EXACT ALGORITHMS FOR TREEWIDTH AND MINIMUM FILL-IN*

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Abstract. We show that the treewidth and the minimum fill-in of an \( n \)-vertex graph can be computed in time \( O(1.8899^n) \). Our results are based on combinatorial proofs that an \( n \)-vertex graph has \( O(1.7087^n) \) minimal separators and \( O(1.8135^n) \) potential maximal cliques. We also show that for the class of asteroidal triple–free graphs the running time of our algorithms can be reduced to \( O(1.4142^n) \).

Key words. exact exponential algorithm, treewidth, fill-in, minimal separators, potential maximal clique, minimal triangulation

AMS subject classifications. 05C35, 05C85, 68R10, 68W40

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1. Introduction. Exact exponential algorithms. The interest in exact (fast) exponential algorithms dates back to Held and Karp’s paper [28] on the travelling salesman problem in the early 1960s. We mention just a few examples: an \( O(1.4422^n) \) time algorithm for knapsack (Horowitz and Sahni [29]); \( O(1.2600^n) \) and \( O(1.2109^n) \) time algorithms for independent set (Tarjan and Trojanowski [43] and Robson [40]); 3-coloring in time \( O(1.4422^n) \) (Lawler [35]); and 3-SAT in time \( O(1.6181^n) \) (Monien and Speckenmeyer [36]).

Nowadays, it is commonly believed that NP-hard problems cannot be solved in polynomial time. For a number of NP-hard problems, we even have strong evidence that they cannot be solved in subexponential time [30]. In order to obtain exact solutions to these problems, the only hope is to design exact algorithms with good exponential running times. In recent years there has been emerging interest in attacking this question for concrete combinatorial problems: there are, for example, an \( O^*(2^n) \) time algorithm for coloring (Björklund and Husfeldt [5] and Koivisto [34]); an \( O(1.3289^n) \) time algorithm for 3-coloring (Beigel and Eppstein [3]); an \( O(1.7325^n) \) time algorithm for Max-Cut (Williams [45]); an algorithm for 3-SAT in time \( O(1.4726^n) \) (Brueggemann and Kern [15]); and an \( O(1.5129^n) \) time algorithm for dominating set (Fomin, Grandoni, and Kratsch [23]).

There are several explanations for the current revival of interest in fast exponential algorithms within the algorithmic community.

• The design and analysis of exact algorithms leads to a better understanding of NP-hard problems and initiates interesting new combinatorial and algorithmic challenges.

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1Because \( c^n\cdot n^{O(1)} = O((c+\epsilon)^n) \) for any \( \epsilon > 0 \), we omit polynomial factors in the big-Oh notation every time we round the base of the exponent. We also use a modified big-Oh notation that suppresses all other (polynomially bounded) terms. Thus for functions \( f \) and \( g \) we write \( f(n) = O^*(g(n)) \) if \( f(n) = O(g(n) \cdot n^{O(1)}) \).
• For certain applications it is important to find exact solutions. With the increased speed of modern computers, fast algorithms, even though they have exponential running times in the worst case, may actually lead to practical algorithms for certain NP-hard problems, at least for moderate instance sizes.
• Approximation algorithms, randomized algorithms, and different heuristics are not always satisfactory. Each of these approaches has weak points such as the necessity of exact solutions, difficulty of approximation, limited power of the method itself, and many others.
• A reduction of the base of the exponential running time, say from \(O(2^n)\) to \(O(1.8^n)\), increases the size of the instances solvable within a given amount of time by a constant multiplicative factor. However, running a given exponential algorithm on a faster computer can enlarge the mentioned size by only a constant additive factor.

For overviews and introductions to the field see the recent surveys by Fomin, Grandoni, and Kratsch [24], Iwama [31], Schöning [41], and Woeginger [46, 47].

**Treewidth and minimum fill-in.** Treewidth is one of the most basic parameters in graph algorithms [7], and it plays an important role in structural graph theory. It serves as one of the main tools in Robertson and Seymour’s graph minors project [39]. Treewidth also plays a crucial role in parameterized complexity theory [19].

The problems of computing the treewidth and minimum fill-in of a graph are known to be NP-hard even when the input is restricted to complements of bipartite graphs (so called cobipartite graphs) [2, 48]. Despite the importance of treewidth almost nothing is known about how to cope with its intractability. For a long time the best known approximation algorithm for treewidth had a factor \(\log \text{OPT} [1, 11]\) (see also [10]). Recently, Feige, Hajiaghayi, and Lee [21] obtained a factor \(\sqrt{\log \text{OPT}}\) approximation algorithm for treewidth. Furthermore, it is an old open question whether the treewidth can be approximated within a constant factor.

Treewidth is known to be fixed parameter tractable. Moreover, for any fixed \(k\), there is a linear time algorithm to compute the treewidth of graphs of treewidth at most \(k\) (unfortunately there is a huge hidden constant in the running time) [6]. There is a number of algorithms that, for a given graph \(G\) and integer \(k\), either report that the treewidth of \(G\) is at least \(k\) or produce a tree decomposition of width at most \(c_1 k\) in time \(c_2 k \cdot n^{O(1)}\), where \(c_1, c_2\) are some constants (see, e.g., [1]). Fixed parameter algorithms are known for the minimum fill-in problem as well [16, 32].

There exists an exact \(O(2.9512^n)\) time algorithm that computes the treewidth of a graph in polynomial space [9]. We are not aware of any previous work on exact algorithms for the treewidth or minimum fill-in problem that solves the problem in \(O(c^n)\) time where \(c < 2\). There are three relatively simple approaches resulting in time \(O^*(2^n)\) algorithms:

- One can reformulate the problems as problems of finding special vertex elimination orderings and then find an optimal permutation by using the dynamic programming based technique as in the article of Held and Karp [28] for the travelling salesman problem. The algorithm of Bodlaender et al. [9] also uses this approach.
- With some modifications, the algorithm of Arnborg, Corneil, and Proskurowski [2] for a given \(k\) deciding in time \(O(n^{k+1})\) if the treewidth of a graph is at most \(k\) can be used to compute the treewidth (and similarly fill-in) in time \(O^*(2^n)\).
• Both problems can be solved by making use of the game theoretic approach, by finding a specific path in the graph of possible states of a cop and robber game [22]. However, it is not clear whether any of the aforementioned approaches can bring us to an $O(n^c)$ time algorithm for some $c < 2$. Prior to our work, no exact algorithm computing the treewidth or minimum fill-in of a graph in time $O(n^c)$ for some $c < 2$ was known.

Our results. In this paper we obtain the first exact algorithm computing the treewidth in time $O(n^c)$ for $c < 2$. Additionally it can be adapted to solve a number of other minimal triangulation problems such as minimum fill-in.

Our main result is an $O(1.8899^n)$ algorithm computing the treewidth and minimum fill-in of a graph on $n$ vertices. The algorithm can be regarded as dynamic programming across partial solutions and is based on results of Bouchitté and Todinca [13, 14]. The analysis of the running time is difficult and is based on combinatorial properties of special structures in a graph, namely, potential maximal cliques, i.e., vertex subsets in a graph that can be maximal cliques in some minimal triangulation of this graph. (See the next section for the definition.)

More precisely, first we modify the algorithm of Bouchitté and Todinca [13] which computes the treewidth and minimum fill-in of a graph on $n$ vertices. The algorithm can be regarded as dynamic programming across partial solutions and is based on results of Bouchitté and Todinca [13, 14]. The analysis of the running time is difficult and is based on combinatorial properties of special structures in a graph, namely, potential maximal cliques, i.e., vertex subsets in a graph that can be maximal cliques in some minimal triangulation of this graph. (See the next section for the definition.)

More precisely, first we modify the algorithm of Bouchitté and Todinca [13] which computes the treewidth and minimum fill-in of a graph $G$ with the given set $\Pi_G$ of all potential maximal cliques of $G$ and then improve the analysis of its running time to obtain an $O^*(|\Pi_G|)$ time complexity. The most technical and difficult part of the paper is the proof that all potential maximal cliques can be listed in time $O(1.8899^n)$. Very roughly, our listing algorithm is based on the following combinatorial result: every “large” potential maximal clique either is “almost” a minimal separator or can be represented by a “small” vertex subset. The technique developed to prove this combinatorial result can be interesting on its own.

For several special graph classes, for which both problems remain NP-complete, we are able to prove that our approach guarantees significantly better bounds. To exemplify this we show that, for the class of asteroidal triple (AT)-free graphs, the number of minimal separators and the number of potential maximal cliques, and thus the running time of our algorithm, is $O^*(2^n/2)$.

This paper is organized as follows. In section 2 we give basic definitions. In section 3 we show how Bouchitté and Todinca’s approach can be used to compute the treewidth and fill-in in time linear in the number of potential maximal cliques. In section 4 we prove that every graph on $n$ vertices has $O(n \cdot 1.7087^n)$ minimal separators. In section 5 we show that an $n$-vertex graph contains $O(1.8135^n)$ potential maximal cliques. This bound is of only combinatorial interest because it is not constructive, in a sense, that we do not know how to use this bound to list all potential maximal cliques in time $O(1.8135^n)$. In order to obtain a fast algorithm computing the treewidth and the fill-in of a graph, we need an algorithm listing all potential maximal cliques. In the remaining part of section 5 we derive the most difficult and important algorithmic result of this paper, namely, that all potential maximal cliques of a graph can be listed in time $O(1.8899^n)$. This result is based on a novel characterization of potential maximal cliques. Combined with the results from section 3, this yields the main result of the paper, that the treewidth and minimum fill-in can be computed in time $O(1.8899^n)$. In section 6 we design a faster $O^*(2^n/2)$ time algorithm for AT-free graphs. We conclude with open problems and final remarks in section 7.

2. Basic definitions. We denote by $G = (V, E)$ a finite, undirected, and simple graph with $|V| = n$ vertices and $|E| = m$ edges. For any nonempty subset $W \subseteq V$,
the subgraph of $G$ induced by $W$ is denoted by $G[W]$. For $S \subseteq V$ we often use $G \setminus S$ to denote $G[V \setminus S]$. The neighborhood of a vertex $v$ is $N(v) = \{u \in V : \{u, v\} \in E\}$, $N[v] = N(v) \cup \{v\}$, and for a vertex set $S \subseteq V$ we set $N(S) = \bigcup_{v \in S} N(v) \setminus S$, $N[S] = N(S) \cup S$. A clique $C$ of a graph $G$ is a subset of $V$ such that all the vertices of $C$ are pairwise adjacent. By $\omega(G)$ we denote the maximum clique-size of a graph $G$.

