Multiple source, single sink maximum flow in a planar graph

Glencora Borradaile
Oregon State University

Christian Wulff-Nilsen
University of Copenhagen

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Abstract

We give an $O(n^{1.5}\log n)$ time algorithm for finding the maximum flow in a directed planar graph with multiple sources and a single sink. The techniques generalize to a subquadratic time algorithm for bounded genus graphs.

1 Introduction

In general graphs, multiple source flow problems are reduced to single-source problems but connecting a super source to the sources with infinite capacity arcs. In planar graphs this reduction destroys the planarity. Using the maximum flow algorithms for general graphs, a multiple-source, single (or multiple) sink max flow in directed planar graphs can be solved in $O(n^2\log n)$ time using Goldberg and Tarjan’s preflow push algorithm [7] or $O(n^{1.5}\log n\log U)$ where $U$ is the maximum edge capacity using Goldberg and Rao’s binary blocking flow algorithm [5]. In this paper we give an $O(n^{1.5}\log n)$ algorithm for the problem in planar graphs by combining preflow push and augmenting path algorithms.

In planar graphs, multiple source and sink flow problems have been studied by Miller and Naor, giving subquadratic-time algorithms for the case when the sources and sinks are on a common face and for the feasibility problem (the amount of flow out of every source and into every sink is known) [11]. The maximum single source, single sink flow in a directed planar graph can be found in $O(n\log n)$ time [1].

2 Definitions

We are given a directed planar graph $G$ with arc capacities a set of source vertices $S$ and sink vertices $T$. A flow is an assignment of values to arcs not exceeding the capacity such that the flow entering a non-source, non-sink vertex is equal to the flow leaving the same vertex. A flow is maximum if it maximizes the amount of flow leaving $S$ and (equivalently) there is no residual path from any vertex in $S$ to any vertex in $T$. An arc is residual if the flow is less than the capacity. The reverse of an arc is residual if there is flow on the arc. For more formal definitions, see [1].

A preflow is a flow that allows excess inflow at vertices that are not sink vertices. A maximum preflow is a preflow that maximizes the flow into the sinks. We will use the following lemma, which holds for general graphs. The forward direction follows from the definition of maximum preflows. The reverse direction follows from the Max Flow, Min Cut Theorem: if there are no residual paths from sources or vertices with excess inflow to sinks, then there is a saturated cut separating the sources from the sinks and the preflow cannot be increased (see [9]).

Lemma 2.1. A preflow is maximum if and only if there is no residual path from a source to a sink or from a vertex with excess inflow to a sink.

We also assume for this draft that the reader is familiar with recursive subdivisions of planar graphs. A piece is a subgraph resulting from this decomposition. See [3] for details.
3 Algorithm for a piece not containing sinks

Our multiple-source, single sink algorithm (Section 4) is a recursive algorithm using a recursive decomposition based on cycle separators. In this section, we present a solution to the non-trivial subproblem in which we need to find a maximum flow from a set of sources in a piece (of the recursive decomposition) to sinks outside the piece. More formally:

**Input:** a piece \( P \) with a constant number of holes, a set \( S_P \) of sources in \( P \), and a set \( T_P \) of sinks none of which are internal vertices of \( P \).

**Output:** The max flow from \( S_P \) to \( T_P \).

The algorithm proceeds in phases, mimicking a preflow algorithm in two pushes: one to the boundary of the piece (Phase 1) and one from the boundary to the sinks (Phase 2). In a final phase, the (maximum) preflow is converted to a maximum flow.

**Phase 1:** Find the max flow in \( P \) from \( S_P \setminus \partial P \) to \( \partial P \).

**Phase 2:** For each vertex \( p \) in \( \partial P \), if \( p \) is a source, compute a max \( pT_p \)-flow. Otherwise, push as much flow from \( p \) to \( T_P \) as possible while not exceeding the inflow to \( p \).

**Phase 3:** Convert the above maximum preflow into a maximum flow.

3.1 Correctness

We need the following lemma, which holds for general graphs.

**Lemma 3.1.** In a graph, let \( A, B \) be a partition of the vertices such that all the arcs from \( A \) to \( B \) are saturated and the sinks \( t \) are vertices of \( B \). Augmenting a preflow cannot make this cut residual.

**Proof.** By assumption, there cannot be an augmenting path from a vertex in \( A \) to a sink, so consider an augmenting path \( P \) from a vertex \( b \in B \) to \( t \). Assume for the sake of contradiction that there is a residual \( f \) that is a source, compute a max \( pT_p \)-flow. Otherwise, push as much flow from \( p \) to \( T_P \) as possible while not exceeding the inflow to \( p \).

