Loop-Free Link-Traffic in Optimal Routing for Multi-Class Networks*

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Abstract

Individually and overall optimal routing problems are considered for networks with nodes interconnected by a generally configured manner and with multiple classes of users. The two problems are formulated along with the discussions on mutual equivalence between both problems, on the existence and uniqueness of solutions, and on the relation between the formulations with path and link flow patterns. We show that a link-traffic loop-free property holds within each class for both individually and overall optimal routing in a wide range of networks, and show the condition that characterizes the category of networks for which the property holds.

Keywords: communications network, transportation network, overall optimization, Wardrop equilibrium, existence and uniqueness, path flow pattern, link flow pattern

1 Introduction

We have two typical approaches for optimal routing in networks. (1) One arises in a context of minimizing the overall cost (overall mean delay) of all users (e.g., packets) from its arrival (origin) node of each user to its destination node through a number of links over the entire network. The optimal routing policy with this framework is called the *overall optimal routing policy*. (2) A second approach is a distributed one in which one seeks for a set of routing strategies for all users such that no user can decrease its cost (expected delay) by deviating from its strategy unilaterally. This could be viewed as a result of leaving each user the decision on which path to route. This approach is called the *individually optimal routing* or *selfish routing* [19, 20]. The situation where each user has unilaterally minimized its cost is called a Wardrop equilibrium [17, 22] or a Nash equilibrium where no user has any incentive to make a unilateral decision to change its route.

In computer and communication networks, most work has focused on overall optimal routing (e.g., [2, 7, 8]). For networks in general, however, minimizing the cost of each user from its arrival (origin) node to its destination node is a major concern of the user. Thus, individually optimal routing has attracted increasing attention of researchers and practitioners also in computer and communication networks, and some research results have been obtained on it [1, 3, 16, 19, 20].

In most studies on optimal routing problems for communication networks in the literature (e.g., [2, 7, 8, 9, 3, 16]), the link cost is modeled as a simple function dependent only on the link flow itself. We call it the *traditional link-cost model*. In this paper, however, the cost on a link of a network is modeled by a function of the flows of all links of the entire network. We call it a *general link-cost model*. For example, we can consider a wireless communication network, where, when a link connecting two

^{*}A preliminary version of this paper was presented at IEEE CDC'04.

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nodes has more flow and, thus, uses more power, links neighboring the link may have less capacity. This paper studies both the individually and overall optimal routing problems in general link-cost models of generally configured networks with multiple classes of users. We call a network with multi-class users a *multi-class network*. We note, however, that, in these optimization problems, the cost to be optimized depends only on the link flow pattern while the instrument (the set of decision variables) is the path flow pattern.

In this paper, we discuss individually and overall optimal routing problems on which Dafermos has obtained some basic results [6, 4, 5]. Our treatment is, however, more general than hers in the following points: (1) Our model allows each user of a class to enter any origin and leave any destination both available to the class with/without fixing the arrival rate at each origin and the departure rate at each destination. (2) The link-traffic loop-free property is discussed. (3) The relation between the case where the instrument is the path flow pattern and the case where it is the link flow pattern is discussed. In particular, we note that, by definition, a path flow pattern determines a unique link flow pattern whereas it may not be sure whether for a link flow pattern there exists a path flow pattern that induces it, i.e., whether a given link flow pattern is realizable.

We confirm that, for the individually optimal routing problem under our assumptions, there is an overall optimal routing problem associated to it, and that both have the same solution. We discuss the existence and uniqueness of the solutions to the overall and individually optimal routing. Furthermore, we show that the link-traffic loop-free property holds for each class, for the individually and overall optimal routing in general link-cost models of multi-class networks. We pay much attention to the relation between the cases where the sets of the control variables are, respectively, the path and link flow patterns. We show the condition that characterizes the category of networks where the link-traffic loop-free property holds for each class. Some examples are discussed. In contrast, note that, even in the networks where the link-traffic loop-free property holds for each class in overall and individually optimal routing, it does not always hold in noncooperative optimal routing by a finite (but plural) number of decision makers, where the decision makers strive to optimize unilaterally the cost of the users under its control. Such counter-examples are given in [12, 13] (with the definition of class in those papers changed to be the same as the one in this paper). Note, in passing, that overall optimal routing may have only one decision maker and that individually optimal routing has infinitely many infinitesimal decision makers.

The rest of this paper is organized as follows. In the next section, we provide the problem formulation. The relation between the individually and overall optimal routing for multi-class networks is provided in Subsection 2.3. Subsection 2.4 discusses the existence and uniqueness of individually and overall optimal routing for multi-class networks. Section 3 discusses the link-traffic loop-free property for the individually and overall optimal routing for multi-class networks. Some examples are shown in Section 4. Section 5 concludes the paper.

2 Problem Formulation and Solutions

Consider a network consisting of n nodes numbered, 1, 2, ..., n, interconnected in an arbitrary fashion by links. N and L, respectively, denote the sets of nodes and links. There are multi-class users in the network. C denotes the set of user classes. Each class may have a distinct set of links available to the class. We assume that users do not change their classes during their trips from origins and destinations. We call links available to class-k users 'class-k links.' L^k denotes the set of 'class-k links.' Then, $L = \bigcup_{k \in C} L^k$. By a path for a class, say class k, connecting an ordered pair $\omega = (o, d)$, we mean a sequence of class-k links $(v_1, v_2), (v_2, v_3), \ldots, (v_{n'-1}, v_{n'})$ that any class-k user can pass through where $v_1, v_2, \ldots, v_{n'}$ are distinct nodes, $v_1 = o$, and $v_{n'} = d$. Then, the path is denoted by $(o, v_2, \ldots, v_{n'-1}, d)$. We call node o an origin,

the node d a destination and the pair $\omega = (o, d)$ an origin-destination pair (or o-d pair for abbreviation). Each class may have a distinct set of origins and of destinations.

If v_i is the same as node v_j for some i and j such that j < i, we say that the path has a *loop* or cycle. We note, however, that in the optimal solutions such a loop within a path never exists. Therefore, a link appears in a path for a class at most once. On the other hand, although each path has no loop, the network may have a loop as to link flows as discussed in Section 3. We have the following assumptions.

- A1 (1) If there exists a possible series of link connections for a class between an o-d pair, there must exist a path for the class between the o-d pair. (2) If there exists a path for a class between an o-d pair, all possible series of link connections for the class between the o-d pair are also paths of the class between the o-d pair.
- A2 The rates of arrivals at each origin and of departures at each destination are given for each class.

Remark 1 As seen later in Sections 3 and 4, Assumption A1 presents the condition that characterizes the category of networks that have the link-traffic loop-free property within each class for overall and individually optimal routing. Even under the assumptions A1 and A2, we can model the situation where there are particular combinations of origins and destinations such that users arriving at an origin should depart the network only from the destination corresponding to the origin.