**Treewidth and minimum fill-in of graphs.** The notion of treewidth is due to Robertson and Seymour [38]. A tree decomposition of a graph $G = (V, E)$, denoted by $TD(G)$, is a pair $(X, T)$ in which $T = (V_T, E_T)$ is a tree and $X = \{X_i \mid i \in V_T\}$ is a family of subsets of $V$ such that

(i) $\bigcup_{i \in V_T} X_i = V$;

(ii) for each edge $e = \{u, v\} \in E$ there exists an $i \in V_T$ such that both $u$ and $v$ belong to $X_i$; and

(iii) for all $v \in V$, the set of nodes $\{i \in V_T \mid v \in X_i\}$ induces a connected subtree of $T$.

The maximum of $|X_i| - 1$, $i \in V_T$, is called the width of the tree decomposition. The treewidth of a graph $G$, denoted by $tw(G)$, is the minimum width taken over all tree decompositions of $G$.

A graph $H$ is chordal (or triangulated) if every cycle of length at least four has a chord, i.e., an edge between two nonconsecutive vertices of the cycle. A triangulation of a graph $G = (V, E)$ is a chordal graph $H = (V, E')$ such that $E \subseteq E'$. $H$ is a minimal triangulation if, for any intermediate set $E''$ with $E \subseteq E'' \subset E'$, the graph $F = (V, E'')$ is not chordal.

The following result is very useful for our algorithms.

**Theorem 2.1** (folklore). For any graph $G$, $tw(G) \leq k$ if and only if there is a triangulation $H$ of $G$ such that $\omega(H) \leq k + 1$.

Thus the treewidth of a graph $G$ can be defined as the minimum of $\omega(H) - 1$ taken over all triangulations $H$ of $G$, of $\omega(H) - 1$.

The minimum fill-in of a graph $G = (V, E)$, denoted by $mfi(G)$, is the smallest value of $|E_H - E|$, where the minimum is taken over all triangulations $H = (V, E_H)$ of $G$.

In other words, computing the treewidth of $G$ means finding a (minimal) triangulation with the smallest maximum clique-size, while computing the minimum fill-in means finding a (minimal) triangulation with the smallest number of edges. Clearly, in both cases it is sufficient to consider only minimal triangulations of $G$, which makes minimal separators and potential maximal cliques important tools of our algorithmic approach.

**Minimal separators.** Minimal separators and potential maximal cliques are the most important tools used in our proofs. Let $a$ and $b$ be two nonadjacent vertices of a graph $G = (V, E)$. A set of vertices $S \subseteq V$ is an $a, b$-separator if $a$ and $b$ are in different connected components of the graph $G \setminus S$. A connected component $C$ of $G \setminus S$ is a full component (associated to $S$) if $N(C) = S$. $S$ is a minimal $a, b$-separator of $G$ if no proper subset of $S$ is an $a, b$-separator. We say that $S$ is a minimal separator of $G$ if there are two vertices $a$ and $b$ such that $S$ is a minimal $a, b$-separator. Notice that a minimal separator can be strictly included in another one. We denote by $\Delta_G$ the set of all minimal separators of $G$. A set of vertices $\Omega \subseteq V$ of a graph $G$ is called a potential maximal clique if there is a minimal triangulation $H$ of $G$ such that $\Omega$ is a maximal clique of $H$. We denote by $\Pi_G$ the set of all potential maximal cliques of $G$. Clearly, $|\Delta_G| \leq 2^n$ and $|\Pi_G| \leq 2^n$ for every graph $G$ on $n$ vertices, and no better upper bounds had been known prior to our work.
The following result will be used to list all minimal separators of a graph.

**Theorem 2.2** (see [4]). There is an algorithm listing all minimal separators of an input graph \( G \) in \( \mathcal{O}(n^3|\Delta_G|) \) time.

There is a very useful relationship between the minimal separators of a graph and its minimal triangulations. Two minimal separators \( S \) and \( T \) of a graph \( G \) are said to be crossing if \( S \) is a minimal \( u,v \)-separator for a pair of vertices \( u,v \in T \), in which case \( T \) is a minimal \( x,y \)-separator for a pair \( x, y \in S \). (See [33] and [37] for a full proof.)

**Theorem 2.3** (see [37]). The graph \( H \) is a minimal triangulation of the graph \( G \) if and only if there is a maximal set of pairwise noncrossing minimal separators \( \{S_1, S_2, \ldots, S_p\} \) of \( G \) such that \( H \) can be obtained from \( G \) by completing each \( S_i \), \( i \in \{1,2,\ldots,p\} \), into a clique.

Although we do not use this characterization explicitly it is fundamental for our paper.

**Potential maximal cliques.** The following structural characterization of potential maximal cliques is extremely useful for our purposes.

For a set \( K \subseteq V \), a connected component \( C \) of \( G \setminus K \) is a **full component associated to** \( K \) if \( N(C) = K \).

**Theorem 2.4** (see [13]). Let \( K \subseteq V \) be a set of vertices of the graph \( G = (V,E) \). Let \( C(K) = \{C_1(K), \ldots, C_p(K)\} \) be the set of the connected components of \( G \setminus K \) and let \( S(K) = \{S_1(K), S_2(K), \ldots, S_p(K)\} \), where \( S_i(K), i \in \{1,2,\ldots,p\} \), is the set of those vertices of \( K \) which are adjacent to at least one vertex of the component \( C_i(K) \). Then \( K \) is a potential maximal clique of \( G \) if and only if

1. \( G \setminus K \) has no full component associated to \( K \), and
2. the graph on the vertex set \( K \) obtained from \( G[K] \) by completing each \( S_i \in S(K) \) into a clique is a complete graph.

Moreover, if \( K \) is a potential maximal clique, then \( S(K) \) is the set of the minimal separators of \( G \) contained in \( K \).

**Remark 2.5.** By Theorem 2.4, for every potential maximal clique \( \Omega \) of \( G \), the sets \( S_i(\Omega) \) are exactly the minimal separators of \( G \) contained in \( \Omega \). For each minimal separator \( S_i = S_i(\Omega) \), all vertices of \( \Omega \setminus S_i \) are contained in the same component of \( G \setminus S_i \).

The following result is an easy consequence of Theorem 2.4.

**Theorem 2.6** (see [13]). There is an algorithm that, given a graph \( G = (V,E) \) and a set of vertices \( K \subseteq V \), verifies if \( K \) is a potential maximal clique of \( G \). The time complexity of the algorithm is \( \mathcal{O}(nm) \).

According to [14], the number of potential maximal cliques of a graph \( G \) is at least \( |\Delta_G|/n \) and at most \( n|\Delta_G|^2 + n|\Delta_G| + 1 \).

**3. Computing treewidth and minimum fill-in.** We describe a modification of the algorithm of [13] that, given a graph, all its minimal separators, and all its potential maximal cliques, computes the treewidth and the minimum fill-in of the graph. The running time stated in [13] could be reformulated as \( \mathcal{O}(n^2 |\Delta_G| \cdot |\Pi_G|) \). We show how the algorithm can be implemented to run in time \( \mathcal{O}(n^3 \cdot |\Pi_G|) \).

For a minimal separator \( S \) and a component \( C \in C(S) \) of \( G \setminus S \), we say that \( (S,C) \) is a **block** associated to \( S \). We sometimes use the notation \( (S,C) \) to denote the set of vertices \( S \cup C \) of the block. It is easy to notice that if \( X \subseteq V \) corresponds to the set of vertices of a block, then this block \( (S,C) \) is unique: indeed, \( S = N(V \setminus X) \) and \( C = X \setminus S \).
A block \((S, C)\) is called \(\textit{full}\) if \(C\) is a full component associated to \(S\). The graph \(R(S, C) = G_S[S \cup C]\) obtained from \(G[S \cup C]\) by completing \(S\) into a clique is called the \emph{realization} of the block \(B\).

**Theorem 3.1** (see [33]). Let \(G\) be a noncomplete graph. Then

\[
\text{tw}(G) = \min_{S \in \Delta_G} \max_{C \in \mathcal{C}(S)} \text{tw}(R(S, C)),
\]

\[
\text{mfi}(G) = \min_{S \in \Delta_G} \left( \text{fill}(S) + \sum_{C \in \mathcal{C}(S)} \text{mfi}(R(S, C)) \right),
\]

where \(\text{fill}(S)\) is the number of nonedges of \(G[S]\).

**Remark 3.2.** In the equations of Theorem 3.1 we may take the minimum only over the inclusion-minimal separators of \(G\). Then all the blocks in the equations are full.

Unfortunately, Theorem 3.1 is not sufficient for computing the treewidth and the minimum fill-in. Therefore we now express the treewidth and the minimum fill-in of realizations of full blocks from realizations of smaller full blocks. Let \(\Omega\) be a potential maximal clique of \(G\). We say that a block \((S', C')\) is \emph{associated} to \(\Omega\) if \(C'\) is a component of \(G \setminus \Omega\) and \(S' = N(C')\).

**Theorem 3.3** (see [13]). Let \((S, C)\) be a full block of \(G\). Then

\[
\text{tw}(R(S, C)) = \min_{S \subset \Omega \subseteq (S, C)} \max(|\Omega| - 1, \text{tw}(R(S_i, C_i))),
\]

\[
\text{mfi}(R(S, C)) = \min_{S \subset \Omega \subseteq (S, C)} \left( \text{fill}(\Omega) - \text{fill}(S) + \sum \text{mfi}(R(S_i, C_i)) \right),
\]

where the minimum is taken over all potential maximal cliques \(\Omega\) such that \(S \subset \Omega \subseteq (S, C)\) and \((S_i, C_i)\) are the blocks associated to \(\Omega\) in \(G\) such that \(S_i \cup C_i \subset S \cup C\).

**Theorem 3.4.** There is an algorithm that, given a graph \(G\) together with the list of its minimal separators \(\Delta_G\) and the list of its potential maximal cliques \(\Pi_G\), computes the treewidth and the minimum fill-in of \(G\) in \(O(n^3 |\Pi_G|)\) time. Moreover, the algorithm constructs optimal triangulations for the treewidth and the minimum fill-in.

