# **Linear-Time Counting Algorithms for Independent Sets in Chordal Graphs**

Yoshio Okamoto, Takeaki Uno, and Ryuhei Uehara

### Abstract

We study some counting and enumeration problems for chordal graphs, especially concerning independent sets. We first provide the following efficient algorithms for a chordal graph: (1) a linear-time algorithm for counting the number of independent sets; (2) a linear-time algorithm for counting the number of maximum independent sets; (3) a polynomial-time algorithm for counting the number of independent sets of a fixed size. With similar ideas, we show that enumeration (namely, listing) of the independent sets, the maximum independent sets, and the independent sets of a fixed size in a chordal graph can be done in constant amortized time per output. On the other hand, we prove that the following problems for a chordal graph are #P-complete: (1) counting the number of maximal independent sets. With similar ideas, we also show that finding a minimum weighted maximal independent set in a chordal graph is NP-hard, and even hard to approximate.

**Keywords:** chordal graph, counting, enumeration, independent set, NP-completeness, #P-completeness, polynomial time algorithm.

# 1 Introduction

How can we cope with computationally hard graph problems? There are several possible answers, and one of them is to utilize the special graph structures arising from a particular context. This has been motivating the study of special graph classes in algorithmic graph theory [3, 13]. This paper deals with counting and enumeration problems from this perspective. Recently, counting and enumeration of some specified sets in a graph have been widely investigated, e.g., in the data mining area. In general, however, from the graph-theoretic point of view, those problems are hard even if input graphs are quite restricted. For example, counting the number of independent sets in a planar bipartite graph of maximum degree 4 is #P-complete [21]. Therefore, we wonder what kind of graph structures makes counting and enumeration problems tractable.

In this paper, we consider chordal graphs. A *chordal graph* is a graph in which every cycle of length at least four has a chord. From the practical point of view, chordal graphs have numerous applications in, for example, sparse matrix computation (e.g., see Blair & Peyton [2]), relational databases [1], and computational biology [4]. Chordal graphs have been widely investigated, and they are sometimes called triangulated graphs, or rigid circuit graphs (see, e.g., Golumbic's book [13, Epilogue 2004]). A chordal graph has various characterizations; for example, a chordal graph is an intersection graph of subtrees of a tree, and a graph is chordal if and only if it admits a special vertex ordering, called perfect elimination ordering [3]. Also, the class of chordal graphs forms a wide subclass of perfect graphs [13].

It is known that many graph optimization problems can be solved in polynomial time for chordal graphs; to list a few of them, the maximum weighted clique problem, the maximum weighted independent set problem set problem [12], the minimum maximal independent set problem [8]. There are also parallel algorithms to solve some of these problems efficiently [14]. However, relatively fewer problems have been studied for enumeration and counting in chordal graphs; the only algorithms we are aware of are the enumeration algorithms for all maximal cliques [11], all maximal independent sets [7, 16], all minimum separators and minimal separators [5], and all perfect elimination orderings [6].

In this paper, we investigate the problems concerning the number of independent sets in a chordal graph. Table 1 lists the results of the paper. We first give the following efficient algorithms for a chordal graph; (1) a linear-time

<sup>\*</sup>Department of Information and Computer Sciences, Toyohashi University of Technology, Hibarigaoka 1-1, Toyohashi, Aichi 441-8580, Japan. E-mail: okamotoy@ics.tut.ac.jp

<sup>&</sup>lt;sup>†</sup>National Institute of Informatics, Hitotsubashi 2-1-2, Chiyoda-ku, Tokyo 101-8430, Japan. E-mail: uno@nii.jp

<sup>\$</sup>School of Information Science, JAIST, Asahidai 1-1, Nomi, Ishikawa 923-1292, Japan. E-mail: uehara@jaist.ac.jp

Table 1: Summary of the results. We denote the number of vertices and edges by n and m respectively. The running times for enumeration algorithms refer to amortized time per output.

Chordal graphs	Counting	[ref.]	Enumeration	[ref.]
independent sets	O(n+m)	[this paper]	O(1)	[this paper]
maximum independent sets	O(n+m)	[this paper]	<i>O</i> (1)	[this paper]
independent sets of size k	$O(k^2(n+m))$	[this paper]	<i>O</i> (1)	[this paper]
maximal independent sets	#P-complete	[this paper]	O(n+m)	[7, 16]
minimum maximal independent sets	#P-complete	[this paper]		

algorithm to count the number of independent sets, (2) a linear-time algorithm to count the number of maximum independent sets, and (3) a polynomial-time algorithm to count the number of independent sets of a given size. The running time of the third algorithm is linear when the size is constant. Note that in general counting the number of independent sets and the number of maximum independent sets in a graph is #P-complete [17], and counting the number of independent sets of size k in a graph is #W[1]-complete [9] (namely, intractable in a parameterized sense). Let us also note that the time complexity here refers to the arithmetic operations, not to the bit operations.

The basic idea of these efficient algorithms is to invoke a clique tree associated with a chordal graph and perform a bottom-up computation via dynamic programming on the clique tree. A clique tree is based on the characterization of a chordal graph as an intersection graph of subtrees of a tree. Since a clique tree can be constructed in linear time and the structure of clique tree is simple, this approach leads to simple and efficient algorithms for the problems above. However, a careful analysis is necessary to obtain the linear-time complexity.

Along the same idea, we can also enumerate all independent sets, all maximum independent sets, and all independent sets of constant size in a chordal graph in O(1) amortized time per output.

On the other hand, we show that the following counting problems are #P-complete: (1) counting the number of maximal independent sets in a chordal graph, and (2) counting the number of minimum maximal independent sets in a chordal graph. Using a modified reduction, we furthermore show that the problem to find a minimum weighted maximal independent set is NP-hard. We also show that the problem is even hard to approximate. More precisely speaking, there is no randomized polynomial-time approximation algorithm to find such a set within a factor of  $c \ln |V|$ , for some constant c, unless NP  $\subseteq$  ZTIME( $n^{O(\log \log n)}$ ). This is in contrast with a linear-time algorithm by Farber that finds a minimum weighted maximal independent set in a chordal graph when the weights are 0 or 1 [8].

The organization of the paper is as follows. Section 2 introduces the concept of a clique tree. In Section 3, we devise a linear-time algorithm for counting the number of independent sets, and in Section 4, we discuss how to count the maximum independent sets in linear time. In Section 5, we provide an efficient algorithm for counting the number of independent sets of each size simultaneously. In Section 6, we briefly describe how to apply our method for counting to enumeration, which leads to constant amortized time algorithms. In Section 7, we prove that counting the number of maximal independent sets and counting the number of minimum maximal independent sets are hard. In Section 8, we modify the reduction in Section 7 to show that it is hard to find a minimum weighted maximal independent set, and even hard to approximate.