In Assumption A2, it may look unnatural that the rate of the departure at each destination is given even though each user can leave the network at any available destination. We see below, however, that the assumption A2 is most general. Consider a network, named \mathcal{M} , where each class-k user can leave the network at any available destination without fixing the class-k departure rate at each destination. We imagine another network, named \mathcal{M}' , where one class-k 'final' destination is added to the network \mathcal{M} and that each of the class-k destinations in \mathcal{M} is connected to the class-k final destination via a class-k zero-cost link in the network \mathcal{M}' for every class k. (Later in this section, we will describe zero-cost links along with the definition of link-cost functions, G_{ij}^k .) Then, the imagined network \mathcal{M}' can be regarded as the one with multiple-origins and one common destination for class k as shown in Section 4.1.1 and satisfies Assumptions A1 and A2. The optimal solutions of the networks \mathcal{M} and \mathcal{M}' should be identical.

In a similar way, we can consider a network where each class-k user can enter the network at any origin without fixing the class-k arrival rate at each origin but with the departure rate at each class-k destination being fixed. We can also consider a network where each class-k user can enter at any origin and leave from any destination without fixing the arrival and departure rates at any origin and destination for class k and with fixing the total arrival and departure rates for the class k, respectively. The former network is equivalent to the network where one 'initial' origin is added and connected to each origin via a zero-cost link for class-k. The latter network is equivalent to the network where one initial origin and one final destination are added for class-k.

We therefore see that the assumption A2 is most general and covers all three kinds of networks each of which is equivalent to the corresponding one of the three networks mentioned above, respectively. \Box

For simplicity, we assume that a node cannot be both an origin and a destination at the same time for the same class. The sets of all origin and destination nodes for class k are denoted by O^k and \mathcal{D}^k , respectively. The sets of all possible paths which originate from an $o \in O^k$ and which are destined for a $d \in \mathcal{D}^k$, for class k, are denoted by \mathcal{P}^k_{o-} and \mathcal{P}^k_{-d} , respectively. The set of all paths in the network for class k is denoted by \mathcal{P}^k , each element of which must appear in a \mathcal{P}^k_{o-} and in a \mathcal{P}^k_{-d} , i.e., $\mathcal{P}^k = \bigcup_{o \in O^k} \mathcal{P}^k_{o-} = \bigcup_{d \in \mathcal{D}^k} \mathcal{P}^k_{-d}$). For every origin $o \in O^k$ and for every destination $d \in \mathcal{D}^k$, respectively, let r^k_{o-} and r^k_{-d} ($k \in C$) be the nonnegative external class-k user traffic demands that originates at node o for all destinations $d \in \mathcal{D}^k$, and that is destined for node d from all origins $o \in O^k$.

For a path $p \in \mathcal{P}_{o-}^k$, y_p^k denotes the part of r_{o-}^k which flows through path p. Similarly for a path $p \in \mathcal{P}_{-d}$. y_p^k is called the class-k path flow through the path p. We have the following relations.

$$\sum_{p \in \mathcal{P}^k} y_p^k = r_{o-}^k, \quad o \in \mathcal{O}^k, \tag{1}$$

$$\sum_{p \in \mathcal{P}^k} y_p^k = r_{-d}^k, \quad d \in \mathcal{D}^k, \tag{2}$$

$$y_p^k \ge 0, \quad p \in \mathcal{P}^k, \ k \in C.$$
 (3)

Naturally, $\sum_{o \in O^k} r_{o-}^k = \sum_{d \in \mathcal{D}^k} r_{-d}^k$. Denote the *path flow pattern* by $y = [y^k]$, where $y^k = [y_p^k]$. By a *feasible path flow pattern*, we mean y which satisfies relations (1), (2), and (3). Denote by FS_y the set of feasible path flow patterns. Clearly, FS_y is convex, closed, and bounded.

Denote by x_{ij}^k the class-k user flow rate, also called the class-k flow, through link (i, j). Let $\mathbf{x} = [\mathbf{x}^k]$ where $\mathbf{x}^k = [x_{ij}^k]$. Furthermore, let $\mathbf{x} = [\mathbf{x}_{ij}]$ where $\mathbf{x}_{ij} = [x_{ij}^k]$. We call \mathbf{x} the link flow pattern. Since a link appears in a path at most once, a class-k link flow is expressed by class-k path flows as follows.

$$x_{ij}^k = \sum_{p \in \mathcal{P}^k} \delta_{ij}^p y_p^k, \quad (i, j) \in L^k, \ k \in C,$$

$$\tag{4}$$

where

$$\delta_{ij}^p = \begin{cases} 1, & \text{if link } (i, j) \text{ is contained in path } p, \\ 0, & \text{otherwise.} \end{cases}$$

If link (i, j) is included in path p, we also express it as $(i, j) \in p$. From (4), we notice that a path flow pattern y induces a unique link flow pattern x, while it is possible that more than one path flow pattern y induces the same link flow pattern x. Moreover, for given x, it may not be sure whether there exists a path flow pattern y that induces x.

Let G_{ij}^k be the class-k link cost of sending a class-k user from node i to node j through link (i,j). G_{ij}^k is a function of all link flows x. We assume that, for most of the link costs, $G_{ij}^k(x)$ is a positive and differentiable function that is convex in x and, in particular, that $G_{ij}^k(x)$ is strictly convex in x_{ij}^k . We also consider the possible existence of zero-cost links the flows of which do not influence other link costs, that is, for some i', j', k', $G_{i'j'}^{k'}(x) = 0$ for all x, and $x_{i'j'}^{k'}$ does not affect any other G_{ij}^k . x_s denotes the vector that consists of the elements x_{ij}^k such that each corresponding $G_{ij}^k(x)$ is strictly convex in x_{ij}^k , and x_{-s} denotes the vector that consists of the elements $x_{i'j'}^{k'}$ each of which is the flow through the class-k' zero-cost link (i', j').

 $D_p^k(x)$ denotes the class-k cost of a path p. Then,

$$D_p^k(\mathbf{x}) = \sum_{(i,j)\in L^k} \delta_{ij}^p G_{ij}^k(\mathbf{x}), \ p \in \mathcal{P}^k, \ k \in C.$$
 (5)

2.1 Overall Optimal Routing for Multi-Class Networks

By using (4) and (5), the overall cost of users over all classes is expressed as

$$D(\mathbf{x}) = \frac{1}{R} \sum_{k \in C} \sum_{p \in \mathcal{P}^k} y_p^k D_p^k(\mathbf{x}) = \frac{1}{R} \sum_{k \in C} \sum_{(i,j) \in L^k} x_{ij}^k G_{ij}^k(\mathbf{x}),$$

where $R = \sum_{k \in C} \sum_{o \in O^k} r_{o-}^k = \sum_{k \in C} \sum_{d \in \mathcal{D}^k} r_{-d}^k$.

