Finding the Smallest H-Subgraph in Real Weighted Graphs and Related Problems

Virginia Vassilevska¹, Ryan Williams¹, and Raphael Yuster²

¹ Computer Science Department, Carnegie Mellon University, Pittsburgh, PA {virgi, ryanw}@cs.cmu.edu

Abstract. Let G be a graph with real weights assigned to the vertices (edges). The weight of a subgraph of G is the sum of the weights of its vertices (edges). The MIN H-SUBGRAPH problem is to find a minimum weight subgraph isomorphic to H, if one exists. Our main results are new algorithms for the MIN H-SUBGRAPH problem. The only operations we allow on real numbers are additions and comparisons. Our algorithms are based, in part, on fast matrix multiplication.

For vertex-weighted graphs with n vertices we obtain the following results. We present an $O(n^{t(\omega,h)})$ time algorithm for MIN H-SUBGRAPH in case H is a fixed graph with h vertices and $\omega < 2.376$ is the exponent of matrix multiplication. The value of $t(\omega,h)$ is determined by solving a small integer program. In particular, the smallest triangle can be found in $O(n^{2+1/(4-\omega)}) \leq o(n^{2.616})$ time, the smallest K_4 in $O(n^{\omega+1})$ time, the smallest K_7 in $O(n^{4+3/(4-\omega)})$ time. As h grows, $t(\omega,h)$ converges to $3h/(6-\omega) < 0.828h$. Interestingly, only for h=4,5,8 the running time of our algorithm essentially matches that of the (unweighted) H-subgraph detection problem. Already for triangles, our results improve upon the main result of [VW06]. Using rectangular matrix multiplication, the value of $t(\omega,h)$ can be improved; for example, the runtime for triangles becomes $O(n^{2.575})$. We also present an algorithm whose running time is a function of m, the number of edges. In particular, the smallest triangle can be found in $O(m^{(18-4\omega)/(13-3\omega)}) \leq o(m^{1.45})$ time.

For edge-weighted graphs we present an $O(m^{2-1/k}\log n)$ time algorithm that finds the smallest cycle of length 2k or 2k-1. This running time is identical, up to a logarithmic factor, to the running time of the algorithm of Alon et al. for the unweighted case. Using the color coding method and a recent algorithm of Chan for distance products, we obtain an $O(n^3/\log n)$ time randomized algorithm for finding the smallest cycle of any fixed length.

1 Introduction

Finding cliques or other types of subgraphs in a larger graph are classical problems in complexity theory and algorithmic combinatorics. Finding a maximum clique is NP-Hard, and also hard to approximate [Ha98]. This problem is also conjectured to be *not* fixed parameter tractable [DF95]. The problem of finding

² Department of Mathematics, University of Haifa, Haifa, Israel raphy@math.haifa.ac.il

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(induced) subgraphs on k vertices in an n-vertex graph has been studied extensively (see, e.g., [AYZ95, AYZ97, CN85, EG04, KKM00, NP85, YZ04]). All known algorithms for finding an induced subgraph on k vertices have running time $n^{\Theta(k)}$. Many of these algorithms use fast matrix multiplication to obtain improved exponents.

The main contribution of this paper is a set of improved algorithms for finding an (induced) k-vertex subgraph in a real vertex-weighted or edge-weighted graph. More formally, let G be a graph with real weights assigned to the vertices (edges). The weight of a subgraph of G is the sum of the weights of its vertices (edges). The MIN H-SUBGRAPH problem is to find an H-subgraph of minimum weight, if one exists. Some of our algorithms are based, in part, on fast matrix multiplication. In several cases, our algorithms use fast rectangular matrix multiplication algorithms. However, for simplicity reasons, we express most of our time bounds in terms of ω , the exponent of fast square matrix multiplications. The best bound currently available on ω is $\omega < 2.376$, obtained by Coppersmith and Winograd [CW90]. This is done by reducing each rectangular matrix product into a collection of smaller square matrix products. Slightly improved bounds can be obtained by using the best available rectangular matrix multiplication algorithms of Coppersmith [Cop97] and Huang and Pan [HP98]. In all of our algorithms we assume that the graphs are undirected, for simplicity. All of our results are applicable to directed graphs as well. Likewise, all of our results on the MIN-H-SUBGRAPH problem hold for the analogous MAX-H-SUBGRAPH problem. As usual, we use the addition-comparison model for handling real numbers. That is, real numbers are only allowed to be compared or added.

Our first algorithm applies to *vertex-weighted* graphs. In order to describe its complexity we need to define a small integer optimization problem. Let $h \geq 3$ be a positive integer. The function $t(\omega, h)$ is defined by the following optimization program.