**Phase 3:** Convert the above maximum preflow into a maximum flow.

3.2 Running time

We analyze the running time of Phases 2 and 3. Phase 1 will be computed recursively; the analysis of the recursive algorithm is in Section 4.

Phase 2 can be solved in \( O(|\partial P|n \log n) = O(\sqrt{|P|}n \log n) \) time by computing a maximum single-source, single-sink flow (in the entire graph) for each \( p \in \partial P \) and \( t \in T_P \). If \( p \) is a source then compute the max \( pt \)-flow algorithm for each \( t \in T_P \). If \( p \) is a non-source vertex with excess \( f \) from Phase 1, augment the graph with a source \( s \) connected to \( p \) by an arc with capacity \( f \) and compute the maximum \( st \) flow for each \( t \in T_P \). Each max flow computation takes \( O(n \log n) \) time in planar graphs. Hence, Phase 2 runs in \( O(\sqrt{|P|}n \log n) \) time.
Maximum preflows can be converted into maximum flows in sparse graphs in $O(n \log n)$ time (see Goldberg and Tarjan [6, 7]; the conversion is only made explicit in the earlier conference version): first eliminate cycles of flow ($O(n \log n)$ time) and then eliminate excess inflow from vertices by processing them in reverse topological order in the flow’s support graph ($O(n)$ time). In fact, our maximum preflow is already acyclic, since the flows computed in Phase 2 are acyclic; converting to a max flow will, in fact, only take linear time.

The total time required for Phases 2 and 3 is therefore $O(\sqrt{|P|}n \log n)$.

# 4 Multiple-source, single-sink algorithm

In this section, the overall max flow algorithm is described. Since we apply recursion, we need to consider a slightly more general problem than multiple source, single sink max flow in order to ensure that subproblems are of the same form: we need to find a max flow from a set of sources to at most a constant $t$ number of sinks.

Let us regard the entire graph as a piece $P$ with no boundary vertices. We define each sink to be a boundary vertex of $P$. Each such sink now defines a degenerate hole of $P$ and $P$ contains no sinks in its interior. By assumption, the number of sinks is bounded by $t$ so the number of holes of $P$ is at most $t$.

Next, we obtain a subdivision of $P$ into a constant number $p$ of subpieces; each subpiece has at most $c_p n$ vertices and $\sqrt{c_p n}$ boundary vertices (constant $c_p < 1$). This is done with the recursive $r$-division algorithm of Frederickson (see Lemmas 1 and 2 in [3]) but with Miller’s cycle separator theorem [10] instead of the separator theorem of Lipton and Tarjan [9]. Through the recursion, we ensure that the number of holes in each piece is bounded by $t - 1$. If we pick $t$ to be a sufficiently large constant, we can ensure this bound with the approach introduced by Fukcharoenphol and Rao [3].

We now run the algorithm of the previous section on each subpiece $P'$. To solve Phase 1 for $P'$, we define a new planar graph $P''$ from $P'$ by introducing a super sink for each hole and adding an edge of infinite capacity from each boundary vertex of that hole to the super sink. The same is done for the external face of $P'$ if it contains boundary vertices. Now, solving Phase 1 for $P'$ corresponds to finding a max flow from the sources in $P''$ to its super sinks. Since $P'$ has at most $t - 1$ holes, $P''$ has at most $t$ super sinks and so we can recurse on $P''$ to solve this problem (note that only the super sinks of $P''$ and not the sinks of $P$ belonging to $P''$ are regarded as sinks in the recursive call).

## 4.1 Running Time

As described in Section 3.2, Phases 2 and 3 run in $O(n^{3/2} \log n)$ time.

We define a recurrence relationship to bound the time for Phase 1 and the entire algorithm. For $n$ larger than some constant, the size of each piece in the division (including the super sinks added to the piece) is at most $c'_p n$, $c_p < c'_p < 1$.

We repeat the recursion until each piece has size at most some constant $r$. Let $N$ be the total size of all pieces at the leaves of the recursion tree. Using ideas of Frederickson [4] (see Lemmas 1 and 2 and their proofs in that paper for details), $N = kn$ for some constant $k$.

In the analysis, we assume that the total size of pieces at any recursion level is $N$ by regarding the vertices in pieces at the leaves of the recursion tree to be present in every level. This allows us to regard the pieces in any level as being pairwise vertex-disjoint by counting a vertex according to its multiplicity, once for each piece to which it belongs. Furthermore, we may assume that no new vertices (i.e. super sinks) are introduced in the recursive steps since all $N$ vertices (including all added super sinks) are present in any recursion level. Finally, we may assume that the size of a piece in a division is exactly $c'_p n$ since larger pieces will only increase running time.

