Degree-Preserving Trees

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We consider the degree-preserving spanning tree (DPST) problem: Given a connected graph G, find a spanning tree T of G such that as many vertices of Tas possible have the same degree in T as in G. This problem is a graph-theoretical translation of a problem arising in the system-theoretical context of identifiability in networks, a concept which has applications in, for example, water distribution networks and electrical networks. We show that the DPST problem is NP-complete, even when restricted to split graphs or bipartite planar graphs, but that it can be solved in polynomial time for graphs with a bounded asteroidal number and for graphs with a bounded treewidth. For the class of interval graphs, we give a linear time algorithm. For the class of cocomparability graphs, we give an $O(n^4)$ algorithm. Furthermore, we present linear time approximation algorithms for planar graphs of a worst-case performance ratio of 1 - ϵ for every ϵ > 0. © 2000 John Wiley & Sons, Inc.

Keywords: graphs; algorithms; complexity; spanning tree; degree condition; planar graphs; cocomparability graphs; interval graphs; treewidth; asteroidal number; AT-free graphs; NPcompleteness; approximation

1. THE PRACTICAL NICHE

Analysis of communication or distribution networks is often concerned with finding spanning trees (or forests) of those networks fulfilling certain criteria. Also, in other contexts, spanning trees show up as important tools in modeling and analyzing problems. Therefore, a myriad of problems on spanning trees has been studied in the literature (see, e.g., [11, 12, 17, 18]). This paper deals with a virtually unexplored problem concerning spanning trees which we call the degree-preserving spanning tree (DPST) problem: Given a connected graph G, find a spanning tree T of G with a maximum number of degreepreserving vertices, that is, with a maximum number of vertices having the same degree in T as in G.

Some closely related questions were studied by Lewinter et al. [1, 7, 24, 25] from a purely theoretical point of view. They published a number of short notes on the subject, but we have not found any paper in which the DPST problem was studied extensively.

Our attention was initially turned to this problem through a practical application in water-distribution networks (see [26]), which makes the DPST problem a nice example of theory and practice going hand-in-hand. We briefly describe the application:

Suppose that we have to determine (or control) all flows in a water-distribution network by installing and using a small number of flowmeters and/or pressure gauges. The network can be regarded as an undirected connected graph G and the flow through each edge of G is described by an orientation of that edge and a nonnegative flow value. Since the sum of all flow values of edges entering a vertex is always the same as the sum of all flow values of edges leaving that vertex, except for possible sources and sinks, it is not difficult to derive all flows in the network from the flows through all edges of a cotree C of G (i.e., C is obtained from G by removing the edges of a spanning tree). Hence, it would suffice to install flowmeters at the edges of C. However,

Received October 19, 1998; accepted March 2, 1999

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the costs of installing a flowmeter are much higher than those of installing a water-pressure gauge at some vertex. Alternatively, we can derive the flow through an edge from the water-pressure drop between the two incident vertices. If we only use pressure gauges, and want to minimize the costs, the problem becomes that of finding a cotree whose edges are incident with a minimum number of vertices (in order to minimize the number of pressure gauges that have to be installed) or, equivalently, of finding a spanning tree T whose complement in G has as many isolated vertices as possible, that is, T has a maximum number of degree-preserving vertices. Recently, Rahal [28] independently discovered the cotree approach in his investigation of a steady-state formulation for water-distribution networks.

Our problem of determining all flows in the network with minimal costs of measuring (installing pressure gauges) is a so-called identifiability problem (see Walter [30]). The concrete water-distribution network that we considered has 80 vertices and 98 edges, making it a very sparse network. Our network is planar and it has outerplanarity 2. Especially, this latter fact enables us to solve the DPST problem in our case by a linear time algorithm; see Section 4.

Although applications of DPSTs so far have been in the area of water-distribution networks, this approach generalizes immediately to all networks in which Kirchhoff's laws [which (in an electrical context) state that the sum of currents in a (nonsource, nonsink) vertex is zero and that the sum of voltages over a cycle is zero] are valid and a bijective relation exists between the flow variable and the effort variable. The most obvious example of such a system is an electrical network, with the current I being the flow variable, the voltage V being the effort variable, and the bijection being Ohm's law $V = I \cdot R$, but the model can be applied to a wide variety of domains such as mechanics (flow = velocity, effort = force), thermodynamics (f = heat flow, e = temperature), acoustics (f = acoustics volume flow rate, e = acoustic pressure), and, in our case, the hydraulics of the water-distribution network (f = hydraulic volume flow rate, e = hydraulic pressure).

In a slightly more abstract sense, the DPST problem can be thought of as finding a tree of a network in which as many vertices as possible remain "undamaged." This idea may have applications in, for instance, constraint satisfaction problems where a minimal blocking set of constraints needs to be found or a system that needs to be made free from feedback (i.e., cycles) without damaging too many vertices.

Lewinter [24] introduced the concept of degreepreserving spanning trees and he proved that the number of degree-preserving vertices interpolates on the set of spanning trees of a given connected graph G. In other words, if spanning trees exist with k and l degreepreserving vertices, respectively, and k < l, then there exists a spanning tree with exactly *m* degree-preserving vertices for every *m* with k < m < l. He later generalized the degree-preserving concept to that of deficiency [1, 7]: A vertex is *k*-deficient if its degrees in the graph and the spanning tree differ by exactly *k*, and a spanning tree is *k*-deficient if the maximum deficiency of its vertices is *k*.

The rest of the paper is organized as follows: We start with some terminology and preliminary results in Section 2. Section 3 deals with the complexity of the DPST problem. It is shown that the problem is NP-complete in general and that it remains NP-complete even when restricted to split graphs or bipartite planar graphs. In Section 4, it is shown how the DPST problem can be reformulated in Monadic Second Order Logic, thereby proving that the problem is solvable in linear time for graphs with a bounded treewidth. Especially, this case is very interesting from a practical point of view, since water-distribution networks (and other supply networks, such as telephone, data, electricity, and, more recently, ISDN networks) tend to be of a "treelike structure," simply due to the high costs involved in installing and maintaining such networks. (Indeed, these costs are in most situations the bottleneck for the structure of the network to be installed.) This requirement of a treelike structure directly imposes the study of the problem for graphs with a relatively small treewidth. In Section 5, we apply an idea of Baker [3] to establish linear time approximation algorithms for the DPST problem when restricted to planar graphs.

In the rest of the paper, we consider the DPST problem when the input graphs are restricted to some other classes of graphs. In Section 6, we present a linear time algorithm for interval graphs. In Section 7, we obtain an $O(n^4)$ algorithm for cocomparability graphs using the fact that the maximum degree-preserving tree in a 2edge connected cocomparability graph corresponds to a set of vertices inducing a disjoint union of paths. In Section 8, we show that the DPST problem can be solved by a polynomial time algorithm for graphs with bounded asteroidal number.

We recently became aware of a paper by Bathia et al., to appear in [4], who independently obtained similar results, in particular on NP-completeness, approximation for planar graphs, and the linear time solution for graphs of bounded treewidth. Furthermore, they gave an almost linear time general approximation algorithm with an $O(\sqrt{n})$ approximation ratio. This is (in a sense) the best possible, since for planar graphs, we show that it is NP-complete for planar graphs, in general, to have an approximation within a factor $n^{1/2-\epsilon}$ for any given $\epsilon > 0$ (see Section 3).

2. PRELIMINARIES

In this paper, a graph is a pair G = (V, E), where V

is a finite set (the *vertices* of *G*) and $E \subseteq V \times V$ is a set of two-element (unordered) subsets of *V* (the *edges* of *G*). We write $v \in e$ if a vertex $v \in V$ is *incident* with an edge $e \in E$.

Throughout, let G = (V, E) be a graph and let n = |V|and m = |E|. For a nonempty subset $S \subseteq V$, we use G[S]to denote the subgraph of *G* induced by the vertices of *S*. For a subset $S \subseteq V$, we also write G - S for $G[V \setminus S]$, and for a vertex *x* of *G*, we write G - x instead of $G - \{x\}$.

For a vertex x of G, we use $N_G(x)$ to denote the set of neighbors of x in G, and we write $N_G[x] = \{x\} \cup N_G(x)$ for the closed neighborhood of x in G; the *degree* of x in G is $d_G(x) = |N_G(x)|$. A *pendant vertex* or *leaf* of G is a vertex with degree one in G. We omit the subscript G from the above expressions if it is clear which graph G we consider. For a graph G = (V, E) and $W \subseteq V$, we define $N[W] = \bigcup_{w \in W} N[w]$ and $N(W) = N[W] \setminus W$.

For a graph G let $Comp(G) = \{C : G[C] \text{ is a component of } G\}$. A cut vertex is a vertex x of G such that G - x has more components than has G. A *block* of G is a maximal subgraph of G without cut vertices, that is, a vertex, an edge, or a maximal 2-connected subgraph.

