# The maximum integer multiterminal flow problem in directed graphs

Cédric Bentz<sup>\*</sup>

#### Abstract

Given an arc-capacitated digraph and k terminal vertices, the directed maximum integer multiterminal flow problem is to route the maximum number of flow units between the terminals. We introduce a new parameter  $k_L \leq k$  for this problem and study its complexity with respect to  $k_L$ .

**Keywords**: integer multiterminal flow, directed graphs, *APX*-hardness, approximation algorithms.

## 1 Introduction

Routing problems in networks are commonly modelled by flow or multicommodity flow problems. Given an edge-capacitated graph (directed or undirected), the goal is to route flow units (requests) between prespecified vertices. When one seeks to route the maximum number of flow units from a unique source to a unique sink, the problem is the famous *maximum flow problem*. The Ford-Fulkerson's theorem [9] gives a good characterization for this case, which is efficiently solvable [1]. In particular, this theorem states that, if the capacities are integral, the value of a maximum integer flow is equal to the value of a minimum cut, i.e., to the value of a minimum weight set of edges whose removal separates the source from the sink. Unfortunately, this does not hold for more general variants.

One of the most studied variant is the maximum integer multiflow problem: given an edge-capacitated graph G = (V, E) and a list of source-sink pairs, the goal is to simultaneously route the maximum number of flow units, each unit being routed from one source to its corresponding sink, while respecting the capacity constraints on the edges. For two source-sink pairs, this problem is strongly  $\mathcal{NP}$ -hard both in directed acyclic graphs and in undirected graphs [8] (more precisely, given an unweighted undirected or directed acyclic graph, two pairs  $(s_1, s'_1)$  and  $(s_2, s'_2)$  and a (polynomially

 $<sup>^{*}{\</sup>rm LRI},$  Université Paris-Sud and CNRS, 91405 Orsay Cedex, France (e-mail: cedric.bentz@lri.fr)

bounded) demand d, it is  $\mathcal{NP}$ -complete to decide whether we can simultaneously route d+1 disjoint paths, d paths from  $s_1$  to  $s'_1$  and one from  $s_2$  to  $s'_2$ ), and even APX-hard in undirected and directed graphs [16, p. 489] (implying that there is no PTAS if  $\mathcal{P} \neq \mathcal{NP}$ ; see Section 2 for a definition). Moreover, if  $\mathcal{P} \neq \mathcal{NP}$ , it cannot be approximated efficiently within  $|E|^{\frac{1}{2}-\epsilon}$  in digraphs [16] and within  $(\log |E|)^{\frac{1}{3}-\epsilon}$  in undirected graphs [2], for any  $\epsilon > 0$  (recall that, for a maximization (resp. minimization) problem, an  $\alpha$ -approximation algorithm is a polynomial-time algorithm that always outputs a feasible solution whose value is at least  $1/\alpha$  times (resp. at most  $\alpha$  times) the value of an optimal solution). On the positive side, an  $O\left(\sqrt{|E|}\right)$ -approximation algorithm is known for general digraphs [18], and an  $O\left(\sqrt{|V|}\right)$ -approximation is known for directed acyclic graphs [4, 20] and undirected graphs [4]. For further results on the tractability and approximability of special cases, see [6, 11].

The corresponding generalization of the problem of finding a minimum cut is the *minimum multicut problem*, which asks to select a minimum weight set of edges whose removal separates each source from its corresponding sink. This problem is also  $\mathcal{NP}$ -hard (even in very restricted classes of graphs, such as unweighted stars [15]), and has a noticeable relationship with the former: the continuous relaxations of the linear programming formulations of the two problems are dual [14]. In particular, this interesting property has been used to design good approximation algorithms for both problems [14, 15, 22]. Further results and references concerning the maximum integer multiflow and minimum multicut problems can be found in [1, 6, 22].