Due to space limitation, some proofs are postponed to Appendix A.

# 2 Preliminaries

In this article, we assume that the reader has a moderate familiarity with graph theory. This section aims at fixing the notation and introducing a chordal graph and concepts around that. Let G = (V, E) be a graph, which we always assume to be simple and finite, and also we assume that graphs are connected without loss of generality. The *neighborhood* of a vertex v in a graph G = (V, E) is the set  $N_G(v) = \{u \in V \mid \{u, v\} \in E\}$ . For a vertex subset U of V, we denote by  $N_G(U)$  the set  $\{v \in V \mid v \in N(u) \text{ for some } u \in U\}$ . If no confusion can arise we will omit the subscript G. We denote the closed neighborhood  $N(v) \cup \{v\}$  by N[v]. A vertex set I is an *independent set* of G if any pari of vertices in I is not an edge of G, and a vertex set C is a *clique* if every pair of vertices in C is an edge of C. An independent set is *maximum* if it has the largest size among all independent sets. An independent set is *maximal* if none of its proper supersets is an independent set. An independent set is *minimum maximal* if it is maximal and has the smallest size among all maximal independent sets. A maximum clique, a maximal clique and a minimum maximal clique are defined analogously. An edge which joins two vertices of a cycle but is not

itself an edge of the cycle is a *chord* of the cycle. A graph is *chordal* if each cycle of length at least 4 has a chord. To a chordal graph G = (V, E), we associate a tree T, called a *clique tree* of G, satisfying the following two properties. (A) The nodes of T are the maximal cliques of G. (B) For every vertex V of G, the subgraph  $T_V$  of T induced by the maximal cliques containing V is a tree. (In the literature, the condition (A) is sometimes weakened as each node is a vertex subset of G.) It is well known that a graph is chordal if and only if it has a clique tree, and in such a case a clique tree can be constructed in linear time. Some details are explained in books [3, 19]. The following property is important in the running time analysis of our algorithms.

**Lemma 1** Let G = (V, E) be a chordal graph, and denote by  $\mathcal{K}$  the family of maximal cliques of G. Then, it holds that  $\sum_{K \in \mathcal{K}} |K| = O(|V| + |E|)$ .

# 3 Linear-Time Algorithm to Count the Independent Sets

In this section, we describe an algorithm for counting the number of independent sets in a chordal graph. The basic idea of our algorithm is to divide the input graph into subgraphs induced by subtrees of the clique tree. Any two of these subtrees share a vertex of a clique if they are disjoint in the clique tree. This property is very powerful for counting the number of independent sets since any independent set can include at most one vertex of a clique. We compute the number of independent sets including each vertex of the clique, or no vertex of the clique by using the recursions.

First, we introduce some notations and state some lemmas. Given a chordal graph G = (V, E), we construct a clique tree T of G. We now pick up any node in the clique tree T, regard the node as the root of T, and denote it by  $K_r$ . This is what we call a *rooted clique tree*. For a maximal clique K in a chordal graph G and a rooted clique tree T of G, a maximal clique K' in G is a *descendant* of K (with respect to T) if K' is a descendant of K in T. For convenience, we consider K itself a descendant of K as well, and when no confusion arises we omit saying "with respect to T." Let PRT(K) be the parent of K in T. For convenience, we define  $PRT(K_r)$  by  $\emptyset$ . We denote by T(K) the subtree of T rooted at the node corresponding to the maximal clique K. Let G(K) denote the subgraph of G induced by the vertices included in at least one node in T(K). Observe that G(K) is a chordal graph of which T(K) is a clique tree.

For a graph G, let IS(G) be the family of independent sets in G. For a vertex v, let IS(G, v) be the family of independent sets in G including v, i.e.,  $IS(G, v) := \{S \mid S \in IS(G), v \in S\}$ . For a vertex set U, let  $\overline{IS}(G, U)$  be the family of independent sets in G including no vertex of U, i.e.,  $\overline{IS}(G, U) := \{S \mid S \in IS(G), S \cap U = \emptyset\}$ .

**Lemma 2** Let G be a chordal graph and T be a rooted clique tree of G. Choose a maximal clique K of G, and let  $K_1, \ldots, K_\ell$  be the children of K in T. (If K is a leaf of the clique tree, we set  $\ell := 0$ .) Furthermore let  $v \in K$  and  $S \subseteq V(G(K))$ . Then,  $S \in IS(G(K), v)$  if and only if S is represented by the union of  $\{v\}$  and  $S_1, \ldots, S_\ell$  such that  $S_i \in IS(G(K_i), v)$  if V belongs to V, and V if V belongs to V if V if V belongs to V if V belongs to V if V if V if V belongs to V if V if

Proof. Assume that  $S \in IS(G(K), v)$ . Let  $S_i := S \cap G(K_i)$  for every  $i \in \{1, ..., \ell\}$ . Then, S includes the union of  $\{v\}$  and  $S_1, ..., S_\ell$ . Let us show the converse inclusion. Choose an arbitrary vertex  $x \in S$ . If x = v, then x is certainly included in the union of  $\{v\}$  and  $S_1, ..., S_\ell$ . Otherwise, we have  $x \in V(G(K)) \setminus K$ . Since  $V(G(K)) = K \cup \bigcup_{i=1}^{\ell} V(G(K_i))$ , the vertex x belongs to  $S_i$  for some  $i \in \{1, ..., \ell\}$ . Therefore, S is included in the union of  $\{v\}$  and  $S_1, ..., S_\ell$ . Now, we need to show that for every  $i \in \{1, ..., \ell\}$  the set  $S_i$  satisfies the property required in the lemma. Fix  $i \in \{1, ..., \ell\}$ . If v belongs to  $K_i$ , then  $S_i$  belongs to  $K_i$ , then  $K_i$  belongs to  $K_i$ . Thus the required property is satisfied. This completes the proof of the only-if part.

