An Approximation Scheme for Optimal Multicast Routing with Multiple QoS Constrains

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Abstract

This paper studies the problem of optimal multicast routing with multiple Quality-of-Service (QoS) constrains (OMRMC), which is regarded as a critical component in networks. However, it is not very efficient and effective for the existing algorithms. An undirected and connected graph G with m edges and n nodes is modeled to find a multicast routing tree from source to destination subset D in this paper, and a novel approximation algorithm called APP-OMRMC is proposed for OMRMC. The theoretical validations for the proposed method are presented to show its efficiency. After that, the different networks with different types of service are evaluated. Experimental results show that the proposed method APP-OMRMC is capable of finding the optimal (or minimum-cost) multicast routing tree with the approximation factor $(1+\varepsilon)$ and time complexity of $O(m \tau^{K-1})$.

Keywords: multicast routing, multiple QoS constrains, approximation algorithm

1. Introduction

Over the last decade, multicast routing (or Steiner) trees used for efficient one-to-many communications applications (*e.g.* web broadcasting, video teleconferencing, VOIP and HDTV) have emerged on the Internet [1-2]. Considering different quality-of-service (QoS) requirements, one of the solutions is minimum-cost multicast QoS routing. It is the problem of finding a tree in a graph from source to a subset of vertices called terminals, such that the total edge cost of which is minimal. More and more practical problems can be modeled as multicast routing tree problems nowadays, such as VLSI layout, network design, *etc*.

The problem of QoS multicast routing is challenging and usually known as multicast routing with multiple constrains problem (MRMC). As the best of our knowledge, finding the minimal-cost multicast routing tree with multiple QoS constrains is one of the Steiner tree problem which is NP-hard [3]. Currently, there has been much work involved in designing heuristic solutions for this problem [4-10]. Approximation algorithms for the node-weighted Steiner problem with fixed costs on each node and each edge have been presented in [11]. G. Xue *et al.* proposed various polynomial time algorithms [7-8] for multiple constrained optimal paths (MCOP) problem based on Yuan X's study [4]. Their algorithms resort to rounding and scaling to guarantee a solution within a factor $(1+\varepsilon)$ of the optimal solution. Other works that consider graph models with fixed edge weighs and employing the approximation algorithm for finding the optimal multicast routing tree or the optimal path can be found in [10-17].

These above mentioned researches mainly focus on solving MCOP by the efficiency approximation algorithm. Rather than previous research model, OMRMC is investigated from an approximation perspective in this paper. As far as we know, Xue's algorithm [7] is currently the fastest for path establishment in unicast routing with multiple QoS constrains. However, the solution to multicast routing problem is rarely or inefficient for the existing

ISSN: 2233-7857 IJFGCN Copyright © 2016 SERSC literature. In this study, a novel approximation algorithm called APP-OMRMC is proposed for OMRMC.

The rest of this paper is organized as follows. The problem to be studied is defined formally in Section 2, along with notations that will be used in later sections. In Section 3, the approximation scheme for the optimal multicast routing with multiple QoS constrains and its theoretical analysis are presented. In Section 4, experimental results obtained from different kinds of networks are reported to verifying the algorithm. Section 5 draws the conclusion of this paper.

2. Preliminaries

2.1. Problem Formulation

A communication network with K QoS constraints can be represented by a connected graph G(V, E, W, L), where V is the set of n vertices, E is the set of m edges, W is the set of weights and L is the set of constrains. Each edge has $K(K \ge 2)$ weights, and $W = \{w_i(e)\}, w_i(e) \ge 0$ is the kth weight of edge $e(\forall e \in E, 1 \le i \le K), L = (L_1, L_2, ..., L_K)$ be the K constraints. Then, it has the following definitions.

Definition 1. MRMC (Multicast Routing with Multiple Constrains Problem). Consider an undirected graph G(V, E, W, L), Let $s \in V$ be the source node and $D \subseteq V(s \notin D)$ be the set of terminal nodes. Denote $T_j(s, D)$, $j \in N$ as a multicast routing tree from source s to destination subset s in s, such that s is a sub-graph of s. Denote s in s is a sub-graph of s in s in

It is said to be a feasible multicast routing tree for T_j $s.t.w_i(T_j) \le L_i$. The entire feasible multicast routing tree in G(V, E, W, L) are denoted to be $\{T_j\}$.