Thus, considering (4), the overall optimal routing problem is expressed as follows.

$$\min_{\mathbf{y}} D(\mathbf{x}(\mathbf{y})) \text{ subject to } \mathbf{y} \in FS_{\mathbf{y}}. \tag{6}$$

We have assumed that $G_{ij}^k(x)$ is convex in x and, in particular, strictly convex in x_{ij}^k , for all $i, (j \neq i), k$. We also assume that D(x) is convex in x and strictly convex in x_s . Then, we see that D(x) is convex in y, which can be easily shown as follows: Indeed, the relation (4) can be regarded as a linear transformation: $y \to x$ and we denote this by x = x(y). Then, $\alpha D(x(y_1)) + (1 - \alpha)D(x(y_2)) \geq D(\alpha x(y_1) + (1 - \alpha)x(y_2)) = D(x(\alpha y_1 + (1 - \alpha)y_2))$, where the inequality follows from the convexity of $D(\cdot)$ and the equality follows from the transformation linearity. Thus, we see that D(x(y)) is convex in y.

Define $g_{ij}(x)$ as follows.

$$g_{ij}^{k}(\mathbf{x}) = R \frac{\partial}{\partial x_{ij}^{k}} D(\mathbf{x}). \tag{7}$$

Consider the following Lagrangian function:

$$L(\boldsymbol{y}, \boldsymbol{\phi}) = RD(\boldsymbol{x}) + \sum_{k \in C} [\sum_{o \in O^k} \phi_{o-}^k (r_{o-}^k - \sum_{p \in \mathcal{P}_{o-}^k} y_p^k) + \sum_{d \in \mathcal{D}^k} \phi_{-d}^k (r_{-d}^k - \sum_{p \in \mathcal{P}_{-d}^k} y_p^k)],$$

where ϕ_{o-}^k and ϕ_{-d}^k are Lagrange multipliers. Then, the path flow pattern y that satisfies the following relation derived from the Kuhn-Tucker condition is a solution for overall optimal routing if such a flow pattern y exists,

$$\sum_{(i,j)\in p} g_{ij}^k(\mathbf{x}) = \beta_{o,d}^k, \text{ for } y_p > 0,$$

$$\sum_{(i,j)\in p} g_{ij}^k(\mathbf{x}) \ge \beta_{o,d}^k, \text{ for } y_p = 0,$$

$$p \in \mathcal{P}^k, o \in O^k, d \in \mathcal{D}^k, k \in C, \mathbf{y} \in FS_{\mathbf{y}},$$
(8)

where $\beta^k_{o,d} = \phi^k_{o-} + \phi^k_{-d}$. We recall that $\mathcal{P}^k = \cup_{o \in \mathcal{O}^k} \mathcal{P}^k_{o-} = \cup_{d \in \mathcal{D}^k} \mathcal{P}^k_{-d}$.

2.2 Individually Optimal Routing for Multi-Class Networks

Informally, we define the individually optimal routing to be such that each individual user routes itself so as to minimize its own cost from the arrival at its origin node to the departure from its destination node, given the expected link cost of each link. In the equilibrium that the routing policy results in, every user of all classes may feel that its own cost is minimized and has no incentive to make a unilateral decision to change its route. In other words, the link flow pattern x of individually optimal routing is a Wardrop equilibrium [22]. or a *Nash equilibrium* point in the sense of noncooperative game [15]. Thus, we define the equilibrium condition of the individually optimal routing as follows.

Definition I: A path flow pattern y is said to satisfy the equilibrium condition of the individually optimal routing if and only if the following relation holds.

$$D_{p}^{k}(\mathbf{x}) = \sum_{(i,j)\in p} G_{ij}^{k}(\mathbf{x}) = A_{o,d}^{k}, \text{ for } y_{p} > 0,$$

$$D_{p}^{k}(\mathbf{x}) = \sum_{(i,j)\in p} G_{ij}^{k}(\mathbf{x}) \ge A_{o,d}^{k}, \text{ for } y_{p} = 0,$$

$$p \in \mathcal{P}^{k}, o \in O^{k}, d \in \mathcal{D}^{k}, k \in C, \mathbf{y} \in FS_{\mathbf{y}}.$$

$$(9)$$

We recall that $\mathcal{P}^k = \bigcup_{o \in O^k} \mathcal{P}^k_{o-} = \bigcup_{d \in \mathcal{D}^k} \mathcal{P}^k_{-d}$. We call the path flow pattern y the solution of the individually optimal routing if it satisfies the above equilibrium condition. It is a Wardrop equilibrium [22].

Remark 2 The above definition I and the assumptions A1 and A2 imply the situation where it only holds that, for each combination of the origin and the destination, the paths used have equal costs that are not less than those of the unused paths. But, this situation may not reflect the freedom of each user of a class to choose one destination among those available to the class. In order that truly individual decisions may be realized, we may use the framework mentioned in the last paragraph of Remark 1. □

2.3 Relation between Individually and Overall Optimal Routing for Multi-Class Networks

We construct a new overall cost function $\hat{D}(x)$ for the same network as that of (6), such that

$$G_{ij}^{k}(\mathbf{x}) = R \frac{\partial \hat{D}(\mathbf{x})}{\partial x_{ij}^{k}}, \quad (i, j) \in L^{k}, k \in C.$$
(10)

According to the line of research by Dafermos, et al. [4, 5, 17], we consider whether and under what condition such a function $\hat{D}(x)$ exists. Since link-cost function $G_{ij}^k(x)$ $((i, j) \in L^k, k \in C)$ is differentiable, $\partial G_{ij}^k/\partial x_{lm}^{k'}$ $((i, j) \in L^k, (l, m) \in L^{k'}, k, k' \in C)$ exists. Then, we assume that the matrix of partial derivatives

of link-cost functions,
$$\Lambda(x) = \left[\frac{\partial G_{ij}^k}{\partial x_{lm}^{k'}}\right]$$
, is symmetric (i.e., $\frac{\partial G_{ij}^k}{\partial x_{lm}^{k'}} = \frac{\partial G_{lm}^{k'}}{\partial x_{ij}^k}$ for all $(i, j) \in L^k$, $(l, m) \in L^k$