Next, we prove the if part. Assume that S is the union of  $\{v\}$  and  $S_1, \ldots, S_\ell$  satisfying that  $S_i \in IS(G(K_i), v)$  if  $v \in K_i$ , and  $S_i \in \overline{IS}(G(K_i), K \cap K_i)$  otherwise. When  $v \in K_i$ , since v is adjacent to all vertices of  $K \setminus \{v\}$ , every vertex in  $S_i \setminus \{v\}$  belongs to  $V(G(K_i)) \setminus K$ . When  $v \notin K_i$ , by the definition of  $\overline{IS}(G(K_i), K \cap K_i)$ , every vertex in  $S_i \setminus \{v\}$  belongs to  $V(G(K_i)) \setminus K$ . Therefore, for each  $i \in \{1, \ldots, \ell\}$  it holds that  $S_i \setminus \{v\} \subseteq V(G(K_i)) \setminus K$ . This implies that  $S \setminus \{v\} \subseteq V(G(K)) \setminus K$ . Now, we show that for every  $i, j \in \{1, \ldots, \ell\}$ ,  $i \neq j$ ,  $(S_i \setminus \{v\}) \cup (S_j \setminus \{v\})$  is independent. To show that, suppose not. Since  $S_i$  and  $S_j$  are independent, there must be an edge  $\{x, y\} \in E$  such that  $x \in S_i \setminus \{v\}$  and  $y \in S_j \setminus \{v\}$ . Since  $\{x, y\}$  is an edge of G, it is included in some maximal clique of G. Then, the definition of a clique tree implies that x or y belongs to K. Without loss of generality, assume that x belongs to K. (Remember that  $x \in S_i \setminus \{v\}$ .) If  $S_i \in IS(G(K_i), v)$ , then  $S_i \cap K \supseteq \{v, x\}$ . This is a contradiction to  $S_i$  being independent. If  $S_i \in \overline{IS}(G(K_i), K \cap K_i)$ , then  $S_i$  cannot contain any vertex of K, particularly x. This is also a contradiction. Thus

the claim is verified, and it implies that  $S \setminus \{v\}$  is an independent set of G(K). Together with the observation that no vertex of  $G(K_i) \setminus K$  is adjacent to v if  $v \notin K_i$ , this further implies that S is an independent set of G(K). Since  $v \in S$ , this shows that  $S \in \mathcal{IS}(G(K), v)$ .

To show the uniqueness, suppose that S is the union of  $\{v\}, S_1, \ldots, S_\ell$  and also the union of  $\{v\}, S_1', \ldots, S_\ell'$  such that there exists i with  $S_i \neq S_i'$ . Without loss of generality assume that  $S_i \setminus S_i' \neq \emptyset$ . Choose a vertex  $u \in S_i \setminus S_i'$ , where  $u \neq v$ . Then, there must exist  $j \neq i$  with  $u \in S_i'$ . Hence, there exists a node  $L \in T(K_i)$  such that  $u \in L$  and a node  $L' \in T(K_j)$  such that  $u \in L'$ . (Note that L and L' are maximal cliques of G.) Then, by Property (B) in the definition of a clique tree, the nodes on the path connecting L and L' in T contain u. In particular we have  $u \in K$ . Therefore, u and v belong to the clique K and at the same time they belong to the independent set S. This is a contradiction.

By a close inspection of the proof above, we can observe that for every  $i, j \in \{1, ..., \ell\}$ ,  $i \neq j$ , it holds that  $V(G(K_i)) \setminus K$  is disjoint from  $V(G(K_j)) \setminus K$ . This property gives a nice decomposition of the problem into several independent parts, and enables us to perform the dynamic programming on a clique tree.

By similar discussion as above, we obtain the following lemma.

**Lemma 3** Let *G* be a chordal graph and *T* be a rooted clique tree of *G*. Choose a maximal clique *K* of *G*, and let  $K_1, \ldots, K_\ell$  be the children of *K* in *T*. (If *K* is a leaf of the clique tree, we set  $\ell := 0$ .)

- 1. We have  $S \in \overline{IS}(G(K), K)$  if and only if S is the union of  $S_1, \ldots, S_l$  such that  $S_i \in \overline{IS}(G(K_i), K \cap K_i)$ . Furthermore, such a representation is unique.
- 2. For each  $i \in \{1, ..., \ell\}$ , we have  $S_i \in \overline{IS}(G(K_i), K \cap K_i)$  if and only if  $S_i$  belongs either to  $\overline{IS}(G(K_i), \nu)$  for some  $\nu \in K_i \setminus K$  or to  $\overline{IS}(G(K_i), K_i)$ . Furthermore,  $S_i$  belongs to exactly one of them.

From these lemmas, we have the following recursive equations for IS.

**Equations 1** Let G be a chordal graph and T be a rooted clique tree of G. For a maximal clique K of G which is not a leaf of the clique tree, let  $K_1, \ldots, K_\ell$  be the children of K in T. Furthermore, let  $v \in K$ . Then, the following identities hold. (We remind that  $\dot{\cup}$  means "disjoint union.")

$$IS(G(K)) = \overline{IS}(G(K), K) \dot{\cup} \bigcup_{v \in K} IS(G(K), v);$$

$$IS(G(K), v) = \{S \cup \{v\} \mid S = \bigcup_{i=1}^{\ell} S_i, S_i \in \left\{ \frac{IS(G(K_i), v)}{\overline{IS}(G(K_i), K \cap K_i)} & \text{if } v \in K_i \\ \overline{IS}(G(K), K) = \{S \mid S = \bigcup_{i=1}^{\ell} S_i, S_i \in \overline{IS}(G(K_i), K \cap K_i)\};$$

$$\overline{IS}(G(K_i), K \cap K_i) = \overline{IS}(G(K_i), K_i) \dot{\cup} \bigcup_{u \in K_i \setminus K} IS(G(K_i), u) & \text{for each } i \in \{1, \dots, \ell\}.$$

These equations lead us to the algorithm in Figure 1 to count the number of independent sets in a chordal graph. For a maximal clique K of a chordal graph G, we denote the set of children of K in a rooted clique tree of G by CHD(K).

**Theorem 4** The algorithm #IndSets outputs the number of independent sets in a chordal graph G = (V, E) in O(|V| + |E|) time.