Definition 2. OMRMC (Optimal Multicast Routing with Multiple Constrains Problem). It is to find an optimal multicast routing T^* among feasible paths $\{T_j\}$ in G for OMRMC from s to D and the smallest value of $\theta \in (0,1]$ such that $w_i(T^*) \le \theta \cdot L_i$.

Definition 3. α -Approximation Algorithm ($\alpha \ge 1$). An algorithm is a α -approximation algorithm for OMRMC if the algorithm generates a multicast routing tree T^{opt} from s to D such that $w_i(T^{opt}) \le \alpha \cdot \theta \cdot L_i$, the running time of the algorithm is bounded by a polynomial in the input size of the instance as well as $1/\alpha$.

2.2. Deterministic Algorithm with Auxiliary Graph

Xue *et al.* [7] provided an important methodology in their works, which is constructing an auxiliary graph. It transforms from an undirected graph G to a directed graph $G^{K,\tau}$. Each vertex $v \in G$ is associated with $(1+\tau)^{K-1}$ vertices (since each weight has been normalized $w_i(e)/L_i$, it could regard $\tau = \lceil (n-1)/\varepsilon \rceil$ (ε is the approximation factor) as the maximum integer of all constraints.) in $G^{K,\tau}$, i.e., $P(u,C_2,...,C_i)$ and $C_i \in [0,\tau]$ are used for recording the first weight and the ith weight of path length from source s to node u respectively, where $2 \le i \le K$. To each undirected edge (u,v) in G, it constructs the directed edge in $G^{K,\tau}$ from $P(u,C_2,...,C_i)$ to $P(v,D_2,...,D_i)$ and $D_i = C_i + w_i(u,v)$. Note that the infeasible paths of latter K-1 metrics have been filtered. Therefore, an optimal path in G corresponds to an optimal path from P(s,0,...,0) to $P(t,\tau,...,\tau)$ in $G^{K,\tau}$.

It uses the space to trade for the time for the basic idea behind Xue's algorithm. As far as we know, for path calculation in multiple constrained QoS routing, Xue's approximation

scheme with time complexity of $O(m\tau^{K-1})$ is currently the fastest one. As it observed from the examples, with the Xue's deterministic algorithm with auxiliary graph, it is feasible and meaningful to seek an effective and efficient algorithm to approximate the optimal multicast routing with multiple constrains.

3. The Proposed APP-OMRMC

3.1. An Approximation for Optimal Multicast Routing with Multiple Constrains

The detailed steps of APP-OMRMC are presented as below.

Algorithm APP-OMRMC

Input: G(V, E, W, L)

source node s and destination nodes set D, and the approximation ratio ε

Output: Topt

Step 1:

For each $e \in E$ in G(V, E, W, L), compute the new weights $w_i^N(e) = \left| \frac{w_i(e)}{L_i} \cdot \gamma \right|$,

i=1,...,K and set $\tau=L_1^N=...=L_K^N=\lceil\gamma\rceil$, $\gamma=\frac{n-1}{\varepsilon}>0$ such that G(V,E,W,L) is turned to be $G^N(V,E,W^N,\tau)$;

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Step 2:

Extend the graph $G^N(V, E, W^N, \tau)$ to the directed $G_K^N(V_K^N, E_K^N, W_K^N, \tau)$ with vertex set V_K^K and edge set E_K^N . To a given vertex $u \in V$ is extended to a vertex group $P(u, C_2, ..., C_i)$, where $0 \le C_i \le \tau, 2 \le i \le K$. To a given edge $(u, v) \in E$, E_K^N contains direct edge from vertex $P(u, C_2, ..., C_i)$ to vertex $P(v, \lambda_2, ..., \lambda_i)$, such that $\lambda_i = C_i + w_i^N(u, v)$. All such edges have the same length $w_1^N(u, v)$. In addition, E_K^N contains zero-length edges from $P(u, C_2, ..., C_i)$ to $P(u, C_2 + 1, ..., C_i)$;

Step 3:

Let $A(v, C_2, ..., C_i)$ be the prior adjacent node to $P(v, C_2, ..., C_i)$ in multicast routing tree T^{opt} . Initialize the total variables as follow.