 $L^{k'}, k, k' \in C$). Moreover, we consider a submatrix, $\Lambda_s(x)$, of $\Lambda(x)$ that contains the ((ijk), (i'j'k'))-th elements such that both x_{ij}^k and $x_{i'j'}^{k'}$ are in x_s . We note that the elements of $\Lambda(x)$ that are not in $\Lambda_s(x)$ are all zero. We assume that $\Lambda_s(x)$ is positive definite. Then, $\hat{D}(x)$ is strictly convex in x_s , $\Lambda(x)$ is semi-positive definite, and $\hat{D}(x)$ is convex in x. In the traditional link-cost models, $\Lambda(x)$ is also symmetric and semi-positive-definite (see, e.g., [1, 19]). Denote by $(x_{-(ijk)}, x_{ij}^{\prime k})$ the vector with the component x_{ij}^k of x replaced by $x_{ij}^{\prime k}$. If $\Lambda(x)$ is symmetric, the following satisfies (10).

$$\hat{D}(\mathbf{x}) = \frac{1}{R \sum_{k \in C} |L^k|} \left\{ \sum_{k \in C} \sum_{(i,j) \in L^k} \int_0^{x_{ij}^k} G_{ij}^k(\mathbf{x}_{-(ijk)}, x_{ij}^{\prime k}) dx_{ij}^{\prime k} \right\}. \tag{11}$$

We define

$$\hat{G}_{ij}^{k}(\mathbf{x}) = \frac{1}{x_{ij}^{k} \sum_{k \in C} |L^{k}|} \int_{0}^{x_{ij}^{k}} G_{ij}^{k}((\mathbf{x}_{-(ijk)}, x_{ij}^{\prime k})) dx_{ij}^{\prime k}, \ (i, j) \in L^{k}, \ k \in C.$$
(12)

Then, we regard \hat{G}_{ij}^k as a new class-k link cost on link (i, j). Thus,

$$\hat{D}(\boldsymbol{x}) = \frac{1}{R} \sum_{k \in C} \sum_{(i,j) \in L^k} x_{ij}^k \hat{G}_{ij}^k(\boldsymbol{x}).$$

We recall that, for $x_{i'j'}^{k'} \in \mathbf{x}_{-s}$, $\hat{G}_{i'j'}^{k'} = 0$ and $x_{i'j'}^{k'}$ would not influence other \hat{G}_{ij}^{k} ($i' \neq i$ or $j' \neq j$). Thus, we have the following overall optimization problem, considering (4).

$$\min_{\mathbf{y}} \ \hat{D}(\mathbf{x}(\mathbf{y})) \text{ subject to } (4) \text{ and } \mathbf{y} \in FS_{\mathbf{y}}. \tag{13}$$

We call the overall optimization problem (13) an *associate problem* to the individually optimal routing problem. We note that it is another overall optimal routing problem. Consider the following Lagrangian function:

$$\hat{L}(\boldsymbol{y}, \boldsymbol{\phi}) = R\hat{D}(\boldsymbol{x}) + \sum_{k \in C} [\sum_{o \in O^k} \hat{\phi}^k_{o-} (r^k_{o-} - \sum_{p \in \mathcal{P}^k_{o-}} y^k_p) + \sum_{d \in \mathcal{D}^k} \hat{\phi}^k_{-d} (r^k_{-d} - \sum_{p \in \mathcal{P}^k_{-d}} y^k_p)],$$

where $\hat{\phi}_{o-}^k$ and $\hat{\phi}_{-d}^k$ are Lagrange multipliers. Then, the path flow pattern y that satisfies the following relation derived from the Kuhn-Tucker condition as to the above Lagrangian function is an overall optimal solution to the associate problem if such a flow pattern y exists.

$$\sum_{(i,j)\in p} G_{ij}^{k}(\mathbf{x}) = A_{o,d}^{k}, \text{ for } y_{p} > 0,$$

$$\sum_{(i,j)\in p} G_{ij}^{k}(\mathbf{x}) \ge A_{o,d}^{k}, \text{ for } y_{p} = 0,$$

$$p \in \mathcal{P}^{k}, o \in \mathcal{O}^{k}, d \in \mathcal{D}^{k}, k \in C, \mathbf{y} \in FS_{\mathbf{y}},$$

$$(14)$$

where $A_{o,d}^k = \hat{\phi}_{o-}^k + \hat{\phi}_{-d}^k$. We see that the above is equivalent to the condition (9) describing the solution of the corresponding individually optimal routing problem. Then, we have the following lemma.

Lemma 1 A path flow pattern y is a solution to associate problem (13) if and only if it satisfies the equilibrium condition (9).

On the other hand, if we regard $g_{ij}^k(\mathbf{x})$ as a new class-k link cost on link (i, j), then we have the equilibrium condition (8) of individual optimization that is an *associate condition* to the overall optimal routing problem for multi-class networks with $G_{ij}^k(\mathbf{x})$ being the class-k cost on link (i, j).

Corollary 1 A path flow pattern y satisfies the associate condition (8) if and only if it is a solution to the overall optimization problem (6).

2.4 Existence and Uniqueness

In this section, we study the existence and uniqueness of the solutions to individually and overall optimal routing problems for multi-class networks. We first discuss the existence and uniqueness of the solutions to the overall optimal routing problem. Then, by noting that the individually optimal routing problem can be transformed into its associate overall optimal routing problem (13), we investigate the existence and uniqueness of the solution to the associate problem (13) and, then, to the individually optimal routing problem (9).

Denote the set of feasible link flow patterns by FS_x . That is,

$$FS_x = \{x \mid \text{ There exists } y \text{ such that } x \text{ and } y \text{ satisfy (4) and } y \in FS_y\}.$$

Clearly, the set $\{(x, y) \mid (x, y) \text{ satisfies } (4) \text{ and } y \in FS_y\}$ is convex, closed, and bounded, Then, by noting that the orthogonal projection of a convex set onto a subspace is another convex set (see, e.g., [18]), the set FS_x is convex in x and a closed and bounded hyperplane (see, e.g., [18]). Note that D(x) in (6) is continuous in x and, thus, in y, and that the feasible set FS_y is closed and bounded. Then, there exists a solution of path flow patterns y to (6), according to the Weierstrass theorem (e.g., [10, 11]). Since D(x) is continuous and convex in x and strictly convex in x_s , we have the following.

Theorem 1 For the overall optimal routing problem for multi-class networks (6), an optimal path flow pattern y exists and, in particular, the resulting x_s is unique.

