

Computing the Largest Bond of a Graph

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Abstract

A bond of a graph G is an inclusion-wise minimal disconnecting set of G , i.e., bonds are cut-sets that determine cuts $[S, V \setminus S]$ of G such that $G[S]$ and $G[V \setminus S]$ are both connected. Given $s, t \in V(G)$, an st -bond of G is a bond whose removal disconnects s and t . Contrasting with the large number of studies related to maximum cuts, there are very few results regarding the largest bond of general graphs. In this paper, we aim to reduce this gap on the complexity of computing the largest bond and the largest st -bond of a graph. Although cuts and bonds are similar, we remark that computing the largest bond of a graph tends to be harder than computing its maximum cut. We show that LARGEST BOND remains NP-hard even for planar bipartite graphs, and it does not admit a constant-factor approximation algorithm, unless $P = NP$. We also show that LARGEST BOND and LARGEST st -BOND on graphs of clique-width w cannot be solved in time $f(w) \times n^{o(w)}$ unless the Exponential Time Hypothesis fails, but they can be solved in time $f(w) \times n^{O(w)}$. In addition, we show that both problems are fixed-parameter tractable when parameterized by the size of the solution, but they do not admit polynomial kernels unless $NP \subseteq \text{coNP/poly}$.

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1 Introduction

Let $G = (V, E)$ be a simple, connected, undirected graph. A *disconnecting set* of G is a set of edges $F \subseteq E(G)$ whose removal disconnects G . The edge-connectivity of G is $\kappa'(G) = \min\{|F| : F \text{ is a disconnecting set of } G\}$. A cut $[S, T]$ of G is a partition of V into two subsets S and $T = V \setminus S$. The cut-set $\partial(S)$ of a cut $[S, T]$ is the set of edges that have one endpoint in S and the other endpoint in T ; these edges are said to cross the cut. In a connected graph, each cut-set determines a unique cut. Note that every cut-set is a disconnecting set, but the converse is not true. An inclusion-wise minimal disconnecting set of a graph is called a *bond*. It is easy to see that every bond is a cut-set, but there are cut-sets that are not bonds. More precisely, a nonempty set of edges F of G is a bond if and only if F determines a cut $[S, T]$ of G such that $G[S]$ and $G[T]$ are both connected. Let $s, t \in V(G)$. An *st-bond* of G is a bond whose removal disconnects s and t .

In this paper, we are interested in the complexity aspects of the following problem.

LARGEST BOND

Instance: A graph $G = (V, E)$; a positive integer k .

Question: Is there a proper subset $S \subset V(G)$ such that $G[S]$ and $G[V \setminus S]$ are connected and $|\partial(S)| \geq k$?

We also consider LARGEST *st*-BOND, where given a graph $G = (V, E)$, vertices $s, t \in V(G)$, and a positive integer k , we are asked whether G has an *st*-bond of size at least k .

A minimum (maximum) cut of a graph G is a cut with cut-set of minimum (maximum) size. Every minimum cut is a bond, thus a minimum bond is also a minimum cut of G , and it can be found in polynomial time using the classical Edmonds–Karp algorithm [11]. Besides that, minimum *st*-bonds are well-known structures, since they are precisely the *st*-cuts involved in the Gomory–Hu trees [20].

Regarding bonds on planar graphs, a folklore theorem states that if G is a connected planar graph, then a set of edges is a cycle in G if and only if it corresponds to a bond in the dual graph of G [18]. Note that each cycle separates the faces of G into the faces in the interior of the cycle and the faces of the exterior of the cycle, and the duals of the cycle edges are exactly the edges that cross from the interior to the exterior [28]. Consequently, the girth of a planar graph equals the edge connectivity of its dual [4].