For every node $v \in V$ and $C_i = 0$ to τ , $2 \le i \le k$ do

$$P(v, C_2, ..., C_i) \leftarrow \infty$$
; $P(v, C_2, ..., C_i) \leftarrow 0$;
 $A(v, C_2, ..., C_i) \leftarrow NULL$; $T^{opt} \leftarrow \{s\}$;

Step 4

for every adjacent node $v(v \notin T^{opt})$ of node u, where $u \in T^{opt}$, $v \notin T^{opt}$ and $C_i \le \tau$ do

if
$$P(v, D_2, ..., D_i) > P(u, C_2, ..., C_i) + w_1(u, v)$$
 where $D_i = C_i + w_i(u, v), 2 \le i \le K$

then
$$P(v, D_2, ..., D_i) \leftarrow P(u, C_2, ..., C_i) + w_1(u, v)$$

$$A(v, D_2, ..., D_i) \leftarrow u$$

 $T^{opt} \leftarrow \text{node v and edge (u, v)};$

end if

Step 5:

if
$$w_1^N(\mathbf{T}^{opt}) > \tau$$

then return No feasible multicast routing tree, Exit;

Step 6:

else return T^{opt} by omitting the later K-1 components of each extended node on the finding multicast routing tree;

end if

3.2. The Analysis of APP-OMRMC

Theorem 1. APP-OMRMC obtains a feasible multicast routing T^{opt} from source s to destination set D which minimizes $\max_{1 \le i \le K} w_i^N(T_j)$ among all multicast routing trees in $G_K^N(V_K^N, E_K^N, W_K^N, \tau)$. The worst case time complexity of the algorithm is $O(m\tau^{K-1})$.

Proof. Since the Step 4 searches $P(v, D_2, ..., D_i)$ in lexicographic order for every adjacent node $v(v \notin T^{opt})$ of node $u(u \in T^{opt})$ in algorithm APP-OMRMC, and guarantees D_i is not more than the largest integer τ , which means that APP-OMRMC searches a shortest path in the space of $\max_{1 \le i \le K} w_i^N(T_j)$, therefore APP-OMRMC obtains a multicast routing tree T^{opt} from source to destination set D that minimizes $\max_{1 \le i \le K} w_i^N(T_j)$ among all multicast routing trees in G_k^N .

During the running of APP-OMRMC, G_K^N have $O(n \cdot Q^{K-1})$ vertices and $O(2m \cdot Q^{K-1} + n \cdot Q^{K-1})$ edges, which confirm our observation that the space complexity is fairly large. Moreover, G_K^N contains a directed edge from $P(u, C_2, ..., C_i)$ to $P(v, D_2, ..., D_i)$ with length $w_i(u, v)$. Therefore, the optimal multicast routing tree T^{opt} found in APP-OMRMC minimizes $\max_{1 \le i \le K} w_i^N(T_i)$ among multicast routing trees in G_K^N .

Since each $w_i(e)$ is a positive real-value, for $2 \le i \le K$, $e \in E$, the existence of directed edge from $P(u, C_2, ..., C_i)$ to $P(v, D_2, ..., D_i)$ in G_K^N implies that $C_i < D_i$ for $2 \le i \le K$, $e \in E$. Therefore, the graph G_K^N is acyclic. For an acyclic graph, the worst case time complexity of each running process is $O(2m \cdot \tau^{K-1} + n \cdot \tau^{K-1}) = O(m\tau^{K-1})$. So the worst case time complexity of APP-OMRMC is $O(m\tau^{K-1})$.

According to Theorem 1, it concludes that the time complexity of APP-OMRMC is nearly the same as that of Xue-based scheme in finding the MCOP [7], but it is easier to implement according to the procedure of algorithm APP-OMRMC.

Theorem 2. To any feasible multicast routing tree T_j in G, it is a feasible multicast routing tree in G_K^N .