From these assumptions, it follows that we may restrict our attention to the case $p = 1/c'_p$ and the recurrence relation for the running time of our algorithm becomes

$$T(N) \leq pT(N/p) + c' N^{3/2} \log N,$$

where $c'$ is a constant and the second term is the total time for Phases 2 and 3 and for computing the division of the graph. We will show that $T(N) \leq c N^{3/2} \log N$ for some constant $c$ (which we may assume is true for small $N$). For the induction step, we have

$$T(N) \leq pc(N/p)^{3/2} \log N + c' N^{3/2} \log N = (c/\sqrt{p} + c') N^{3/2} \log N.$$
The desired time bound is achieved by setting $c = c'/(1 - 1/\sqrt{p}) > 0$.

4.2 Correctness

Correctness of the overall algorithm follows from Theorem 4.1, that is, the algorithm is correct over all pieces because we can consider the sources in any order, and we always fully saturate the flow from the source we consider.

**Theorem 4.1.** Given an instance of a multiple source, multiple sink maximum flow problem, one can find the maximum flow by:

- Iterating over the sources $s$ in any order
- Iterating over the sinks $t$ in any order
- Saturating a maximum $st$-flow in the residual graph

**Proof.** We prove by induction over the inner and outer loops of the algorithm. Let $s_1, s_2, \ldots$ be the arbitrary order that the sources are iterated over.

**Outer loops:** Say we are currently considering source $s_j$. Assume that there are no residual paths from source $s_i$ for any $i < j$ to any sink. We will show that no residual $s_j$-to-$t$ path can, by way of augmentation, introduce an $s_i$-to-$t'$ residual path. By the inductive hypothesis, the Max Flow, Min Cut Theorem, and submodularity of cuts, there are non-residual non-crossing cuts $S_i, \bar{S}_i$ and $S'_i, \bar{S}'_i$ with $s_i \in S_i \cap \bar{S}'_i$ and $t \in \bar{S}_i \cap \bar{S}'_i$. Let $R$ be a residual $s_j$-to-$t'$ path. If $R$ could introduce a residual $s_i$-to-$t'$ path, then it must be that $s_j \in \bar{S}'_i$ and $t \in S_i$. By the submodularity of cuts, it must be that $S_i \subset S'_i$. It follows that $S_i, \bar{S}_i$ is a non-residual cut that also separates $s_i$ and $t'$ and augmenting $R$ cannot introduce a residual $s_i$-to-$t'$ path.

**Inner loops:** Suppose we are currently considering source $s_i$. Let $t_1, t_2, \ldots$ be the arbitrary order that the sinks are iterated over for source $s_i$. Suppose we are saturating the max $s_i t_j$ flow. By the above argument, doing so will not introduce a residual source-to-sink path for any source $s_k$, $k < \ell$. Assume that there are no residual $s_i$-to-$t_j$ paths for any $i < j$. By this inductive hypothesis and the Max Flow, Min Cut Theorem, there are non-residual cuts $S_i, \bar{S}_i$ such that $s_i \in S_i$ and $t_j \in \bar{S}_i$. If there is a non-zero $s_i$-to-$t_j$ flow in the residual graph, then it must be that $t_j \in S_i$ for every $i < j$. It follows that no $s_i$-to-$t_j$ flow can leave $S_i$.

Note that saturating an arbitrary order of the source, sink pairs will in general not result in a maximum flow, see Figure 1.

![Figure 1: A counterexample to arbitrarily saturating source sink pairs: the maximum $\{s, s'\}$-to-$\{t, t'\}$ flow has value 3. If the source, sink pairs are maximized in the order $(s, t), (s', t'), (s, t'), (s', t)$, the value of the flow found is only 2.](image)

5 Bounded Genus Graphs

Since our algorithm does not rely too much on planarity, it is easy to generalize it to bounded genus graphs. For such graphs, we can use the $O(n \log^2 n \log^2 C)$ time max $st$-flow algorithm due to Erickson, Chambers and Nayyeri (C is the sum of capacities). To begin the decomposition, we may first planarize the graph with a set of $g$ simple cycles, each of length $O(\sqrt{n})$ (e.g., construction by Hutchinson and Miller). This will allow us to continue with a decomposition based on planar separators and introduce only $O(g)$ holes in which we can embed a super sink while maintaining planarity as required for Phase 1. This gives an $O(n^{3/2} \log^2 n \log^2 C)$ time algorithm; of course, using the binary blocking flow algorithm would be asymptotically faster.
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