**Proof.** W.l.o.g. we may assume that the input graph \(G\) is connected (otherwise we can run the algorithm for each connected component of \(G\)).

The algorithm for computing the treewidth and the minimum fill-in of a graph, using its minimal separators and its potential maximal cliques, is given below. It is a slightly different version of the algorithm given in [13].

**Input:** \(G\), all its potential maximal cliques and all its minimal separators

**Output:** \(\text{tw}(G)\) and \(\text{mfi}(G)\)

**begin**

compute all the full blocks \((S, C)\) and sort them by the number of vertices

for each full block \((S, C)\) taken in increasing order

\[
\text{tw}(R(S, C)) := |S \cup C| - 1 \text{ if } (S, C) \text{ is inclusion-minimal}
\]

and \(\text{tw}(R(S, C)) := \infty\) otherwise

\[
\text{mfi}(R(S, C)) := \text{fill}(S \cup C) \text{ if } (S, C) \text{ is inclusion-minimal}
\]

and \(\text{mfi}(R(S, C)) := \infty\) otherwise

for each p.m.c. \(\Omega\) with \(S \subset \Omega \subseteq (S, C)\)

**end**
compute the blocks \((S_i, C_i)\) associated to \(\Omega\) s.t. \(S_i \cup C_i \subset S \cup C\)
\[
tw(R(S, C)) := \min(\tw(R(S, C)), \\
\max(|\Omega| - 1, \tw(R(S, C_i))))
\]
\[
mfi(R(S, C)) := \min(mfi(R(S, C))), \\
\fill(\Omega) - \sum_{i}(mfi(R(S, C_i)))
\]
\end{for}

For the sake of completeness we shortly discuss the correctness proof. According to Theorem 3.3, at the end of the outer for loop the values of \(\tw(R(S, C))\) and \(mfi(R(S, C))\) are correctly computed for each full block \((S, C)\) of \(G\). Then the treewidth and the minimum fill-in of the graph are computed using Theorem 3.1 and the fact that in Theorem 3.1 one can work only with inclusion-minimal separators.

Let us show how the algorithm can be implemented such that its running time is \(O(n^3 \cdot |\Pi_G|)\).

To store and manipulate the minimal separators, potential maximal cliques, and blocks we use data structures that allow us to search, to insert, or to check whether an element is inclusion-minimal in \(O(n)\) time.

During a preprocessing step, we realize the following operations.

- Compute the list of all full blocks and, for each minimal separator \(S\), store a pointer towards each full block of type \((S, C)\). These operations are performed as follows. For each minimal separator \(S\), we compute the connected components of \(G \setminus S\). For each such component \(C\), if \(N(C) = S\), then the block \((S, C)\) is full, so we add it to the list of full blocks and store the pointer from \(S\) to \((S, C)\). Note that this procedure will generate all the full blocks, and each of them is encountered exactly once. For a given minimal separator \(S\), there are at most \(n\) full blocks associated to it, and thus at most \(n\) pointers to be stored. The insertion of these blocks into the list of full blocks requires \(O(n)\) time for each block. Hence the whole step costs \(O(n^2 |\Delta_G|)\) time.

- For each potential maximal clique \(\Omega\), store a pointer to each full block associated to it as follows: compute the components \(C_i\) of \(G \setminus \Omega\), and then \((N(C_i), C_i)\) are precisely the blocks associated to \(\Omega\). In particular there are at most \(n\) such blocks. This computation can be done in \(O(n^2)\) time for each potential maximal clique, hence globally in \(O(n^2 |\Pi_G|)\) time.

- Compute all the good triples \((S, C, \Omega)\), where \((S, C)\) is a full block and \(\Omega\) is a potential maximal clique such that \(S \subset \Omega \subseteq S \cup C\). Moreover, for each good triple we store a pointer from \((S, C)\) to \(\Omega\). By Theorem 2.4, there are at most \(n\) minimal separators \(S \subset \Omega\), each of them being the neighborhood of a component of \(G \setminus \Omega\), and for each such \(S\) there is exactly one component \(G' \setminus S\) intersecting \(\Omega\) (in particular there are at most \(n|\Pi_G|\) good triples). For each component \(C'\) of \(G' \setminus \Omega\) we take \(S = N(C')\), find the component \(C\) of \(G \setminus S\) intersecting \(\Omega\), and store the pointer from \((S, C)\) to \(\Omega\). Thus this computation takes \(O(nm)\) time for each potential maximal clique, hence \(O(nm |\Pi_G|)\) globally.

Hence this preprocessing step costs \(O(n^2 |\Delta_G| + nm |\Pi_G|)\). Sorting the blocks by their size can be done in \(O(n |\Delta_G|)\) time using a bucket sort.
Observe that the cost of one iteration of the inner for loop is \( \mathcal{O}(n^2) \), by the fact that there are at most \( n \) blocks associated to a potential maximal clique. With the data structures obtained during the preprocessing step, each full block \((S,C)\) keeps a pointer towards each potential maximal clique \( \Omega \) such that \((S,C,\Omega)\) form a good triple. Thus the number of iterations of the two nested loops is exactly the number of good triples, that is, at most \( n|\Pi_G|\). It follows that the two loops cost \( \mathcal{O}(n^2|\Pi_G|) \) time.

After the execution of the loops, computing the set \( \Delta_G^* \) of inclusion-minimal separators costs \( \mathcal{O}(n|\Delta_G^*|) \) time. Each inclusion-minimal separator \( S \) keeps the list of its associated blocks, obtained during the preprocessing step. Computing the maximum required by the two last instructions costs \( \mathcal{O}(n) \) time for a given \( S \). This last step costs \( \mathcal{O}(n|\Delta_G^*|) \) time.

Altogether, the algorithm runs in time \( \mathcal{O}(n^2 \cdot |\Delta_G| + n^3 \cdot |\Pi_G|) \). It is known [14] that each minimal separator is contained in at least one potential maximal clique. According to Theorem 2.4, each potential maximal clique contains at most \( n \) minimal separators. Therefore \( |\Pi_G| \geq |\Delta_G|/n \). We conclude that the algorithm runs in \( \mathcal{O}(n^3 \cdot |\Pi_G|) \) time.

The algorithm can be easily transformed in order to output not only the treewidth and the minimum fill-in of the graph, but also optimal triangulations with respect to these parameters. It is sufficient to keep, for each full block, the set of potential maximal cliques realizing the minimum treewidth and fill-in of its realization. At the end of the algorithm, the potential maximal cliques of the chosen blocks will be the maximal cliques of the computed optimal triangulation: optimal tree decomposition or minimum triangulation. \( \square \)

Using Theorem 3.4, the only missing ingredient of our treewidth and minimum fill-in algorithms is an algorithm listing all (minimal separators and) potential maximal cliques of a graph in time \( \mathcal{O}^*(c^n) \) for some \( c < 2 \). This requires exponential upper bounds of the type \( \mathcal{O}^*(c^n) \) for some \( c < 2 \) for the number of minimal separators and for the number of potential maximal cliques in a graph on \( n \) vertices. In the next two sections we discuss this issue.

### 4. The number of minimal separators

In this section we show that any graph with \( n \) vertices has \( \mathcal{O}(1.7087^n) \) minimal separators. For the main algorithm of this paper the upper bound \( \mathcal{O}(1.8899^n) \) would be sufficient. However, bounding the number of minimal separators in a graph is a nice combinatorial problem, and we prefer to give here the best upper bound we were able to find.

Let \( S \) be a separator in a graph \( G = (V,E) \). For \( x \in V \setminus S \), we denote by \( C_x(S) \) the component of \( G \setminus S \) containing \( x \). The following lemma is an exercise in [27].

**Lemma 4.1 (folklore).** A set \( S \) of vertices of \( G \) is a minimal \( a,b \)-separator if and only if \( a \) and \( b \) are in different full components associated to \( S \). In particular, \( S \) is a minimal separator if and only if there are at least two distinct full components associated to \( S \).

Here is one of the main combinatorial results of our paper.

**Theorem 4.2.** For any graph \( G \), \( |\Delta_G| = \mathcal{O}(1.7087^n) \).

**Proof.** For a constant \( \alpha \), \( 0 < \alpha < 1 \), we distinguish two types of minimal separators: small separators, of size at most \( \alpha n \), and big separators, of size more than \( \alpha n \). We denote the cardinalities of these sets by \( \#\text{small sep} \) and \( \#\text{big sep} \). Notice that \( |\Delta_G| = \#\text{small sep} + \#\text{big sep} \).

The number of big separators. Let \( S \) be a minimal separator. By Lemma 4.1, there are at least two full components associated to \( S \). Hence at least one of these full
components has at most \( n(1 - \alpha)/2 \) vertices. For every \( S \in \Delta_G \) we choose one of these full components and call it the small component of \( S \), denoted by \( s(S) \).

By the definition of a full component, \( S = N(s(S)) \). In particular, for distinct minimal separators \( S \) and \( T \), we have that \( s(S) \neq s(T) \). Therefore the number \#big sep of big minimal separators is at most the number of small components, and we conclude that \#big sep does not exceed the number of subsets of \( V \) of cardinality at most \( n(1 - \alpha)/2 \); i.e.,

\[
\#\text{big sep} \leq \sum_{i=1}^{\lceil n(1-\alpha)/2 \rceil} \binom{n}{i}.
\]

By making use of Stirling’s formula we deduce that

\[
\#\text{big sep} \leq \frac{n(1-\alpha)}{2} \left( \pi n(1-\alpha) \frac{1+\alpha}{2} \right)^{-\frac{1}{2}} \left[ \left( \frac{1-\alpha}{2} \right)^{-\frac{1+\alpha}{2}} \left( \frac{1+\alpha}{2} \right)^{-\frac{1+\alpha}{2}} \right]^n.
\]

The number of small separators. To count small separators we use a different technique. Let \( S \) be a minimal separator, let \( x \) be a vertex of a full component \( C_x(S) \) associated to \( S \) with minimum number of vertices, and let \( X \subseteq V \) be a vertex subset. We say that \((x, X)\) is a bad pair associated to \( S \) if \( C_x(S) \subseteq X \subseteq V \setminus S \).

**Claim 1.** Let \( S \neq T \) be two minimal separators and let \((x, X)\) and \((y, Y)\) be two bad pairs associated to \( S \) and \( T \), respectively. Then \((x, X) \neq (y, Y)\).