Definition 1

$$b_1 = \max\{b \in N : \frac{b}{4-\omega} \le \lfloor \frac{h-b}{2} \rfloor\}. \tag{1}$$

$$s_1 = h - b_1 + \frac{b_1}{4 - \omega}. (2)$$

$$s_2(b) = \max\{h - b + \lfloor \frac{h - b}{2} \rfloor , h - (3 - \omega) \lfloor \frac{h - b}{2} \rfloor\}. \tag{3}$$

$$s_2 = \min\{s_2(b) : \lfloor \frac{h-b}{2} \rfloor \le b \le h-2\}. \tag{4}$$

$$t(\omega, h) = \min\{s_1, s_2\}. \tag{5}$$

By using fast rectangular matrix multiplication, an alternative definition for $t(\omega, h)$, resulting in slightly smaller values, can be obtained (note that if $\omega = 2$, as conjectured by many researchers, fast rectangular matrix multiplication has no advantage over fast square matrix multiplication).

Theorem 1. Let H be a fixed graph with h vertices. If G = (V, E) is a graph with n vertices, and $w : V \to \Re$ is a weight function, then an induced H-subgraph of G (if exists) of minimum weight can be found in $O(n^{t(\omega,h)})$ time.

It is easy to establish some small values of $t(\omega, h)$ directly. For h = 3 we have $t(\omega,3)=2+1/(4-\omega)<2.616$ by taking $b_1=1$ in (1). Using fast rectangular matrix multiplication this can be improved to 2.575. In particular, a triangle of minimum weight can be found in $o(n^{2.575})$ time. This should be compared to the $O(n^{\omega}) \leq o(n^{2.376})$ algorithm for detecting a triangle in an unweighted graph. For h = 4 we have $t(\omega, 4) = \omega + 1 < 3.376$ by taking b = 2 in (4). Interestingly, the fastest algorithm for detecting a K_4 , that uses square matrix multiplication, also runs in $O(n^{\omega+1})$ time [NP85]. The same phenomena also happens for h=5where $t(\omega, 5) = \omega + 2 < 4.376$ and for h = 8 where $t(\omega, 8) = 2\omega + 2 < 6.752$, but in no other cases! We also note that $t(\omega, 6) = 4 + 2/(4 - \omega)$, $t(\omega, 7) = 4 + 3/(4 - \omega)$, $t(\omega, 9) = 2\omega + 3$ and $t(\omega, 10) = 6 + 4/(4 - \omega)$. However, a closed formula for $t(\omega, h)$ cannot be given. Already for h = 11, and for infinitely many values thereafter, $t(\omega, h)$ is only piecewise linear in ω . For example, if $7/3 \le \omega < 2.376$ then $t(\omega, 11) = 3\omega + 2$, and if $2 \le \omega \le 7/3$ then $t(\omega, 11) = 6 + 5/(4 - \omega)$. Finally, it is easy to verify that both s_1 in (2) and s_2 in (4) converge to $3h/(6-\omega)$ as h increases. Thus, $t(\omega, h)$ converges to $3h/(6-\omega) < 0.828h$ as h increases.

Prior to a few months ago, the only known algorithm for MIN H-SUBGRAPH in the vertex-weighted case was the naïve $O(n^h)$ algorithm. Very recently, [VW06] gave an $O(n^{h\cdot \frac{\omega+3}{6}}) \leq o(n^{0.896h})$ randomized algorithm, for h divisible by 3. Our algorithms are deterministic, and uniformly improve upon theirs, for all values of h.¹

A slight modification in the algorithm of Theorem 1, without increasing its running time by more than a logarithmic factor, can also answer the decision problem: "is there an H-subgraph whose weight is in the interval $[w_1, w_2]$ where $w_1 \leq w_2$ are two given reals?" Another feature of Theorem 1 is that it makes a relatively small number of comparisons. For example, the smallest triangle can be found by the algorithm using only $O(m + n \log n)$ comparisons, where m is the number of edges of G.

Since Theorem 1 is stated for induced H-subgraphs, it obviously also applies to not-necessarily induced H-subgraphs. However, the latter problem can, in some cases, be solved faster. For example, we show that the $o(n^{2.616})$ time bound for finding the smallest triangle also holds if one searches for the smallest H-subgraph in case H is the complete bipartite graph $K_{2,k}$.

Several *H*-subgraph detection algorithms take advantage of the fact that G may be sparse. Improving a result of Itai and Rodeh [IR78], Alon, Yuster and Zwick obtained an algorithm for detecting a triangle, expressed in terms of m [AYZ97]. The running time of their algorithm is $O(m^{2\omega/(\omega+1)}) \leq o(m^{1.41})$. This is faster than the $O(n^{\omega})$ algorithm when $m = o(n^{(\omega+1)/2})$. The best known

¹ [VW06] also give a deterministic $O(B \cdot n^{(\omega+3)/2}) \le o(B \cdot n^{2.688})$ algorithm, where B is the number of bits needed to represent the (absolute) maximum weight. Note this algorithm is *not* strongly polynomial.

running times in terms of m for $H = K_k$ when $k \ge 4$ are given in [EG04]. Sparseness can also be used to obtain faster algorithms for the vertex-weighted MIN H-SUBGRAPH problem. The triangle algorithm of [VW06] extends to a randomized $O(m^{1.46})$ algorithm. We prove:

Theorem 2. If G = (V, E) is a graph with m edges and no isolated vertices, and $w : V \to \Re$ is a weight function, then a triangle of G with minimum weight (if exists) can be found in $O(m^{(18-4\omega)/(13-3\omega)}) \le o(m^{1.45})$ time.