Definition 1. A nonempty subset $S \subseteq V$ is *realizable* if there exists a spanning forest T of G such that the degree of every vertex $x \in S$ is preserved in T [i.e., if $d_T(x) = d_G(x)$ for every vertex $x \in S$]. If T is such a spanning forest, then we call T an *S*-preserving forest. If, moreover, T is chosen in such a way that |S| is maximum, then we call T a maximum degree-preserving forest, and |S|, the degree-preserving number (of T or G). The DPST problem is the problem to find for a given graph G a maximum degree-preserving spanning forest.

Remark 1. For historic reasons, we call the problem *the degree-preserving spanning tree* problem. In case the graph is not connected, we content ourselves with a spanning forest.

As an example, the degree-preserving number of a tree, a unicyclic graph, and a complete graph ($\neq K_2$, i.e., a single edge) on *n* vertices are, respectively, n, n - 2, and 1.

Notice that to solve the DPST problem it is sufficient to compute a maximum (cardinality) realizable set S, since, given S, an S-preserving spanning forest, is then easy to find. By p(G), we denote the cardinality of a maximum realizable set in G. If a graph G is disconnected, then a maximum realizable set of G is simply the union of maximum realizable sets of all components of G. If a connected graph G (or a component of G) has a bridge e, then to compute a maximum realizable sets S_1 and S_2 for both components. Let T_1 be an S_1 -preserving forest and T_2 be an S_2 -preserving forest. Adding e as an edge between T_1 and T_2 gives a forest T which is $S_1 \cup S_2$ -preserving, and $S_1 \cup S_2$ is a maximum realizable set in G.

Henceforth, in the rest of this paper, we will assume that all graphs are 2-edge connected.

Let W be a set of vertices of a graph G. By G[[W]], we denote the graph with vertex set N[W] containing all edges of G incident with a vertex in W. Take note of the following simple observations:

Lemma 2. Let *S* be a nonempty set of vertices of a graph G = (V, E). Then, *S* is a realizable set of *G* if and only if G[[S]] is a forest.

Proof. If S is realizable, then every edge of G[[S]] must be an edge of T for every S-preserving forest T of G. Conversely, if G[[S]] is a forest, then this forest is clearly S-preserving.

Remark 2. Clearly, the above lemma implies that for any realizable set S of G, G[S] is a forest.

Corollary 3. Let S be a realizable set of vertices of G with $x \in S$. Then, $G[N[x] \cap S]$ is a star [isomorphic to $K_{1,k}$ for some $k \leq d_G(x)$, with center x].

Lemma 4. Let *S* be a realizable set with $x \in S$, and let $y \neq x$ be a vertex with $N(y) \subseteq N[x]$. Then, $\{y\} \cup S \setminus \{x\}$ is realizable.

Proof. Let (V, F) be an *S*-preserving spanning forest of G = (V, E). Then, for $F' = \{\{y, z\} : z \in N_G(y)\} \cup F \setminus \{\{x, z\} : z \in N_G(y)\}$, the graph (V, F') is a (y + (S - x))-preserving spanning forest of G.

3. HARDNESS RESULTS

A graph G = (V, E) is called a *split graph* if V can be partitioned into an independent set I and a clique C of G. Such a split graph is also denoted by G = (I, C, E). It is easy to see that split graphs are *chordal graphs*, that is, graphs that do not contain a chordless cycle of length greater than three.

Theorem 5. For a given split graph H and a given integer k, it is NP-complete to decide whether H contains a realizable set of cardinality k.

Proof. The reduction is from the NP-complete graph problem INDEPENDENT SET. Let G = (V, E) be a graph. We define a split graph H with independent set V and clique $E \times \{1, 2\}$ as follows: A pair $\{v, (e, i)\}$ is an edge of H if and only if $v \in V, e \in E, i \in \{1, 2\}$ and $v \in e$. It is easy to see that a set $W \subseteq V$ is an independent set of G if and only if W is a realizable set in H. Moreover, if G has no isolated vertices (i.e., vertices with degree zero), then for every realizable set W of H with |W| > 1, we have $W \subseteq V$.

A graph G = (V, E) is called *bipartite* if V can be partitioned into two independent sets X and Y of G. Such a bipartite graph is also denoted by G = (X, Y, E). **Theorem 6.** For a given bipartite planar graph B of maximum degree six and a given integer k, it is NP-complete to decide whether B contains a realizable set of cardinality k.

Proof. The reduction is from the INDEPENDENT SET problem restricted to cubic (i.e., 3-regular) planar graphs [18]. Let (G, k) be an instance of this NP-complete problem, where G = (V, E) with |E| = m. We define a bipartite graph $B = (V \cup (E \times \{2, 4, 6, 8\}), E \times \{1, 3, 5, 7\}, F_1 \cup F_2)$, where

$$F_1 = \{\{v, (e, i)\} : v \in V, e \in E, i \in \{1, 5\}, v \in e\}$$

 $F_2 = \{\{(e, 1), (e, 2)\}, \{(e, 2), (e, 3)\}, \{(e, 3), (e, 4)\}, \\ \{(e, 4), (e, 1)\}, \{(e, 5), (e, 6)\}, \{(e, 6), (e, 7)\}, \\ \{(e, 7), (e, 8)\}, \{(e, 8), (e, 5)\} : e \in E\}.$

Obviously, H is a bipartite graph, and H is planar since G is planar. The maximum degree of a vertex in H is six since G is cubic.

We observe that for every edge $e \in E$ and every realizable set *S* of *B*, $|S \cap (\{e\} \times \{1, 2, 3, 4\})| \leq 2$. In what follows, we may assume that $S \subseteq V \cup (E \times \{2, 3, 6, 7\})$ for all realizable sets *S* of *B*, since for every other realizable set *T*, the set $T' = (T \cap V) \cup (E \times \{2, 3, 6, 7\})$ is also realizable and fulfills $|T| \leq |T'|$.

In what follows, we apply the idea of the previous theorem. $W \subseteq V$ is an independent set of *G* if and only if *W* is a realizable set of *B*. Consequently, the planar graph *G* has an independent set of cardinality *k* if and only if the bipartite planar graph *B* has a realizable set of cardinality k + 4m.

Our problem remains NP-complete even when restricted to bipartite planar graphs of maximum degree four or three [15].

The INDEPENDENT SET problem is not only NPcomplete—it is also hard to approximate. More precisely, for every $\epsilon > 0$, there is no polynomial time approximation algorithm for the MAXIMUM INDEPENDENT SET problem with a worst-case performance ratio of $n^{1/2-\epsilon}$ unless P = NP and there is no polynomial time approximation algorithm with a worst-case performance ratio of $n^{1-\epsilon}$ unless ZPP = NP [20], where *n* is the number of vertices of the input graph. We consider the reduction used in the proof of Theorem 5 again. The split graph *H* has at most $|V(G)|^2$ vertices. Hence, for every $\epsilon > 0$, there is no polynomial time algorithm to approximate a maximum realizable set of a given (split) graph within factor $n^{1/4-\epsilon}$ unless P = NP (respectively, within factor $n^{1/2-\epsilon}$ unless ZPP = NP).

However, we cannot conclude the same for the restriction to planar graphs, since the INDEPENDENT SET problem admits a polynomial time approximation scheme for planar graphs [3]. In fact, as we will show in Section 5, the idea of Baker [3] can be applied to establish linear time approximation algorithms for the DPST problem when restricted to planar graphs.

For graphs in general, a polynomial time approximation algorithm with the worst-case performance ratio $O(\sqrt{n})$ will appear in [4].

4. GRAPHS OF BOUNDED TREEWIDTH

In this section, we will prove that the problem of finding a maximum realizable set is solvable in linear time for graphs which have a bounded treewidth. From a practical viewpoint, this is of great interest, since the graphs under consideration are often of a "treelike" structure. Well-known examples of graph classes with a bounded treewidth are forests, series-parallel graphs, Halin graphs, almost trees, *k*-outerplanar graphs, and graphs with bounded bandwidth and cutwidth (see e.g., [29]).

The definition of the treewidth of a graph is usually given in terms of a tree-decomposition (see e.g., [16, 21]). For our purpose, it is more convenient to take as a starting point the observation made in [29] that a graph has treewidth at most k if and only if it is a subgraph of a chordal graph with clique number at most k + 1. If some *constant upper bound* is placed on the treewidth of the graphs under consideration, we say that the class of graphs has a *bounded treewidth*. So, for example, the class of Halin graphs is a class of bounded treewidth graphs, since it can be shown that Halin graphs have a treewidth at most three (see e.g., [29]).

Many optimization problems can be solved "efficiently" for graphs of a bounded treewidth by formulating the problem in a logical language, called *Monadic Second Order Logic* (MSOL). It is known that problems which can be expressed in this way can be solved in linear time for graphs with a bounded treewidth [2].