Proof. To a feasible multicast routing tree T_i in G, it has

$$w_i(\mathbf{T}_j) = \sum_{e \in \mathbf{T}_j} w_i(e) \le L_i \tag{1}$$

It implies that

$$\frac{\sum_{e \in T_j} w_i(e)}{L_i} \cdot \gamma = \sum_{e \in T_j} \frac{w_i(e)}{L_i} \cdot \gamma \le \gamma$$
 (2)

Since
$$w_i^N(\mathbf{e}) = \left| \frac{w_i(\mathbf{e})}{L_i} \cdot \gamma \right| \le \frac{w_i(\mathbf{e})}{L_i} \cdot \gamma$$
 (3)

And
$$\gamma \leq \lceil \gamma \rceil = \tau$$
 (4)

Then it has

$$w_{i}^{N}(\mathbf{T}_{j}) = \sum_{e \in \mathbf{T}_{j}} w_{i}^{N}(\mathbf{e})$$

$$= \sum_{e \in \mathbf{T}_{j}} \left\lfloor \frac{w_{i}(\mathbf{e})}{L_{i}} \cdot \gamma \right\rfloor$$

$$\leq \sum_{e \in \mathbf{T}_{i}} \frac{w_{i}(\mathbf{e})}{L_{i}} \cdot \gamma$$
(5)

According to Equations (2), (4) and (5), it has

$$w_i^N(\mathbf{T}_i) \le \tau \tag{6}$$

Hence, any feasible multicast routing tree T_i in G is a feasible multicast routing tree in G_K^N .

Theorem 3. Algorithm APP-OMRMC finds a $(1+\varepsilon)$ approximation multicast routing tree to OMRMC.

Proof. To the optimal multicast routing tree T^* in G, it has

$$w_i(T^*) \le \theta \cdot L_i \tag{7}$$

$$w_i^N(T^*) = \sum_{e \in T^*} w_i^N(e) = \sum_{e \in T^*} \left\lfloor \frac{w_i(e)}{L_i} \cdot \gamma \right\rfloor$$
 (8)

This implies that

$$w_{i}^{N}(T^{*}) \leq \sum_{e \in T^{*}} \left(\frac{w_{i}(e)}{L_{i}} \cdot \gamma\right)$$

$$= \frac{w_{i}(T^{*})}{L_{i}} \cdot \gamma$$

$$\leq \theta \cdot \gamma$$
(9)

On the other hand, T^{opt} minimizes $\max_{1 \le i \le K} w_i^N(T_j)$ among all multicast routing trees in G_K^N and T^* is a feasible multicast routing tree as mentioned in Theorem 1 and in Theorem 2 respectively, then it has

$$\max_{1 \le i \le K} w_i^N(T^{opt}) \le \max_{1 \le i \le K} w_i^N(T^*)$$
(10)

Since T^{OPT} have |V|-1 hops at most, and there are n nodes in G, it implies that

$$w_{i}^{N}(T^{opt}) = \sum_{e \in T^{opt}} w_{i}^{N}(\mathbf{e}) = \sum_{e \in T^{opt}} \left\lfloor \frac{w_{i}(\mathbf{e})}{L_{i}} \cdot \gamma \right\rfloor$$

$$\geq \sum_{e \in T^{opt}} \left(\frac{w_{i}(\mathbf{e})}{L_{i}} \cdot \gamma - 1 \right)$$

$$= \sum_{e \in T^{opt}} \left(\frac{w_{i}(\mathbf{e})}{L_{i}} \cdot \gamma \right) - \sum_{e \in T^{opt}} 1$$

$$\geq \frac{w_{i}(T^{opt})}{L_{i}} \cdot \gamma - (n-1)$$

$$(11)$$

According to Equations (9), (10) and (11), it has

$$\frac{w_i(T^{opt})}{L_i} \cdot \gamma - (n-1) \le \theta \cdot \gamma \tag{12}$$

It implies that

$$\frac{w_i(T^{opt})}{L_i} \le \theta + \frac{\mathsf{n} - 1}{\gamma} \tag{13}$$

Let $\frac{n-1}{\gamma} = \varepsilon \cdot \theta$, $\varepsilon > 0$, then

$$w_i(T^{opt}) \le (1+\varepsilon) \cdot \theta \cdot L_i, 1 \le i \le K \tag{14}$$

Hence, APP-OMRMC can find a $(1+\varepsilon)$ approximation multicast routing tree to OMRMC, and the time complexity of APP-OMRMC is $O(m\tau^{K-1})$.