We now turn to edge-weighted graphs. An $O(m^{2-1/\lceil k/2 \rceil})$ time algorithm for detecting the existence of a cycle of length k is given in [AYZ97]. A small improvement was obtained later in [YZ04]. However, the algorithms in both papers fail when applied to edge-weighted graphs. Using the color coding method, together with several additional ideas, we obtain a randomized $O(m^{2-1/\lceil k/2 \rceil})$ time algorithm in the edge-weighted case, and an $O(m^{2-1/\lceil k/2 \rceil} \log n)$ deterministic algorithm.

Theorem 3. Let $k \geq 3$ be a fixed integer. If G = (V, E) is a graph with m edges and no isolated vertices, and $w : E \to \Re$ is a weight function, then a minimum weight cycle of length k, if exists, can be found with high probability in $O(m^{2-1/\lceil k/2 \rceil})$ time, and deterministically in $O(m^{2-1/\lceil k/2 \rceil} \log n)$ time.

In a recent result of Chan [Ch05] it is shown that the distance product of two $n \times n$ matrices with real entries can be computed in $O(n^3/\log n)$ time (again, reals are only allowed to be compared or added). [VW06] showed how to reduce the MIN H-SUBGRAPH problem in edge-weighted graphs to the problem of computing a distance product. (The third author independently proved this as well.)

Theorem 4 ([VW06]). Let H be a fixed graph with h vertices. If G = (V, E) is a graph with n vertices, and $w : E \to \Re$ is a weight function, then an induced H-subgraph of G (if exists) of minimum weight can be found in $O(n^h/\log n)$ time.

We can strengthen the above result considerably, in the case where H is a cycle. For (not-necessarily induced) cycles of fixed length we can combine distance products with the color coding method and obtain:

Theorem 5. Let k be a fixed positive integer. If G = (V, E) is a graph with n vertices, and $w : E \to \Re$ is a weight function, a minimum weight cycle with k vertices (if exist) can be found, with high probability, in $O(n^3/\log n)$ time.

In fact, the proof of Theorem 5 shows that a minimum weight cycle with $k = o(\log \log n)$ vertices can be found in (randomized) sub-cubic time.

Finally, we consider the related problem of finding a certain chromatic *H*-subgraph in an edge-colored graph. We consider the two extremal chromatic cases. An *H*-subgraph of an edge-colored graph is called *rainbow* if all the edges have distinct colors. It is called *monochromatic* if all the edges have the same color. Many combinatorial problems are concerned with the existence of rainbow and/or monochromatic subgraphs.

We obtain a new algorithm that finds a rainbow H-subgraph, if it exists.

Theorem 6. Let H be a fixed graph with 3k + j vertices, $j \in \{0, 1, 2\}$. If G = (V, E) is a graph with n vertices, and $c : E \to C$ is an edge-coloring, then a rainbow H-subgraph of G (if exists) can be found in $O(n^{\omega k + j} \log n)$ time.

The running time in Theorem 6 matches, up to a logarithmic factor, the running time of the induced *H*-subgraph detection problem in (uncolored) graphs.

We obtain a new algorithm that finds a monochromatic H-subgraph, if it exists. For fixed H, the running time of our algorithm matches the running time of the (uncolored) H-subgraph detection problem, except for the case $H = K_3$.

Theorem 7. Let H be a fixed connected graph with 3k+j vertices, $j \in \{0,1,2\}$. If G = (V, E) is a graph with n vertices, and $c : E \to C$ is an edge-coloring, then a monochromatic H-subgraph of G (if exists) can be found in $O(n^{\omega k+j})$ time, unless $H = K_3$. A monochromatic triangle can be found in $O(n^{(3+\omega)/2}) \le o(n^{2.688})$ time.

Due to space limitation, the proofs of Theorems 6 and 7 will appear in the journal version of this paper.

The rest of this paper is organized as follows. In Section 2 we focus on vertex-weighted graphs, describe the algorithms proving Theorems 1 and 2, and some of their consequences. Section 3 considers edge-weighted graphs and contains the algorithms proving Theorems 3, 4 and 5. The final section contains some concluding remarks and open problems.

2 Minimal *H*-Subgraphs of Real Vertex-Weighted Graphs

In the proof of Theorem 1 it would be convenient to assume that $H = K_h$ is a clique on h vertices. The proof for all other induced subgraphs with h vertices is only slightly more cumbersome, but essentially the same.

Let G=(V,E) be a graph with real vertex weights, and assume $V=\{1,\ldots,n\}$. For two positive integers a,b, the adjacency system A(G,a,b) is the 0-1 matrix defined as follows. Let S_x be the set of all $\binom{n}{x}$ x-subsets of vertices. The weight w(U) of $U\in S_x$ is the sum of the weights of its elements. We sort the elements of S_x according to their weights. This requires $O(n^x\log n)$ time, assuming x is a constant. Thus, $S_x=\{U_{x,1},\ldots,U_{x,\binom{n}{x}}\}$ where $w(U_{x,i})\leq w(U_{x,i+1})$. The matrix A(G,a,b) has its rows indexed by S_a . More precisely, the j'th row is indexed by $U_{a,j}$. The columns are indexed by S_b where the j'th column is indexed by $U_{b,j}$. We put A(G,a,b)[U,U']=1 if and only if $U\cup U'$ induces a K_{a+b} in G (this implies that $U\cap U'=\emptyset$). Otherwise, A(G,a,b)[U,U']=0. Notice that the construction of A(G,a,b) requires $O(n^{a+b})$ time.