The description below of the logical language and Definition 7 below are taken from [16] (see, e.g., [9] for an older description of the method). We explicitly describe the way our problem can be described in the MSOL language:

For graphs G = (V, E), the MSOL consists of a language in which predicates can be built with the following constituents:

- The logic connectives ∧, ∨, ¬, ⇒ and ⇔ (with their usual meanings),
- Individual variables which may be vertex variables (with domain V), edge variables (with domain E), vertex set variables (with domain P(V), the power set of V), and edge set variables [with domain P(E)],
- The existential and universal quantifiers ranging over variables (∃ and ∀, respectively), and
- the following binary relations:
 - $-v \in W$ (where v is a vertex variable and W a vertex set variable),

- $-e \in F$ (where e is an edge variable and F an edge set variable),
- "v and w are adjacent in G" (where v and w are vertex variables),
- "v is incident with e in G" (where v is a vertex variable and e an edge variable), and
- equality for variables.

It can easily be seen that if *A* and *B* are edge set variables and \mathcal{H} is a predicate in this language, then predicates like $\forall A \subseteq B \ \mathcal{H}$ and $\exists A \subseteq B \ \mathcal{H}$ are also admissible.

Definition 7. We define which graph properties and optimization problems are *MS-definable* as follows:

- An *extended graph property* is a function Q for which there are D_1, \ldots, D_t ($t \ge 0$), such that for each graph G and each $X_i \in D_i, 1 \le i \le t, Q(G, X_1, \ldots, X_t)$ is mapped to the value true or false.
- An extended graph property Q is said to be MSdefinable if there is a predicate $R(Y_1, \ldots, Y_t)$ that is defined in MSOL for graphs, with free variables Y_1, \ldots, Y_t , such that for each graph G and every X_1, \ldots, X_t with $X_i \in D_i$ for each $i, Q(G, X_1, \ldots, X_t)$ has the value true if and only if G satisfies $R(X_1, \ldots, X_t)$.
- An optimization problem is MS-definable if there is an MS-definable extended graph property $Q(G, X_1, ..., X_t)$ and constants $\alpha_1, ..., \alpha_t$ such that the problem is to find for a given graph *G* the maximum value of $\alpha_1|X_1| + \cdots + \alpha_t|X_t|$ for which $Q(G, X_1, ..., X_t)$ evaluates to true.

Now, we consider the optimization problem *P*: Given a graph G = (V, E), find a realizable set $S \subseteq V$ of maximum cardinality. We will show that the problem *P* is MS-definable. For that purpose, we define the following predicates, for $v \in V, e \in E, S \in \mathcal{P}(V)$, and $E', F \in \mathcal{P}(E)$:

- 1. The Boolean expression "vertex v is incident with edge e" is denoted as $v \in e$.
- 2. Vertex *v* is not incident with an edge *e* from an edge set *E*':

$$d_0(v, E') = \neg (\exists e \in E' \ v \in e).$$

3. Vertex v is incident with at least two edges in E': $d_{\geq 2}(v, E') = \exists e_1 \in E' \exists e_2 \in E' (\neg (e_1 = e_2))$

$$\wedge v \in e_1 \wedge v \in e_2$$
).

4. A set *F* of edges contains a cycle:

$$cycle(F) = \exists F' \subseteq F \; (\exists e' \in F' \land \forall v \in V)$$

$$\times (\neg d_0(v, F') \Rightarrow d_{\geq 2}(v, F'))).$$

5. A set $F \subseteq E$ of edges is the edge set of the graph G[[S]]:

weakly(F, S) = $\forall e \in E \ (e \in F \Leftrightarrow \exists v \in S \ v \in e)$.

6. The set of edges of G[[S]] contains no cycle:

$$R(S) = \forall F \subseteq E \text{ (weakly}(F, S) \Rightarrow \neg cycle(F)).$$

Lemma 8. R(S) evaluates to true if and only if $S \subseteq V$ is a realizable set.

Proof. This follows from Lemma 2 and the observation that a graph contains a cycle if and only if it has a (nonempty) subgraph in which all vertices have degree at least 2.

Corollary 9. Let Q be the extended graph property defined by: Q(G, S) is true if and only if $S \subseteq V$ is a realizable set of G. Then, Q is MS-definable.

Proof. This follows directly from Definition 7 and Lemma 8.

Corollary 10. The optimization problem P is MS-definable.

Proof. This follows from Definition 7 and Corollary 9: Take $\alpha_1 = 1$ and the extended graph property Q; then, P is equivalent to finding the maximum value of $\alpha_1 |S|$ for which Q(G, S) evaluates to true.

Arnborg et al. [2] showed that MS-definable optimization problems can be solved in linear time, given a tree decomposition of bounded width of the input graph. Bodlaender [6] proved that for any fixed constant $k \ge 1$ there exists a linear time algorithm that tests whether a given graph has treewidth at most k and, if so, outputs a tree decomposition of the graph with treewidth at most k. A linear time algorithm for the DPST problem can now be described as follows: First, use the algorithm by Bodlaender to find a tree-decomposition of minimum width in linear time. Use this tree-decomposition to find a solution for the DPST problem in linear time using the dynamic programming method described in [2].

Theorem 11. *The DPST problem is solvable in linear time for graphs of a bounded treewidth.*

It is important to notice that the graphs arising from applications such as water-distribution networks are very sparse and are likely to be k-outerplanar for a very small k and, hence, have a small treewidth. However, for larger values of k (say 4 or 5), the linear time algorithm described above is mainly of theoretical interest. In those cases, it is not very practical because of the enormous constants involved.

5. APPROXIMATION FOR PLANAR GRAPHS

In this section, we apply an idea of Baker [3] to establish linear time approximation algorithms for the DPST problem when restricted to planar graphs. We will prove the following theorem:

Theorem 12. For every $\epsilon > 0$, there is a linear time approximation algorithm of a worst-case performance ratio of $1-\epsilon$ for the DPST problem restricted to planar graphs.

Let $W \subseteq V$ be a set of (forbidden) vertices. A realizable set R of G is called a *maximum W*-avoiding realizable set if $R \cap W = \emptyset$ and $|R| \ge |R'|$ for every realizable set R' of G with $R' \cap W = \emptyset$.

Let G = (V, E) be a planar graph given with a fixed embedding in the plane. We partition V into levels L_1, L_2, \ldots, L_d . The level L_1 contains all vertices on the outer face of G. For i > 1, the level L_i contains all vertices on the outer face of $G - \bigcup_{j=1}^{i-1} L_j$. Let d be the largest index such that $L_d \neq \emptyset$. For technical reasons, set $L_i = \emptyset$ for i > d or i < 1. A planar graph is k-outerplanar if and only if it has an embedding defining at most k nonempty levels. We remark that, given a planar graph, a k-outerplanar embedding for which k is minimal can be found in polynomial time [5].

We decompose the planar graph *G* into *k*-outerplanar graphs. Each *k*-outerplanar graph consists of *k* consecutive levels of *G*. More precisely, let *k* and *r* be integers with $1 \le r \le k$. For i = 0, 1, ..., q with $q = \lceil (d-r)/k \rceil$, we define

$$G_{k,r,i} = G\left[\bigcup_{j=(i-1)k+r+1}^{ik+r} L_j\right] \text{ and} \\ W_{k,r,i} = L_{(i-1)k+r+1} \cup L_{ik+r}$$

Note that $W_{k,r,i}$ contains all vertices in the outer and inner levels of $G_{k,r,i}$.

Lemma 13. For i = 0, 1, ..., q, let $R_{k,r,i}$ be a $W_{k,r,i}$ -avoiding realizable set of $G_{k,r,i}$. Then, $\bigcup_{i=0}^{q} R_{k,r,i}$ is a realizable set of G.

Proof. For all *i*, the set $W_{k,r,i}$ contains the vertices on the outer and the inner levels of the *k*-outerplanar graph $G_{k,r,i}$. Hence, the endvertices of an arbitrary edge of $G[[R_{k,r}]]$ belong to the same *k*-outerplanar graph.

Lemma 14. Let *R* be a maximum realizable set of *G*. For every $k \ge 1$, there is an index r(k) with $1 \le r(k) \le k$ such that

$$\left| R \setminus \bigcup_{i=0}^{q} W_{k,r(k),i} \right| \geq \frac{k-2}{k} p(G)$$

Proof. Let *R* be a maximum realizable set of *G* and let $W_{k,r} = \bigcup_{i=0}^{q} W_{k,r,i}$. For every level L_j , j = 1, 2, ..., d, of *G*, there exist at most two $r \in \{1, 2, ..., k\}$ with $L_j \subset W_{k,r}$. Hence, $\sum_{r=1}^{k} |R \cap W_{k,r}| \le 2|R|$, which implies that there is an r = r(k) such that $|R \cap W_{k,r(k)}| \le (2/k)|R|$.

