$  and  $\lambda_e(i)$  for each  $e \in E$  and  $i \in Q$  such that

- (i)  $\sum_{i \in Q} \tilde{\alpha}_i \pi_e(i) + \sum_{i \in Q} \tilde{\beta}_i \lambda_e(i) \leq A(e)$  for each  $e \in E$
- (ii)  $x_{ij}^e \leq \pi_e(i) + \lambda_e(j)$  for each  $e \in E$  and every  $i, j \in Q$

**Proof.** (“only if” direction): Let routing  $x_{ij}^e$  and working capacity reservation  $A(e)$  satisfy constraints (2e) for all traffic matrices in  $M$  (i.e.,  $\sum_{ij} x_{ij}^e d_{ij} \leq A(e)$  for all  $e \in E$  and  $T \in M$ ). Consider a link  $e$ . The problem of finding  $T = \{d_{ij}\}$  that maximizes link load on  $e$  can be formulated as the following linear programming problem.

$$\max \sum_{ij} x_{ij}^e d_{ij} \quad (\text{A.1a})$$

$$\text{s.t. } \sum_{j \in Q} d_{ij} \leq \tilde{\alpha}_i, \quad i \in Q \quad (\text{A.1b})$$

$$\sum_{i \in Q} d_{ij} \leq \tilde{\beta}_j, \quad j \in Q \quad (\text{A.1c})$$

$$d_{ij} \geq 0, \quad i, j \in Q \quad (\text{A.1d})$$

where constraints (A.1b) and (A.1c) are the ingress and egress bandwidth constraints. The dual of the above LP problem for link  $e$  is:

$$\min \sum_i \tilde{\alpha}_i \pi_e(i) + \sum_i \tilde{\beta}_i \lambda_e(i) \quad (\text{A.2a})$$

$$\text{s.t. } \pi_e(i) + \lambda_e(j) \geq x_{ij}^e, \quad i, j \in Q \quad (\text{A.2b})$$

$$\pi, \lambda \geq 0 \quad (\text{A.2c})$$

Since  $\sum_{ij} x_{ij}^e d_{ij} \leq A(e)$ , the dual for any link  $e$  must have optimal value  $\leq A(e)$ . Therefore, the objective function of the dual satisfies (i). Requirement (ii) is trivially satisfied by the dual problem constraint (A.2b).

(“if” direction): Let  $x_{ij}^e$  be a routing, and  $T = \{d_{ij}\}$  be any valid traffic matrix. Also let  $\pi_e(i)$  and  $\lambda_e(i)$  be the weights satisfying requirements (i)-(ii). Consider a link  $e$ . From (ii), we have

$$x_{ij}^e \leq \pi_e(i) + \lambda_e(j)$$

Summing over all node pairs  $(i, j)$ , we have

$$\begin{aligned} \sum_{i,j \in Q} x_{ij}^e d_{ij} &\leq \sum_{i,j \in Q} [\pi_e(i) + \lambda_e(j)] d_{ij} \\ &= \sum_{i \in Q} \pi_e(i) \sum_{j \in Q} d_{ij} + \sum_{j \in Q} \lambda_e(j) \sum_{i \in Q} d_{ij} \\ &\leq \sum_{i \in Q} \tilde{\alpha}_i \pi_e(i) + \sum_{i \in Q} \tilde{\beta}_i \lambda_e(i) \end{aligned}$$

The last inequality comes from the constraints imposed by  $H$  (i.e.,  $\sum_j d_{ij} \leq \tilde{\alpha}_i$  and  $\sum_i d_{ij} \leq \tilde{\beta}_j$ ). From (i), we have

$$\sum_{i,j \in Q} x_{ij}^e d_{ij} \leq \sum_{i \in Q} \tilde{\alpha}_i \pi_e(i) + \sum_{i \in Q} \tilde{\beta}_i \lambda_e(i) \leq A(e)$$

This means that for any traffic matrix constrained by  $H$ , the working traffic on any link is at most  $A(e)$ .  $\square$

## Appendix B. Concavity of $\theta(A)$

**Property 2.** Let  $R$  denote the set of feasible  $A$ . Then  $\theta(A)$  is a concave function on  $R$ .

**Proof.** The property can be proven from the dual form of Eq. (5).

*Dual Problem of Eq. (5):*

$$\min \sum_{e \in E} \omega_e A(e) \quad (\text{B.1a})$$

$$\text{s.t. } \sigma_{ij}^u - \sigma_{ij}^v + \mu_{ij}^e \geq 0, \quad i, j \in Q, \quad e = (u, v) \in E \quad (\text{B.1b})$$

$$\sum_{i, j \in Q} \sigma_{ij}^j \geq 1 \quad (\text{B.1c})$$

$$\sigma_{ij}^i = 0, \quad i, j \in Q \quad (\text{B.1d})$$

$$\tilde{\alpha}_i \omega_e - \sum_{j \in Q} \mu_{ij}^e \geq 0, \quad i \in Q, \quad e \in E \quad (\text{B.1e})$$

$$\tilde{\beta}_j \omega_e - \sum_{i \in Q} \mu_{ij}^e \geq 0, \quad j \in Q, \quad e \in E \quad (\text{B.1f})$$

$$\omega, \mu, \sigma \geq 0 \quad (\text{B.1g})$$

Let  $\tilde{A}, \bar{A} \in R$ , and  $\rho$  be in the range  $0 \leq \rho \leq 1$ . Let  $A = \rho \tilde{A} + (1 - \rho) \bar{A}$ . Then according to the dual form of Eq. (5),

$$\begin{aligned} \theta(A) &= \min \left\{ \sum_{e \in E} \omega_e A(e) : \right. \\ &\quad \text{s.t. B.1b–B.1g} \\ &= \min \left\{ \sum_{e \in E} \omega_e [\rho \tilde{A}(e) + (1 - \rho) \bar{A}(e)] : \right. \\ &\quad \text{s.t. B.1b–B.1g} \\ &= \min \left\{ \rho \sum_{e \in E} \omega_e \tilde{A}(e) + (1 - \rho) \sum_{e \in E} \omega_e \bar{A}(e) : \right. \\ &\quad \text{s.t. B.1b–B.1g} \end{aligned}$$

Let

$$\begin{aligned} \theta(\tilde{A}) &= \min \left\{ \sum_{e \in E} \omega_e \tilde{A}(e) : \right. \\ &\quad \text{s.t. B.1b–B.1g} \\ \theta(\bar{A}) &= \min \left\{ \sum_{e \in E} \omega_e \bar{A}(e) : \right. \\ &\quad \text{s.t. B.1b–B.1g} \end{aligned}$$

From linear programming theory,  $\theta(A) \geq \rho \theta(\tilde{A}) + (1 - \rho) \theta(\bar{A})$  obviously holds. Hence,  $\theta(A)$  is a concave function on  $R$ .  $\square$

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