**Proof.** Since \( C_x(S) \subseteq X \) and \( X \cap S = \emptyset \), we have that the connected component of \( G[X] \) containing \( x \) is \( C_x(S) \). Similarly, the connected component of \( G[Y] \) containing \( y \) is \( C_y(T) \).

Thus if \( x = y \) and \( X = Y \), then \( C_x(S) = C_y(T) \). Since \( C_x(S) \) is a full component associated to \( S \) in \( G \), we have that \( S = N(C_x(S)) \) and \( T = N(C_y(T)) \). Therefore \( S = T \), which is a contradiction. \( \square \)

By Lemma 4.1, there are at least two full components associated to every small separator \( S \). For a full component \( C_x(S) \) associated to \( S \) with the minimum number of vertices, \( |V \setminus (S \cup C_x(S))| \geq n \cdot (1 - \alpha)/2 \). For any \( Z \subseteq V \setminus (S \cup C_x(S)) \), the pair \((x, Z \cup C_x(S))\) is a bad pair associated to \( S \). Therefore there are at least \( 2^n \cdot (1 - \alpha)/2 \) distinct bad pairs associated to \( S \). Hence by Claim 1, the total number of bad pairs is at least \#small sep \cdot \( 2^n \cdot (1 - \alpha)/2 \). On the other hand, the number of bad pairs is at most \( n \cdot 2^n \). We conclude that

\[
\#\text{small sep} \leq n 2^{n(1-\alpha)/2}.
\]

Finally, choosing \( \alpha = 0.5456 \), we obtain

\[
|\Delta_G| = \#\text{small sep} + \#\text{big sep} = \calO(n \cdot 1.7087^n).
\]

Let us note that, by Theorem 2.2, Theorem 4.2 yields that all minimal separators of a graph can be listed in time \( \calO(1.7087^n) \).

**5. The number of potential maximal cliques.** In this section we prove that the number of potential maximal cliques in a graph with \( n \) vertices is \( \calO(1.8135^n) \) and then show that there exists an algorithm to list all potential maximal cliques of any graph in time \( \calO(1.8899^n) \).

We bound the number of potential maximal cliques by counting specific potential maximal cliques called nice potential maximal cliques. Later these nice potential maximal cliques are used to generate and to bound the number of all potential maximal cliques.
DEFINITION 5.1. Let $\Omega$ be a potential maximal clique of a graph $G$, and let $S \subset \Omega$ be a minimal separator of $G$. We say that $S$ is an active separator for $\Omega$ if $\Omega$ is not a clique in the graph $G_{S(\Omega) \setminus \{S\}}$, obtained from $G$ by completing all the minimal separators contained in $\Omega$ except $S$. If $S$ is active, a pair of vertices $x,y \in S$ nonadjacent in $G_{S(\Omega) \setminus \{S\}}$ is called an active pair. Otherwise, $S$ is called inactive for $\Omega$.

DEFINITION 5.2. We say that a potential maximal clique $\Omega$ is nice if at least one of the minimal separators contained in $\Omega$ is active for $\Omega$.

We define $\Pi_n$ as the maximum number of nice potential maximal cliques in a graph on $n$ vertices. By the theorem and lemma below, the number of potential maximal cliques is polynomially bounded by the number of nice potential maximal cliques and minimal separators.

Theorem 5.3 (see [14]). Let $\Omega$ be a potential maximal clique of $G$, and let a be a vertex of $G$ and $G' = G \setminus \{a\}$. Then one of the following cases holds:
1. either $\Omega$ or $\Omega \setminus \{a\}$ is a potential maximal clique of $G'$;
2. $\Omega = S \cup \{a\}$, where $S$ is a minimal separator of $G$;
3. $\Omega$ is nice.

Lemmas 5.4. A graph $G$ on $n$ vertices has at most $n^2 |\Delta_G| + n\Pi_n$ potential maximal cliques.

Proof. Let $x_1,x_2,\ldots,x_n$ be the vertices of $G$ and $G_i = G[x_1,\ldots,x_i]$ for all $i \in \{1,2,\ldots,n\}$. By Theorem 5.3, for each $i \in \{2,3,\ldots,n\}$, $|\Pi_{G_i}| \leq |\Pi_{G_{i-1}}| + n|\Delta_{G_i}| + |\Pi_i|$. By [14, Corollary 4], $|\Delta_{G_i}| \leq |\Delta_G|$ for any $i \in \{1,\ldots,n\}$. This yields that

$$|\Pi_{G_n}| \leq \sum_{i=1}^n n|\Delta_{G_i}| + \Pi_1 \leq n^2 |\Delta_G| + n\Pi_n. \quad \Box$$

5.1. Nonconstructive upper bound on the number of potential maximal cliques. We show that the number of potential maximal cliques in a graph is $O(1.8135^n)$. This bound is obtained by finding an upper bound on the number of nice potential maximal cliques. We do this by computing two numbers (as for the separator bound): the number of nice potential maximal cliques of size less than $an$ and the number of nice potential maximal cliques of size at least $an$ for $0 < \alpha < 1$.

Definition 5.5. We say that the pair $(Z,z)$ is a vertex representation of a potential maximal clique $\Omega$ in $G$ where $Z \subset V$ and $z \in Z$ if $\Omega = N(Z) \cup \{z\}$ and $G[Z]$ is connected.

Lemma 5.6. Let $\Omega$ be a potential maximal clique of $G$ and let $z \in \Omega$. Then $(Z,z)$ is a vertex representation of $\Omega$ if and only if $Z$ is the vertex set of the connected component of $G \setminus (\Omega \setminus \{z\})$ containing $z$.

Proof. Suppose that $(Z,z)$ is a vertex representation of $\Omega$. Vertex $z$ is contained in $Z$, and thus every neighbor of $z$ not in $\Omega$ has to be contained in $Z$. By applying this argument recursively, every connected component $C$ of $G \setminus \Omega$, where $z \in N(C)$, is contained in $Z$. On the other hand, these are the only vertices in $Z$, because the rest are separated by $\Omega \setminus \{z\}$.

Conversely, let $Z$ be the vertex set of the component of $G \setminus (\Omega \setminus \{z\})$ containing $z$. Clearly $N(Z) \cup \{z\}$ is contained in $\Omega$, so it remains to prove that any $x \in \Omega \setminus \{z\}$ is contained in $N(Z)$. By Theorem 2.4, $x$ is adjacent to $z$ or there is a component $C$ of $G \setminus \Omega$ such that both $x$ and $z$ are in $N(C)$. Since $C \subset Z$ the conclusion follows. \hfill \Box

Lemma 5.7. Let $\Omega$ be a nice potential maximal clique of size $an$. Then there exists a vertex representation $(U,u)$ of $\Omega$ such that $|U| \leq 2n(1-\alpha)/3 + 1$. 

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Proof. Let $S$ be a minimal separator active for $\Omega$, let $x, y \in S$ be an active pair, and let $z$ be a vertex contained in $\Omega \setminus S$. By Lemma 5.6, every vertex in a potential maximal clique defines a unique vertex representation. Let $(X, x), (Y, y), (Z, z)$ be the unique vertex representations defined by $\Omega$ and, respectively, $x, y$, and $z$ (Lemma 5.6).

Let us now prove that one of the sets $X, Y, Z$ contains at most $2n(1 - \alpha)/3 + 1$ vertices. By Lemma 5.6, we can observe that if $(U, u)$ is a vertex representation of $\Omega$, then $G[U \setminus \{u\}]$ is formed exactly by the set of connected components of $G \setminus \Omega$ having $u$ in their neighborhood. We partition the connected components of $G \setminus \Omega$ into three sets:

- $A_1 = (X \setminus \{x\}) \cap (Y \setminus \{y\})$,
- $A_2 = (X \setminus \{x\}) \setminus (Y \setminus \{y\})$, and
- $A_3 = (V \setminus \Omega) \setminus (A_1 \cup A_2)$.

Let us emphasize that

- $|A_1 \cup A_2 \cup A_3| = n(1 - \alpha)$ (because $A_1 \cup A_2 \cup A_3 = V \setminus \Omega$ and $|\Omega| = \alpha n$),
- the sets $A_1, A_2, A_3$ are pairwise disjoint,
- $X \setminus \{x\} = A_1 \cup A_2$,
- $Y \setminus \{y\} \subseteq A_1 \cup A_3$, and
- $Z \setminus \{z\} \subseteq A_2 \cup A_3$ (by construction, $G[A_1]$ is the union of components of $G \setminus \Omega$ that see both $x$ and $y$; since $S$ is an active separator for $\Omega$, $x, y$ is an active pair, and $z \notin S$, it follows by Definition 5.1 that none of these components can see $z$; thus $A_1 \cap Z = \emptyset$).

One of the vertex sets $A_1, A_2, A_3$, say $A_1$, is of size at least $n(1 - \alpha)/3$; then $|A_2| + |A_3| \leq 2n(1 - \alpha)/3$. Since $Z \setminus \{z\} \subseteq A_2 \cup A_3$, we have that $|Z \setminus \{z\}| \leq 2n(1 - \alpha)/3$, and thus there exists a vertex representation $(U, z) = (Z, z)$ of $\Omega$ such that $|U| \leq (2n(1 - \alpha)/3) + 1$. □

Lemma 5.8. For a constant $0 < \alpha < 1$ and a graph $G$, the number $\Pi_{\geq \alpha}$ of nice potential maximal cliques of size at least $\alpha n$ is at most $n \sum_{i=1}^{2n(1-\alpha)/3} \binom{n}{i}$.

Proof. By Lemma 5.7, every potential maximal clique $\Omega$ of size at least $\alpha n$ has a vertex representation $(X, x)$ such that $|X \setminus \{x\}| \leq 2n(1 - \alpha)/3$. Thus $\Pi_{\geq \alpha}$ is at most the number of pairs $(X, x)$, where $|X \setminus \{x\}| \leq 2n(1 - \alpha)/3$ and $x \in V \setminus X$, which is at most $n \sum_{i=1}^{2n(1-\alpha)/3} \binom{n}{i}$. □

Lemma 5.9. For a constant $0 < \alpha < 1$ and a graph $G$, the number $\Pi_{< \alpha}$ of nice potential maximal cliques of size less than $\alpha n$ is at most $n \cdot 2^{n(2+\alpha)/3}$.