Although cuts and bonds are similar, computing the largest bond of a graph seems to be harder than computing its maximum cut. MAXIMUM CUT is NP-hard in general [16], but becomes polynomial for planar graphs [21]. On the other hand, finding a longest cycle in a planar graph is NP-hard, implying that finding a largest bond of a planar multigraph (or of a simple edge-weighted planar graph) is NP-hard. In addition, it is well-known that if a simple planar graph is 3-vertex-connected, then its dual is a simple planar graph. In 1976, Garey, Johnson, and Tarjan [17] proved that the problem of establishing whether a 3-vertex-connected planar graph is Hamiltonian is NP-complete, thus, as also noted by Haglin and Venkatesan [22], finding the largest bond of a simple planar graph is also NP-hard, contrasting with the polynomial-time solvability of MAXIMUM CUT on planar graphs.

From the point of view of parameterized complexity, it is well known that MAXIMUM CUT can be solved in FPT time when parametrized by the size of the solution [25], and since every graph has a cut with at least half the edges [12], it follows that it has a linear kernel. Concerning approximation algorithms, a $1/2$ -approximation algorithm can be obtained by randomly partitioning the set vertices into two parts, which induces a cut-set whose expected size is at least half of the number of edges [26]. The best-known result is the seminal work of Goemans and Williamson [19], who gave a 0.878-approximation based on

semidefinite programming. This has the best approximation factor unless the Unique Games Conjecture fails [24]. To the best of our knowledge, there is no algorithmic study regarding the parameterized complexity of computing the largest bond of a graph as well as the approximability of the problem.

A closely related problem is the CONNECTED MAX CUT [23], which asks for a cut $[S, T]$ of a given a graph G such that $G[S]$ is connected, and that the cut-set $\partial(S)$ has size at least k . Observe that a bond induces a feasible solution of CONNECTED MAX CUT, but not the other way around, since $G[T]$ may be disconnected. Indeed, the size of a largest bond can be arbitrarily smaller than the size of the maximum connected cut; take, e.g., a star with n leaves. For CONNECTED MAX CUT on general graphs, there exists a $\Omega(1/\log n)$ -approximation [15], where n is the number of vertices. Also, there is a constant-factor approximation with factor $1/2$ for graphs of bounded treewidth [31], and a polynomial-time approximation scheme for graphs of bounded genus [23].

Recently, Saurabh and Zehavi [30] considered a generalization of CONNECTED MAX CUT, named MULTI-NODE HUB. In this problem, given numbers l and k , the objective is to find a cut $[S, T]$ of G such that $G[S]$ is connected, $|S| = l$ and $|\partial(S)| \geq k$. They observed that the problem is $W[1]$ -hard when parameterized on l , and gave the first parameterized algorithm for the problem with respect to the parameter k . We remark that the $W[1]$ -hardness also holds for LARGEST BOND parameterized by $|S|$.

Since every nonempty bond determines a cut $[S, T]$ such that $G[S]$ and $G[T]$ are both connected, every bond of G has size at most $|E(G)| - |V(G)| + 2$. A graph G has a bond of size $|E(G)| - |V(G)| + 2$ if and only if $V(G)$ can be partitioned into two parts such that each part induces a tree. Such graphs are known as *Yutsis graphs*. The set of planar Yutsis graphs is exactly the dual class of Hamiltonian planar graphs. According to Aldred, Van Dyck, Brinkmann, Fack, and McKay [1], cubic Yutsis graphs appear in the quantum theory of angular momenta as a graphical representation of general recoupling coefficients. They can be manipulated following certain rules in order to generate the so-called summation formulae for the general recoupling coefficient (see [2, 10, 32]).

There are very few results about the largest bond size in general graphs. In 2008, Aldred, Van Dyck, Brinkmann, Fack, and McKay [1] showed that if a Yutsis graph is regular with degree 3, the partition of the vertex set from the largest bond will result in two sets of equal size. In 2015, Ding, Dziobiak and Wu [9] proved that any simple 3-connected graph G will have a largest bond with size at least $\frac{2}{17}\sqrt{\log n}$, where $n = |V(G)|$. In 2017, Flynn [13] verified the conjecture that any simple 3-connected graph G has a largest bond with size at least $\Omega(n^{\log_3 2})$ for a variety of graph classes including planar graphs.