Proof. From Equations 1, the algorithm correctly computes the number of independent sets in G. Let us consider the computation time t(K) taken by a call to #IndSetsIter(K). The overall running time of #IndSets is  $t(K_r) + O(|K_r|)$ . Steps 7 and 8 take O(t(K')) and O(|K'|) time for each  $K' \in CHD(K)$  respectively. Step 9 can be done in O(|CHD(K)|). Next, we analyze the computation time for Step 10. Since  $\left|\overline{IS}(G(K),K)\right| = \prod_{K' \in CHD(K)} \left|\overline{IS}(G(K'),K \cap K')\right|$ , we have that  $\prod_{K' \in CHD(K),v \in K'} \left|\overline{IS}(G(K'),v)\right| \times \prod_{K' \in CHD(K),v \in K'} \left|\overline{IS}(G(K'),K \cap K')\right| = \left|\overline{IS}(G(K),K)\right| \times \frac{\prod_{K' \in CHD(K),v \in K'} \left|\overline{IS}(G(K'),K \cap K')\right|}{\prod_{K' \in CHD(K),v \in K'} \left|\overline{IS}(G(K'),K \cap K')\right|}$ , and we use this equation in Step 10. Then |IS(G(K),v)| can be computed in  $O(|\{K' \in CHD(K) \mid v \in K'\}|)$  time, thus Step 10 can be done in  $O(\sum_{v \in K} |\{K' \in CHD(K) \mid v \in K'\}|)$  time. Therefore, the accumulated time taken by a call to #IndSetsIter( $K_r$ ) is  $\sum_{K' \in CHD(K_r)} (O(t(K')) + O(|K'|)) + O(|CHD(K_r)|) + O(\sum_{v \in K_r} |\{K' \in CHD(K_r) \mid v \in K'\}|)$ . By expanding t(K') inside the sum, we can see that this is at most  $O(\sum_{K \in K} (|K| + \sum_{v \in K} |\{K' \in CHD(K) \mid v \in K'\}|))$ , where K denotes the set of nodes

## Algorithm 1: #IndSets

```
Input : A chordal graph G = (V, E);

Output: The number of independent sets in G;

1 construct a rooted clique tree T of G with root K_r;

2 call #IndSetsIter(K_r);

3 return |\overline{IS}(G, K_r)| + \sum_{v \in K_r} |\overline{IS}(G(K_r), v)|.
```