Proof. We say that $(x, X)$ is a bad pair associated to $\Omega$ if $\Omega = N(C_x) \cup \{x\}$, where $C_x$ is the connected component of $G[X \cup \{x\}]$ containing $x$.

To prove that a bad pair is unique for a potential maximal clique, we let $(x, X)$ be a bad pair associated to $\Omega_x$, and let $(y, Y)$ be a bad pair associated to $\Omega_y$, where $\Omega_x \neq \Omega_y$. We claim that $(x, X) \neq (y, Y)$. Targeting a contradiction, we assume that $x = y$ and that $X = Y$. From the definition of a bad pair, we know that $N(X) \cup \{x\} = N(Y) \cup \{y\}$. But this is a contradiction because $N(X) \cup \{x\} = \Omega_x$, $N(Y) \cup \{y\} = \Omega_y$, and $\Omega_x \neq \Omega_y$.

By Lemma 5.7, every potential maximal clique $\Omega$ of size less than $\alpha n$ has a vertex representation $(U, u)$ such that $|V \setminus (\Omega \cup U)| \geq n(1 - \alpha)/3$. Thus we can create $2^n(1-\alpha)/3$ unique bad pairs $(u, X)$ for $\Omega$ by selecting $X = U \cup Z$, where $Z$ is any of the $2^n(1-\alpha)/3$ subsets of $V \setminus N[U]$. The number of bad pairs is at most $n \cdot 2^n$, and we get that $n \cdot 2^n \geq \Pi_{< \alpha} \cdot 2^{n(1-\alpha)/3}$. □

Lemma 5.10. The number of nice potential maximal cliques in a graph $G$ with $n$ vertices is $O(1.8135^n)$.

Proof. The number of nice potential maximal cliques $\Pi_n$ is at most $\Pi_{\geq \alpha} + \Pi_{< \alpha}$ for $0 \leq \alpha \leq 1$. By using Lemmas 5.8 and 5.9, we have that $\Pi_n \leq n \cdot \sum_{i=1}^{2n(1-\alpha)/3} \binom{n}{i}$.
By making use of Stirling’s formula for $\alpha = 0.5763$, we obtain the bound $O(1.81349^n)$.

**Theorem 5.11.** For every graph $G$ on $n$ vertices, $|\Pi_G| = O(1.8135^n)$.

**Proof.** By Lemma 5.4, $G$ has at most $n^2|\Delta_G| + n\Pi_n$ potential maximal cliques. By Theorem 4.2 and Lemma 5.10, this yields that $\Pi_G = O(n^21.7087^n + n1.81349^n) = O(1.8135^n)$.

### 5.2. Listing potential maximal cliques

Notice that the proof of Lemma 5.9 is nonconstructive; i.e., the proof cannot be turned into an algorithm listing potential maximal cliques, which is required by the algorithms computing treewidth and fill-in.

Roughly speaking, the idea of this subsection is to show that each potential maximal clique of a graph can be identified by a set of vertices of size at most $n/3$. The algorithm for generating all the potential maximal cliques of a graph lists all the sets of vertices of size at most $n/3$ and then, by applying a polynomial time procedure for each set, generates all the potential maximal cliques of the input graph. A potential maximal clique can be recognized by the following three representations.

**Definition 5.12.** Let $\Omega$ be a potential maximal clique of $G$. The triple $(S,a,b)$ is called a separator representation of $\Omega$ if $S$ is a minimal separator of $G$, $a \in S$, $b \in V \setminus S$, and $\Omega = S \cup (N(a) \cap C_b)$, where $C_b$ is the component of $G \setminus S$ containing $b$.

The number of all possible separator representations of a graph is at most $n^2|\Delta_G|$. Unfortunately, not every nice potential maximal clique has a separator representation. The two definitions below allow us to represent a potential maximal clique by using a small vertex set.

**Definition 5.13.** For a potential maximal clique $\Omega$ of $G$, we say that a pair $(X,c)$, where $X \subset V$ and $c \in X$, is a partial representation of $\Omega$ if $\Omega = N(C_c) \cup (X \setminus C_c)$, where $C_c$ is the connected component of $G[X]$ containing $c$.

**Definition 5.14.** For a potential maximal clique $\Omega$ of $G$, we say that a triple $(X,x,c)$, where $X \subset V$ and $x,c \notin X$, is an indirect representation of $\Omega$ if $\Omega = N(C_c \cup D_x \cup \{x\}) \cup \{x\}$, where

- $C_c$ is the connected component of $G \setminus N[X]$ containing $c$;
- $D_x$ is the vertex set of the union of all connected components $C'$ of $G[X]$ such that $x \in N(C')$.

Let us note that for a given vertex set $X$ and two vertices $x,c$ one can check in polynomial time whether the pair $(X,c)$ is a partial representation or if the triple $(X,x,c)$ is a separator representation or indirect representation of a (unique) potential maximal clique $\Omega$.

The next step is to partition the vertex sets of the graph into smaller sets that can be used to create one of the three representations for the nice potential maximal clique. First the following theorem is required.

**Theorem 5.15** (see [14]). Let $\Omega$ be a potential maximal clique of $G$ and $S \subset \Omega$ a minimal separator, active for $\Omega$. Let $(S,C)$ be the block associated to $S$ containing $\Omega$, and let $x,y \in \Omega$ be an active pair. Then $\Omega \setminus S$ is a minimal $x,y$-separator in $G[C \cup \{x,y\}]$.

We are now ready to divide the set of connected components of $G \setminus \Omega$ into subsets.

**Lemma 5.16.** Let $\Omega$ be a nice potential maximal clique, $S$ be a minimal separator active for $\Omega$, and $x,y \in S$ be an active pair, and $C$ be the component of $G \setminus S$ containing $\Omega \setminus S$. There is a partition $(D_x,D_y,D_r)$ of $C \setminus \Omega$ such that $N(D_x \cup \{x\}) \cap C = N(D_y \cup \{y\}) \cap C = \Omega \setminus S$.

**Proof.** By Theorem 5.15, $\Omega \setminus S$ is a minimal $x,y$-separator in $G[C \cup \{x,y\}]$. Let $C_x$ be the full component associated to $\Omega \setminus S$ in $G[C \cup \{x,y\}]$ containing $x$, $D_x = C_x \setminus \{x\}$,
and let $C_y$ be the full component associated to $\Omega \setminus S$ in $G[C \cup \{x, y\}]$ containing $y$, $D_y = C_y \setminus \{y\}$, and $D_z = C \setminus (\Omega \cup D_x \cup D_y)$. Since $D_x \cup \{x\}$ and $D_y \cup \{y\}$ are full components of $\Omega \setminus S$, we have that $N(D_x \cup \{x\}) \cap C = N(D_y \cup \{y\}) \cap C = \Omega \setminus S$. □

By the lemma below we get that every potential maximal clique $\Omega$ that is not defined by the closed neighborhood of a vertex or defined by a separator representation has a neighbor outside of $\Omega$ in a full connected component of $S$.

**Lemma 5.17.** Let $\Omega$ be a potential maximal clique of $G$, $S$ be a minimal separator contained in $\Omega$, and $C$ be the component of $G \setminus S$ intersecting $\Omega$. Then one of the following holds:

1. there is a vertex $a$ such that $\Omega = N[a]$;
2. $\Omega$ has a separator representation;
3. $\Omega = N(C \setminus \Omega)$.

**Proof.** Suppose that there is a vertex $a \in \Omega$ having no neighbor in $C \setminus \Omega$. We consider first the case $a \in \Omega \setminus S$. We claim that in this case $\Omega = N[a]$. Because $a \in \Omega \setminus S \subseteq C$, we conclude that $N[a] \subseteq \Omega$. Thus to prove the claim we need to show that $\Omega \subseteq N[a]$. For sake of contradiction, suppose that there is $b \in \Omega$ which is not adjacent to $a$. By Theorem 2.4, every two nonadjacent vertices of a potential maximal clique are contained in some minimal separator $S_i(\Omega)$. Thus both $a$ and $b$ should have neighbors in a component $C_i(\Omega)$ of $G \setminus \Omega$. Since $a \in \Omega \setminus S \subseteq C$, we have that $C_i(\Omega) \subseteq C \setminus \Omega$. But this contradicts the assumption that $a$ has no neighbors in $C \setminus \Omega$.

The case $a \in S$ is similar. Suppose that $\Omega \setminus S \neq N(a) \cap C$; i.e., there is a vertex $b \in \Omega \setminus S$ nonadjacent to $a$. Then again, $a$ and $b$ are contained in some minimal separator and thus should have neighbors in a component $C_i(\Omega) \subseteq C$ of $G \setminus \Omega$ which is a contradiction.

Since $C$ is a component of $G \setminus S$ and $S$ is contained in $\Omega$, we have that $N(C \setminus \Omega) \subseteq \Omega$. If every vertex of $\Omega$ is adjacent to a vertex of $C \setminus \Omega$, then $\Omega = N(C \setminus \Omega)$. □

We state now the main tool for upper bounding the time required to list the set of nice potential maximal cliques.

**Lemma 5.18.** Let $\Omega$ be a nice potential maximal clique of $G$. Then one of the following holds:

1. there is a vertex $a$ such that $\Omega = N[a]$;
2. $\Omega$ has a separator representation;
3. $\Omega$ has a partial representation $(X, c)$ such that $|X| \leq n/3$;
4. $\Omega$ has an indirect representation $(X, x, c)$ such that $|X| \leq n/3$.

**Proof.** Let $S$ be a minimal separator active for $\Omega$, $x, y \in S$ be an active pair, and $C$ be the component of $G \setminus S$ containing $\Omega \setminus S$. By Lemma 5.16, there is a partition $(D_x, D_y, D_v)$ of $C \setminus \Omega$ such that $N(D_x \cup \{x\}) \cap C = N(D_y \cup \{y\}) \cap C = \Omega \setminus S$. If one of the sets $D_x, D_y$, say $D_x$, is empty, then $N(D_x \cup \{x\}) \cap C = N(x) \cap C = \Omega \setminus S$, and thus the triple $(x, z, S)$ is a separator representation of $\Omega$.