For positive integers a,b,c, so that a+b+c=h, consider the Boolean product $A(G,a,b,c)=A(G,a,b)\times A(G,b,c)$. For $U\in S_a$ and $U'\in S_c$ for which A(G,a,b,c)[U,U']=1, define their smallest witness $\delta(U,U')$ to be the smallest element $U''\in S_b$ for which A(G,a,b)[U,U'']=1 and also A(G,b,c)[U'',U']=1. For each $U\in S_a$ and $U'\in S_c$ with A(G,a,b,c)[U,U']=1 and with $U\cup U'$

inducing a K_{a+c} , if $U'' = \delta(U, U')$ then $U \cup U' \cup U''$ induces a K_h in G whose weight is the smallest of all the K_h copies of G that contain $U \cup U'$. This follows from the fact that S_b is sorted. Thus, by computing the smallest witnesses of all plausible pairs $U \in S_a$ and $U' \in S_c$ we can find a K_h in G with minimum weight, if it exists, or else determine that G does not have K_h as a subgraph.

Let $A=A_{n_1\times n_2}$ and $B=B_{n_2\times n_3}$ be two 0-1 matrices. The smallest witness matrix of AB is the matrix $W=W_{n_1\times n_3}$ defined as follows. W[i,j]=0 if (AB)[i,j]=0. Otherwise, W[i,j] is the smallest index k so that A[i,k]=B[k,j]=1. Let $f(n_1,n_2,n_3)$ be the time required to compute the smallest witness matrix of the product of an $n_1\times n_2$ matrix by an $n_2\times n_3$ matrix. Let $h\geq 3$ be a fixed positive integer. For all possible choices of positive integers a,b,c with a+b+c=h denote

$$f(h,n) = \min_{a+b+c=h} f(n^a, n^b, n^c).$$

Clearly, the time to sort S_b and to construct A(G, a, b) and A(G, b, c) is overwhelmed by $f(n^a, n^b, n^c)$. It follows from the above discussion that:

Lemma 1. Let $h \geq 3$ be a fixed positive integer and let G be a graph with n vertices, each having a real weight. A K_h -subgraph of G with minimum weight, if exists, can be found in O(f(h,n)) time. Furthermore, if $f(n^a, n^b, n^c) = f(h, n)$ then the number of comparisons needed to find a minimum weight K_h is $O(n^b \log n + z(G, a + c))$ where z(G, a + c) is the number of K_{a+c} in G.

In fact, if $b \ge 2$, the number of comparisons in Lemma 1 can be reduced to only $O(n^b + z(G, a + c))$. Sorting S_b reduces to sorting the sums $X + X + \ldots + X$ (X repeated b times) of an n-element set of reals X. Fredman showed in [Fr76a] that this can be achieved with only $O(n^b)$ comparisons.

A simple randomized algorithm for computing (not necessarily first) witnesses for Boolean matrix multiplication, in essentially the same time required to perform the product, is given by Seidel [Sei95]. His algorithm was derandomized by Alon and Naor [AN96]. However, computing the matrix of first witnesses seems to be a more difficult problem. Improving an earlier algorithm of Bender et al. [BFPSS05], Kowaluk and Lingas [KL05] show that $f(3,n) = O(n^{2+1/(4-\omega)}) \le o(n^{2.616})$. This already yields the case h=3 in Theorem 1. We will need to extend and generalize the method from [KL05] in order to obtain upper bounds for f(h,n). Our extension will enable us to answer more general queries such as "is there a K_h whose weight is within a given weight interval?"

Proof of Theorem 1. Let $h \geq 3$ be a fixed integer. Suppose a, b, c are three positive integers with a+b+c=h and suppose that $0 < \mu \leq b$ is a real parameter. For two 0-1 matrices $A = A_{n^a \times n^b}$ and $B = B_{n^b \times n^c}$ the μ -split of A and B is obtained by splitting the columns of A and the rows of B into consecutive parts of size $\lceil n^{\mu} \rceil$ or $\lfloor n^{\mu} \rfloor$ each. In the sequel we ignore floors and ceilings whenever it does not affect the asymptotic nature of our results. This defines a partition of A into $p = n^{b-\mu}$ rectangular matrices A_1, \ldots, A_p , each with n^a rows and n^{μ} columns, and a partition of B into p rectangular matrices B_1, \ldots, B_p , each with

 n^{μ} rows and n^{c} columns. Let $C_{i} = A_{i}B_{i}$ for $i = 1, \ldots, p$. Notice that each element of C_{i} is a nonnegative integer of value at most n^{μ} and that $AB = \sum_{i=1}^{p} C_{i}$. Given the C_{i} , the smallest witness matrix W of the product AB can be computed as follows. To determine W[i,j] we look for the smallest index r for which $C_{r}[i,j] \neq 0$. If no such r exists, then W[i,j] = 0. Otherwise, having found r, we now look for the smallest index r so that r so that r so that r index r is a smallest index r so that r index r

