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EDGE-COLORING BIPARTITE MULTIGRAPHS IN $O(E \log D)$ TIME

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Let V, E, and D denote the cardinality of the vertex set, the cardinality of the edge set, and the maximum degree of a bipartite multigraph G. We show that a minimal edge-coloring of G can be computed in $O(E \log D)$ time. This result follows from an algorithm for finding a matching in a regular bipartite graph in $O(E)$ time.

1. Introduction

The edge-coloring problem is to find a minimal edge-coloring of a given multigraph. An *edge-coloring* of a multigraph $G = (V, E)$ with vertex set V and edge set E is a map $c : E \longrightarrow \mathbb{N}$ giving each edge $e \in E$ a *color* $c(e) \in \mathbb{N}$ such that no two adjacent edges have the same color. If edge-coloring c uses at most k colors, c is called a *k-edge-coloring*. A *color class* of an edge-coloring c of a multigraph G refers to the subset of edges having a certain color in c. An edge-coloring of a multigraph G is said to be *minimal* if there is no edge-coloring of G using fewer colors.

The number of colors used in a minimal edge-coloring of a multigraph G is called the *chromatic index* of G. It is well known that the chromatic index of a simple graph G is equal to D or $D + 1$, where D denotes the maximum degree of any vertex in G [19]. Upper bounds on the chromatic index of multigraphs can be found in [15]. In general it is an NP-complete problem to determine the chromatic index of a multigraph [8]. However,

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there are some classes of graphs for which the chromatic index always is the maximum degree. The most important is the class of bipartite multigraphs. Throughout this paper we restrict ourselves to this class.

The edge-coloring problem for bipartite multigraphs has applications in different areas. Many scheduling problems can be formulated in terms of edge-coloring bipartite multigraphs, for example the class-teacher time-table problem [7] and the file transfer problem [15]. Other applications of edge-coloring bipartite multigraphs are routing permutation networks [14], the k-k routing problem [18], and the simulation of a PRAM on a distributed memory machine [11]. Finally algorithms for edge-coloring bipartite multigraphs can be applied in matching theory.

The best running times previously known for computing a D-edge-coloring were $O(E \log D + V \log V \log D)$ [16] and $O(ED)$ [17]. In Section 4, we present an algorithm which computes a minimal edge-coloring of G in time $O(E \log D)$.

This paper is organized as follows. In Section 2 we briefly report on previous related work. In Section 3 we review some basic techniques. In Section 4 we give our coloring algorithm.

2. Previous work

The first algorithms solving the edge-coloring problem for bipartite graphs were derived from classic graph theory. Splitting G into D matchings using the matching algorithm of [9] combined with a construction of Mendelsohn and Dulmage [13] gives an algorithm running in time $O(E\sqrt{V}D)$. Coloring edge by edge using augmenting paths leads to an $O(EV)$ time algorithm [12]. In 1976 Gabow presented a solution based on a divide and conquer approach that takes $O(E\sqrt{V}\log D)$ time [4]. In 1978 Gabow and Kariv developed an algorithm for edge-coloring bipartite graphs with time complexity $O(\min\{E\sqrt{V\log V}, E(\log V)D, (E+V^2)\log D\})$ [5]. They also used a divide and conquer approach but they make the most effort in the conquer step while Gabow concentrated on the divide step. Furthermore, Gabow and Kariv took advantage of the fact that a graph with maximum degree a power of two can be edge-colored efficiently. Using this fact again in an algorithm which works by repeatedly enlarging a partial edge-coloring until all edges are colored they achieved a running time of $O(E(\log E)^2)$ in 1982 [6]. In the same year Cole and Hopcroft developed an algorithm which takes time $O(E \log D + V \log V (\log D)^3) = O(E \log E)$ [1]. The algorithm combined the idea of Gabow with a faster matching algorithm and the fact that a graph with maximum degree a power of two can be edge-colored efficiently. Im-

proving the matching algorithm led to an $O(E \log D + V \log V (\log D)^2)$ time solution in [2]. Recently, [17] found an $O(E \cdot D)$ time matching algorithm, which leads to an $O(E \cdot D)$ time edge coloring algorithm. In 1995, Cole, Ost and Schirra [16] showed that it sufficed to find one matching; the same result was also found by Kapoor and Rizzi in 1996 [10]. In combination with the matching algorithm in [2], this yielded an $O(E \log D + V \log V \log D)$ time algorithm for edge coloring.

3. Preliminaries

WLOG we assume that G is regular (to achieve this, low degree vertices are combined to give degree at least $D/2$ to all but at most one vertex; then some vertices may be added to one side of the graph to give the same number of vertices on each side; finally $O(E)$ edges are added to give every vertex degree exactly D; cf. [1]). Graphs in which every vertex has the same degree are said to be *regular*.