In this paper, we complement the state of the art on the problem of computing the largest bond of a graph. Preliminarily, we observe that while MAXIMUM CUT is trivial for bipartite graphs, LARGEST BOND remains NP-hard for such a class of graphs, and we also present a general reduction that allows us to observe that LARGEST BOND is NP-hard for several classes for which MAXIMUM CUT is NP-hard. Using this framework, we are able to show that LARGEST BOND on graphs of clique-width w cannot be solved in time $f(w) \times n^{o(w)}$ unless the ETH fails. Moreover, we show that LARGEST BOND does not admit a constant-factor approximation algorithm, unless $P = NP$, and thus is asymptotically harder to approximate than MAXIMUM CUT.

As for positive results, the main contributions of this work concern the parameterized complexity of LARGEST BOND. Using win/win approaches, we consider the strategy of preprocessing the input in order to bound the treewidth of the resulting instance. After that, by presenting a dynamic programming algorithm for LARGEST BOND parameterized by the

treewidth, we show that the problem is fixed-parameter tractable when parameterized by the size of the solution. Finally, we remark that LARGEST BOND and LARGEST st -BOND do not admit polynomial kernels, unless $\text{NP} \subseteq \text{coNP}/\text{poly}$. Due to space, some proofs were omitted.

2 Intractability results

In this section, we discuss aspects of the hardness of computing the largest bond. Notice that LARGEST BOND is Turing reducible to LARGEST st -BOND. Therefore, the results presented in this section also holds for LARGEST st -BOND.

Although MAXIMUM CUT is trivial for bipartite graphs, we first observe that the same does not apply to compute the largest bond. Since a connected planar graph is Eulerian if and only if its dual graph is bipartite; subdivision of edges does not increase the size of the largest bond; and to decide whether a 4-regular planar graph has a Hamiltonian cycle is NP-complete [29]. The following holds.

► **Theorem 1.** LARGEST BOND is NP-complete for planar bipartite graphs.

► **Theorem 2.** Let G be a simple bipartite graph and $\ell \in \mathbb{N}$. To determine the largest bond $\partial(S)$ of G with $|S| = \ell$ is $W[1]$ -hard with respect to ℓ .

Next, we present a general framework for reducibility from MAXIMUM CUT to LARGEST BOND, by defining a special graph operator ψ such that MAXIMUM CUT on a graph class \mathcal{F} is reducible to LARGEST BOND on the image of \mathcal{F} via ψ . An interesting particular case occurs when \mathcal{F} is closed under ψ (for instance, chordal graphs are closed under ψ).

► **Definition 3.** Let G be a graph and let $n = V(G)$. The graph $\psi(G)$ is constructed as follows: (i) create n disjoint copies G_1, G_2, \dots, G_n of G ; (ii) add vertices v_a and v_b ; (iii) add an edge between v_a and v_b ; (iv) add all possible edges between $V(G_1 \cup G_2 \cup \dots \cup G_n)$ and $\{v_a, v_b\}$.

► **Definition 4.** A set of graphs \mathcal{G} is closed under operator ψ if whenever $G \in \mathcal{G}$, then $\psi(G) \in \mathcal{G}$.

From the fact that a graph G has a cut $[S, V(G) \setminus S]$ of size k if and only if $\psi(G)$ has a bond $\partial(S')$ of size at least $nk + n^2 + 1$, the following theorem holds.

► **Theorem 5.** LARGEST BOND is NP-complete for any graph class \mathcal{G} such that: \mathcal{G} is closed under operator ψ ; and MAXCUT is NP-complete for graphs in \mathcal{G} .

► **Corollary 6.** LARGEST BOND is NP-complete for the following classes: chordal graphs; co-comparability graphs; P_5 -free graphs.