4. Simulation and Performance Evaluation

It shall evaluate both the performance of APP-OMRMC and the performance of obtained multicast routing tree experimentally in this section. To evaluate its efficiency, APP-OMRMC was implemented and conducted a performance study. Special network namely CERNet [2] is used in these experiments which are run on an Intel Core Duo CPU 1.66 GHz PC with 2 GB memory. There are 25 nodes and 30 edges in CERNet as showed in Figure 1, and other parameters for the network can be found in the network link address: http://code.google.com/p/efptas/downloads/list. Each link of CERNet has three weights, which corresponds to Cost, Delay and Jitter.

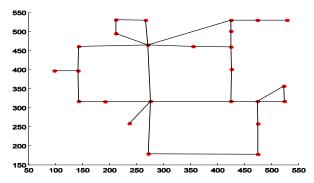


Figure 1. The Network Topology of CERNet

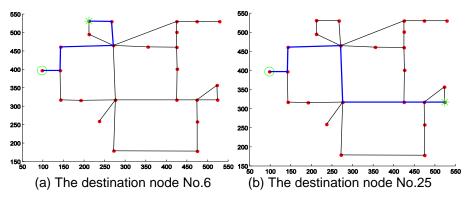


Figure 2. The Minimal Cost Path

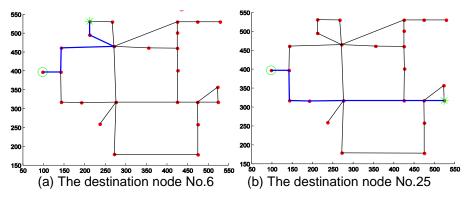


Figure 3. The Minimal Delay Path

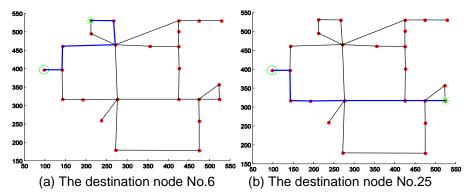


Figure 4. The Minimal Jitter Path

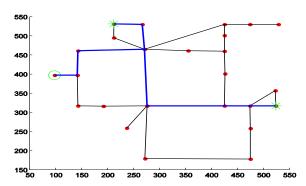


Figure 5. The Optimal Multicast Routing Tree with Three Constrains

All the red points are denoted as nodes of the network, the black wires as the links, the green circle as the source node s (No.16) and the green star as the destination node set D (No.6 and No.25) in the network CERNet respectively for Figure 2- Figure 5. The blue paths in these figures indicate the minimal or the optimal paths from source to destination. Figure 2(a) and Figure 2(b) show the minimal cost paths from source node (No.16) to destination nodes (No.16 and No.25) respectively in the network, each link of which is with the weight cost. As the same as Figure 2(a) and Figure 2(b), the minimal delay path is show in Figure 3 (a) and Figure 3(b), each link of which is with the weight delay. Figure 4 illustrates the minimal delay jitter path, and the link is with the weight delay jitter. As expected, the optimal multicast routing tree with three constrains by the algorithm APP-OMRMC is able to be found, and the optimal multicast routing tree described in Figure 5, for which the source node (No.16) and the destination node set D (No.6 and No.25), shows the corresponding results.

5. Conclusion

It discusses the problem of optimal multicast routing with multiple in this paper. An approximation scheme for the OMRMC (APP-OMRMC) is proposed, which is based on the technique of auxiliary graph construction, scaling and rounding. The algorithm can achieve the low complexity and obtain the approximation optimal multicast routing tree, respectively. According to the experiment on the random network, it shows that the proposed method APP-OMRMC is capable of finding the optimal (or minimum-cost) multicast routing tree with the approximation factor $(1+\varepsilon)$.

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