## **Procedure** #IndSetsIter(*K*)

```
Input: A maximal clique K of the chordal graph G;

4 if K is a leaf of T then

5 | set |\overline{IS}(G(K), K)| := 0 and |IS(K, v)| := 1 for each v \in K;

6 else

7 | foreach child K' of K do call #IndSetsIter(K');

8 | foreach child K' of K do compute |\overline{IS}(G(K'), K \cap K')| by |\overline{IS}(G(K'), K')| + \sum_{u \in K' \setminus K} |IS(G(K'), u)|;

9 | compute |\overline{IS}(G(K), K)| by \prod_{K' \in CHD(K)} |\overline{IS}(G(K'), K \cap K')|;

10 | foreach v \in K do compute |IS(G(K), v)| by \prod_{K' \in CHD(K), v \in K'} |IS(G(K'), v)| \times \prod_{K' \in CHD(K), v \notin K'} |\overline{IS}(G(K'), K \cap K')|.
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Figure 1: Algorithm to count the number of independent sets in a chordal graph.

in the clique tree, i.e., the family of maximal cliques of G. By Lemma 1, we have  $\sum_{K \in \mathcal{K}} |K| = O(|V| + |E|)$ . Furthermore, it follows that  $\sum_{K \in \mathcal{K}} \sum_{v \in K} |\{K' \in CHD(K) \mid v \in K'\}\} = \sum_{v \in V} |\{K' \in \mathcal{K} \mid v \in K'\}\} = \sum_{K \in \mathcal{K}} |K| = O(|V| + |E|)$  again by Lemma 1. Hence, the overall running time is O(|V| + |E|).

# 4 Linear-Time Algorithm to Count the Maximum Independent Sets

In this section, we modify Algorithm #IndSets to count the number of maximum independent sets in a chordal graph. For a set family S, we denote by  $\max(S)$  the cardinality of a largest set in S, and  $\operatorname{argmax}(S)$  denotes the family of largest sets in S. For a graph G, let MIS(G) be the family of maximum independent sets in G. For a vertex v, let MIS(G,v) be the family of maximum independent sets in G including V, i.e.,  $MIS(G,v) := \{S \in MIS(G) \mid v \in S\}$ . For a vertex set U, let  $\overline{MIS}(G,U)$  be the family of maximum independent sets in G including no vertex of U, i.e.,  $\overline{MIS}(G,U) := \{S \in MIS(G) \mid S \cap U = \emptyset\}$ .

From lemmas stated in the previous section and Equations 1, we immediately have the following equations.

**Equations 2** With the same set-up as Equations 1, the following identities hold.

$$\begin{split} \mathcal{M}IS(G(K)) &= \operatorname{argmax}(\overline{\mathcal{M}IS}(G(K),K) \cup \bigcup_{v \in K} \mathcal{M}IS(G(K),v)); \\ \mathcal{M}IS(G(K),v) &= \operatorname{argmax}(\{S \mid S = \bigcup_{i=1}^{\ell} S_i, S_i \in \left\{ \begin{array}{l} \mathcal{M}IS(G(K_i),v) & \text{if } v \in K_i \\ \overline{\mathcal{M}IS}(G(K_i),K \cap K_i) & \text{otherwise} \end{array} \right\}\}); \\ \overline{\mathcal{M}IS}(G(K),K) &= \operatorname{argmax}(\{S \mid S = \bigcup_{i=1}^{\ell} S_i, S_i \in \overline{\mathcal{M}IS}(G(K_i),K \cap K_i)\}); \\ \overline{\mathcal{M}IS}(G(K_i),K \cap K_i) &= \operatorname{argmax}(\overline{\mathcal{M}IS}(G(K_i),K_i) \cup \bigcup_{u \in K_i \setminus K} \mathcal{M}IS(G(K_i),u)). \end{split}$$

Since the sets of each family on the left hand side have the same size in each equation, the cardinality of the set can be computed in the same order as Algorithm #IndSets. For example,  $\mathcal{MIS}(G(K))$  can be computed as follows.

- 1. Set N := 0 and  $M := \max(\overline{\mathcal{MIS}}(G(K), K) \cup \bigcup_{v \in K} \mathcal{MIS}(G(K), v));$
- 2. if the size of a member of  $\overline{\mathcal{MIS}}(G(K), K)$  is equal to M, then  $N := N + \left| \overline{\mathcal{MIS}}(G(K), K) \right|$ ;
- 3. for each  $v \in K$ , if the size of a member of  $\mathcal{MIS}(G(K), v)$  is equal to M, then  $N := N + |\mathcal{MIS}(G(K), v)|$ ;
- 4. output N.

In this way we have the following theorem.

**Theorem 5** The number of maximum independent sets in a chordal graph G = (V, E) can be computed in O(|V| + |V|)|E|) time.

#### 5 Efficient Algorithm to Count the Independent Sets of Size k

In this section, we modify Algorithm #IndSets to count the number of independent sets of size k. For a graph G and a number k, let IS(G;k) be the family of independent sets in G of size k. For a vertex v, let IS(G,v;k)be the family of independent sets in G of size k including v, i.e.,  $IS(G, v; k) := \{S \in IS(G; k) \mid v \in S\}$ . For a vertex set U, let  $\overline{IS}(G,U;k)$  be the family of independent sets in G of size k including no vertex of U, i.e.,  $\overline{IS}(G, U; k) = \{S \in IS(G; k) \mid S \cap U = \emptyset\}.$ 

From lemmas stated in Section 3 and Equations 1, we immediately obtain the following equations.

## **Equations 3**

$$IS(G(K);k) = \overline{IS}(G(K),K;k) \cup \bigcup_{v \in K} IS(G(K),v;k);$$

$$IS(G(K),v;k) = \{S \mid S = \bigcup_{i=1}^{\ell} S_i, |S| = k, S_i \in \left\{ \frac{IS(G(K_i),v)}{\overline{IS}(G(K_i),K \cap K_i)} \quad \text{if } v \in K_i \\ \overline{IS}(G(K),K;k) = \{S \mid S = \bigcup_{i=1}^{\ell} S_i, |S| = k, S_i \in \overline{IS}(G(K_i),K \cap K_i)\};$$

$$\overline{IS}(G(K_i),K \cap K_i;k) = \overline{IS}(G(K_i),K_i;k) \cup \bigcup_{u \in K_i \setminus K} IS(G(K_i),u;k).$$

In contrast to Equations 1, the second and third equations of Equations 3 do not give a straightforward way to compute |IS(G(K), v; k)| and  $|\overline{IS}(G(K), K; k)|$ , respectively, since we have to count the number of combinations of  $S_1, \ldots, S_\ell$  which generate an independent set of size k. To compute them, we use a more detailed algorithm.

Here we only explain a method to compute |IS(G(K), v; k)| since  $|\overline{IS}(G(K), K; k)|$  can be computed in a similar way. Fix an arbitrary vertex  $v \in K$ . Then, according to v, we give indices to the children of K such that  $K_1, \ldots, K_p$  include v and  $K_{p+1}, \ldots, K_\ell$  do not. For  $k' \leq k$  and  $\ell' \leq p$ , let  $\text{Num}(\ell'; k') := \{S \mid S = \bigcup_{i=1}^{\ell'} S_i, S_i \in K_1, \ldots, K_p \}$  $\mathcal{IS}(K_i, v), |S| = k'$ . For  $k' \le k$  and  $\ell' \ge p + 1$ , let  $\overline{\text{Num}}(\ell'; \underline{k'}) := \{S \mid S = \bigcup_{i=\ell'}^{\ell} S_i, S_i \in \mathcal{IS}(K_i, K_i \setminus K), |S| = k'\}$ . Then, it holds that  $|IS(G(K), v; k)| = \sum_{h=0}^{k} (|\text{Num}(p; h)| \times |\overline{\text{Num}}(p+1; k-h)|)$ . For each  $\ell'$  and k',  $|\text{Num}(\ell'; k')|$  can be computed in  $O(k \times p)$  time based on the following recursive equation:

$$\left| \mathrm{Num}(\ell';k') \right| = \left\{ \begin{array}{ll} \sum_{h=0}^{k'} \left| \mathrm{Num}(\ell'-1;h) \right| \times \left| \mathcal{IS}(G(K_{\ell'}),v;k'-h) \right| & \text{if } \ell' > 1, \\ \left| \mathcal{IS}(G(K_1),v;k') \right| & \text{otherwise.} \end{array} \right.$$

Similarly,  $|\overline{\text{Num}}(\ell';k')|$  can be computed in O(k') time. The computation of  $|\overline{\text{Num}}(\ell';k')|$  and  $|\overline{\text{Num}}(\ell';k')|$  for all combinations of  $\ell'$  and k' can be done in  $O(k^2|\mathsf{CHD}(K)|)$  time, thus we can count the number of independent sets of size k in a chordal graph in  $O(k^2|V|^2)$  time. In the following, we reduce the computation time by the same technique used in the previous sections.

Observe that  $\left|\overline{IS}(G(K), K; k')\right| = \sum_{h=0}^{k'} \left|\overline{\text{Num}}(p; h)\right| \times \left|\overline{\text{Num}}(p+1; k'-h)\right|$ , which gives  $\left|\overline{\text{Num}}(p+1; k')\right| \times \left|\overline{\text{Num}}(p+1; k'-h)\right|$  $\left|\overline{\mathrm{Num}}(p;0)\right| = \left|\overline{IS}(G(K),K;k')\right| - \sum_{h=1}^{k'} \left|\overline{\mathrm{Num}}(p;h)\right| \times \left|\overline{\mathrm{Num}}(p+1;k'-h)\right|$ . This implies that we can compute  $\left|\overline{\mathrm{Num}}(k';p+1)\right|$  from  $\left|\overline{IS}(G(K),K;h)\right|$  and  $\left|\overline{\mathrm{Num}}(p;h)\right|$  in the increasing order of k'. The computation time for this task is  $O(k \times p)$ .

In summary, we can compute |IS(G(K), v; k')| for all  $v \in K$  and k' $\{0, \dots, k\}$  in  $O(k^2 \sum_{v \in K} |\{K' \in CHD(K) \mid v \in K'\}|)$  time. Therefore, the total computation time over all iterations can be bounded in the same way as the above section, and we obtain the following theorem.

1. The number of independent sets of size k in a chordal graph G = (V, E) can be computed in  $O(k^2(|V| + |E|))$  time.

2. The numbers of independent sets of all sizes from 0 to |V| in a chordal graph G = (V, E) can be simultaneously computed in  $O(|V|^2(|V| + |E|))$  time.

#### **Enumeration** 6

Equations 1 in Section 3 directly give the following algorithm for enumerating the independent sets of a given chordal graph, in which each procedure corresponds to an equation of Equations 1.

## **Algorithm 3**: EnumlS(G)

```
Input: a chordal graph G = (V, E);
  Output: all independent sets in G;
1 construct a clique tree T of G with root K;
```

- 2 **foreach**  $u \in K$  **do** enumerate all independent sets in TS(G, u) by EnumlS2(K, u);
- 3 enumerate all independent sets in  $\overline{IS}(G, K)$  by EnumIS3(K).

## **Procedure** EnumIS2(K, u)