The uniqueness of x_s is shown by contradiction as follows. Suppose that x_s is not unique and that both $x^1 = (x_s^1, x_{-s}^1)$ and $x^2 = (x_s^2, x_{-s}^2)$ give the minimum D_{min} of D(x), for $x_s^1 \neq x_s^2$. Then, from the convexity of the feasible region of x, $\alpha x^1 + (1 - \alpha)x^2$, for some α (0 < α < 1), is also in the feasible region, and

$$D(\alpha \mathbf{x}^{1} + (1 - \alpha)\mathbf{x}^{2}) = D(\alpha \mathbf{x}_{s}^{1} + (1 - \alpha)\mathbf{x}_{s}^{2}, \alpha \mathbf{x}_{-s}^{1} + (1 - \alpha)\mathbf{x}_{-s}^{2})$$

$$< \alpha D(\mathbf{x}_{s}^{1}, \alpha \mathbf{x}_{-s}^{1} + (1 - \alpha)\mathbf{x}_{-s}^{2}) + (1 - \alpha)D(\mathbf{x}_{s}^{2}, \alpha \mathbf{x}_{-s}^{1} + (1 - \alpha)\mathbf{x}_{-s}^{2})$$

$$= \alpha D_{min} + (1 - \alpha)D_{min} = D_{min},$$

where the inequality follows from the strict convexity of D(x) in x_s and the second-last equality follows from the meaning of x_{-s} . The above relation contradicts the assumption that D_{min} is the minimum of D(x), and we see that the x that minimizes D(x) has a unique x_s .

For the individually optimal routing problem, we note that $\hat{D}(x)$ in (13) is continuous in x and, thus, in y, and that FS_y is closed and bounded. Then, similarly as above, there exists a solution of y to (13) according to the Weierstrass theorem (e.g., [10, 11]). With Lemma 1 and by noting that $\hat{D}(x)$ is convex in x and strictly convex in x_s , we have the existence and uniqueness of a solution to individually optimal routing as follows.

Theorem 2 For the individually optimal routing problem, there exists a solution y to (13) and thus that satisfies (9), and, in particular, the resulting x_s is unique.

For the overall (respectively, individually) optimal routing problem, consider the following optimization problem that involves only the link flow pattern x and does not involve the path flow pattern y (with D(x) to be replaced by $\hat{D}(x)$ for the individually optimal routing problem).

$$\min_{\mathbf{x}} D(\mathbf{x}) \text{ subject to } \mathbf{x} \in FS_{\mathbf{x}}. \tag{15}$$

Similarly as above, we see that there exists a solution of x to (15) (respectively, to the problem with D(x) being replaced by $\hat{D}(x)$), according to the Weierstrass theorem (e.g., [10, 11]). The optimization problem (15) (respectively, the one with D(x) being replaced by $\hat{D}(x)$) is another nonlinear convex optimization problem, but, clearly, (15) (respectively, the one with D(x) being replaced by $\hat{D}(x)$) gives the solution x that is the same as the link flow pattern x that the solutions y to (6) (respectively, (13), thus (9)) result in.

3 Link-Traffic Loop-Free Property

In this section, we show a property that holds for the individually/overall optimal routing for multi-class networks, called the *link-traffic loop-free property*. The link-traffic loop-free property is such that there exists no loop that consists of a sequence of links $(v_1, v_2), (v_2, v_3), \ldots, (v_{n'-1}, v_{n'})$ where $v_1, v_2, \ldots, v_{n'}$ are distinct nodes while $v_1 = v_{n'}$ such that class-k link flow $x_{v_1,v_2}^k > 0$, $x_{v_2,v_3}^k > 0$, $x_{v_{n'-1},v_{n'}}^k > 0$ (Figure 1). Although it is evident that no path has a loop, it is not clear whether there exists no loop for link flows of each class. For example, if Assumption A1 does not hold, as shown by the example given later, there may exist loops for link flows in the network.

From relations (1), (2), and (4), we have the following flow-balance relation.

$$r_{i-}^{k} (\text{if } i \in O^{k}) + \sum_{l \in V_{i}^{k}} x_{li}^{k} = r_{-i}^{k} (\text{if } i \in \mathcal{D}^{k}) + \sum_{l \in V_{i}^{k}} x_{il}^{k}, \quad i = 1, 2, \dots, n-1, \ k \in C,$$

$$(16)$$

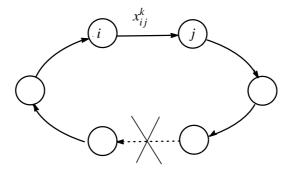


Figure 1: Link-traffic loop-free property in individually/overall optimal routing

where V_i^k is the set of immediately neighboring nodes of node i for class k, i.e., $V_i^k = \{j | (i, j) \in L^k, or (j, i) \in L^k\}$. The constraint with respect to i = n can be derived by summing up both sides of the above constraints for i = 1, 2, ..., n - 1. Define FSI as follows:

$$FSI = \{x \mid x \text{ satisfies (16) and } x \ge 0\}.$$

Note that the set of FSI is convex, closed, and bounded. Note, furthermore, that FSI includes but may not be identical to FS_x .

We have the associate problem for the individually optimal routing (and the overall optimal routing problem) with the following new constraint (with $\hat{D}(x)$ and $\hat{G}_{ij}^k(x)$, respectively, to be replaced by D(x) and $G_{ij}^k(x)$ for the overall optimal routing problem).

$$\min_{\mathbf{r}} \hat{D}(\mathbf{x}) \text{ subject to } \mathbf{x} \in FSI.$$
 (17)

The necessary and sufficient condition that a solution to the above individually/overall-optimal routing problem satisfies is given as follows.

Lemma 2 The link flow pattern x is an optimal solution to the individually (and overall) optimal routing problem with constraint $x \in FSI$ (17) if and only if x satisfies the following set of relations (with $G_{ij}^k(x)$ to be replaced by $g_{ij}^k(x)$ for the overall optimal routing problem).

$$\alpha_i^k - G_{ij}^k(\mathbf{x}) = \alpha_j^k, \text{ for } x_{ij}^k > 0, \ \ (i, j) \in L^k, \ k \in C,$$
 (18)

$$\alpha_i^k - G_{ij}^k(\mathbf{x}) \le \alpha_j^k, \text{ for } x_{ij}^k = 0, \ \ (i,j) \in L^k, \ k \in C,$$
 (19)

subject to $\mathbf{x} \in FSI$, where α_i^k $(i \in N, k \in C)$ are Lagrange multipliers.

PROOF: We show the case of individual optimization. The case of overall optimization is shown in a similar way. To obtain an optimal solution to problem (17), we form the Lagrangian function as follows,

$$H(\mathbf{x}, \alpha) = R\hat{D}(\mathbf{x}) + \sum_{k \in C} \sum_{i=1}^{n-1} \alpha_i^k [r_{i-}^k (\text{if } i \in O^k) + \sum_{l \in V_i} x_{li}^k - r_{-i}^k (\text{if } i \in \mathcal{D}^k) - \sum_{l \in V_i} x_{il}^k], \tag{20}$$

where α_i^k are Lagrange multipliers.