Another generalization of the maximum flow problem is the maximum integer multiterminal flow problem (MAXIMTF): given an edge-capacitated graph and a set  $T = \{t_1, \ldots, t_k\}$  of terminal vertices, MAXIMTF is to route the maximum number of flow units between the terminals. Note that this problem is a particular maximum integer multiflow problem in which the source-sink pairs are  $(t_i, t_j)$  for  $i \neq j$ . The associated minimum multiterminal cut problem (MINMTC) is to select a minimum weight set of edges whose removal separates  $t_i$  from  $t_j$  for  $i \neq j$ . Note that MAXIMTF and MINMTC also have the duality relationship mentioned above. MINMTC has been widely studied in the undirected case [5, 6, 7, 23], and the directed case has also received some attention: Garg et al. [13] show that it is  $\mathcal{NP}$ hard even for k = 2 and give an  $2 \log_2 k$ -approximation algorithm, and Naor and Zosin [19] give a 2-approximation algorithm. However, the algorithm of Garg et al. has an interesting property: it computes a multiterminal cut whose value is at most  $2\log_2 k$  times the value of an integer multiterminal flow, and hence is an  $2\log_2 k$ -approximation for both MINMTC and MAXIMTF (while the algorithm of Naor and Zosin does not provide an approximate solution for MAXIMTF). Costa et al. [6] show that MAXIMTF

and MINMTC are polynomial-time solvable in acyclic directed graphs by using a simple reduction to a maximum flow and a minimum cut problem, respectively. To the best of our knowledge, these are the only results about MAXIMTF in directed graphs. In undirected graphs, MAXIMTF has recently be shown to be polynomial-time solvable via the ellipsoid method [17]: the resolution is based on the related Mader's theorem on T-paths [21, Chap. 73]. Algorithmic aspects of special cases have also been studied (see [3] and [12]). However, it can be noticed that, for all the problems mentioned above, the general directed case is "harder" than the undirected one, since there exists a linear reduction from the latter to the former: simply replace each edge by the gadget given in [21, (70.9) on p. 1224].

The motivation of this paper is to explore further the complexity of MAXIMTF in directed graphs. We say that a terminal is *lonely* if it lies on at least one directed cycle containing no other terminal, and we let  $T_L$  denote the set of lonely terminals and  $k_L = |T_L|$  (note that  $k_L = 0$  in directed acyclic graphs). We shall see that  $k_L$  is a particularly interesting parameter for better understanding the complexity and approximability of MAXIMTF in digraphs ( $k_L$  is small when either k or the number of directed cycles is small, and both parameters can be arbitrarily larger than  $k_L$ ): this paper gives a complete classification of the tractable and intractable cases of MAXIMTF in digraphs with respect to this parameter. Moreover, some of our results will extend to MINMTC.

Intuitively, the condition  $k_L = 0$  ensures that any flow unit routed from a terminal  $t_i$  to one of the k terminals (including  $t_i$  itself) will go through at least one terminal different from  $t_i$ ; therefore, it can be assumed to be routed from one terminal to another (different) terminal.

We first show that MAXIMTF is  $\mathcal{NP}$ -hard to approximate within  $2 - \epsilon$ for any  $\epsilon > 0$  in directed graphs, even if  $k_L = k = 2$  or if  $k_L = 1$  and k = 3(Section 2). Then, we prove MAXIMTF to be tractable when  $k_L = 0$  by reducing it to a simple maximum flow problem, and improve the  $2\log_2 k$ approximation algorithm of Garg et al. [13] by providing an  $2\log_2(k_L + 2)$ approximation algorithm for the general case (Section 3). We also show the tightness of our analysis. Eventually, we show the case  $k_L = 1$  and k = 2 to be polynomial-time solvable (Section 4). We leave as open the problem of deciding whether there exists an O(1)-approximation algorithm for MAXIMTF in digraphs.

Note that, throughout this paper, we consider only *simple* graphs. We call *Directed* MAXIMTF the problem MAXIMTF defined in directed graphs.

### 2 APX-hardness proof

We show in this section that Directed MAXIMTF is APX-hard (i.e., there exists an  $\alpha > 1$  such that there is no  $\alpha$ -approximation algorithm for

this problem), even if  $k = k_L = 2$  (or  $k_L = 1$  and k = 3). First, notice that, when k = 2 (i.e.,  $T = \{t_1, t_2\}$ ), Directed MAXIMTF is equivalent (in digraphs) to the maximum integer multiflow problem with two source-sink pairs  $(s_1, s'_1)$  and  $(s_2, s'_2)$ . Indeed, given  $(s_1, s'_1)$  and  $(s_2, s'_2)$ , we can obtain an equivalent instance of Directed MAXIMTF by defining two new terminals,  $t_1$  and  $t_2$ , and by linking (by arcs with sufficiently large capacities)  $t_1$  to  $s_1$ ,  $s'_2$  to  $t_1$ ,  $t_2$  to  $s_2$  and  $s'_1$  to  $t_2$ : any flow unit routed between  $t_1$  and  $t_2$  is either routed from  $s_1$  to  $s'_1$  or from  $s_2$  to  $s'_2$ . Conversely, given two terminals  $t_1$  and  $t_2$ , we can define  $s_1 = t_1$ ,  $s'_1 = t_2$ ,  $s_2 = t_2$  and  $s'_2 = t_1$ . Obviously, this transformation does not apply to the undirected case.