We now determine a choice of parameters a, b, c, μ so that the time to compute C_1, \ldots, C_p and the time to compute the first witnesses matrix W, is $O(n^{t(\omega,h)})$. By Lemma 1, this suffices in order to prove the theorem. We will only consider $\mu \leq \min\{a, b, c\}$. Taking larger values of μ results in worse running times. The rectangular product C_i can be computed by performing $O(n^{a-\mu}n^{c-\mu})$ products of square matrices of order n^{μ} . Thus, the time required to compute C_i is

$$O(n^{a-\mu}n^{c-\mu}n^{\omega\mu}) = O(n^{a+c+(\omega-2)\mu}).$$

Since there are p such products, and since each of the n^{a+c} witnesses can be computed in $O(p+n^{\mu})$ time, the overall running time is

$$O(pn^{a+c+(\omega-2)\mu} + n^{a+c}(p+n^{\mu})) = O(n^{h-(3-\omega)\mu} + n^{h-\mu} + n^{h-b+\mu})$$
$$= O(n^{h-(3-\omega)\mu} + n^{h-b+\mu}).$$
(6)

Optimizing on μ we get $\mu = b/(4-\omega)$. Thus, if, indeed, $b/(4-\omega) \leq \min\{a,c\}$ then the time needed to find W is $O(n^{h-b+b/(4-\omega)})$. Of course, we would like to take b as large as possible under these constraints. Let, therefore, b_1 be the largest integer b so that $b/(4-\omega) \leq \lfloor (h-b)/2 \rfloor$. For such a b_1 we can take $a = \lfloor (h-b_1)/2 \rfloor$ and $c = \lceil (h-b_1)/2 \rceil$ and, indeed, $\mu \leq \min\{a,c\}$. Thus, (6) gives that the running time to compute W is

$$O(n^{h-b_1+b_1/(4-\omega)}).$$

This justifies s_1 appearing in (2) in the definition of $t(\omega, h)$. There may be cases where we can do better, whenever $b/(4-\omega) > \min\{a, c\}$. We shall only consider the cases where $a = \mu = \lfloor (h-b)/2 \rfloor \le b$ (other cases result in worse running times). In this case $c = \lceil (h-b)/2 \rceil$ and, using (6), the running time is

$$O(n^{h-(3-\omega)\lfloor \frac{h-b}{2} \rfloor} + n^{h-b+\lfloor \frac{h-b}{2} \rfloor}).$$

This justifies s_2 appearing in (4) in the definition of $t(\omega, h)$. Since $t(\omega, h) = \min\{s_1, s_2\}$ we have proved that W can be computed in $O(n^{t(\omega,h)})$ time.

As can be seen from Lemma 1 and the remark following it, the number of comparisons that the algorithm performs is relatively small. For example, in the case h=3 we have a=b=c=1 and hence the number of comparisons is $O(n\log n+m)$. In all the three cases h=4,5,6 the value b=2 yields $t(\omega,h)$. Hence, the number of comparisons is $O(n^2)$ for h=4, $O(n^2+mn)$ for h=5 and $O(n^2+m^2)$ for h=6.

Suppose $w:\{1,\ldots,n^b\}\to\Re$ so that $w(k)\leq w(k+1)$. The use of the μ -split in the proof of Theorem 1 enables us to determine, for each i,j and for a real interval I(i,j), whether or not there exists an index k so that A[i,k]=B[k,j]=1 and $w(k)\in I(i,j)$. This is done by performing a binary search within the $p=n^{b-\mu}$ matrices C_i,\ldots,C_p . The running time in (6) only increases by a $\log n$ factor. We therefore obtain the following corollary.

Corollary 1. Let H be a fixed graph with h vertices, and let $I \subset \Re$. If G = (V, E) is a graph with n vertices, and $w : V \to \Re$ is a weight function, then, deciding whether G contains an induced H-subgraph with total weight in I can be done $O(n^{t(\omega,h)} \log n)$ time.

Proof of Theorem 2. We partition the vertex set V into two parts $V = X \cup Y$ according to a parameter Δ . The vertices in X have degree at most Δ . The vertices in Y have degree larger than Δ . Notice that $|Y| < 2m/\Delta$. In $O(m\Delta)$ time we can scan all triangles that contain a vertex from X. In particular, we can find a smallest triangle containing a vertex from X. By Theorem 1, a smallest triangle induced by Y can be found in $O((m/\Delta)^{t(\omega,3)}) = O((m/\Delta)^{2+1/(4-\omega)})$ time. Therefore, a smallest triangle in G can be found in

$$O\left(m\Delta + \left(\frac{m}{\Delta}\right)^{2+1/(4-\omega)}\right)$$

time. By choosing $\Delta = m^{(5-\omega)/(13-3\omega)}$ the result follows.