Since function $\hat{D}(x)$ is continuous and convex in x (and strictly convex in x_s) and FSI is convex, closed, and bounded, there exists a solution (that has a unique x_s) to problem (17) similarly as Theorem

2. Thus, the link flow pattern x that satisfies the following Kuhn-Tucker condition is an optimal solution (that has a unique x_s) to problem (17) (see, e.g., [10]).

$$\frac{\partial H}{\partial x_{ij}^k} = G_{ij}^k(\mathbf{x}) + \alpha_j^k - \alpha_i^k \ge 0,$$

$$x_{ij}^k \frac{\partial H}{\partial x_{ij}^k} = x_{ij}^k (G_{ij}^k(\mathbf{x}) + \alpha_j^k - \alpha_i^k) = 0,$$

$$x_{ij}^k \ge 0, (i, j) \in L^k, k \in C,$$

$$r_{i-}^{k}$$
 (if $i \in O^{k}$) + $\sum_{l \in V_{i}} x_{li}^{k} = r_{-i}^{k}$ (if $i \in \mathcal{D}^{k}$) + $\sum_{l \in V_{i}} x_{il}^{k}$, $i = 1, 2, ..., n - 1, k \in C$.

Rearranging the above relations, we have,

$$\alpha_i^k - G_{ij}^k(\mathbf{x}) = \alpha_j^k, \text{ for } x_{ij}^k > 0,$$

$$\alpha_i^k - G_{ij}^k(\mathbf{x}) \le \alpha_j^k, \text{ for } x_{ij}^k = 0,$$

$$(i, j) \in L^k, k \in C, \quad \mathbf{x} \in FSI.$$

The above set of relations is equivalent to the set of relations (18), (19) and $x \in FSI$, and it is the necessary and sufficient condition for a link flow pattern x to be a solution to the individually optimal routing problem (17). \square

With Lemma 2, we proceed to have the link-traffic loop-free property in the individually/overall optimal routing for problem (17).

Lemma 3 The class-k link traffic in a solution to the individually (and overall) optimal routing problem with constraint $\mathbf{x} \in FSI$ (17) is loop free for all $k \in C$. That is, there exists no class-k link traffic such that $x_{v_1v_2}^k > 0$, $x_{v_2v_3}^k > 0$, ..., $x_{v_{m-1}v_m}^k > 0$, $x_{v_mv_1}^k > 0$ (for all $k \in C$), where v_1, v_2, \ldots, v_m are distinct nodes in the solution to the individually (and overall) optimal routing problem and where at least one of the links involved is not a zero-cost link.

PROOF: We show the case of individual optimization. It is proved by contradiction. Assume that there exists class-k link traffic in the solution to individually optimal routing problem (17) for multi-class users such as $x_{\nu_1\nu_2}^k > 0$, $x_{\nu_2\nu_3}^k > 0$, ... $x_{\nu_{m-1}\nu_m}^k > 0$, $x_{\nu_m\nu_1}^k > 0$ ($k \in C$), where $\nu_1, \nu_2, \ldots, \nu_m$ are distinct nodes. According to Lemma 2, in the solution, we have

$$\alpha_{\nu_{1}}^{k} - G_{\nu_{1}\nu_{2}}^{k}(\mathbf{x}) = \alpha_{\nu_{2}}^{k},$$

$$\vdots$$

$$\alpha_{\nu_{m}}^{k} - G_{\nu_{m}\nu_{1}}^{k}(\mathbf{x}) = \alpha_{\nu_{1}}^{k}.$$
(21)

Then, we have

$$G_{\nu_1\nu_2}^k(\mathbf{x}) + G_{\nu_2\nu_3}^k(\mathbf{x}) + \dots + G_{\nu_m\nu_1}^k(\mathbf{x}) = 0,$$
 (22)

which contradicts the fact that $G_{ij}^k(\mathbf{x}) > 0$ if $x_{ij}^k > 0$ for at least one of (non-zero-cost) links involved.

The case of overall optimization is shown in a similar way as above by replacing $G_{ij}^k(\mathbf{x})$ by $g_{ij}^k(\mathbf{x})$. \square

Since the constraint $x \in FSI$ of the optimization problems may be weaker than the set of constraints $x \in FS_x$, there may be the possibility that a link flow solution x ($x \in FSI$) may not be realized by any path flow pattern. In the following, however, we confirm that there exists a path flow pattern y that results in any loop-free link-flow pattern x such that $x \in FSI$.

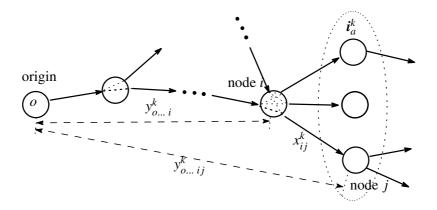


Figure 2: Assigning flows to paths on the basis of link flows

Proposition 1 There exists a path flow pattern y satisfying the constraint $y \in FS_y$ that results in a link-traffic loop-free flow pattern x satisfying the constraint $x \in FSI$. That is, for networks with a link-traffic loop-free flow pattern, $FSI = FS_x$.

PROOF: Consider an arbitrary loop-free link-flow pattern x that satisfies the constraint $x \in FSI$. We show how to make a path flow pattern $y \in FSI$, that results in the loop-free link-flow pattern $x \in FSI$.

We consider the following for each class. Consider a path $(o, v_1, v_2, ..., v_i, v_{i+1}, ..., d)$ where o is an origin node, d is a destination node, and $v_1, v_2, ..., v_i, v_{i+1}, ...$ are called 'intermediate nodes.' Then, we call the sequence $(o, v_1, v_2, ..., v_i)$ an *intermediate path* at node v_i of path $(o, v_1, v_2, ..., v_i, v_{i+1}, ..., d)$. Naturally, there may be multiple intermediate paths at each node including those coming from different origins. Furthermore, we also say that the sequence $(o, v_1, v_2, ..., v_i)$ is the *intermediate path* at node v_i that is included in the intermediate path $(o, v_1, v_2, ..., v_i, ..., v_i)$ longer than it, for i < j and $v_i \ne d$.

We can assign a path flow pattern y such that the constraint $x \in FSI$ is satisfied, as follows. We note that the flow through each intermediate path must be the sum of the flows of the paths that go though the intermediate path for each class. The allotment of the flow to an intermediate path of node v_{i+1} is done by splitting the flow to the intermediate paths of node v_i , in a way proportional to the link flow $x_{v_i,v_{i+1}}^k$ on the link (v_i, v_{i+1}) , say, for class k (in the case where node v_i is neither an origin nor a destination for the class, and other cases are also treated in a formal manner below). In that sense, we obtain a proportionally fair allotment.