```
Input: A maximal clique K of G, a vertex u \in K;
4 if K has no child then
       output \{u\}; //output an independent set if the bottom level is reached
5
6 else
       foreach child K_i of K such that u \in K_i do enumerate all independent sets in IS(G(K_i), u) by EnumlS2(K_i, u);
7
       foreach child K_i of K such that u \notin K_i do enumerate all independent sets in \overline{IS}(G(K_i), K \cap K_i) by Enum|S4(K_i);
8
       output all independent sets in IS(G(K), u) by combining the independent sets in IS(G(K_i), u) and in
       IS(G(K_i), K \cap K_i) for all i, j;
```

## **Procedure** EnumlS3(K)

```
Input: A maximal clique K of G;
10 if K has no child then
        output \emptyset; //output an independent set if the bottom level is reached
11
12 else
        foreach child K_i of K do enumerate all independent sets in \overline{IS}(G(K_i), K \cap K_i) by EnumIS4(K_i);
13
        output all independent sets in \overline{IS}(G(K), K) by combining the independent sets in \overline{IS}(G(K_i), K \cap K_i);
14
```

## **Procedure** EnumlS4(K)

```
Input: A maximal clique K of G;
```

- 15 call EnumIS3(K);
- 16 foreach  $u \in K \setminus PRT(K)$  do enumerate all independent sets in IS(G(K), u) by EnumlS2(G(K), u);
- 17 **output** all independent sets in  $\overline{IS}(G(K), K \cap PRT(K))$  by combining the independent sets in IS(G(K), u);

From the lemmas and theorems in the previous sections, EnumIS(G) surely enumerates all independent sets in G. However, we cannot bound its time complexity by constant for each output. In the following, we present a slight modification to obtain a constant-time enumeration algorithm.

Let us consider the computation tree of this algorithm. A computation tree is a rooted-tree representation of a recursive structure, in which the vertices are recursive calls, and the edges connect two vertices if and only if one vertex recursively calls the other. We define an *iteration* of the algorithm by the operations done in a vertex of the computation tree. In other words, an iteration is the computation in some procedure P recursively called by another procedure, in which the computation in the recursive calls generated by P is excluded.

We first reduce the number of iterations by the following two modifications. (1) If an iteration I generated by an iteration  $I_p$  recursively calls just one iteration  $I_c$ , then we modify the algorithm so that  $I_p$  recursively calls  $I_c$ directly. (2) If an iteration I outputs just one independent set, then merge I and the iteration which recursively calls I into one.

For a given chordal graph G = (V, E) and a rooted clique tree of G, the number of possible inputs for each procedure is at most O(|E|), as in our counting algorithms. Thus, we can enumerate all of these cases in O(|E|)time, and keep the results of modifications (1) and (2) in the memory. It can be done as a preprocessing within O(|E|) time.

By these modifications, we can see that any iteration which is a leaf of the computation tree outputs at least two independent sets, thus the number of iterations is not greater than the number of independent sets in G. We can also see that if an iteration outputs just one independent set, then, the input clique must be a leaf of the clique tree. Hence, the size of the output independent set is at most one.

We next consider how to compute all combinations of independent sets in, for example, Step 9 of the algorithm. In the procedures, the independent sets for K are generated by combining the independent recursive calls for several maximal cliques, say  $K_1$  and  $K_2$ . This step can be implemented as follows. First, we compute an independent set  $I_1$  for  $K_1$ , and for this  $I_1$ , we compute all independent sets  $I_2$  for  $K_2$ , and output  $I_1 \cup I_2$ . Next we compute another independent set for  $K_1$ , and compute all independent sets  $I_2$  for  $K_2$ , and output  $I_1 \cup I_2$ , then compute yet another independent set for  $K_1$ , and so on. Then the computation time in one iteration is proportional to (the number of recursive calls generated) times (the maximum number of vertices added to the current independent set). Because of modification (2), any iteration adds at most one vertex to the current independent set. Therefore, the total time complexity of the algorithm is linear in the number of independent sets.

**Theorem 7** All independent sets in a chordal graph can be enumerated in constant time for each on average with additional O(|V| + |E|) time for preprocessing.

Similar algorithms can be developed to enumerate the maximum independent sets and the independent sets of size k. However, some iterations may add to the current independent set several vertices not bounded by a constant. Since there are at most |E| kinds of inputs for each procedure, we can enumerate all such sets of vertices that will be added in an iteration, and put an identical name to each set of vertices in short time. By adding the name instead of adding vertices in a vertex set, we can execute the addition in constant time. Thus, the maximum independent sets and the independent sets of size k can be enumerated in constant time for each on average with additional  $O((|V| + |E|)|V|^2)$  time for preprocessing.

# 7 Hardness of Counting the Maximal Independent Sets

In this section, we show the hardness results for counting the number of maximal independent sets in a chordal graph. Although finding a maximal independent set is easy even in a general graph, we show that the counting version of the problem is actually hard.

Theorem 8 Counting the number of maximal independent sets in a chordal graph is #P-complete.

The proof is based on a reduction from the counting problem of the number of set covers. Let X be a finite set, and  $S \subseteq 2^X$  be a family of subsets of X. A *set cover* of X is a subfamily  $\mathcal{F} \subseteq S$  such that  $\bigcup \mathcal{F} = X$ . Counting the number of set covers is #P-complete [17].

Proof of Theorem 8. The membership in #P is immediate. To show the #P-hardness, we use a polynomial-time reduction of the problem for counting the number of set covers to our problem.