Let $k \ge 1$. For every r = 1, 2, ..., k and every i = 1, 2, ..., q, let $R_{k,r,i}$ be a maximum $W_{k,r,i}$ -avoiding realizable set of $G_{k,r,i}$. By Lemma 13, $R_{k,r} = \bigcup_{i=0}^{q} R_{k,r,i}$ is a realizable set of *G*. Consequently,

$$\max\{|R_{k,r}| : 1 \le r \le k\} \ge \frac{k-2}{k}p(G)$$

For every k, we are able to develop an exact linear time algorithm computing a maximum W-avoiding realizable set for k-outerplanar graphs. Using a variant of the method of Section 4, using labels to indicate the vertices of W, it can be shown that a linear time algorithm exists (see, e.g., [9]). Notice that the treewidth of a kouterplanar graph is at most 3k - 1 (see [29]). Consequently, for every fixed k, we can obtain a linear time approximation algorithm of a worst-case performance ratio of (k - 2)/k.

Remark 3. A detailed analysis of the dynamic programming algorithm for the DPST problem on *k*outerplanar graphs could give a polynomial time approximation algorithm with better worst-case performance ratio for the DPST problem on planar graphs.

6. INTERVAL GRAPHS

Definition 15. A graph is *chordal* if it contains no induced cycle of length more than three.

There are many characterizations of chordal graphs, for example, using perfect elimination schemes, intersection models of subtrees of a tree, and the existence of simplicial vertices. For an introduction into this graph class, we refer to [19].

Notice that for chordal graphs, in general, the problem of finding a maximum realizable set is NP-complete, since the class of split graphs is a proper subclass of the class of chordal graphs. However, for the class of interval graphs, which is another important subclass of the class of chordal graphs, we can give a fast algorithm.

Our first result shows that for chordal graphs we can restrict our search for realizable sets to independent sets:

Theorem 16. If G is a 2-edge connected chordal graph, then any realizable set S of G is an independent set of G.

Proof. Let G = (V, E) be a 2-edge connected chordal graph and assume that $\{x, y\} \in E$ for two distinct vertices $x, y \in S$. Since G is 2-edge connected, $\{x, y\}$ is contained in a cycle of G, and, since G is chordal, this implies that $\{x, y\}$ is contained in some triangle of G. This contradicts Lemma 2.

Remark 4. Notice that the condition that *S* is independent is, in general, not sufficient. A counterexample is the diamond (i.e., $K_4 - e$). It has an independent set with two vertices, but clearly this set is *not* realizable.

We will use the above observations and the following properties of 2-edge connected interval graphs:

Definition 17. An interval graph is a graph for which one can associate with each vertex an interval on the real line such that two vertices are adjacent if and only if their corresponding intervals have a nonempty intersection. Interval graphs can be recognized in linear time, and given an interval graph, an interval model for it can be found in linear time [8, 19]. In the following, we assume that an interval model of the graph is given, and we identify the vertices of the graph with the corresponding intervals. Without loss of generality, we may assume that no two intervals have an endpoint in common.

Definition 18. An interval and its corresponding vertex are called *minimal* if the interval is minimal with respect to inclusion, that is, if it does not contain any other interval.

Lemma 19. Let G be a 2-edge connected interval graph. Then, there exists a maximum realizable set S of G such that for every vertex $p \in S$ the corresponding interval is minimal.

Proof. Let *S* be a maximum realizable set containing a vertex *x* which is not minimal. Then, there exists an interval *y* contained in the interval *x*. By Theorem 16, we know that a realizable set can contain only one of *x* and *y* and, hence, $y \notin S$. Now, $N(y) \subseteq N[x]$, and, hence, by Lemma 4, there exists a maximum realizable set $S' = \{y\} \cup S \setminus \{x\}$. Repeating the arguments, we can prove the assertion of the lemma.

Consider the ordering of the minimal intervals defined by the left endpoints.

Lemma 20. Let G be a 2-edge connected interval graph with corresponding interval model and let x be the first minimal interval (i.e., with the leftmost left endpoint). There exists a maximum realizable set S of G with $x \in S$.

Proof. Consider a maximum realizable set *S* of *G* containing only minimal intervals. If $x \in S$, there is nothing to prove. Otherwise, let *y* be the first interval in *S*. The other intervals of *S* lie totally to the right of *y* because *S* is an independent set by Theorem 16. The right endpoint of *y* must be to the right of the right endpoint of *x* since the interval *x* is minimal. It follows that $S' = \{y\} \cup S \setminus \{x\}$ is also realizable, since *x* lies totally left of $S \setminus \{y\}$ and $N(z) \cap N(x) \subseteq N(z) \cap N(y)$ for all $z \in S \setminus \{y\}$.

Theorem 21. There is a linear time algorithm to compute a maximum realizable set S for a given interval graph G.

Proof. Locate the set of bridges B in G and compute maximum cardinality realizable sets for each component of G - B. This can be done as follows:

Consider an interval model for a 2-edge connected component. First, mark the minimal intervals. Take the minimal interval with the leftmost left endpoint as the first element of S. Consider the endpoints one by one,

from left to right. We keep track of the last minimal interval in S which is totally left of the current position. We also keep a counter for the number of intervals that have one endpoint to the left of the current position and that overlap with the last interval in S. If we encounter a left endpoint of a minimal interval which starts to the right of the last interval in S so far, and if there is at most one interval overlapping the current position and the last interval of S, then we put this new minimal interval in S.

Let S' be a maximum realizable set such that $S \neq S'$. By the previous lemmas, we may assume that S' contains minimal intervals only and that S and S' have a common first interval. Suppose that y is the first interval of S' which is not in S and that x_1, x_2, \ldots, x_p are common intervals of S and S' and $x_{p+1} \neq y$ is the next interval of S chosen by the above procedure. We complete the proof by showing that y in S' can be replaced by x_{p+1} . This follows by the same arguments as in the proof of Lemma 20 and the following observations: By the choice of x_1, x_2, \ldots, x_p , for all $i, j \in \{1, \ldots, p\}$ with $i \neq j, x_i$ and x_i have at most one common neighbor and $N(x_{p+1}) \cap N(x_i) \subseteq N(x_{p+1}) \cap N(x_{i+1}) (i = 1, \dots, p-1).$ If the addition of x_{p+1} to $\{x_1, \ldots, x_p\}$ would cause a cycle in $G[[\{x_1, x_2, \dots, x_p, x_{p+1}\}]]$, then such a cycle would already exist in $G[[\{x_1, \ldots, x_p\}]]$, a contradiction to the choice of x_1, x_2, \ldots, x_p .

7. COCOMPARABILITY GRAPHS

Definition 22. A graph G = (V, E) is a *cocomparability* graph if and only if there is an ordering $v_1, v_2, ..., v_n$ of V such that i < j < k and $\{v_i, v_k\} \in E$ implies either $\{v_i, v_j\} \in E$ or $\{v_j, v_k\} \in E$. Hence, $N(v_j) \cap \{v_i, v_k\} \neq \emptyset$ for all j with i < j < k. Such an ordering is called a *cocomparability ordering*.

For $w, w' \in V$, we shall write w < w' if w is on the left of w' in the ordering, that is, $w = v_i, w' = v_j$ and i < j.

Given a cocomparability graph G = (V, E), a cocomparability ordering can be computed in linear time [27]. In this section, we consider a cocomparability graph G = (V, E) with a fixed cocomparability ordering.

Lemma 23. Let P be a path with endvertices v_i and $v_k, i < k$, in a cocomparability graph G. Then, i < j < k implies that v_j has a neighbor in P.

Proof. Assume that i < j < k and v_j does not belong to *P*. Then, *P* contains an edge $\{v_h, v_l\}$ such that h < j < l. Hence, v_h or v_l is adjacent to v_j .

We will use chordless paths of a cocomparability graph in our algorithm to solve the DPST problem.

Lemma 24 [22]. Let $P = (p_1, p_2, ..., p_k), k \ge 1$, be a chordless path in a cocomparability graph G with $p_1 < p_k$. Then, $p_i < p_{i+2}$ for all i with $1 \le i \le k - 2$.

Consider a chordless path $P = (p_1, p_2, ..., p_k)$ with $p_1 < p_k$ and traverse P from p_1 to p_k . Then, each traversed edge is either a forward edge, that is, the next vertex is further to the right than is any previous vertex, or a backward edge, that is, the next vertex is to the left of the previous vertex but to the right of all other previous vertices. By Lemma 24, there cannot be two consecutive backward edges.