Suppose that none of the first two conditions of the lemma holds. Then $D_x$ and $D_y$ are nonempty. In order to argue that $\Omega$ has a partial representation $(X, c)$ or an indirect representation $(X, x, c)$ such that $|X| \leq n/3$, we partition the graph further. Let $R = \Omega \setminus S$, and let $D_S$ be the union of all full components associated to $S$ in $G \setminus \Omega$. The vertex set $D_S$ is the union of vertex sets of all connected components $C'$ of $G \setminus (\Omega \cup D_S)$ such that $x$ is contained in the neighborhood of $C'$. Thus a connected component $C'$ of $G \setminus (\Omega \cup D_S)$ is contained in $D_x$ if and only if $x \in N(C')$. Similarly, a connected component $C'$ of $G \setminus (\Omega \cup D_S)$ is contained in $D_y$ if and only if $y \in N(C')$. We also define $D_z = V \setminus (\Omega \cup D_S \cup D_x \cup D_y)$, which is the set of vertices of the components of $G \setminus (\Omega \cup D_S)$ which are not in $D_x$ and $D_y$.  

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We partition $S$ in the following sets:
- $S_x = (S \setminus N(D_x)) \cap N(D_y)$;
- $S_y = (S \setminus N(D_y)) \cap N(D_x)$;
- $S_{xy} = S \setminus (N(D_y) \cup N(D_x))$;
- $S_{xy} = S \cap N(Y) \cap N(D_x)$.

Thus $S_x$ is the set of vertices in $S$ with no neighbor in $D_x$ and with at least one neighbor in $D_y$, $S_y$ is the set of vertices in $S$ with no neighbor in $D_y$ and with at least one neighbor in $D_x$, $S_{xy}$ is the set of vertices in $S$ with neighbors in neither $D_x$ nor $D_y$, and finally, $S_{xy}$ is the set of vertices in $S$ with neighbors in both $D_x$ and $D_y$.

Notice that the vertex sets $S_x, D_x, D_y, D_r, R, S_y, S_y, S_{xy}$, and $S_{xy}$ are pairwise disjoint. The set $S_{xy}$ is mentioned only to complete the partition of $S$ and will not be used in the rest of the proof.

Both for size requirements and because of the definition of indirect representation we cannot use the sets $S_x, S_y$, and $S_{xy}$ directly; they have to be represented by the sets $Z_x, Z_y$, and $Z_r$, which are subsets of the vertex sets $D_x, D_y$, and $D_r$.

By the definition of $S_x$ and $S_y$ it follows that there exist two vertex sets $Z_x \subseteq D_y$ and $Z_y \subseteq D_x$ such that $S_x \subseteq N(Z_y)$ and $S_y \subseteq N(Z_x)$. Let $Z$ be such a set of minimum cardinality. By Lemma 5.17, $\Omega = N(D_x \cup D_y \cup D_r)$ since cases 1 and 2 of Lemma 5.17 correspond to cases 1 and 2 of the lemma we are proving. Thus, there exists a vertex set $Z_x \subseteq D_r$ such that $S_{xy} \subseteq N(Z_x)$. Let $Z$ be such a set of minimum cardinality. A sketch of how these vertex sets relate to each other is given in Figure 5.1.

Fig. 5.1. The figure shows a sketch of how the vertex sets $D_x, D_y, D_r, R, S_y, S_y, S_{xy}$, and $S_{xy}$ partition the graph $G$, and how the sets $Z_x, Z_y$, and $Z_r$ relate to this partition.

Let $C^*$ be a connected component of $G[D_S]$ (let us remind that $N(C^*) = S$). We define the following sets:
- $X_1 = C^* \cup R$;
- $X_2 = D_x \cup Z_y \cup Z_r$;
- $X_3 = D_y \cup Z_y \cup Z_r$.

First we claim that:
- the pair $(X_1, c)$, where $c \in C^*$, is a partial representation of $\Omega$;
- the triple $(X_2, x, c)$, where $c \in C^*$, is an indirect representation of $\Omega$;
- the triple $(X_3, y, c)$, where $c \in C^*$, is an indirect representation of $\Omega$.

In fact, the pair $(X_1, c) = (C^* \cup R, c)$ is a partial representation of $\Omega$ because $N(C^*) \cap R = \emptyset$, $C^*$ induces a connected graph, and $\Omega = N(C^*) \cup R$.

To prove that $(X_2, x, c) = (D_x \cup Z_y \cup Z_r, x, c)$ is an indirect representation of $\Omega$, we have to show that $\Omega = N(C \cup D_y \cup \{x\}) \cup \{x\}$, where $C_c$ is the connected component of $G \setminus N[X_c]$ containing $c$, and $D_z$ is the vertex set of the union of all connected components $C'$ of $G[X_c]$ such that $x \in N(C')$. Notice that $(S \cup C^*) \cap X_2 = \emptyset$ and
that $S \subseteq N(X_2)$ since $S \subseteq N(D_x \cup Z_\tau \cup Z_\tau)$ and $X_2 = D_x \cup Z_\tau \cup Z_\tau$. Hence the connected component $C_c$ of $G \setminus N[X_3]$ containing $c$ is $C^*$.

Every connected component $C'$ of $G[X_3]$ is contained in $D_x$, $Z_\tau$, or $Z_\tau$ since $\Omega \cap (D_x \cup Z_\tau \cup Z_\tau) = \emptyset$ and $\Omega$ separates $D_x$, $Z_\tau$, and $Z_\tau$. From the definition of $D_x$ it follows that $x \in N(C')$ for every connected component $C'$ of $G[D_x]$, and from the definition of $D_y$ and $D_r$ it follows that $x \not\in N(C')$ for every connected component $C'$ of $G[Z_\tau \cup Z_\tau]$. We can now conclude that $D_x$ is the vertex set of the union of all connected components $C'$ of $G[X_3]$ such that $x \in N(C')$. It remains to prove that $\Omega \setminus S = R$ is a subset of $N(D_x \cup \{x\})$ and $N(D_y \cup \{y\})$, and we remember that $N(C^*) = S$. From this observation it follows that $\Omega = N(C^* \cup D_x \cup \{x\}) \cup \{x\}$ since $N(C^* \cup D_x \cup \{x\}) = (S \cup R) \setminus \{x\}$.

By similar arguments, $(X_3, y, c)$ is an indirect representation of $\Omega$.

To conclude the proof of the lemma, we argue that at least one of the vertex sets $X_1$, $X_2$, or $X_3$ used to represent $\Omega$ contains at most $n/3$ vertices.

We partition the graph into the following three sets:

- $V_1 = D_S \cup R$
- $V_2 = D_x \cup S_\tau \cup S_{\tau^*}$
- $V_3 = D_y \cup S_\tau \cup D_r$

These sets are pairwise disjoint, and at least one of them is of size at most $n/3$; to prove the lemma we show that $|X_1| \leq |V_1|$, $|X_2| \leq |V_2|$, and $|X_3| \leq |V_3|$.\[|X_1| \leq |V_1|.\] Since $C^* \subseteq D_S$, we have that $X_1 = C^* \cup R \subseteq V_1 = D_S \cup R$.

- $|X_2| \leq |V_2|$. To prove the inequality we need the additional result

\[
|Z_\tau| \leq |S_\tau|, \quad |Z_\tau| \leq |S_{\tau^*}|, \quad \text{and } |Z_\tau| \leq |S_{\tau^*}|.
\]

In fact, since $Z_\tau$ is the smallest subset of $D_y$ such that $S_\tau \subseteq N(Z_\tau)$, we have that for any vertex $u \in Z_\tau$, $S_\tau \not\subseteq N(Z_\tau \setminus \{u\})$. Thus $u$ has a private neighbor in $S_\tau$, or in other words there exists $v \in S_\tau$ such that $\{u\} = N(v) \cap Z_\tau$. Therefore $S_\tau$ contains at least one vertex for every vertex in $Z_\tau$, which yields $|Z_\tau| \leq |S_\tau|$. The proofs of inequalities $|Z_\tau| \leq |S_{\tau^*}|$ and $|Z_\tau| \leq |S_{\tau^*}|$ are similar.

Now the proof of $|X_3| \leq |V_3|$, which is equivalent to $|D_x \cup S_\tau \cup Z_\tau| \leq |D_x \cup S_\tau \cup S_{\tau^*}|$, follows from (5.1) and the fact that all subsets on each side of the inequality are pairwise disjoint.

- $|X_3| \leq |V_3|$. This inequality is equivalent to $|D_y \cup Z_\tau \cup \{x\}| \leq |D_y \cup S_\tau \cup D_r|$. Again, the sets on each side of the inequality are pairwise disjoint. $|Z_\tau| \leq |D_r|$ because $Z_\tau \subseteq D_x$, and $|Z_\tau| \leq |S_\tau|$ by (5.1).

Thus $\min\{|X_1|, |X_2|, |X_3|\} \leq n/3$, which concludes the proof of the lemma. \[
\text{Lemma 5.19. The set of nice potential maximal cliques in a graph } G \text{ on } n \text{ vertices can be listed in } \mathcal{O}^*(\binom{n}{n/3}) \text{ time.}
\]

Proof. By Lemma 5.18, the number of possible partial representations $(X, c)$ and indirect representations $(X, x, c)$ with $|X| \leq n/3$ is at most \[2n^2 \sum_{i=1}^{n/3} \binom{n}{i}.\]

By Theorem 4.2, the number of all possible separator representations is at most \[n^2 |\Delta_G| \leq n^2 \binom{n}{n/3},\]

and we deduce that the number of nice potential maximal cliques is at most \[2n^2 \sum_{i=1}^{n/3} \binom{n}{i}.\] Moreover, these potential maximal cliques can be computed in $\mathcal{O}^*(\binom{n}{n/3})$ time as follows. We enumerate all the triples $(S, a, b)$ where $S$ is a minimal separator and $a, b$ are vertices, and check if the triple is the separator representation of a potential maximal clique $\Omega$; if so, we store this potential maximal clique. We also enumerate all the potential maximal cliques of type $N[a], a \in V(G)$ in polynomial time. Finally, by listing all the sets $X$ of at most $n/3$ vertices and all the couples
of vertices \((x, c)\), we compute all the nice potential maximal cliques with a partial representation \((X, c)\) or an indirect representation \((X, x, c)\).

**Theorem 5.20.** There is an algorithm to list all potential maximal cliques of a graph \(G\) on \(n\) vertices in time \(O(1.8899^n)\).