Let X be a finite set and  $S \subseteq 2^X$  be a family of subsets of X, and consider them as an instance of the set cover problem. Let us put  $S := \{S_1, \ldots, S_t\}$ . From X and S, we construct a chordal graph G = (V, E) in the following way.

We set  $V := X \cup S \cup S'$ , where  $S' := \{S'_1, \dots, S'_t\}$ . Namely, S' is a copy of S. Now, we draw edges. There are three kinds of edges. (1) We connect every pair of vertices in X by an edge. (2) For every  $S \in S$ , we connect  $x \in X$  and S by an edge if and only if  $x \in S$ . (3) For every  $S \in S$ , we connect S and S' (a copy of S) by an edge. Formally speaking, we define  $E := \{\{x,y\} \mid x,y \in X\} \cup \{\{x,S\} \mid x \in X, S \in S, x \in S\} \cup \{\{S,S'\} \mid S \in S\}$ . This completes our construction. Note that this construction can be done in polynomial time.

First, let us check that the constructed graph G is indeed chordal. Let C be a cycle of length at least four in G. Since the degree of a vertex in S' is one, they do not take part in any cycle of G. So forget them. Since S is an independent set of G, vertices in S cannot appear along C in a consecutive manner. Then, since the length of C is at least four, there have to be at least two vertices of X which appear in C not consecutively. Then, these two vertices give a chord since X is a clique of G. Hence, G is chordal.

Now, we look at the relation between the set covers of X and the maximal independent sets of G. Let U be a maximal independent set of G. We distinguish two cases.

Case 1. Consider the case in which U contains a vertex  $x \in X$ . Since X is a clique of G, U cannot contain any other vertices of X. Let  $G_x := G \setminus N_G[x]$ . (Remember that  $N_G[x]$  is the closed neighborhood of x, i.e., the set of vertices adjacent to x in G and x itself.) By the construction, we have that  $V(G_x) = \{S \in S \mid x \notin S\} \cup S'$  and  $E(G_x) = \{\{S, S'\} \mid S \in S, x \notin S\}$ . Then, a vertex  $S' \in S'$  such that  $x \in S$  is an isolated vertex of  $G_x$ . Therefore, this vertex must belong to U by the maximality of U. For each  $S \in S$  such that  $x \notin S$ , U must contain either S or S', but not both. This means that the number of maximal independent sets containing x is exactly  $2^{|\{S \in S\}|x\notin S'\}|}$ .

Case 2. Consider the case in which U contains no vertex of X. Then, for each  $S \in S$ , due to the maximality, U must contain either S or S'. Furthermore,  $U \cap S$  has to be a set cover of X (otherwise an element of X not covered

by  $U \cap S$  could be included in U). Hence, the number of maximal independent sets containing no vertex of X is equal to the number of set covers of X.

To summarize, we obtained that the number of maximal independent sets of G is equal to the number of set covers of X plus  $\sum_{x \in X} 2^{|\{S \in S\}|x \notin S\}|}$ . Since the last sum can be computed in polynomial time, this concludes the reduction.

As a variation, let us consider the problem for counting the minimum maximal independent sets in a chordal graph. Note that a minimum maximal independent set in a chordal graph can be found in polynomial time [8]. In contrast to that, the counting version is hard.

**Theorem 9** Counting the minimum maximal independent sets in a chordal graph is #P-complete.

Proof. We use the same reduction as in the proof of Theorem 8. Look at the case distinction in that proof again. The maximal independent sets arising from Case 1 have |S| + 1 elements, while the maximal independent sets from Case 2 have |S| elements. Therefore, the minimum maximal independent sets of the graph G constructed in that proof are exactly the maximal independent sets arising from Case 2, which precisely correspond to the set covers of X.

# 8 Hardness of Finding a Minimum Weighted Maximal Independent Set

In this section, we consider an optimization problem to find a minimum weighted maximal independent set in a chordal graph. Namely, given a chordal graph G and a weight for each vertex, we are asked to find a maximal independent set of G with minimum weight. Here, the weight of a vertex subset is the sum of the weights of its vertices.

Notice that there is a linear-time algorithm for this problem when the weight of each vertex is zero or one [8]. On the contrary, we show that the problem is actually hard when the weight is arbitrary.

Theorem 10 Finding a minimum weighted maximal independent set in a chordal graph is NP-hard.

The proof is similar to what we saw in the previous section. We use the optimization version of the set cover problem, namely the minimum set cover problem. It is known that the minimum set cover problem is NP-hard.

Proof of Theorem 10. For a given instance of the minimum set cover problem, we use the same construction of a graph G as in the proof of Theorem 8. We define a weight function w as follows: w(x) := 2|S| + 1 for every  $x \in X$ ; w(S) := 2 for every  $S \in S$ ; w(S') := 1 for every  $S' \in S'$ . This completes the construction.

Now, observe that S is a maximal independent set of the constructed graph G, and the weight of S is 2|S|. Therefore, no element of X takes part in any minimum weighted maximal independent set of G. Then, from the discussion in the proof of Theorem 8, if M is a maximal independent set of G satisfying  $M \cap X = \emptyset$ , then  $M \cap S$  is a set cover of X. The weight of M is  $|M \cap S| + |S|$ . Therefore, if M is a minimum weighted independent set of G, then M minimizes  $|M \cap S|$ , which is the size of a set cover. Hence,  $M \cap S$  is a minimum set cover. This concludes the reduction.

We can further show the hardness to get an approximation algorithm running in polynomial time. The precise statement is as follows.

**Theorem 11** There is no randomized polynomial-time algorithm for the minimum weight maximal independent set problem in a chordal graph with approximation ratio  $c \ln |V|$ , for some fixed constant c, unless NP  $\subseteq$  ZTIME $(n^{O(\log \log n)})$ .

Remark that  $\mathsf{ZTIME}(t)$  is the class of languages which have a randomized algorithm running in expected time t with zero error.

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# **A Omitted Proofs**