Nearly all previous algorithms for edge-coloring bipartite multigraphs use a divide and conquer approach: They partition G into subgraphs that are recursively colored with different color sets. To obtain a minimal edge-coloring it is crucial that the maximum degrees of the subgraphs sum up to D. Hence the edges of G have to be divided so that each of the resulting subgraphs is regular as well.

A *matching* in G is a subset M of E with the property that no two edges have a common endpoint. M is said to be *full* if it contains, as an endpoint, every vertex of G. Obviously each color class of a D-edge-coloring of G is a full matching in G.

Euler partitions are a helpful tool in forming regular decompositions. An *Euler partition* of a graph G is a partition of its edges into open and closed paths, such that each vertex of odd degree is at the end of exactly one open path, and each vertex of even degree is at the end of no open path. An *Euler split* of bipartite graph G splits G into two bipartite graphs $G_1 = (V, E_1)$ and $G_2 = (V, E_2)$ where E_1 and E_2 are formed by scanning the paths of an Euler partition of G and alternately placing one edge into E_1 and one edge into E_2. Any vertex of even degree in G will have the same degree in both G_1 and G_2, while any vertex of odd degree in G will have degrees in G_1 and G_2 differing by one. This implies that if G is regular and D is even, then all the vertices in G will have degree $D/2$ in each of G_1 and G_2 and this is the maximum degree of any vertex in G_1 or G_2. Hence the decomposition is regular.

It immediately follows that for a graph with $D = 2^i$, $i \in \mathbb{N}_0$, a minimal edge-coloring of G can be computed by splitting the graph recursively by Euler splits until all resulting subgraphs have degree one. This gives a decomposition into D matchings that can trivially be 1-colored. This algorithm runs in time $O(E \log D)$.

If G has an odd maximum degree a decomposition of G into G_1 and G_2 obtained by an Euler split is not necessarily regular. This follows from the fact that each vertex of maximum degree in G will have degree $\lceil D/2 \rceil$ in G_1 and degree $\lfloor D/2 \rfloor$ in G_2 or vice versa. Hence G_1 and G_2 might both have maximum degree $\lceil D/2 \rceil$. This means that in general using repeated Euler splits to color the edges of G yields a $(2D-2)$-edge-coloring only. Every split of a (sub)graph with an odd maximum degree might require an additional color since the sum of the maximum degrees of the resulting subgraphs might exceed the maximum degree of the split graph by one.

4. Computing a minimal edge-coloring

The basic approach of the algorithm in [1] is to attempt to follow the Euler split method even when D is not a power of 2. When D is even, using an Euler split, G is partitioned into two graphs each of degree $D/2$. The two graphs are colored recursively. When D is odd, a regular matching M is computed; it is colored and removed from G. The graph $G - M$ has even degree and it is colored using the method for even D.

Assuming the matching algorithm runs in time $O(E)$, as shown below, we have

Theorem 1. *The edge coloring algorithm runs in time $O(E \log D)$.*

Proof. Aside recursive calls, the main procedure uses $O(E)$ time, to possibly find a matching and to compute an Euler Partition. ∎

It remains to describe the matching algorithm. The basic approach used in [17] and [2] is one of weight redistribution. Each edge is given a non-negative integer weight, initially one. The weights are redistributed so that the weight incident on each vertex remains unchanged at D and the edge weights remain integers in the range $[0, D]$. The goal is to obtain one edge of weight D incident on each vertex, i.e. a matching. The method used is to repeatedly find a cycle C in the graph, then to alternately increment and decrement the weight of the edges on the cycle by Δ so as to cause at least one edge to have weight 0 or D while keeping all weights in the range $[0, D]$. Any edges with weight 0 or D are removed from the graph. This process

is continued until all edges have weight 0 or D. Using DFS, this process is readily implemented in time $O(L)$ for each length L cycle found.

The key observation in [17] is that by choosing the reweighing to go in the "right" direction, this takes time $O(E \cdot D)$ overall. Suppose an edge e_i has weight $wt(e_i) = w_i$. Edge e_i is given potential w_i^2. Suppose WLOG that on cycle C of length $2l$, $\sum_{i=1}^{l} w_{2i} \geq \sum_{i=1}^{l} w_{2i-1}$. Suppose the weight of the even index edges is incremented by one. Then the gain in potential is

$$\sum_{i=1}^{l}[(w_{2i}+1)^2 - w_{2i}^2] + \sum_{i=1}^{l}[(w_{2i-1}-1)^2 - w_{2i-1}^2] =$$
$$= \sum_{i=1}^{l}(2w_{2i}+1) - \sum_{i=1}^{l}(2w_{2i-1}-1) \geq 2l$$

Since the final potential is $VD^2/2$, the method runs in time $O(VD^2) = O(E \cdot D)$ time. We speed this up by a $\Theta(D)$ factor.