**Proof.** Let \(x_1, x_2, \ldots, x_n\) be the vertices of \(G\) and \(G_i = G\{x_1, \ldots, x_i\}\) for all \(i \in \{1, 2, \ldots, n\}\). Theorem 5.3 and Lemma 5.19 imply that \(\|\Pi_{G_i}\| \leq \|\Pi_{G_{i-1}}\| + n|\Delta_{G_i}| + 2n^2 \sum_{i=1}^{n/3} \binom{n}{i}\) for all \(i \in \{2, 3, \ldots, n\}\). By Theorem 4.2, \(\|\Pi_G\| \leq 2n^3 \sum_{i=1}^{n/3} \binom{n}{i}\).

Clearly, if we have the potential maximal cliques of \(G_{i-1}\), the potential maximal cliques of \(G_i\) can be computed in \(O^*(|\Pi_{G_{i-1}}| + \binom{n}{n/3})\) time by making use of Theorems 5.3 and 4.2 and Lemma 5.19. The graph \(G_1\) has a unique potential maximal clique, namely, \(\{x_1\}\). Therefore \(\Pi_G\) can be listed in time \(O^*(\binom{n}{n/3})\) time which is \(O(1.8899^n)\). □

Theorems 3.4 and 5.20 imply the main result of this paper.

**Theorem 5.21.** For a graph \(G\) on \(n\) vertices, the treewidth and the minimum fill-in of \(G\) can be computed in \(O(1.8899^n)\) time.

### 6. AT-free graphs

In this section we establish exact algorithms to compute the treewidth and the minimum fill-in of AT-free graphs which are faster than the ones obtained for general graphs in the previous section. Both algorithms are based on new upper bounds on the number of minimal separators and the number of potential maximal cliques in AT-free graphs.

Three pairwise nonadjacent vertices of a graph \(G\) form an *asteroidal triple* (AT) if any two of them are connected by a path avoiding the neighborhood of the third vertex. Graphs without asteroidal triples are called *AT-free*.

Corneil, Olariu, and Stewart studied structural properties of AT-free graphs in their fundamental paper [17]. Among other results, they showed that every connected AT-free graph has a dominating pair, where two vertices \(x\) and \(y\) of \(G\) form a *dominating pair* (DP for short) if the vertex set of each \(x, y\)-path is a dominating set of \(G\).

AT-free graphs contain cocomparability graphs, permutation graphs, interval graphs, and cobipartite graphs. Thus the treewidth problem and the minimum fill-in problem remain NP-hard when restricted to AT-free graphs [2, 48].

**Remark 6.1.** There is a well-known cobipartite (and thus AT-free) graph consisting of two cliques of size \(n/2\) and a perfect matching between them which has precisely \(2n^2/2 - 2\) minimal separators. It is not hard to show that this is indeed the largest number of minimal separators of a cobipartite graph on \(n\) vertices.

First we show that \(\|\Pi_G\| = O^*(|\Delta_G|)\) for AT-free graphs, improving a result in [13, Corollary 5.2]. This also establishes an algorithm to list the potential maximal cliques of an AT-free graph in \(O^*(|\Delta_G|)\) time. Then we prove that an AT-free graph on \(n\) vertices has at most \(2^{n/2 + 3}\) minimal separators.

First let us summarize some structural properties of potential maximal cliques in AT-free graphs.

**Lemma 6.2** (Proposition 5.1 of [13]). Let \(\Omega\) be a potential maximal clique of an AT-free graph \(G\). Then the set \(S(\Omega)\) of minimal separators contained in \(\Omega\) has at most two inclusion-maximal elements.

**Lemma 6.3** (Theorem 3.10 of [13]). Let \(G\) be a graph and \(\Omega\) be a potential maximal clique of \(G\) such that \(S(\Omega)\) has a unique inclusion-maximal element \(S\). Then \(\Omega \setminus S\) is a connected component of \(G \setminus S\).

Let \(S\) and \(T\) be two noncrossing minimal separators of \(G\), incomparable with respect to inclusion. Thus \(S\) meets a unique component of \(G \setminus T\), say \(C_S(T)\), and \(T\)
meets a unique component of $G \setminus S$, say $C_T(S)$. We define the piece between $S$ and $T$ as $P(S,T) = S \cup T \cup (C_T(S) \cap C_T(T))$.

**Lemma 6.4** (Theorem 3.11 of [13]). Let $G$ be a graph and $\Omega$ be a potential maximal clique of $G$ such that $S(\Omega)$ has exactly two inclusion-maximal elements $S$ and $T$. Then $\Omega = P(S,T)$.

**Lemma 6.5.** Let $G$ be an AT-free graph and $\Omega$ be a potential maximal clique of $G$ such that $S(\Omega)$ has two inclusion-maximal elements $S$ and $T$. Choose $s \in S \setminus T$. Then $\Omega = S \setminus (N(s) \cap C_T(S))$.

**Proof.** By Lemma 6.4, $\Omega = P(S,T)$. Clearly $s$ is in the unique component $C_S(T)$ of $G \setminus T$ meeting $S$, so $N(s) \cap C_T(S) \subseteq P(S,T)$. Consequently, $S \cup (N(s) \cap C_T(S)) \subseteq \Omega$.

Conversely, suppose there is a vertex $t \in \Omega$, not contained in $S \cup (N(s) \cap C_T(S))$. Let $S' = (S \setminus \{s\}) \cup (N(s) \cap C_T(S))$. Clearly $S'$ separates $s$ and any vertex of $C_T(S) \setminus S'$ in $G$; in particular $S'$ separates $s$ and $t$. It follows that there is a minimal separator $S'' \subseteq S'$ of $G$, contained in $\Omega$ and separating two vertices of $\Omega$. According to Theorem 2.4, for each minimal separator $U$ contained in $\Omega$, $\Omega$ intersects exactly one component of $G \setminus U$, which is a contradiction. \(\square\)

**Theorem 6.6.** An AT-free graph $G$ has at most $n^2|\Delta_G| + n|\Delta_G| + 1$ potential maximal cliques. Furthermore, there is an algorithm to list the potential maximal cliques of an AT-free graph in $O^*(|\Delta_G|)$ time.

**Proof.** If $G$ has no minimal separator, then $G$ is a complete graph, and its vertex set is the unique potential maximal clique of $G$.

Suppose now that $G$ is not complete. Fix a minimal separator $S$ of $G$. By Lemma 6.3, the number of potential maximal cliques $\Omega$ such that $S$ is the unique inclusion-maximal element of $S(\Omega)$ is bounded by the number of connected components of $G \setminus S$. Hence, there are at most $n$ such potential maximal cliques.

Now let us consider the potential maximal cliques $\Omega$ for which $S$ is one of the two inclusion-maximal separators contained in $S(\Omega)$. For any component $C$ of $G \setminus S$, there are, by Lemma 6.5, at most $|S|$ such potential maximal cliques contained in $S \cup C$. It follows that there are at most $n^2$ potential maximal cliques of this type.

Therefore, $G$ contains at most $(n^2 + n)|\Delta_G| + 1$ potential maximal cliques. These combinatorial arguments can easily be transformed into an algorithm listing the potential maximal cliques of an AT-free graph in time $O^*(|\Delta_G|)$. \(\square\)

Hence Theorem 3.4 implies that to construct an $O(1.4142^n)$ algorithm computing the treewidth and the minimum fill-in of an AT-free graph it is enough to prove that the number of minimal separators in an AT-free graph is $O(1.4142^n)$.

Our proof that the number of minimal separators in an AT-free graph is at most $2n^2/2^{+3}$ relies on properties of 2LexBFS, i.e., a combination of two runs of lexicographic breadth-first-search (also called 2-sweep LexBFS), on AT-free graphs established by Corneil, Olariu, and Stewart in [18].

**Definition 6.7.** A vertex ordering $x_n, x_{n-1}, \ldots, x_1$ is said to be a 2LexBFS ordering of $G$ if some 2LexBFS($G$) returns the vertices in this order (starting with $x_n$) during the second sweep of LexBFS on $G$ where $x_n$ is supposed to be the last vertex of the first sweep of LexBFS on $G$.

We shall write $u \prec v$ if $u = x_i$, $v = x_j$, and $i < j$. A 2LexBFS ordering and the levels $L_0 = \{x_n\}, L_1 = N(x_n), \ldots, L_i = \{x_j : d(x_j, x_n) = i\}, \ldots, L_r$, are called a 2LexBFS scheme of $G$. Consider any 2LexBFS scheme. Clearly all neighbors of a vertex $v \in L_i$ are contained in $L_{i-1} \cup L_i \cup L_{i+1}$. For a vertex $v \in L_i$ we denote $N(v) \cap L_{i-1}$ by $N_1(v)$, and we denote $N(v) \cap L_{i+1}$ by $N_1(v)$.

**Theorem 6.8** (see [18]). Every 2LexBFS ordering $x_n, x_{n-1}, \ldots, x_1$ of a connected AT-free graph has the dominating pair-property (DP-property); i.e., for all
i ∈ {1, 2, ..., n}, (x_n, x_i) is a dominating pair of the graph G[{x_i, x_{i+1}, ..., x_n}].

The following easy consequence of Theorem 6.8 is useful.

Lemma 6.9. Let x_n, x_{n-1}, ..., x_1 be a 2LexBFS ordering of an AT-free graph G, and let L_0, L_1, ..., L_r be the corresponding 2LexBFS scheme. Let s ∈ r, x_s, x_{s+1} ∈ L_i and \{x_r, x_s\} ∉ E. Then N^s_i(x_r) ⊆ N^s_i(x_s).

Proof. Let w ∈ N^s_i(x_r) \ N^s_i(x_s). Then the path x_r, w, u_{i-2}, ..., u_1, x_n with u_j ∈ L_j and u_{j-1} ∈ N^s_i(u_j) for all j = i - 2, ..., 1 contains no neighbor of x_s, contradicting the DP-property of a 2LexBFS scheme of an AT-free graph. □

Theorem 6.10. An AT-free graph on n vertices has at most 2^{n/2+3} minimal separators.

Proof. Let G be an AT-free graph. Let x_n, x_{n-1}, ..., x_1 be a 2LexBFS ordering of G, and let L_0, L_1, ..., L_r be the levels of the corresponding 2LexBFS scheme.