2.1 Algorithmic lower bound for clique-width parameterization

In the '90s, Courcelle, Makowsky, and Rotics [6] proved that all problems expressible in MS1-logic are fixed-parameter tractable when parameterized by the clique-width of a graph and the logical expression size. The applicability of this meta-theorem has made clique-width become one of the most studied parameters in parameterized complexity. However, although several problems are MS1-expressible, this is not the case with MAXIMUM CUT.

In 2014, Fomin, Golovach, Lokshtanov and Saurabh [14] showed that MAXIMUM CUT on a graph of clique-width w cannot be solved in time $f(w) \times n^{o(w)}$ for any function f of w unless Exponential Time Hypothesis (ETH) fails. Using operator ψ , we are able to extend this result to LARGEST BOND.

► **Lemma 7.** LARGEST BOND on graphs of clique-width w cannot be solved in time $f(w) \times n^{o(w)}$ unless the ETH fails.

Proof. MAXIMUM CUT cannot be solved in time $f(w) \times n^{o(w)}$ on graphs of clique-width w , unless Exponential Time Hypothesis (ETH) fails [14]. Therefore, by the polynomial-time reduction presented in Theorem 5, it is enough to show that the clique-width of $\psi(G)$ is upper bounded by a linear function of the clique-width of G .

If G has clique-width w , then the disjoint union $H_1 = G_1 \oplus G_2 \oplus \dots \oplus G_n$ has clique-width w . Suppose that all vertices in H_1 have label 1. Now, let H_2 be the graph isomorphic to a K_2 such that $V(H) = \{v_a, v_b\}$, and v_a, v_b are labeled with 2. In order to construct $\psi(G)$ from $H_1 \oplus H_2$ it is enough to apply the join $\eta(1, 2)$. Thus, $\psi(G)$ has clique-width equals w . ◀

2.2 Inapproximability

While the maximum cut of a graph has at least a constant fraction of the edges, the size of the largest bond can be arbitrarily smaller than the number of edges; take, e.g., a cycle on n edges, for which a largest bond has size 2. This discrepancy is also reflected on the approximability of the problems. Indeed, we show that LARGEST BOND is strictly harder to approximate than MAXIMUM CUT. To simplify the presentation, we consider a weighted version of the problem in which edges are allowed to have weights 0 or 1; the hardness results will follow for the unweighted case as well. In the BINARY WEIGHTED LARGEST BOND, the input is given by a connected weighted graph H with weights $w : E(H) \rightarrow \{0, 1\}$. The objective is to find a bond whose total weight is maximum.

Let G be a graph on n vertices and whose maximum cut has size k . Next, we define the G -edge embedding operator ξ_G . Given a connected weighted graph H , the weighted graph $\xi_G(H)$ is constructed by replacing each edge $\{u, v\} \in E(H)$ with weight 1 by a copy of G , denoted by G_{uv} , whose edges have weight 1, and, for each vertex t of G_{uv} , new edges $\{u, t\}$ and $\{v, t\}$, both with weight 0.

We can also apply the G -edge embedding operation on the graph $\xi_G(H)$, then on $\xi_G(\xi_G(H))$, and so on. In what follows, for an integer $h \geq 0$, denote by $\xi_G^h(H)$ the graph resulting from the operation that receives a graph H and applies ξ_G successively h times. Notice that $\xi_G^h(H)$ can be constructed in $\mathcal{O}(|V(G)|^{h+1})$ time. For some j , $0 \leq j \leq h - 1$, observe that an edge $\{u, v\} \in E(\xi_G^j(H))$ will be replaced by a series of vertices added in iterations $j + 1, j + 2, \dots, h$. These vertices will be called the *descendants* of $\{u, v\}$, and will be denoted by V_{uv} .

Let K_2 be the graph composed of a single edge $\{u, v\}$, and consider the problem of finding a bond of $\xi_G(K_2)$ with maximum weight. Since edges connecting u or v have weight 0, one can assume that u and v are in different sides of the bond, and the problem reduces to finding a maximum cut of G . In other words, the operator ξ_G embeds an instance G of MAXIMUM CUT into an edge $\{u, v\}$ of K_2 .