As in [1] our algorithm starts by reducing the number of edges to $O(V \log D)$ in $O(E)$ time. This is done by creating edge sets with edges of weight $0, 2, \cdots, 2^i, \cdots$, respectively, for $2^i \leq D$, where all the edges sets, except for that of edges of weight 0, have no cycles. To this end, the cycle finding and reweighting is applied to the weight one edges until no cycles remain among weight one edges; this takes time $O(E)$. In turn, the cycle finding is applied to the weight 2^i edges, for $i = 1, 2, \ldots$, respectively, for $2 \leq 2^i \leq D$, and this takes time $O(E/2^i)$.

For the remaining edge collection of size $O(V \log D)$ the goal is to process long cycles in sublinear time. To this end, edges are formed into paths, called *chains*, of length at most s edges, s a parameter to be specified. Chains have an implicit direction. The chains will be vertex disjoint. The algorithm will need to perform the following operations on chains, each in time $O(\log s)$.

- Concatenation.
- Split.
- Reverse.
- Find the total weight of the even index edges and of the odd index edges.
- Find a minimum weight edge of even (resp. odd) index and optionally delete it.
- Given a vertex, find the chain it belongs to, if any.

These operations are readily implemented using a balanced tree; indeed a splay tree, with just an amortized time bound, would suffice for our application.

A cycle will comprise an alternating sequence of chains and edges; further, all but at most two chains on the cycle will have length exactly s. To process and reweight the edges on a cycle, the cycle is traversed a chain at a time to find the weight of even index edges and the weight of odd index edges; the minimum weights of edges of odd and even index are also found (two weights in all). WLOG suppose the even index edges have lesser weight; their weight is decremented maximally. Each chain is split at each edge now having weight 0; this also separates any edges of weight D. Then the remainder of the cycle is removed from the path. The process of growing a cycle continues from the remainder of the path, which could be a single vertex.

A path is built by DFS. Whenever possible, the path will be extended by a chain rather than a single edge. If an edge is added which is incident on a vertex v in the interior of a chain, the chain is split at v and the larger portion is added to the path. The shorter portion edge with endpoint v is removed from the chain, and the remainder of this portion forms a new shorter chain. On the path being built, by means of concatentions and splits, the path is maintained as an alternation of chains of length s and singleton edges plus possibly a chain of length less than s at the end of the path.

Because of the vertex disjointness of chains, a cycle can be formed only by the addition of an edge and not a chain to the path. But this edge may be added to a chain of length less than s and may be incident on the interior of a chain. Thus a cycle may have up to two chains of length less than s, as claimed earlier.

To help detect a cycle, for each chain a bit is kept indicating if it is on the current path. Whenever an edge or chain is added to the path, the new final vertex on the path is tested to see if it belongs to a chain already on the path. We choose $s = D^2$. This yields:

Theorem 2. *The matching algorithm runs in time $O(E)$.*

Proof. It suffices to bound the time spent finding cycles. To this end, we bound the number of distinct chains created, including singleton edges, since for each chain just one join or split will occur, taking $O(\log s)$ time. First, we bound the number of chains created from finding cycles. A cycle of length L will yield at most

$$\lceil L/s \rceil + 1 + \text{ number of weight 0 edges deleted}$$

chains. Since at least one edge is deleted for each cycle found this is bounded by

$$L/s + 3 \cdot \text{ (number of weight 0 edges deleted).}$$

By the potential argument of [17] reviewed above, the total length of all the cycles is at most $VD^2/4$ (a cycle of length L yields an increase of at least $2L$ in potential). Summing over all cycles gives the following bound on the number of chains created by finding cycles: $VD^2/4s + 3V(\log D + 1) = O(V \log D)$, if $s \geq D^2$. But each such chain can be split only $O(\log s)$ times in creating cycles, giving a bound of $O(V \log D \log s)$ on the total number of chains created over the course of the algorithm, including the original $O(V \log D)$ edges. For $s = D^2$, this implies that the time to process the chains is bounded by $O(V \log^3 D) = O(E)$.

5. Further Comments

The overall $O(E \log D)$ running time of the algorithm arises from two components: the cost of finding Euler splits, and the cost of matching. While asymptotically the same, for moderate values of D the cost of matching appears to dominate. Variants of the algorithm with the same overall running time exist, in which just $O(\log D)$ matchings are found, with overall cost $O(E)$, while the number of Euler splits that are performed increases by a constant factor. In fact, the number of matchings being computed can be reduced to 1, but with a further increase in the number of Euler splits being performed. (Details can be found in [3].) For moderate values of D, "back of the envelope" estimates suggest the first variant will be the fastest.

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