The results in Theorems 1 and 2 are useful not only for real vertex weights, but also when the weights are large integers. Consider, for example, the graph parameter $\beta(G, H)$, the H edge-covering number of G. We define $\beta(G, H) = 0$ if G has no H-subgraph. Otherwise, $\beta(G, H)$ is the maximum number of edges incident with an H-subgraph of G. To determine $\beta(G, K_k)$ we assign to each vertex a weight equal to its degree. We now use the algorithm of Theorem 1 to find the maximum weighted K_k . If the weight of the maximum weighted K_k is W, then $\beta(G, K_k) = W - {k \choose 2}$. In particular, $\beta(G, K_k)$ can be computed in $O(n^{t(\omega,k)})$ time.

Finally, we note that Theorems 1 and 2 apply also when the weight of an H-subgraph is not necessarily defined as the sum of the weights of its vertices. Suppose that the weight of a triangle (x,y,z) is defined by a function f(x,y,z) that is monotone in each variable separately. For example, we may consider f(x,y,z)=xyz, f(x,y,z)=xy+xz+yz etc. Assuming that f(x,y,z) can be computed in constant time given x,y,z, it is easy to modify Theorems 1 and 2 to find a triangle whose weight is minimal with respect to f in $O(n^{2+1/(4-\omega)})$ time and $O(m^{(18-4\omega)/(13-3\omega)})$ time, respectively.

We conclude this section with the following proposition.

Proposition 1. If G = (V, E) is a graph with n vertices, and $w : V \to \Re$ is a weight function, then a (not necessarily induced) minimum weight $K_{2,k}$ -subgraph can be found in $O(n^{2+1/(4-\omega)})$.

Proof. To find the smallest $K_{2,k}$ we simply need to find, for any two vertices i, j, the first k smallest weighted vertices v_1, \ldots, v_k so that each v_i is a common neighbor of i and j. As in Lemma 1, this reduces to finding the first k smallest witnesses of a 0-1 matrix product. A simple modification of the algorithm in Theorem 1 achieves this goal in the same running time (recall that k is fixed).

3 Minimal H-Subgraphs of Real Edge-Weighted Graphs

Given a vertex-colored graph G with n vertices, an H-subgraph of G is called colorful if each vertex of H has a distinct color. The color coding method presented in [AYZ95] is based upon two important facts. The first one is that, in many cases, finding a colorful H-subgraph is easier than finding an H-subgraph in an uncolored graph. The second one is that in a random vertex coloring with k colors, an H-subgraph with k vertices becomes colorful with probability $k!/k^k > e^{-k}$ and, furthermore, there is a derandomization technique that constructs a family of not too many colorings, so that each H-subgraph is colorful in at least one of the colorings. The derandomization technique, described in [AYZ95], constructs a family of colorings of size $O(\log n)$ whenever k is fixed.

By the color coding method, in order to prove Theorem 3, it suffices to prove that, given a coloring of the vertices of the graph with k colors, a colorful cycle of length k of minimum weight (if exists) can be found in $O(m^{2-1/\lceil k/2 \rceil})$ time.

Proof of Theorem 3. Assume that the vertices of G are colored with the colors $1,\ldots,k$. We first show that for each vertex u, a minimum weight colorful cycle of length k that passes through u can be found in O(m) time. For a permutation π of $1,\ldots,k$, we show that a minimum weight cycle of the form $u=v_1,v_2,\ldots,v_k$ in which the color of v_i is $\pi(i)$ can be found in O(m) time. Without loss of generality, assume π is the identity. For $j=2,\ldots,k$ let V_j be the set of vertices whose color is j so that there is a path from u to $v\in V_j$ colored consecutively by the colors $1,\ldots,j$. Let S(v) be the set of vertices of such a path with minimum possible weight. Denote this weight by w(v). Clearly, V_j can be created from V_{j-1} in O(m) time by examining the neighbors of each $v\in V_{j-1}$ colored with j. Now, let $w_u=\min_{v\in v_k}w(v)+w(v,u)$. Thus, w_u is the minimum weight of a cycle passing through u, of the desired form, and a cycle with this weight can be retrieved as well.

We prove the theorem when k is even. The odd case is similar. Let $\Delta = m^{2/k}$. There are at most $2m/\Delta = O(m^{1-2/k})$ vertices with degree at least Δ . For each vertex u with degree at least Δ we find a minimum weight colorful cycle of length k that passes through u. This can be done in $O(m^{2-2/k})$ time. It now suffices to find a minimum weight colorful cycle of length k in the subgraph G' of G induced by the vertices with maximum degree less than Δ . Consider a permutation π of $1, \ldots, k$. For a pair of vertices x, y, let S_1 be the set of all paths of length k/2 colored consecutively by $\pi(1), \ldots, \pi(k/2), \pi(k/2+1)$. There are at most $m\Delta^{k/2-1} = m^{2-2/k}$ such paths and they can be found using the greedy algorithm in $O(m^{2-2/k})$ time. Similarly, let S_2 be the set of all paths of length k/2 colored consecutively by $\pi(k/2+1), \ldots, \pi(k), \pi(1)$. If u, v are endpoints of at least one path in S_1 then let $f_1(\{u,v\})$ be the minimum weight of such a path. Similarly define $f_2(\{u,v\})$. We

can therefore find, in $O(m^{2-2/k})$ a pair u, v (if exists) so that $f_1(\{u, v\}) + f_2(\{u, v\})$ is minimized. By performing this procedure for each permutation, we find a minimum weight colorful cycle of length k in G'.