Let *S* be a subset of vertices such that each component of *G*[*S*] is a chordless path. An *S*-path is either the vertex set or the corresponding chordless path of a component of *G*[*S*], depending on the context. We say that a vertex *w* is a common neighbor of two different *S*-paths *S'* and *S''* if $w \notin S' \cup S''$ and *w* is adjacent to a vertex $s' \in S'$ and to a vertex $s'' \in S''$.

Our algorithm is based on the following characterization of realizable sets:

Theorem 25. Let G = (V, E) be a 2-edge connected cocomparability graph. A set $S \subseteq V$ is realizable if and only if

- 1. G[S] is a union of chordless paths,
- 2. Two vertices of an S-path have no common neighbor outside S, and
- 3. Different S-paths have at most one common neighbor.

Proof. Assume that *S* is realizable. First, consider condition 1. Suppose, on the contrary, that the vertices $c, x, y, z \in S$ induce a claw in *G* with central vertex *c*. There is a vertex $x' \in N(x) \setminus \{c\}$ since *G* is 2-edge connected. Moreover, x' is not adjacent to c, y, or z since *S* is realizable. Similarly, there exist vertices

$$y' \in N(y) \setminus N(\{c, x, z\})$$
 and $z' \in N(z) \setminus N(\{c, x, y\})$.

If $\{x', y', z'\}$ is an independent set of *G*, then these three vertices form a so-called asteroidal triple (see also Section 8), which is impossible in cocomparability graphs. Hence, we may assume that $\{x', y'\} \in E$. But, now, x', x, c, y, and y' induce a chordless 5-cycle in *G*, which is also impossible. Consequently, for 2-edge connected cocomparability graphs, condition 1 holds. It is easy to check that a realizable set *S* must satisfy conditions 2 and 3.

Assume that a set *S* satisfies the three conditions. Consider a shortest cycle *C* in G[[S]]. By conditions 2 and 3, *C* contains vertices from at least three *S*-paths. Since *C* is a shortest cycle, each vertex of *C* that does not belong to *S* has exactly two neighbors in *S* which belong to *C* by condition 2. Hence, all three vertices of *C* in different *S*-paths form an asteroidal triple. This proves the theorem

since cocomparability graphs do not contain asteroidal triples.

Combining Lemma 23 and Theorem 25, we obtain

Lemma 26. Let *S* be a realizable set of a 2-connected cocomparability graph *G* with $v_i, v_j, v_k \in S$ and i < j < k. If v_i and v_k belong to one *S*-path, then v_j belongs to the same *S*-path.

Our dynamic programming algorithm is based on Theorem 25. We still assume that the input graph is 2edge connected. The algorithm constructs a set S such that G[S] is a union of chordless paths. Thus, the algorithm can be considered as a procedure to construct a particular collection of chordless paths of G. For two chordless paths T' and T'' of G, we define T' < T'' if t' < t'' for all $t' \in T'$ and all $t'' \in T''$. Notice that $S_i < S_j$ or $S_j < S_i$ for any two different S-paths of a realizable set S by Lemma 26.

Lemma 27. Let T', T'', and \tilde{T} be chordless paths of a cocomparability graph G such that $T' < \tilde{T} < T''$ and there is no edge between either T' or T'' and \tilde{T} . Then, $t' \in T'$ and $t'' \in T''$ imply that $\{t', t''\} \notin E$. Furthermore, if T' and T'' have a common neighbor, then every $\tilde{t} \in \tilde{T}$ is adjacent to every common neighbor of T' and T''.

Proof. Let $\tilde{t} \in \tilde{T}, t' \in T'$ and $t'' \in T''$. Then, by the definition of a cocomparability ordering, $t' < \tilde{t} < t''$ and $\{t', t''\} \in E$ imply that either $\{\tilde{t}, t'\} \in E$ or $\{\tilde{t}, t''\} \in E$, contradicting the choice of \tilde{T} .

Now, let *w* be a common neighbor of *T'* and *T''*. Then, there are $t' \in T'$ and $t'' \in T''$ such that $\{w, t'\} \in E$ and $\{w, t''\} \in E$. Hence, (t', w, t'') is a path in *G*. Since $t' < \tilde{t} < t''$ for all $\tilde{t} \in \tilde{T}$, Lemma 23 implies that $\{w, \tilde{t}\} \in E$.

Our algorithm constructs a maximum realizable set S of a given cocomparability graph G = (V, E) with cocomparability ordering v_1, v_2, \ldots, v_n . It uses a dynamic programming approach with a linear scan through the cocomparability ordering. This technique has been used in previous algorithms for cocomparability graphs (see, e.g., [22]).

A subsolution constructed by the algorithm is a realizable set *S*. Subsolutions are stored as states $Z \in \{0, 1, ..., n\}^4$ with $Z = [z_1, z_2, z_3, z_4]$ such that z_4, z_3 and z_2 are the indices of the last, second last, and third last vertex of *S*, respectively, in the order of the path or, in case of different *S*-paths, according to the cocomparability ordering. ($z_j = 0$ if the corresponding vertex of *S* does not exist.) Finally, z_1 indicates the maximum number of vertices in a realizable set with last vertices z_2, z_3 , and z_4 .

The algorithm starts with a preprocessing in which it computes A^2 in time $O(n^{2.376})$ by matrix multiplication [14], where A is the adjacency matrix of G for which

A(i, j) = 1 if $i \neq j$ and $\{v_i, v_j\} \in E$, and A(i, j) = 0 otherwise. Consequently, during the dynamic programming part of the algorithm, the number of common neighbors of two vertices v_i and v_j can be computed in constant time.

The dynamic programming algorithm starts with the subsolution $S = \emptyset$ and state Z = [0, 0, 0, 0]. It works in rounds j = 0, 1, ..., n - 1, such that in round j the successors of all existing states Z with $j = \max(z_3, z_4)$ are computed. As typical for the dynamic programming approach, the algorithm maintains the following invariant: If Z is a state computed by the algorithm, then there is a realizable set S corresponding to Z, that is, $|S| = z_1$ and z_2, z_3, z_4 are the last vertices of S. Furthermore, for any realizable set S of G, the algorithm computes a state Z such that S corresponds to Z with the possible exception of $z_1 > |S|$.

Now, Z' is a successor of the state $Z = [z_1, z_2, z_3, z_4]$ if Z' is a state corresponding to a realizable set $S \cup \{v_k\}$, where S is a realizable set corresponding to Z and v_k is added by a backward or a forward step. This means that v_k is a vertex with $z_3 < k < z_4$ in a backward step and max $(z_3, z_4) < k$ in a forward step.

Any round *j* of our algorithm consists of two phases: In the first phase, all successors via a backward step of previously computed states *Z* with $j = \max(z_3, z_4)$ are computed. In the second phase, all successors via a forward step of previously computed states *Z* with $j = \max(z_3, z_4)$ are computed. Notice that this implies that in the second phase all successors of states obtained during the first phase are computed. To justify the correctness of our algorithm, we show that it is enough to know the last three vertices of any realizable set *S* for deciding whether $S \cup \{v_k\}$ is realizable or not.

Lemma 28. Let $P = (p_1, p_2, ..., p_k), k \ge 4$, be a chordless path and $p_1 < p_k$. Let u be a vertex with $p_{k-1} < u$. Assume that $\tilde{P} = (p_{k-2}, p_{k-1}, p_k, u)$ is a chordless path. Then, $P' = (p_1, p_2, ..., p_k, u)$ is also a chordless path. Furthermore, if no vertex of \tilde{P} has a common neighbor with u outside \tilde{P} , then no vertex of P' has a common neighbor with u outside P'.

Proof. Suppose that $\tilde{P} = (p_{k-2}, p_{k-1}, p_k, u)$ is chordless but P' is not. Let l be the largest index l < k for which $\{u, p_l\} \in E$. By our assumption, l < k - 2. Thus, $(p_l, p_{l+1}, \ldots, p_k, u, p_l)$ is a chordless cycle of length at least 5 in G. This is a contradiction since a chordless cycle of a cocomparability graph has length at most 4. Suppose that w is a common neighbor of u and a vertex in P', while the only neighbor of w in \tilde{P} is u. Let p_l be the rightmost neighbor of w in P'. Hence, l < k - 2 and $(p_l, p_{l+1}, \ldots, p_k, u, w, p_l)$ is a chordless cycle of length at least 6 in G—a contradiction.

Lemma 29. Let $P = (p_1, p_2, \dots, p_k)$ be a chordless path and $p_1 < p_k$. Let $L(P) = \{p_{k-1}, p_k\}$ if $k \ge 2$,

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and $L(P) = \{p_k\}$ if k = 1. Let u be a vertex with $\max(p_{k-1}, p_k) < u$. If u is nonadjacent to the vertices of L(P), then u has no neighbor in P. Furthermore, let w be a common neighbor of u and a vertex of P. Then, u has a common neighbor with a vertex of L(P), unless either |L(P)| = 2 and u is adjacent to both vertices of L(P) or w has at least two neighbors in P.