More precisely, given a link-traffic loop-free flow pattern, the path flow pattern for class $k, k \in C$, can be obtained as follows: Since the used links has the loop-free property, a 'partial order' relation among class-k nodes holds, that is, from the origins down to the destinations. Therefore, there must exist at least one origin that receives no class-k flows from any nodes but only sends class-k flows to other nodes. We call it a *pure origin* for class k. Similarly, there must exist at least one destination that sends no class-k flows to any nodes but only receives class-k flows from other nodes. We call it a *pure destination* for class k. Denote by \mathbf{i}_a^k the set of nodes that may receive class-k flow directly from node k (Fig. 2).

- (i) The case where node i is a pure origin o. A node j directly connected to a pure origin o ($j \in o_a^k$) has the class-k link flow x_{oj}^k . There must be only one class-k intermediate path from the origin at a node $j \in o_a^k$, that passes through one class-k link to the node j from the origin, and, thus, the allotment of class-k flow to the class-k intermediate path is straightforward. Clearly, this allotment is relevant to (4) and (1) but does not violate them since $x \in FSI$ must hold.
- (ii) The case where node i is neither an origin nor a destination: Each of the class-k intermediate paths at

node $j, j \in \mathbf{i}_a^k$, that go through node i, will be allotted the ratio $\frac{x_{ij}^k}{\sum_{l \in \mathbf{i}_a^k} x_{il}^k}$ of the flow of the corresponding class-k intermediate path at node i. For example, if the class-k intermediate path (o, \dots, i) included in a class-k intermediate path (o, \dots, i, j) has the flow $y_{o\dots i}^k$, the class-k intermediate path (o, \dots, i, j) is to be allotted the flow

$$y_{o...ij}^{k} = y_{o...i}^{k} \frac{x_{ij}^{k}}{\sum_{l \in i_{o}^{k}} x_{il}^{k}}.$$

Clearly, this allotment is relevant to (4) and (1) but does not violate them since $x \in FSI$ must hold.

(iii) The case where the node i is not a pure origin but an origin that has the external arrival rate r_{i-}^k : Each of the class-k intermediate paths at node j, $j \in i_a^k$, that go through node i, will be allotted the ratio

$$\frac{\sum_{l \in i_a^k} x_{il}^k - r_{i-}^k}{\sum_{l \in i_a^k} x_{il}^k} \frac{x_{ij}^k}{\sum_{l \in i_a^k} x_{il}^k}$$

of the flow of the corresponding class-k intermediate path at node i. In addition, a new set of intermediate paths starting at node i is added to the group of node-j intermediate paths, and each is allotted the flow $r_{i-}^k \frac{x_{ij}^k}{\sum_{l \in i_a^k} x_{il}^k}$. Clearly, this allotment is relevant to (4) and (1) but does not violate them since $x \in FSI$ must hold.

(iv) The case where the node i is not a pure destination but a destination that has the departure rate r_{-i}^k : Each of the class-k intermediate paths at node j, $j \in i_a^k$, that go through node i, will be allotted the ratio

$$\frac{\sum_{l \in i_a^k} x_{il}^k}{\sum_{l \in i_a^k} x_{il}^k + r_{-i}^k} \frac{x_{ij}^k}{\sum_{l \in i_a^k} x_{il}^k}$$

of the flow of the corresponding class-k intermediate path at node i. In addition, we have a set of complete paths ending at node i, and each is allotted the ratio $\frac{r_{-i}^k}{\sum_{l \in i_a^k} x_{il}^k + r_{-i}^k}$ of the flow of the corresponding intermediate path of node i. Clearly, this allotment is relevant to (4) and (2) but does not violate them since $\mathbf{x} \in FSI$ must hold.

(v) The case where node i is a pure destination for class k. All the class-k paths that reach this node terminate at this node, and no further path-flow allotment for class k is needed anymore. Clearly, we see that, at this node, class-k path flow allotment so far is relevant to (4) and (2) but does not violate them since $x \in FSI$ must hold.

We therefore see that, at every step of the above five allotments in cases (i), (ii), (iii), (iv), and (v), the set of constraints (4) and $y \in FS_y$ is satisfied.

Therefore, for an arbitrary $x \in FSI$ with the loop-free property, starting from nodes that directly receives class-k flows only from origins that receive no class-k flows from other nodes, we can proceed the steps of allotting the amount of class-k flows to intermediate paths, and finally we can complete the assignment y of class-k path flows, for all $k \in C$, that result in the above-mentioned $x \in FSI$.

We note that the above obtained y satisfies both the constraints (4) and $y \in FS_y$ and, thus, that the above mentioned x satisfies the constraint $x \in FS_x$.

Since FSI includes FS_x , then, for networks with a link-traffic loop-free flow pattern, $FSI = FS_x$. \square

From the above Proposition 1 and Lemma 3, we can confirm the following.

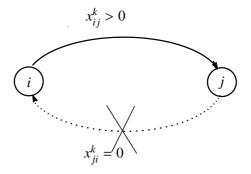


Figure 3: One-way traffic property in individually and overall optimal routing

Lemma 4 For the solution x with the link-traffic loop-free property for the individually/overall optimal routing with constraint $x \in FSI$ (17) there exists a y that satisfies both the set of constraints (4) and $y \in FS_y$, i.e., $FSI = FS_x$.

We, therefore, see that the solution x with the link-traffic loop-free property for the individually/overall optimal routing with constraint $x \in FSI$ (17) is also the solution for the individually (and overall) optimal routing problem (6,13,17), since its solution exists as discussed in Section 2.4.

Then, we have the following theorem.

Theorem 3 (**Link-traffic loop-free property**) The class-k link traffic in a solution to the individually (and overall) optimal routing problem (6,13,17) for a multi-class network is loop free for all $k \in C$. That is, there exists no class-k link traffic such that $x_{v_1v_2}^k > 0$, $x_{v_2v_3}^k > 0$, ..., $x_{v_{m-1}v_m}^k > 0$, $x_{v_mv_1}^k > 0$ (for all $k \in C$), where v_1, v_2, \ldots, v_m are distinct nodes in the solution to the individually (and overall) optimal routing problem and where at least one of the links involved is not a zero-cost link.

Now consider the case where two nodes are connected by two links. We have following result.

Corollary 2 (One-way traffic property) For any optimal solution x to an individually/overall optimal routing problem (6,13,17) for multi-class networks, the following relations hold true,

$$x_{ij}^k = 0, \quad \text{if } x_{ji}^k > 0,$$
 (23)

where either (i, j) or (j, i) is not a zero-cost link for class k, for (i, j), $(j, i) \in L^k$, $k \in C$.

PROOF: It is a direct result from Theorem 3. □

The property shown in the corollary 2 is called the *one-way traffic property* for the individually/overall optimal routing in multi-class networks. The physical meaning is clear. It shows that the traffic from the node i to node j, x_{ij}^k , and the user flow rate from the node j to node i, x_{ji}^k cannot be positive both at the same time as shown in Figure 3.