Let $A = A_{n_1 \times n_2}$ and $B = B_{n_2 \times n_3}$ be two matrices with entries in $\Re \cup \infty$. The distance product $C = A \star B$ is an $n_1 \times n_3$ matrix with $C[i,j] = \min_{k=1,\dots,n_2} A[i,k] + B[k,j]$. Clearly, C can be computed in $O(n_1n_2n_3)$ time in the addition-comparison model. However, Fredman showed in [Fr76] that the distance product of two square matrices of order n can be performed in $O(n^3(\log\log n/\log n)^{1/3})$ time. Following a sequence of improvements over Fredman's result, Chan gave an $O(n^3/\log n)$ time algorithm for distance products. By partitioning the matrices into blocks it is obvious that Chan's algorithm computes the distance product of an $n_1 \times n_2$ matrix and an $n_2 \times n_3$ matrix in $O(n_1n_2n_3/\log \min\{n_1,n_2,n_3\})$ time. Distance products can be used to solve the MIN H-SUBGRAPH problem in edge weighted graphs.

Proof of Theorem 4. We prove the theorem for $H = K_h$. The proof for other induced H-subgraphs is essentially the same. Partition h into a sum of three positive integers a + b + c = h. Let S_a be the set of all K_a -subgraphs of G. Notice that $|S_a| < n^a$ and that each $U \in S_a$ is an a-set. Similarly define S_b and S_c . We define A to be the matrix whose rows are indexed by S_a and whose columns are indexed by S_b . The entry A[U,U'] is defined to by ∞ if $U \cup U'$ does not induce a K_{a+b} . Otherwise, it is defined to be the sum of the weights of the edges induced by $U \cup U'$. We define B to be the matrix whose rows are indexed by S_b and whose columns are indexed by S_c . The entry A[U,U'] is defined to by ∞ if $U \cup U'$ does not induce a K_{b+c} . Otherwise, it is defined to be the sum of the weights of the edges induced by $U \cup U'$ with at least one endpoint in U'. Notice the difference in the definitions of A and B. Let $C = A \star B$. The time to compute C using Chan's algorithm is $O(n^h/\log n)$. Now, for each $U \in S_a$ and $U' \in S_c$ so that $U \cup U'$ induces a K_{a+c} , let w(U, U') be the sum of the weights of the edges with one endpoint in U and the other in U' plus the value of C[U, U']. If w(U, U') is finite then it is the weight of the smallest K_h that contains $U \cup U'$. Otherwise, no K_h contains $U \cup U'$.

The weighted DENSE k-SUBGRAPH problem (see, e.g., [FKP01]) is to find a k-vertex subgraph with maximum total edge weight. A simple modification of the algorithm of Theorem 4 solves this problem in $O(n^k/\log n)$ time. To our knowledge, this is the first non-trivial algorithm for this problem. Note that the maximum total weight of a k-subgraph can potentially be much larger than a k-clique's total weight.

Proof of Theorem 5. We use the color coding method, and an idea similar to Lemma 3.2 in [AYZ95]. Given a coloring of the vertices with k colors, it suffices to show how to find the smallest colorful path of length k-1 connecting any pair of vertices in $2^{O(k)}n^3/\log n$ time. It will be convenient to assume that k is a power of two, and use recursion. Let C_1 be a set of k/2 distinct colors, and let C_2 be the complementary set of colors. Let V_i be the set of vertices

colored by colors from C_i for i=1,2. Let G_i be the subgraph induced by V_i . Recursively find, for each pair of vertices in G_i , the minimum weight colorful path of length k/2-1. We record this information in matrices A_1, A_2 , where the rows and columns of A_i are indexed by V_i . Let B be the matrix whose rows are indexed by V_1 and whose columns are indexed by V_2 where B[u,v]=w(u,v). The distance product $D_{C_1,C_2}=(A_1\star B)\star A_2$ gives, for each pair of vertices of G, all shortest paths of length k-1 where the first k/2 vertices are colored by colors from C_1 and the last k/2 vertices are colored by colors from C_2 . By considering all $\binom{k}{k/2} < 2^k$ possible choices for (C_1,C_2) , and computing D_{C_1,C_2} for each choice, we can obtain an $n\times n$ matrix D where D[u,v] is the shortest colorful path of length k-1 between u and v. The number of distance products computed using this approach satisfies the recurrence $t(k) \leq 2^k t(k/2)$. Thus, the overall running time is $2^{O(k)}n^3/\log n$.

The proof of Theorem 5 shows that, as long as $k = o(\log \log n)$, a cycle with k vertices and minimum weight can be found, with high probability, in $o(n^3)$ time. The previous best known algorithm (to our knowledge) for finding a minimum weight cycle of length k, in real weighted graphs, has running time $O(k!n^32^k)$ [PV91].