Proof of Lemma 1. To prove Lemma 1, we use more notions. Given a graph G = (V, E), a vertex  $v \in V$  is *simplicial* in G if  $N_G(v)$  is a clique in G. An ordering  $v_1, \ldots, v_n$  of the vertices of V is a *perfect elimination ordering* of G if  $v_i$  is simplicial in the subgraph induced by  $\{v_i, v_{i+1}, \ldots, v_n\}$  for all  $i = 1, \ldots, n$ . It is known that a graph is chordal if and only if it has a perfect elimination ordering [3, Section 1.2].

Take any perfect elimination ordering  $v_1, v_2, \ldots, v_n$  of G. Let  $C(v_i) := N_G[v_i] \cap \{v_i, v_{i+1}, \ldots, v_n\}$ . It is known that for every maximal clique K of G there exists a vertex  $v_i \in V$  such that  $K = C(v_i)$  holds [11]. Since  $C(v_i) \subseteq N_G[v_i]$ , we have  $|C(v_i)| \le |N_G[v_i]| = 1 + \deg_G(v_i)$ . Putting together, we obtain  $\sum_{K \in \mathcal{K}} |K| \le \sum_{v \in V} |C(v)| \le \sum_{v \in V} (1 + \deg_G(v)) = |V| + 2|E| = O(|V| + |E|)$ .

Proof of Lemma 3. 1. Similar to Lemma 2. We omit.

2. First, assume that  $S_i \in \mathcal{IS}(G(K_i), v)$  for some  $v \in K_i \setminus K$ . Since  $K_i$  is a clique,  $S_i$  cannot include any vertex of  $K_i \setminus \{v\}$ , particularly of  $K \cap K_i$ . Therefore,  $S_i \in \overline{\mathcal{IS}}(G(K_i), K \cap K_i)$ . Secondly, assume that  $S_i \in \overline{\mathcal{IS}}(G(K_i), K_i)$ . Then,  $S_i$  includes no vertex of  $K_i \cap K$ , since  $K_i \cap K \subseteq K_i$ . Hence,  $S_i \in \overline{\mathcal{IS}}(G(K_i), K \cap K_i)$ . This proves the if part.

Let us prove the only-if part and the uniqueness. Assume that  $S_i$  belongs to  $\overline{IS}(G(K_i), K_i \cap K)$ . When  $S_i$  includes a vertex v of  $K_i \setminus K$ , we have  $S_i \in \overline{IS}(G(K_i), v)$ . Note that v is a unique element in  $S_i \cap (K_i \setminus K)$  since  $S_i$  is an independent set and  $K_i \setminus K$  is a clique. Therefore,  $S_i \notin \overline{IS}(G(K_i), u)$  for  $u \in (K_i \setminus K) \setminus \{v\}$ . When  $S_i$  includes no vertex of  $K_i \setminus K$ , it follows that  $S_i \in \overline{IS}(G(K_i), K_i)$ .

Proof of Theorem 11. We use the following variant of the minimum set cover problem in which the intersection of every two sets has at most one element. We call the problem the *minimum set cover problem with intersection 1*.

Anil Kumar, Arya & Ramesh [15] showed that the minimum set cover problem with intersection 1 cannot be approximated by any randomized polynomial-time algorithm with approximation ratio  $c' \ln |X|$ , for some constant c', unless  $NP \subseteq ZTIME(n^{O(\log \log n)})$ . We use this fact in our proof.

Before proving the theorem, we need a lemma which bounds the size of S in an instance of the minimum set cover problem with intersection 1. This is an easy special case of a theorem by Frankl & Wilson [10], which is well known in extremal combinatorics. Hence we are not going to prove it.

**Lemma 12** Let *X* be a finite set and  $S \subseteq 2^X$  be a family of subsets of *X* such that  $|A \cap B| \le 1$  for every  $A, B \in S$  and  $\emptyset \notin S$ . Then,  $|S| \le {|X|+1 \choose 2}$ .

Now we are ready to prove Theorem 11.

Suppose that there exists a randomized polynomial-time algorithm A with approximation ratio  $c \ln |V|$  for the minimum weighted maximal independent set problem in a chordal graph. (The constant c will be determined later.) We use the algorithm A to get a polynomial-time algorithm with approximation ratio  $c' \ln |X|$  for the minimum set cover problem with intersection 1. Then, this will imply that  $NP \subseteq ZTIME(n^{O(\log \log n)})$ .

Let X be a finite set and  $S \subseteq 2^X$  be a nonempty family of subsets of X such that  $|A \cap B| \le 1$  for every  $A, B \in S$  and  $\emptyset \notin S$ . (We can assume  $\emptyset \notin S$  without loss of generality for an instance of the minimum set cover problem with intersection 1 because the empty set covers no element.) From them, we construct a graph G exactly in the same way as in the proof of Theorem 8. Setting  $\alpha := \lceil c \ln(|X|(|X|+2)) \rceil |S|$ , we define a weight w as follows:  $w(x) := 2\alpha^2 + 1$  for every  $x \in X$ ;  $w(S) := 2\alpha$  for every  $S \in S$ ; w(S') := 1 for every  $S' \in S'$ . This is our construction. (Note that this construction can be done in polynomial time.)

Denote by OPT an arbitrary (fixed) minimum weighted maximal independent set of G, by APX an output of the algorithm A for G, and by w(OPT) and w(APX) the weights of them. Since the number of vertices in G is |X|+2|S|, which is at most  $|X|+2\binom{|X|+1}{2}=|X|(|X|+2)$  by Lemma 12, it follows that  $w(APX) \le c \ln(|X|(|X|+2))w(OPT)$ .

As in the proof of Theorem 10, S is a maximal independent set of G and its weight is  $2\alpha |S|$ . Therefore, it holds that  $w(OPT) \le 2\alpha |S|$ .

Now, suppose that there exists an element  $x \in X$  which is contained in APX. Then,  $w(APX) \ge w(x) = 2\alpha^2 + 1$ . This implies that  $2\alpha^2 < w(APX) \le c \ln(|X|(|X|+2))w(OPT) \le \lceil c \ln(|X|(|X|+2))\rceil \times 2\alpha |S| = 2\alpha^2$ . This is a contradiction. Thus, no element  $x \in X$  belongs to APX. This means that APX  $\cap S$  is a set cover of X. Let  $C := APX \cap S$  and we show that C approximates the optimal value for the minimum set cover problem within a factor of  $c' \ln |X|$ .

Again, by the same argument as in the proof of Theorem 10, we get  $w(APX) = (2\alpha - 1)|C| + |S|$ . Let  $C^*$  be a minimum set cover of X. Then, similarly we get  $w(OPT) = (2\alpha - 1)|C^*| + |S|$ . Since  $w(APX) \le c \ln(|X|(|X| + 2))w(OPT)$ , it follows that  $(2\alpha - 1)|C| + |S| \le c \ln(|X|(|X| + 2))((2\alpha - 1)|C^*| + |S|) = c \ln(|X|(|X| + 2))((2\alpha - 1)|C^*| + |\alpha|)$ .

Hence, we obtain  $|C| \le c \ln(|X|(|X|+2))|C^*| + \frac{\alpha-|S|}{2\alpha-1} \le c \ln(|X|(|X|+2))|C^*| + \frac{\alpha-\frac{1}{2}}{2\alpha-1} = c \ln(|X|(|X|+2))|C^*| + \frac{1}{2} \le c \ln(|X|(|X|+2))|C^*| + \frac{1}{2} \ln(|X|(|X|+2))|C^*| + \frac{1}{2} \ln(|X|(|X|+2))|C^*| \le (c+\frac{1}{2}) \ln(|X|^3)|C^*| = ((3c+\frac{3}{2}) \ln|X|)|C^*|.$  Setting  $c = \frac{c'}{3} - \frac{1}{2}$  gives an approximation ratio of  $c' \ln |X|$ .