Let S be any minimal separator of G. Let C and C' be two (not necessarily full) components of G \ S. We claim that at most one level of the 2LexBFS scheme may contain vertices of C and C'. Suppose not. Let L_i and L_{i+1} be levels containing vertices of C and C'. Then there are edges {u, v} in C and {w, x} in C' such that u, w ∈ L_i and v, x ∈ L_{i+1}. W.l.o.g. assume u < w. Then Lemma 6.9 implies that w and v are adjacent, a contradiction.

Let C and C' be two (not necessarily full) components of G \ S such that both contain vertices of some level of the 2LexBFS scheme, say L_i. Furthermore, assume C ∩ L_{i-1} ≠ ∅ and C' ∩ L_{i-1} = ∅. Hence there is an edge \{u, v\} in C such that u ∈ L_i and v ∈ L_{i-1}. Then for each w ∈ C' holds w < u. Otherwise u < w, w ∈ L_i, and Lemma 6.9 would imply that u and v are adjacent, a contradiction.

Finally we claim that in this case c' < c for each vertex c ∈ C and each vertex c' ∈ C'. This is obviously true if one of c and c' is not in L_i. It remains to consider the case c ∈ L_i, c' ∈ L_i. To the contrary assume c < c'. Since C contains vertices of L_i and L_{i-1}, there is a path in C starting in c passing through vertices of C ∩ L_i only until it passes through an edge \{u, v\} in C with u ∈ L_i and v ∈ L_{i-1}. This path can be extended to a path from c to x_n that does not contain a neighbor of c' although c < c', a contradiction to the DP-property.

Now we are able to upper bound the number of those minimal separators in an AT-free graph in which no full component contains only vertices of one level. Simply divide the vertex set into two halves: A = \{x_n, x_{n-1}, ..., x_{[n/2]+1}\} and B = \{x_{[n/2]}, ..., x_1\}. Now consider two full components C and C' of a minimal separator S of G, i.e., S = N(C) = N(C'). Then either C or C' is a subset of either A or B, and surely each of C and C' uniquely determines S. Hence we simply consider all subsets of A and all subsets of B as possible full components of a minimal separator of G. Consequently, there are at most 2^{n/2+1} minimal separators of this type.

It remains to upper bound the number of all those minimal separators S of an AT-free graph G for which each full component is neither a subset of A nor a subset of B. Hence at least one component of S contains only vertices from one level of the 2LexBFS scheme.

Let S be such a minimal separator of G. Let C and C' be two full components of G \ S. W.l.o.g. assume C ⊆ L_i. Hence x_{[n/2]} ∈ L_i, and thus the level L_i is uniquely determined.

C' ∩ \bigcup_{j=0}^{i-1} L_j = ∅ since otherwise c < c' for all c ∈ C and all c' ∈ C', and either C or C' is a subset of A or B. Similarly C' must contain vertices of L_i. Consequently, C' ⊆ \bigcup_{j=i}^{r} L_j. It is easy to see that C ⊆ L_i and S = N(C) imply N(C') = S ⊆ \bigcup_{j=i-1}^{r} L_j. Furthermore, N(C) = N(C') = S implies S ∩ L_{i-1} = N_i^s(C ∩ L_i) = N_i^s(C' ∩ L_i).

Now let us consider the graph G' = G \ \bigcup_{j=0}^{i-1} L_j. Then S' = S \ \bigcup_{j=0}^{i-1} L_j is...
a separator of \( G' \); \( C \) and \( C' \) are components of \( G' \setminus S' \). Furthermore, every vertex of \( S' \subseteq S \) has a neighbor in \( C \) and \( C' \), and thus \( S' \) is a minimal separator of \( G' \). Consequently, every minimal separator \( S \) of \( G \) for which no full component is a subset of \( A \) or \( B \) corresponds uniquely to a minimal separator of \( G' \). Notice that \( G' \) has at most \( n - 1 \) vertices since we remove at least one vertex of \( L_{i-1} \) from \( G \) to obtain \( G' \).

Let \( f(n) \) be a function such that \( f(n) \) is an upper bound for the number of minimal separators in an \( n \)-vertex AT-free graph. Then we establish the recurrence \( f(n) \geq 2^{n/2+1} + f(n-1) \) and conclude with \( f(n) = 4 \cdot 2^{n/2+1} = 8 \cdot 2^{n/2} \).

Combining Theorems 3.4, 6.6, and 6.10, we obtain algorithms for AT-free graphs that are faster than the corresponding ones for general graphs.

**Theorem 6.11.** There are algorithms to compute the treewidth and the minimum fill-in of an AT-free graph in \( O(1.4142^n) \) time.

**7. Open problems and final remarks. Planar graphs.** The computational complexity of treewidth restricted to planar graphs is a longstanding open problem in graph algorithms. The treewidth of planar graphs can be approximated within a constant factor of 1.5. More precisely, Seymour and Thomas [42] gave a polynomial algorithm for computing the branchwidth of planar graphs, and the latter parameter differs by at most a factor of 1.5 from the treewidth.

In the case of planar graphs with \( n \) vertices, the treewidth is at most \( O(\sqrt{n}) \).

**Theorem 7.1 (see [26]).** For any planar graph \( G \) on \( n \) vertices, \( \text{tw}(G) \leq 3.182\sqrt{n} + O(1) \).

Also, given a graph \( G \) and a number \( k \), one can decide if \( \text{tw}(G) \leq k \) in \( O^*(n^k) \) time, either using the algorithm of Arnborg Corneil, and Proskurowski [2] or our technique, restricted to potential maximal cliques of size at most \( k + 1 \).

Consequently, the treewidth of planar graphs can be computed in time \( O^*(n^{3.182}\sqrt{n}) \) = \( 2^{O(\sqrt{n} \log n)} \).

Unfortunately, although the structure of potential maximal cliques in planar graphs is very particular [12], our approach cannot be used for obtaining algorithms of running time \( 2^{O(\sqrt{n})} \) for planar treewidth. This is because the number of “large” potential maximal cliques in planar graphs can be “large.”

**Claim 2.** For any integer \( N \), there is a planar graph on \( n > N \) vertices with at least \( 2^{0.49\sqrt{n} \log n} \) potential maximal cliques of size at least \( 2\sqrt{n} + 2 \).

**Proof.** Consider the planar graph \( G_p \) depicted in Figure 7.1. It has \( n = p^2 + p + 3 \) vertices. The set of vertices \( S = \{a_1, b_{i_1}, a_2, b_{i_2}, \ldots, a_p, b_{i_p}, a_{p+1}\} \) forms a \( c, \) \( d \)-minimal separator for any values \( i_1, i_2, \ldots, i_p \) between 1 and \( p \). By making use of Theorem 2.4, it is not hard to see that \( S \cup \{c\} \) is a potential maximal clique of size \( p + 1 \) in \( G_p \). Consequently, \( G \) has at least \( p^\alpha \) potential maximal cliques. If \( p \geq 2 \), we have \( p > \sqrt{n} - 1 \); thus the number of potential maximal cliques is at least \( (\sqrt{n} - 1)^{\sqrt{n} - 1} \).

Since we do not know if the treewidth of a planar graph can be computed in polynomial time, an interesting task is to design an algorithm of running time \( 2^{O(\sqrt{n})} \). As we mentioned, this will need new techniques.

**Combinatorial bounds.** The running time estimation of our algorithms is based on combinatorial upper bounds on the number of minimal separators and an upper bound for the time to list all potential maximal cliques of a graph. Finding better bounds on the number of minimal separators and potential maximal cliques in a graph is an interesting combinatorial challenge.

**How many potential maximal cliques can be in a graph?** We have shown that the number of potential maximal cliques in a graph on \( n \) vertices is at most \( O(1.8135^n) \). Unfortunately, it is not clear if the same bound can be obtained by an algorithm.
listing potential maximal cliques. Of course, such an algorithm can be used to speed
up our algorithm for treewidth and fill-in. A related interesting question is whether
it is possible to list potential maximal cliques with polynomial time delay.

How many minimal separators can be in a graph? We are aware of the following
construction providing the lower bound $3^{n/3} \approx 1.4422^n$ on the number of minimal
separators: Let $G$ be a graph on $n = 3k + 2$ vertices. $G$ has two vertices $a, b$ that
are connected by $k$ vertex disjoint paths of length 4. Every minimal $a, b$-separator in
$G$ contains exactly one inner vertex of each $a, b$-path. Thus the number of minimal
separators in $G$ is at least $3^{n/3} \approx 1.4422^n$. However, the gap between the lower bound
and the upper bound $O(1.7087^n)$ from Theorem 4.2 is still big. A related question is
whether it is possible to list the minimal separators with polynomial delay.

For some special graph classes, the use of minimal separators can imply faster
algorithms for triangulation problems. For example, we have shown that every AT-
free graph on $n$ vertices has at most $2^{n/2 + 3}$ minimal separators and that this upper
bound is tight up to a multiplicative constant factor. The interesting question here
is whether similar techniques can be used for other graph classes, such as bipartite
graphs and graphs of small degree.

Related problems. Our algorithms for treewidth and minimum fill-in can also
be used for solving other problems that can be expressed in terms of minimal trian-
gulations such as finding a tree decomposition of minimum cost [8] or computing
treewidth of weighted graphs. However, there are two “width” parameters related
to treewidth, namely bandwidth and pathwidth, and one parameter called profile,
related to minimum fill-in, that do not fit into this framework. Bandwidth can be
computed in time $O^*(10^n)$ [20], and reducing Feige’s bounds is a challenging problem.
Pathwidth (and profile) can be expressed as vertex ordering problems and thus solved
in $O^*(2^n)$ time by applying a dynamic programming approach similar to Held and
Karp’s approach [28] for the travelling salesman problem. Let us note that reach-

\begin{figure}[h]
\centering
\includegraphics[width=\textwidth]{planar_graphs}
\caption{Planar graphs with many large potential maximal cliques.}
\end{figure}
ing time complexity $O^*(c^n)$ for any constant $c < 2$, even for the Hamiltonian cycle problem, is a longstanding problem. So it is unlikely that some modification of Held and Karp’s approach would provide us with a better exact algorithm for pathwidth or profile. It is tempting to ask if one can reach time complexity $O^*(c^n)$, for any constant $c < 2$, for these problems.

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EXACT ALGORITHMS FOR TREEWIDTH AND MINIMUM FILL-IN