4 Examples

4.1 The Cases where Assumption A1 and A2 hold naturally

As to many networks, we can naturally assume the assumption A1. As we noted before, however, Assumption A2 looks somewhat awkward. In the following two cases, however, the assumption A2 holds naturally.

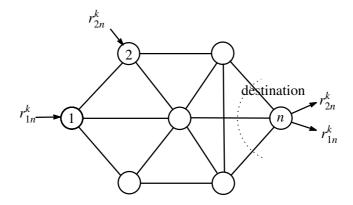


Figure 4: A network with one common destination and multiple origins

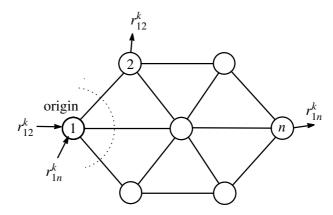


Figure 5: A network with one common origin and multiple destinations

4.1.1 Multi-class routing in networks with a common destination

Consider the individually/overall optimal routing problem for a multi-class network with one common destination and multiple origins for a class (we call it the problem with a common destination for the sake of brevity). as shown in Figure 4. Note that the problems of load balancing in distributed computer systems [14, 21] are equivalent to the routing problems in the networks with one common destination and multiple origins. Clearly, the assumption A2 holds naturally for the networks. The two link-traffic loop-free properties, Theorem 3 and Corollary 2 shown in the above section hold for the networks under individually and overall optimal routing.

In contrast, consider a case of noncooperative optimal routing with a finite (but plural) number of players for this model, i.e., users are divided into groups each of which is controlled by a decision maker that strives to optimize unilaterally the cost for its group only. Link-traffic loops have been found in the above-mentioned load-balancing problems (shown in [12, 13] if the definition of class given in those papers is changed to be the same as the one given in this paper).

4.1.2 Multi-class routing in networks with a common origin

We proceed to consider another network where there are multiple destinations but only one common origin (we call it the network with a common origin for the sake of brevity) as shown in Figure 5. The two link-traffic loop-free properties, Theorem 3 and Corollary 2 shown in the above section hold for the

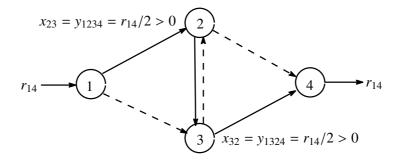


Figure 6: A network with one origin and one destination that satisfies Assumption A1(1) but does not satisfy A1(2)

networks for individually and overall optimal routing. Clearly, the assumption A2 holds naturally for the networks also.

4.2 Examples where Assumption A1 Does Not Hold

In this section, we examine two examples wherein either (1) or (2) in the assumption A1 is violated whereas the assumption A2 holds. We see that in both examples the link-traffic loop-free property does not hold. Therefore, we see that Assumption A1 is the condition that characterizes the category of networks for which the link-traffic loop-free property holds in individually and overall optimal routing in multi-class networks.

4.2.1 An example where (1) holds but (2) does not hold in the assumption A1

Consider a single-class network consisting of four nodes 1, 2, 3, and 4 (|C| = 1) and a single pair of origin 1 and destination 4, shown in Figure 6. Nodes 2 and 3 are connected by links (2,3) and (3,2). We consider the case where we have only two paths (1,2,3,4) and (1,3,2,4) connecting the O-D pair (1,4), but (1,2,4) and (1,3,4) are not paths connecting the O-D pair (1,4), which violates (2) in the assumption A1. We assume that the cost of each link depends only on the flow of the link and that $G_{12}(x) = G_{13}(x)$, $G_{23}(x) = G_{32}(x) > 0$, and $G_{24}(x) = G_{34}(x)$ where x denotes the flow through each link. Let the arrival rate at the origin be $r_{14} > 0$. Then, the optimal path flows of the two paths are identical, and $x_{23} = y_{1234} = r_{14}/2 = y_{1324} = x_{32} > 0$, which means that the network has a link-traffic loop. On the other hand, if (2) in the assumption A1 is to hold, then paths (1,2,4) and (1,3,4) need to be additionally available. Then, in optimal routing, only paths (1,2,4) and (1,3,4) are used, and the network has no link-traffic loop.

4.2.2 An example where (2) holds but (1) does not hold in the assumption A1

Consider a single-class network consisting of four nodes 1, 2, 3, and 4 (|C| = 1) shown in Figure 7. We consider the case where we have two distinct O-D pairs 1-3 and 4-2. That is, (1,2,3) and (4,3,2) are only paths. On the other hand, there exists no path connecting each of O-D pairs 1-2 and 4-3, respectively, even though O-D pairs 1-2 and 4-3 have links (1,2) and (4,3) connecting the origin and the destination, which violates (1) in the assumption A1. Let the arrival rates at origins be such that $r_{13} > 0$ and $r_{42} > 0$. Thus, there are two origin nodes (i.e., nodes 1 and 4) and two destination nodes (i.e., nodes 3 and 2) in the network. It is clear that we have only one solution such that $r_{23} = r_{12} > 0$ and $r_{32} = r_{42} > 0$, which is the optimal solution to individually/overall optimal routing problem under

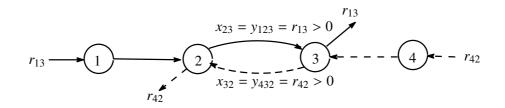


Figure 7: A network with two origins and two destinations that satisfies Assumption A1(2) but does not satisfy A1(1)

the set of constraints (4) and $y \in FS_y$. In this example, it is clear that we have a link-traffic loop, i.e., $x_{23} > 0$ and $x_{32} > 0$. On the other hand, if (1) in the assumption A1 holds, both of (1,2) and (4,3) can be paths, and the solution under the constraint $x \in FS_x$ is such that $x_{23} = r_{13} - r_{42} \ge 0$ and $x_{32} = 0$ if $r_{13} \ge r_{42}$ (under Assumption A2), which shows the freedom of link-traffic loops that holds under Assumption A1.

5 Conclusion

In this paper, we have studied both individually and overall optimal routing problems for multi-class networks with generalized link-cost functions and network configurations. We have seen that there is an associate overall optimal routing problem to each individually optimal routing problem for multi-class networks with the same solution under some condition. We have discussed the existence and uniqueness of the solutions to overall and individually optimal routing. Furthermore, we have shown that the link-traffic loop-free property holds for the individually and overall optimal routing in a wide range of networks. While doing so, we have discussed the relation between the formulations with path and link flow patterns. We have shown the condition that characterizes the category of multi-class networks that have the link-traffic loop-free property for overall and individually optimal routing.

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