Proof. Let $p_i \notin L(P)$ be a vertex of the path *P* with $\{u, p_i\} \in E$. Then, $k \ge 3, i \le k - 2$ and $p_i < p_k$. Hence, $p_i < p_k < u$ and $\{u, p_i\} \in E$ implies that $\{u, p_k\} \in E$, since $\{p_i, p_k\} \notin E$ by the choice of *P*. Hence, *u* has a neighbor in L(P).

Let *w* be a common neighbor of *u* and a vertex $p_i \notin L(P)$ of the path *P*. Then, $k \ge 3, i \le k - 2$, and $p_i < p_k$. If $\max(p_{k-1}, p_k) < w$, then *w* has a neighbor in L(P) as shown above.

Now, assume that *u* is not adjacent to both vertices of L(P) and that *w* has exactly one neighbor in *P*, implying that *w* has no neighbor in L(P). If $w < \min(p_{k-1}, p_k)$, then $\{w, u\} \in E$ implies that both p_{k-1} and p_k are adjacent either to *u* or to *w*. By our assumption, *u* is not adjacent to p_{k-1} or p_k . Thus, *w* has two neighbors in *P*—a contradiction. Finally, *w* cannot be between p_{k-1} and p_k in the cocomparability ordering since $\{p_{k-1}, p_k\} \in E$ implies that *w* is adjacent to a vertex of L(P), a contradiction.

Clearly, if u and all vertices of P are contained in a realizable set S of G, then neither u nor w can be adjacent to two vertices of P.

Summarizing, we obtain

Proposition 30. Let *S* be a realizable set of a cocomparability graph *G*. Let $L_3(S)$ be the set of the last three vertices of *S*, if $|S| \ge 3$. Otherwise, let $L_3(S) = S$. Let *u* be a vertex such that either s < u for all $s \in S$ (forward edge or nonedge) or s < u for all but one $s \in S$ (backward edge). Then, $S \cup \{u\}$ is realizable in *G* if $L_3(S) \cup \{u\}$ is realizable.

Proof. By Lemma 27, when we add vertex u to a realizable set S, we only have to consider the component of $G[S \cup \{u\}]$ containing u and the previous S-path in the cocomparability ordering. Then, by Lemmas 28 and 29, checking $L_3(S) \cup \{u\}$ is sufficient if u creates a new component or is added to an isolated vertex of G[S]. By Lemma 29, when adding u to an S-path of more than two vertices, the previous component need not be checked. Hence, by Lemma 28, checking $L_3(S) \cup \{u\}$ suffices.

The proposition immediately guarantees the correctness of our dynamic programming algorithm: In a backward step, $Z' = [z_1 + 1, z_3, z_4, k]$ is a successor of a state $Z = [z_1, z_2, z_3, z_4]$ if $z_3 < k < z_4, \{v_{z_4}, v_k\} \in E$ and $\{v_{z_2}, v_{z_3}, v_{z_4}, v_k\}$ is realizable. In a forward step, Z' = $[z_1+1, z_3, z_4, k]$ is a successor of a state $Z = [z_1, z_2, z_3, z_4]$ if max $(z_3, z_4) < k$ and $\{v_{z_2}, v_{z_3}, v_{z_4}, v_k\}$ is realizable.

Consider the running time: The test whether a set of up to four vertices is realizable can be done in constant time, since, by Theorem 25, it requires only adjacency tests and the computation of the number of common neighbors for vertices in the set. States are maintained as follows: There is a three-dimensional array B[0..n, 0..n, 0..n] initialized to be zero. Whenever a new state $Z' = [z'_1, z'_2, z'_3, z'_4]$ has been computed as a successor, then z'_1 is stored in $B(z'_2, z'_3, z'_4)$, if z'_1 is larger than the current value of $B(z'_2, z'_3, z'_4)$ (which means that we found a better subsolution). Hence, during the algorithm, the O(n) successors of $O(n^3)$ different states are computed. The algorithm uses a variable max to maintain the largest first entry of any state computed. Hence, the value of max upon termination is the degree-preserving number of the input graph. Consequently, the running time of the algorithm is $O(n^4)$.

Using a standard pointer structure, the algorithm can be implemented to compute within the same time a maximum realizable set and this can easily be transformed into a maximum degree-preserving forest. Hence, we may conclude.

Theorem 31. There is an algorithm to compute a maximum degree-preserving forest of a cocomparability graph in time $O(n^4)$.

8. GRAPHS WITH BOUNDED ASTEROIDAL NUMBER

We remind the reader that we still assume that G is 2-edge connected.

Definition 32. A set $A \subseteq V$ is called an *asteroidal set* if for every vertex $a \in A$ the set $A \setminus \{a\}$ is contained in a component of G - N[a]. The *asteroidal number* of a graph G, denoted by an(G), is the maximum cardinality of an asteroidal set in G.

Clearly, every asteroidal set of *G* is an independent set of *G*. On the other hand, every independent set of cardinality at most two is asteroidal in *G*. An asteroidal set of cardinality three is called an *asteroidal triple* (AT). The class of AT-free graphs contains all graphs *G* with $an(G) \le 2$. This class, studied in detail in [13], contains all cocomparability graphs. The intersection with the class of chordal graphs gives exactly all interval graphs [23].

In this section, we consider graphs with a bounded asteroidal number.

For a vertex $w \in W$, a neighbor $u \in N(w)$ is called a *W*-private neighbor of *w* if $u \notin N[W \setminus \{w\}]$. We define $M(c) = \{v \in V : N[v] \subseteq N[c]\} \setminus \{c\}$. Note that if *S* is a realizable set containing *c* then $S \cap M(c) = \emptyset$. The next lemma bounds $|S \cap N(c)|$. **Lemma 33.** Let c be a vertex in a realizable set S of a graph G with $\operatorname{an}(G) \leq k$. Then, $|S \cap N(c)| \leq k^2$.

Clearly, S contains no elements of M(c). To prove the lemma, we need the following claim:

Claim. Let *G* be a graph on *n* vertices such that every independent set of *G* contains at most *a* vertices and every block of *G* contains at most *c* vertices. Then, $n \le a \cdot c$.

Proof. For fixed values *a* and *c*, we choose a graph G = (V, E) such that it has maximum number of vertices and, among those, one with maximum number of edges. This implies that every block of *G* is complete.

Let Z be the set of cut vertices of G and $N = V \setminus Z$.

We consider a block G[B] of G with $B \cap N \neq \emptyset$. By maximality, we have |B| = c. Every maximal independent set of G contains exactly one vertex of B since G[B] is a complete graph. By the induction hypothesis applied to G - B, we know that $n - c \leq (a - 1)c$. Consequently, $n \leq ac$.

It remains to show that *G* has a block containing a vertex in *N*. Therefore, we consider a graph $(\mathcal{B} \cup Z, F)$, where $\mathcal{B} = \{B : G[B] \text{ is a block of } G\}$ and $F = \{\{B, x\} : B \in \mathcal{B} \text{ and } x \in B \cap Z\}$. It is well known that this graph is a forest. Every leaf of this forest corresponds with a block containing a vertex in *N*.

Now, we are able to prove the lemma:

Proof. We define $X = S \cap N(c)$ and choose a set $Y \subseteq N(X) \setminus N[c]$ such that every vertex in X has exactly one X-private neighbor in Y. Notice that |X| = |Y| and every vertex of Y has exactly one neighbor in X. Then, every independent set of G[Y] is an asteroidal set of G, and for every block G[T] of G[Y], the set $N(T) \cap X$ is an asteroidal set of G. Thus, $\operatorname{an}(G) \leq k$ and our claim implies that $|X| = |Y| \leq k^2$.

Definition 34. Let A be an asteroidal set of G = (V, E)and $C \in \text{Comp}(G - N[A]) \cup \{\emptyset\}$. The pair (A, C) is called a *lump* of G if either

- $A = \emptyset$ and $C \in \text{Comp}(G)$ or
- $A = \{a\}$ and $C \in \text{Comp}(G N[a]) \setminus \text{Comp}(G)$ or
- $|A| \ge 2$ and $C = \bigcap_{a \in A} C_a$, where C_a is the set in Comp(G N[a]) with $A \setminus \{a\} \subseteq C_a$.

We give some examples: For every asteroidal set A of G = (V, E) with |A| > 1, there is exactly one set $C \subseteq V$ such that (A, C) is a lump of G.

If $\operatorname{an}(G) = 1$, then G is complete and (\emptyset, V) is the unique lump of G.

For all $C \in Comp(G)$, every lump of G[C] is a lump of G. All other lumps of G are of the type $(\{x, y\}, \emptyset)$ for two vertices x and y in different components of G.