We will use the fact that, if  $\mathcal{P} \neq \mathcal{NP}$ , there is no  $\rho$ -approximation algorithm for the maximum integer multiflow problem in digraphs even with two source-sink pairs, for any  $\rho < 2$  (recall that this problem is APX – hard even in undirected graphs [16, p. 489]). Indeed, the problem P2: given a digraph G and two vertex pairs  $(s_1, s'_1)$  and  $(s_2, s'_2)$ , decide whether there simultaneously exist in G two paths, one from  $s_1$  to  $s'_1$ , and the other from  $s_2$  to  $s'_2$  is  $\mathcal{NP}$ -complete [10] (an instance of P2 is called *feasible* if two such paths exist). This result was used in [16] to prove a strong inapproximability result for the maximum integer multiflow problem in digraphs.

Now, let  $\mathcal{I}$  be any instance of P2. If there exists an approximation algorithm for the maximum integer multiflow problem with a ratio  $\rho < 2$ , such an algorithm would output a solution of value 1 if  $\mathcal{I}$  is not feasible, and of value 2 otherwise (since  $\frac{2}{\rho} > 1$ ): this would give a polynomial-time algorithm for solving P2. This implies:

**Theorem 1.** Directed MAXIMTF is  $\mathcal{NP}$ -hard to approximate within  $2 - \epsilon$  for any  $\epsilon > 0$ , even when  $k_L = k = 2$ .

Note that this lower bound is tight for k = 2, since applying Garg et al.'s algorithm yields a 2-approximation in this case [13]. Moreover, this result essentially matches the complexity result (namely, *APX*-hardness) for the associated cut problem MINMTC in directed graphs [13]. Now, we prove the same result for the case k = 3,  $k_L = 1$ . The proof is quite similar: given two vertex pairs  $(s_1, s'_1)$  and  $(s_2, s'_2)$ , define three new terminals  $t_1, t_2, t_3$ , and add four arcs,  $(t_1, s_1), (s'_2, t_1), (t_2, s_2), (s'_1, t_3)$ ; then, any path leaving  $t_1$ is routed towards  $t_3$ . Note that  $T_L = \{t_1\}$ , since the only terminal lying on a directed cycle is  $t_1$ . Hence:

**Theorem 2.** Directed MAXIMTF is  $\mathcal{NP}$ -hard to approximate within  $2 - \epsilon$  for any  $\epsilon > 0$ , even when k = 3 and  $k_L = 1$ .

We shall deal with the case  $k_L = 0$  (which generalizes the acyclic case) in Section 3 and with the case  $k_L = 1$  and k = 2 in Section 4.

### **3** Exact and approximation algorithms

From the previous section, Directed MAXIMTF is APX-hard for  $k_L \geq 1$ . Hence, if  $\mathcal{P} \neq \mathcal{NP}$ , the only efficient algorithms one can expect to design are approximation algorithms. In this section, we improve the  $2\log_2 k$ approximation algorithm of Garg et al. [13] and give an  $2\log_2(k_L + 2)$ approximation algorithm for Directed MAXIMTF.

The basic idea of our algorithm is to combine the algorithm of Garg et al. with an improvement of an idea used in [6, Proposition 3]. The main idea of the proof of [6, Proposition 3] (that shows that MAXIMTF and MINMTC are polynomial-time solvable in directed acyclic graphs) is to split up each terminal vertex  $t_i$  into two new vertices,  $t'_i$  and  $t''_i$ , such that all the vertices in  $\Gamma^-(t_i)$  are linked to  $t'_i$  and  $t''_i$  is linked only to the vertices in  $\Gamma^+(t_i)$  (where, for a digraph G = (V, A) and  $v \in V$ ,  $\Gamma^+(v) = \{u \in V \text{ such that } (v, u) \in A\}$ and  $\Gamma^-(v) = \{u \in V \text{ such that } (u, v) \in A\}$ ). Then, we add two new vertices,  $\sigma$  and  $\tau$ , and link (by arcs with large capacities) every  $t'_i$  to  $\tau$  and  $\sigma$  to every  $t''_i$ . Finally, we compute a maximum flow between  $\sigma$  and  $\tau$  (obviously, we assume that the capacities are integral). The obtained flow is a valid integer multiterminal flow for the initial instance if, in the modified instance, no flow unit is routed from  $t''_i$  to  $t'_i$  for some i.