This suggests the following strategy to solve an instance of MAXIMUM CUT. For some constant integer $h \geq 1$, calculate $H = \xi_G^h(K_2)$, and obtain a bond F of H with maximum weight. Note that, to solve H , one must solve embedded instances of MAXIMUM CUT in multiple levels simultaneously. For a level j , $1 \leq j \leq h - 1$, each edge $\{u, v\} \in E(\xi_G^j(K_2))$ with weight 1 will be replaced by a graph G_{uv} which is isomorphic to G . In Lemma 9 below, we argue that F is such that either $V(G_{uv}) \cup \{u, v\}$ are all in the same side of the cut, or u and v are in distinct sides. In the latter case, the edges of F that separate u and v will induce a cut of G .

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In the remaining of this section, we consider a constant integer $h \geq 0$. Then, we define $H^j = \xi_G^j(K_2)$ for every j , $0 \leq j \leq h$, and $H = H^h$. Also, we write $[S, T]$ to denote the cut induced by a bond F of H .

► **Definition 8.** Let F be a bond of H with cut $[S, T]$. We say that an edge $\{u, v\} \in E(H^j)$ with weight 1 is nice for F if either

- $|\{u, v\} \cap S| = 1$, or
- $(\{u, v\} \cup V_{uv}) \subseteq S$, or
- $(\{u, v\} \cup V_{uv}) \subseteq T$.

Also, we say that F is nice if, for every j , $0 \leq j \leq h - 1$, and every edge $\{u, v\} \in E(H^j)$ with weight 1, $\{u, v\}$ is nice for F .

► **Lemma 9.** There is a polynomial-time algorithm that receives a bond F , and finds a nice bond F' such that $w(F') = w(F)$.

In the following, assume that F is a nice bond with cut $[S, T]$. Consider a level j , $0 \leq j \leq h$, and an edge $\{u, v\} \in E(H^j)$ with weight 1 such that $|\{u, v\} \cap S| = 1$. If $j < h$, then we define F_{uv} to be the subset of edges in F which are incident with some vertex of V_{uv} ; if $j = h$, then we define $F_{uv} = \{\{u, v\}\}$. Note that, because F is nice, if $|\{u, v\} \cap S| \neq 1$, then no edge of F is incident with V_{uv} .

Suppose now that $|\{u, v\} \cap S| = 1$ for some edge $\{u, v\} \in E(H^j)$ with weight 1 and $0 \leq j \leq h - 1$. In this case, F induces a cut-set of G_{uv} . Namely, define $\hat{S}_{uv} := S \cap V(G_{uv})$ and $\hat{T}_{uv} := T \cap V(G_{uv})$ and let \hat{F}_{uv} be the cut-set of G_{uv} corresponding to cut $[\hat{S}_{uv}, \hat{T}_{uv}]$.

Observe that for distinct edges $\{u, v\}$ and $\{s, t\}$, it is possible that $|\hat{F}_{uv}| \neq |\hat{F}_{st}|$. We will consider bonds F for which all induced cut-sets \hat{F}_{uv} have the same size.

► **Definition 10.** Let ℓ be a positive integer. A bond F of H with cut $[S, T]$ is said to be ℓ -uniform if, (i) F is nice, and (ii) for every j , $0 \leq j \leq h - 1$, and every edge $\{u, v\} \in E(H^j)$ with weight 1 such that $|\{u, v\} \cap S| = 1$, $|\hat{F}_{uv}| = \ell$.

An ℓ -uniform bond induces a cut-set of G of size ℓ .

► **Lemma 11.** Suppose F is an ℓ -uniform bond of H . One can find in polynomial time a cut-set L of G with $|L| = \ell$.

Proof. Let u, v be the vertices of K_2 to which ξ_G was applied. Since F is ℓ -uniform, $|\hat{F}_{uv}| = \ell$. Note that \hat{F}_{uv} induces a cut-set of size ℓ on G . ◀

In the opposite direction, a cut of G induces an ℓ -uniform bond of H .