If G is isomorphic to $K_{n,m}$, $n \ge 1$ and $m \ge 2$, and (A, C) is a lump of G with $A \ne \emptyset$ or $C \ne V$, then there

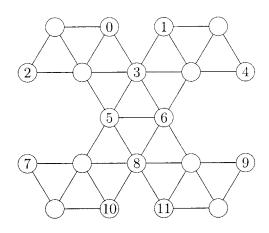


FIG. 1. The set $A = \{2, 4, 7, 9\}$ is an asteroidal set in this graph. Since $|A| \ge 2$, there is a unique set *C* such that (A, C) is a lump. Here, $C = \{0, 1, 3, 5, 6, 8, 10, 11\}$. The vertex $3 \in C$ decomposes (A, C)into three smaller lumps, namely, $(\{2, 3\}, \emptyset)$, $(\{3, 4,\}, \emptyset)$, and $(\{3, 7, 9\}, \{8, 10, 11\})$. The vertex $6 \in C$ decomposes (A, C) into four smaller lumps, namely, $(\{2, 6\}, \{0\})$, $(\{4, 6\}, \{1\})$, $(\{6, 7\}, \{10\})$, and $(\{6, 9\}, \{11\})$.

exist two different nonadjacent vertices x and y of G such that $A = \{x\}$ and $C = \{y\}$ or $A = \{x, y\}$ and $C = \emptyset$.

Lemma 35. Every graph G on n vertices with $\operatorname{an}(G) = k \neq 2$ has at most n^k lumps. Every AT-free graph G on n vertices has at most $\frac{3}{2}n^2$ lumps.

Proof. The number of lumps (A, C) with $A = \emptyset$ is bounded by *n*. The number of lumps (A, C) with |A| = 1is bounded by n(n-1), since for every vertex *a*, at most n-1 components of G-N[a] exist. The number of lumps (A, C) with |A| = i is bounded by $\binom{n}{i}$ for $2 \le i \le k$, since in this case, *C* is uniquely determined by *A*. For k > 2, this sums up to at most n^k since $n \ge 2k$.

The following theorem shows how to decompose lumps into smaller lumps. The basic technique was developed in [10]:

Theorem 36. Let (A, C) be a lump of G. For every $c \in C$, there exist unique partitions \mathscr{A} of A and \mathscr{C} of $C \setminus N[c]$ such that

- 1. For every set $B \in \mathcal{A}$, either $(B \cup \{c\}, \emptyset)$ is a lump of G or there is a set $D \in \mathcal{C}$ such that $(B \cup \{c\}, D)$ is a lump of G, and
- 2. For every set $D \in \mathcal{C}$, either ({c}, D) is a lump of G or there is a set $B \in \mathcal{A}$ such that $(B \cup \{c\}, D)$ is a lump of G.

Proof. Let (A, C) be a lump of G = (V, E) and $c \in C$. We define

$$\mathcal{A} = \{A \cap D : D \in \operatorname{Comp}(G - N[c])\} \setminus \{\emptyset\},\$$

$$\mathcal{C}_1 = \{D : D \in \operatorname{Comp}(G - N[c]) \text{ and } A \cap D = \emptyset\},\$$

$$\mathcal{C}_2 = \{D : (B \cup \{c\}, D) \text{ is a lump of } G \text{ and } B \in \mathcal{A}, \text{ and}\$$

$$\mathcal{C} = \mathcal{C}_1 \cup \mathcal{C}_2 \setminus \{\emptyset\}.$$

 \mathscr{A} is a partition of *A* since $\operatorname{Comp}(G - N[c])$ is a partition of $V \setminus N[c]$. By the definition of a lump (Definition 34), $B \cup \{c\}$ is an asteroidal set of *G* for every $B \in \mathscr{A}$. \mathscr{C} is a partition of $C \setminus N[c]$, since for every set $D \in \operatorname{Comp}(G - N[c])$, either $C \cap D = \emptyset$ or there is exactly one set in \mathscr{C} containing $C \cap D$. Consider the latter case in detail: Either $A \cap D = \emptyset$, $D \in \mathscr{C}_1$, and $(\{c\}, D)$ is a lump or $A \cap D = B \neq \emptyset$, implying that $B \in \mathscr{A}$ and $(B \cup \{c\}, C \cap D)$ is a lump.

Now, we consider any pair $(\mathscr{A}', \mathscr{C}')$ of partitions of Aand C satisfying the conditions of the theorem. Then, \mathscr{A}' is a refinement of \mathscr{A} , that is, for every set $A' \in \mathscr{A}'$, there is a set $B \in \mathscr{A}$ such that $A' \subseteq B$. Otherwise, there would be a set $A' \in \mathscr{A}'$ such that $A' \cup \{c\}$ is not asteroidal, contradicting the first assumption. Furthermore, let $A' \in$ \mathscr{A}' and $B \in \mathscr{A}$ be sets such that $b \in B \setminus A'$. The first condition implies that $b \in D'$ for the lump $(A' \cup \{c\}, D')$ of G contradicting $b \notin C$. Consequently, $\mathscr{A}' = \mathscr{A}$. Now, the first condition implies that $\mathscr{C}' = \mathscr{C}$, since $\mathscr{C}_1 \subseteq \mathscr{C}'$ by the second condition.

Let (A, C) be a lump of G. For a fixed vertex $c \in C$, let \mathscr{A} and \mathscr{C} be the partitions given in the proof of the previous theorem. We define the *decomposition* of (A, C)to be the set of lumps $(A' \cup \{c\}, C')$ of G with $A' \in$ $\mathscr{A} \cup \{\emptyset\}$ and $C' \in \mathscr{C} \cup \{\emptyset\}$. This set of lumps is denoted by Dec(A, C, c). See Fig. 1 and Fig. 2.

Our algorithm is a dynamic programming algorithm on the lumps of *G*. Let (A, C) be a lump and let *S* be a realizable set of *G* such that $A \subseteq S \subseteq N[A] \cup C$. Let $B = S \cap N(A)$. If $C \cap S = \emptyset$, then $S = A \cup B$. Otherwise, let $c \in C \cap S$ and consider the decomposition $\text{Dec}(A, C, c) = \{(A_i, C_i) : 1 \le i \le l\}$. Let $U = S \cap N(c)$ and $S_i = S \cap (N[A_i] \cup C_i)$ for $1 \le i \le l$. Now, it is easy to express $S_i \cap N(A_i)$ and $S_i \cap C_i$ in terms of *A*, *B*, and *U*. However, our task is the other way around: Given the decomposition Dec(A, C, c) and the sets $S_i, 1 \le i \le l$, we have to assemble these sets to a realizable set *S* of *G*. The next lemma gives a condition that ensures that realizable sets S_i can be joined to a realizable set *S*:

Lemma 37. Let (A, C) be a lump of G and $B \subseteq N(A)$. Let c be a vertex in C and $U \subseteq N(c)$ such that $U \cup \{c\}$ is realizable and $N(A) \cap U = N(c) \cap B$. Let Dec(A, C, c) =

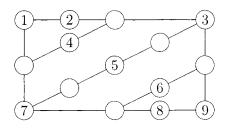


FIG. 2. The set $A = \{3, 5, 7\}$ is an asteroidal set in this graph. Since $|A| \ge 2$ there is a unique set $C = \{1, 2, 4, 6, 8, 9\}$ such that (A, C) is a lump. The vertex $6 \in C$ decomposes (A, C) into two smaller lumps, namely, $(A \cup \{6\}, \{1, 2, 4\})$ and $(\{6\}, \{8, 9\})$.

 $\{(A_i, C_i) : 1 \le i \le l\}$. For all realizable sets $S_i \subseteq N[A_i] \cup C_i$ of G such that $A_i \subseteq S_i$ and $S_i \cap N(A_i) = (B \cup U) \cap N(A_i)$, for all $1 \le i \le l$, then the set $S = \bigcup_{i=1}^l S_i$ is realizable in $G, A \cup \{c\} \subseteq S$, and $S \cap N(A) = B$.

Proof. Let *D* be shorthand for N(A) and $D_i = N(A_i)$ for $1 \le i \le l$.

First, we observe that $S_i \cap D_i \cap D_j = (B \cup U) \cap D_i \cap D_j = S_j \cap D_i \cap D_j$ for all indices *i* and *j* with $1 \le i, j \le l$. This implies that $S_i = S \cap (A_i \cup C_i \cup D_i)$ for all *i*.

Now, we apply Lemma 2 to show that *S* is a realizable set of *G*. We consider a chordless cycle *Z* in *G*[[*S*]]. *Z* contains a vertex $z \notin N[c]$ since $U \cup \{c\}$ is realizable in *G*. Let *i* be an index such that $z \in S_i$ and let $(b, \ldots, z, \ldots, d)$ be a longest subpath of *Z* contained in *G*[[*S_i*]]. Then, $b, d \in U$ since $C_i = C \cap \text{Comp}_z(G - N[c])$. Now, $(c, b, \ldots, z, \ldots, d)$ is a cycle in *G*[[*S_i*]] since $c \in A_i \subseteq S_i$, contradicting the fact that S_i is realizable. Hence, such a cycle *Z* cannot exist, and by Lemma 2, the set *S* is realizable in *G*.