4 Concluding Remarks and Open Problems

We presented several algorithms for MIN H-SUBGRAPH in both real vertex weighted or real edge weighted graphs, and results for the related problem of finding monochromatic or rainbow H-subgraphs in edge-colored graphs. It may be possible to improve upon the running times of some of our algorithms. More specifically, we raise the following open problems.

- (i) Can the exponent $t(\omega, 3)$ in Theorem 1 be improved? If so, this would immediately imply an improved algorithm for first witnesses.
- (ii) Can the logarithmic factor in Theorem 3 be eliminated? We know from [AYZ97] that this is the case in the unweighted version of the problem. Can the logarithmic factor in Theorem 6 be eliminated?
- (iii) Can monochromatic triangles be detected faster than the $O(n^{(3+\omega)/2})$ algorithm of Theorem 7? In particular, can they be detected in $O(n^{\omega})$ time?

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References

- [AN96] N. Alon and M. Naor. Derandomization, witnesses for Boolean matrix multiplication and construction of perfect hash functions. Algorithmica, 16:434–449, 1996.
- [AYZ95] N. Alon, R. Yuster, and U. Zwick. Color-coding. *Journal of the ACM*, 42:844–856, 1995.

- [AYZ97] N. Alon, R. Yuster, and U. Zwick. Finding and counting given length cycles. Algorithmica, 17:209–223, 1997.
- [BFPSS05] M. Bender, M. Farach-Colton, G. Pemmasani, S. Skiena, and P. Sumazin. Lowest common ancestors in trees and directed acyclic graphs. J. Algorithms, 57(2):75–94, 2005.
- [Ch05] T.M. Chan. All-Pairs Shortest Paths with Real Weights in $O(n^3/\log n)$ Time. In *Proc. of the 9th WADS*, Lecture Notes in Computer Science 3608, Springer (2005), 318–324.
- [CN85] N. Chiba and L. Nishizeki. Arboricity and subgraph listing algorithms. SIAM Journal on Computing, 14:210–223, 1985.
- [Cop97] D. Coppersmith. Rectangular matrix multiplication revisited. Journal of Complexity, 13:42–49, 1997.
- [CW90] D. Coppersmith and S. Winograd. Matrix multiplication via arithmetic progressions. J. Symbol. Comput., 9:251–280, 1990.
- [DF95] R.G. Downey and M.R. Fellows. Fixed-parameter tractability and completeness II. On completeness for W[1]. Theoret. Comput. Sci., 141(1-2):109-131, 1995.
- [EG04] F. Eisenbrand and F. Grandoni. On the complexity of fixed parameter clique and dominating set. *Theoret. Comput. Sci.*, 326(1-3):57–67, 2004.
- [FKP01] U. Feige, G. Kortsarz and D. Peleg. The Dense k-Subgraph Problem. Algorithmica, 29(3):410-421, 2001.
- [Fr76] M.L. Fredman. New bounds on the complexity of the shortest path problem. SIAM Journal on Computing, 5:49–60, 1976.
- [Fr76a] M.L. Fredman. How good is the information theory bound in sorting? Theoret. Comput. Sci., 1:355–361, 1976.
- [Ha98] J. Håstad. Clique is hard to approximate within $n^{1-\epsilon}$. Acta Math., $182(1):105-142,\ 1998.$
- [HP98] X. Huang and V.Y. Pan. Fast rectangular matrix multiplications and applications. *Journal of Complexity*, 14:257–299, 1998.
- [IR78] A. Itai and M. Rodeh. Finding a minimum circuit in a graph. SIAM Journal on Computing, 7:413–423, 1978.
- [KKM00] T. Kloks, D. Kratsch, and H. Müller. Finding and counting small induced subgraphs efficiently. *Inf. Process. Lett.*, 74(3-4):115–121, 2000.
- [KL05] M. Kowaluk and A. Lingas. LCA Queries in Directed Acyclic Graphs. In Proc. of the 32nd ICALP, Lecture Notes in Computer Science 3580, Springer (2005), 241–248.
- [NP85] J. Nešetřil and S. Poljak. On the complexity of the subgraph problem. Comment. Math. Univ. Carolin., 26(2):415–419, 1985.
- [PV91] J. Plehn and B. Voigt. Finding Minimally Weighted Subgraphs. In Proceedings of the 16th International Workshop on Graph-Theoretic Concepts in Computer Science (WG), Springer-Verlag, 1991.
- [Sei95] R. Seidel. On the All-Pairs-Shortest-Path Problem in Unweighted Undirected Graphs. J. Comput. Syst. Sci., 51(3):400-403, 1995.
- [VW06] V. Vassilevska and R. Williams. Finding a maximum weight triangle in $n^{3-\delta}$ time, with applications. In *Proceedings of the 38th Annual ACM Symposium on Theory of Computing (STOC)*, to appear.
- [YZ04] R. Yuster and U. Zwick. Detecting short directed cycles using rectangular matrix multiplication and dynamic programming. In *Proc. of the 15th ACM-SIAM Symposium on Discrete Algorithms (SODA)*, ACM/SIAM (2004), 247–253.