We can obtain an interesting strengthening of [6, Proposition 3] by noticing that, if there is no lonely terminal, then, by splitting up the terminals as explained, there will remain no directed path from  $t''_i$  to  $t'_i$  for each *i*, and hence we can solve MAXIMTF and MINMTC using the above technique.

**Theorem 3.** MINMTC and MAXIMTF are polynomial-time solvable in directed graphs if  $k_L = 0$ , by using a max flow-min cut algorithm.

Actually, if we want to guarantee that, after splitting up each terminal, the modified graph does not admit a directed path from  $t''_i$  to  $t'_i$  for some *i* (otherwise, we cannot be sure that the flow we will compute in the modified graph will be a valid multiterminal flow in the initial graph), this is essentially the best (i.e., weakest) assumption that can be made.

**Theorem 4.** After splitting up all the terminals, there is no directed path between  $t''_i$  and  $t'_i$  for each *i* if and only if  $k_L = 0$ .

*Proof.* Follows directly from the definition of  $T_L$  and the way we split up the terminals.

Theorems 3 and 4 show the importance of the parameter  $k_L$  for both MAXIMTF and MINMTC. Moreover, this suggests the following approach (algorithm  $\mathcal{A}$ ) for finding approximate solutions for these two problems:

1. If  $k \ge k_L + 2$ , split up each terminal  $t_i \in T - T_L$  into  $t'_i$  and  $t''_i$  as explained above, add the two vertices  $\sigma$  and  $\tau$ , and link (by arcs with sufficiently large capacities) every  $t'_i$  to  $\tau$  and  $\sigma$  to every  $t''_i$ ;

2. Compute a solution for this new instance (i.e., where the terminal set is  $T_L(\bigcup \{\sigma, \tau\})$  by using the algorithm of Garg et al. [13].

Hence, the main difference with the algorithm in [13] is that, before using their divide-and-conquer strategy, we "replace" the terminals in  $T \setminus T_L$  by only two terminals,  $\sigma$  and  $\tau$ . This implies that we use Garg et al.'s algorithm on an instance with  $k_L + 2$  terminals, and so we obtain an approximation factor of  $2\log_2(k_L + 2)$  (instead of  $2\log_2 k$ ). The key point is that any flow unit routed during the algorithm from  $\sigma$  to  $\tau$  via  $t'_i$  and  $t'_i$  for some *i* with  $t_i \notin T_L$  (if any) can be re-routed either between  $\sigma$  and a terminal in  $T_L$  or between a terminal in  $T_L$  and  $\tau$  (because  $t_i \in T - T_L$ ). This implies:

**Theorem 5.** Algorithm  $\mathcal{A}$  is an  $2\log_2(\min(k_L+2,k))$ -approximation algorithm for MAXIMTF in directed graphs.

Moreover, the above transformation can be of independent interest, since it can always be applied in order to reduce the instance size (i.e., to reduce the number of terminals from k to  $k_L + 2$ ) in any procedure computing an integer multiterminal flow (e.g., an exact implicit enumeration procedure).