By Theorem 36, $c \in A_i$ and $A_i \subseteq S_i$ for all i = 1, ..., kimplies that $A \cup \{c\} \subseteq S$. Finally, $S_i \cap D_i = (B \cup U) \cap D_i$ for all $i, 1 \leq i \leq l$, implies that $S \cap D = B$ since $U \subseteq N(c)$ and $N(A) \cap U = N(c) \cap B$.

Let (A, C) be a lump of G. For all $B \subseteq N(A)$, we define

$$p(A, B, C) = \max\{|S \cap C| : A \subseteq S, S \cap N(A) = B,$$
and S is realizable in G}.

Note that $p(A, B, C) = -\infty$ if no such realizable set exists.

Lemma 38. Let (A, C) be a lump of a graph G with $\operatorname{an}(G) = k$. Then, for all $B \subseteq N(A)$, we have

$$p(G) = \sum_{C \in \text{Comp}(G)} p(\emptyset, \emptyset, C)$$
 and

 $p(A, B, C) = \begin{cases} \max(0, \max_{c \in C} p'(c)) & \text{if } A \cup B \text{ is realizable,} \\ -\infty & \text{otherwise.} \end{cases}$

Here,

$$p'(c) = 1 + \max_{U \in \mathcal{U}(c)} \sum_{(A',C') \in \text{Dec}(A,C,c)} p(A', (B \cup U) \cap N(A'), C')$$

$$\mathcal{U}(c) = \{ U : U \subseteq N(c) \setminus M(c), B \cap N(c) = U \cap N(A), \\ and \ U \cup \{c\} \text{ is realizable in } G \}.$$

Proof. The formula for p(G) follows directly from $p(G[C]) = p(\emptyset, \emptyset, C)$ for all components G[C] of G. Next, we consider the trivial cases concerning the formula for p(A, B, C). If there is no realizable set S with $A \cup B \subseteq S$, then $p(A, B, C) = -\infty$. Otherwise, if $C \cap S = \emptyset$ for all these realizable sets S, then p(A, B, C) = 0. In the formula, we have $\max_{c \in C} p'(c) = -\infty$ if $C = \emptyset$. Otherwise, for every $c \in C$, there is a block $(A', C') \in \text{Dec}(A, C, c)$ such that $A' \cup (B \cap N(A))$ is not realizable, since $C \cap S$

= \emptyset for all *S*. Since every subset of a realizable set is realizable, this implies that $A' \cup B'$ is not realizable for all $c \in C$ and all $U \in \mathcal{U}(c)$ and $B' = (B \cup U) \cap N(A)$. Again, this implies that $\max_{c \in C} p'(c) = -\infty$. Hence, our formula is correct if $A \cup B$ is realizable, but $C \cap S = \emptyset$ for all realizable set *S* with $A \subseteq S$ and $S \cap N(A) = B$. This completes the base step of an induction.

Now, we assume that a realizable set *S* exists with $A \subseteq S, S \cap N(A) = B$, and $C \cap S \neq \emptyset$. We choose *S* such that $|S \cap C|$ is maximum. Let *c* be an arbitrary vertex in $C \cap S$. Let $\text{Dec}(A, C, c) = \{(A_i, C_i) : i = 1, ..., l\}$. Let $U = S \cap N(c) \setminus M(c)$. Then, $U \in \mathcal{U}(c)$.

We consider an arbitrary lump $(A_i, C_i) \in \text{Dec}(A, C, c)$ and define $B_i = (B \cup U) \cap N(A_i)$. First, let $S_i = S \cap (N[A_i] \cup C_i)$. Then, $p(A_i, B_i, C_i) \ge |S_i \cap C_i|$ by the induction hypothesis. By induction, we obtain $|S \cap C| \ge p'(c)$. This proves that $p(A, B, C) \ge \max_{c \in C} p'(c)$.

To prove the other inequality, we choose realizable sets $T_i \subseteq N[A_i] \cup C_i$ with $A_i \subseteq T_i$ and $T_i \cap N(A_i) =$ $(B \cup U) \cap N(A_i)$ such that $|T_i \cap C_i|$ is maximum. By Lemma 37, the set $T = \{c\} \cup \bigcup_{i=1}^{l} T_i$ is realizable, $A \subseteq T$, and $T \cap N(A) = B$. Now, $|S \cap C| \leq |T \cap C|$ since $|S_i \cap C_i| \leq |T_i \cap C_i|$ for all *i*. This proves that $p(A, B, C) \leq \max_{c \in C} p'(c)$.

Now, it is easy to derive a recursive algorithm from the formulas in Lemma 38. We consider the running time of our algorithm on an input graph G = (V, E)with |V| = n, |E| = m, and $\operatorname{an}(G) = k$. We use a data structure to memorize values of p(A, B, C) already computed. This data structure supports the following operations:

- store(A, B, C, p) stores the value p for the lump (A, C) and the set $B \subseteq N(A)$,
- present(A, B, C) returns true, if an operation store(A, B, C, p) has been performed before, for any value of p, and false otherwise, and
- value(A, B, C) returns the value p of the (last) store(A, B, C, p) operation, if present(A, B, C) = true.

All three operations can be executed by iterated search for a vertex in the universe V. A single search can be done in time $O(\log n)$ by standard techniques. To find a whole triple (A, B, C), we need at most |A| + |B| + 1single searches. If (A, C) is a lump, then A, B, and C are pairwise disjoint and, consequently, $|A| + |B| + 1 \le n$. This implies that the operations store, present, and value can be executed in time each of $O(n \log n)$.

Consider the algorithm in Table I:

This algorithm calls value(A, B, C) only if present(A, B, C) = true. Furthermore, if store(A, B, C) is called, then we have present(A, B, C) = false, that is, for each triple (A, B, C), store is called at most once. The number of such triples is bounded by the number of lumps (A, C) times the number of subsets $B \subseteq N(A)$. By Lemma 33, $|B \cap N(a)| \leq k^2$ for every

```
procedure main;
begin
  p \leftarrow 0:
  for C \in \text{Comp}(G) do p \leftarrow p + \text{access}(\emptyset, \emptyset, C);
  return(p)
end.
procedure access(A, B, C);
begin
  if not present(A, B, C) then compute(A, B, C);
  return(value(A, B, C))
end:
procedure compute(A, B, C);
begin
  if A \cup B is realizable in G
     then
         begin
         p \leftarrow 0;
         for c \in C do
           begin
               Q \leftarrow \{B \cap N(c)\};
               while Q \neq \emptyset do
               begin
                  choose a set U \in Q of minimum cardinality;
                  Q \leftarrow Q \setminus \{U\};
                  r \leftarrow 1 + |U \setminus N(A)|;
                  for (A', C') \in \text{Dec}(A, C, c) do
                  r \leftarrow r + \operatorname{access}(A', (B \cup U) \cap N(A'), C');
                  p \leftarrow \max\{p, r\};
                  for x \in N(c) \setminus (U \cup M(c) \cup N(A)) do
                    if U \cup \{c, x\} is realizable then Q \leftarrow Q \cup
                     \{U \cup \{x\}\};
                  end
           end
         end
     else p \leftarrow -\infty;
   store(A, B, C, p)
end:
```

 $a \in A$. By Lemma 35, we have to evaluate compute for at most $O(2^{k^3}n^k)$ triples (A, B, C).

Finally, we consider the running time of a single call of compute(A, B, C) without counting the running time of those recursive calls of procedure compute for which present(A', B', C') = false when compute(A', B', C') is called. We consider at most n vertices $c \in C$. By Lemma 33, we know $|U| \leq k^2$ for every $U \in \mathcal{U}(c)$. Hence, $\mathcal{U}(c)$ can be computed in time $O(2^{k^2}n)$. Clearly, |Dec(A, C, c)| < n. Consequently, we have at most $2^{k^3}n^2$ calls access(A', B', C'). Thus, the total time for one call of compute is bounded by $O(2^{k^3}n^3\log n)$.

Theorem 39. There is an algorithm to solve the degreepreserving spanning tree problem for any graph G in time $O(2^{k^3}n^{k+3}\log n)$, where $k = \operatorname{an}(G)$.

9. OPEN PROBLEMS

It follows from Theorem 6 that the DPST problem is NP-complete for the class of bipartite planar graphs with maximum degree six. It can be easily seen that a proper subclass of this class is the class of grid graphs (a grid graph is a vertex induced finite subgraph of the infinite grid). So, an interesting question is whether the DPST problem restricted to grid graphs remains NP-complete.

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