Actually, we can prove that our analysis of the approximation ratio of algorithm  $\mathcal{A}$  is tight. To do this, we use an instance built on an undirected tree with  $2^p$  vertices. We are given a path  $P = u_0, u_1, \ldots, u_p$ , and, for each  $i \in \{0, \ldots, p-1\}, 2^i - 1$  leaves are linked by an edge of capacity 1 to  $u_i$  (let this set of leaves be  $\mathcal{L}_i$ ). All the vertices are terminal vertices (and hence  $k = 2^p$ , i.e.,  $p = \log_2 k$ , and the p edges of P are valued by a big integer N. To transform this undirected graph into a directed one, we replace each edge by the gadget given in [21, (70.9) on p. 1224]: each arc of the gadget has the capacity of the initial edge. In this instance, all the terminals are lonely (i.e.,  $k = k_L$ ), and hence our algorithm simply consists in applying Garg et al.'s algorithm. Hence, we will prove the tightness of their analysis, and this will imply the tightness of ours. We assume without loss of generality that Garg et al.'s algorithm always breaks ties in the worst possible way, and that it computes a solution by iteratively separating  $u_{p-i+1}$  and  $\mathcal{L}_{p-i}$  from  $u_{p-i}$ : indeed, this will result in a binary search on k. Moreover, the first max flow computed at the  $i^{th}$  iteration will consist in routing N units of flow between  $u_0$  and  $u_{p-i+1}$  and one unit of flow between  $u_{p-i}$  and each leaf in  $\mathcal{L}_{p-i}$  (the second max flow is symmetric; thus, it has the same value). The min cut will consist in cutting two different arcs (adjacent to the same terminal) on the gadget linking  $u_{p-i}$  to  $u_{p-i+1}$  and on the gadget linking  $u_{p-i}$  to each leaf in  $\mathcal{L}_{p-i}$ . The algorithm of Garg et al. outputs a flow of value at most  $N + (2^{p-1}-1) + (2^{p-2}-1) + \dots + (2^{1}-1) = N + 2^{p} - p - 1$  (even if it combines flows obtained in different iterations in the best possible way) and a cut of value  $2(pN + 2^p - p - 1)$ . The ratio between these two values is equal to  $\frac{2(pN+2^p-p-1)}{N+2^p-p-1}$ , and tends to  $2p = 2\log_2 k_L$  as N increases, establishing the

tightness of Garg et al.'s analysis. However, this does not imply that one cannot hope for a better approximation ratio by using a different algorithm.

Eventually, note that tighter ratios can be obtained for special cases (for example, when  $k_L = 1$ , our analysis yields a 2-approximation).

#### 4 $k_L = 1$ and k = 2: A tractable case

Directed MAXIMTF is APX-hard even when  $k = k_L = 2$  and when  $k_L = 1$  and k = 3 (see Section 2), and polynomial-time solvable when  $k_L = 0$  (see Section 3). In this section, we settle the last case and show that Directed MAXIMTF is polynomial-time solvable when  $k_L = 1$  and k = 2.

Let  $T = \{t_1, t_2\}$  and  $T_L = \{t_1\}$ . From the definition of  $T_L$ , there exists at least one directed cycle containing  $t_1$  but not  $t_2$ , and there exists no directed cycle containing  $t_2$  but not  $t_1$ . We are going to show that, in fact, this instance can be transformed into an equivalent one in which  $k_L = 0$ ; then, we will conclude by using Theorem 3. We show the following lemma:

**Lemma 1.** Let  $\mathcal{I}$  be an instance of Directed MAXIMTF with  $T = \{t_1, t_2\}$ and  $T_L = \{t_1\}$ . On any directed cycle containing  $t_1$  but not  $t_2$ , there is a removable arc, i.e., an arc not used by at least one optimal solution for  $\mathcal{I}$ .

Proof. Let  $C = \{t_1, u_1, u_2, \ldots, u_p, t_1\}$  be a directed cycle not containing  $t_2$ . We show that, on C, there exists an arc lying neither on an elementary path from  $t_1$  to  $t_2$  nor on an elementary path from  $t_2$  to  $t_1$ . Let i be such that there is a path from  $u_i$  to  $t_2$  not containing  $t_1$ , but there is no path from  $u_{i+1}$  to  $t_2$  not containing  $t_1$ . If i exists, then the arc  $(u_i, u_{i+1})$  can be removed. Indeed, it cannot lie on an elementary path from  $t_1$  to  $t_2$  (since there is no path from  $u_{i+1}$  to  $t_2$  not containing  $t_1$ ), and it cannot lie on an elementary path from  $t_2$  to  $t_1$  (since otherwise there is a path from  $t_2$  to  $u_i$  not containing  $t_1$ , and  $t_2 \in T_L$ ). If i does not exist, then we can remove either  $(t_1, u_1)$  (if there is no path from  $u_1$  to  $t_2$  not containing  $t_1$ ) or  $(u_p, t_1)$ (if there is a path from  $u_p$  to  $t_2$  not containing  $t_1$ ). Lemma 1 follows.

The proof of Lemma 1 also provides an algorithm to solve MAXIMTF in this case, by iteratively removing arcs until there remains no lonely terminal.

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