► **Lemma 12.** Suppose L is a cut-set of G with $|L| = \ell$. One can find in polynomial time an ℓ -uniform bond F of H with $w(F) = \ell^h$.

Proof. For each $j \geq 0$, we construct a bond F^j of H^j . For $j = 0$, let F^0 be the set containing the unique edge of $H^0 = K_2$. Suppose now that we already constructed a bond F^{j-1} of H^{j-1} . For each edge $\{u, v\} \in F^{j-1}$, let L_{uv} be the set of edges of G_{uv} corresponding to L . Define $F^j := \cup_{\{u, v\} \in F^{j-1}} L_{uv}$. One can verify that indeed F^j is a bond of H^j , and that $w(F_j) = |L| \times w(F_{j-1}) = \ell^j$. ◀

► **Lemma 13.** There is a polynomial-time algorithm that receives a bond F of H , and finds an ℓ -uniform bond F' of H such that $w(F') = \ell^h \geq w(F)$.

► **Lemma 14.** Let F^* be a bond of H with maximum weight. Then $w(F^*) = k^h$.

Proof. We assume that F^* is ℓ -uniform such that $w(F^*) = \ell^h$ for some ℓ ; if this is not the case, then use Lemma 13.

Since F^* is ℓ -uniform, using Lemma 12 one obtains a cut-set L of G with size ℓ , then $\ell \leq k$, and thus $w(F^*) \leq k^h$.

Conversely, let L be a cut-set of G with size k . Using Lemma 12 for L , we obtain a bond F of H with weight k^h , and thus $w(F^*) \geq k^h$. ◀

► **Lemma 15.** *If there exists a constant-factor approximation algorithm for WEIGHTED LARGEST BOND, then $P = NP$.*

Proof. Consider a graph G whose maximum cut has size k . Construct graph H and obtain a bond F of H using an α -approximation, for some constant $0 < \alpha < 1$. Using the algorithm of Lemma 13, obtain an ℓ -uniform bond F' of H such that $w(F') = \ell^h \geq w(F)$. Using Lemma 14 and the fact that F' is an α -approximation, $\ell^h \geq \alpha \times k^h$. Using Lemma 11, one can obtain a cut-set L of G with size $\ell \geq \alpha^{\frac{1}{h}} k$.

For any constant ε , $0 < \varepsilon < 1$, we can set $h = \lceil \log_{1-\varepsilon} \alpha \rceil$, such that the cut-set L has size at least $\ell \geq (1 - \varepsilon)k$. Since MAXIMUM CUT is APX-hard, this implies $P = NP$. ◀

► **Theorem 16.** *If there exists a constant-factor approximation algorithm for LARGEST BOND, then $P = NP$.*

Proof. We show that if there exists an α -approximation algorithm for LARGEST BOND, for constant $0 < \alpha < 1$, then there is an $\alpha/2$ -approximation algorithm for the BINARY WEIGHTED LARGEST BOND, so the theorem will follow from Lemma 15.

Let H be a weighted graph whose edge weights are all 0 or 1. Let m be the number of edges with weight 0, and let l be the weight of a bond of H with maximum weight. Assume $l \geq 2/\alpha$, as otherwise, one can find an optimal solution in polynomial time by enumerating sets of up to $2/\alpha$ edges.

Construct an unweighted graph G as follows. Start with a copy of H and, for each edge $\{u, v\} \in E(H)$ with weight 1, replace $\{u, v\} \in E(G)$ by m parallel edges. Finally, to obtain a simple graph, subdivide each edge of G . If F is a bond of G , then one can construct a bond F' of H by undoing the subdivision and removing the parallel edges. Each edge of F' has weight 1, with exception of at most m edges. Thus, $w(F') \geq (|F| - m)/m$.

Observe that an optimal bond of H induces a bond of G with size at least ml . Thus, if F is an α -approximation for G , then $|F| \geq \alpha ml$ and therefore

$$w(F') \geq \frac{\alpha ml - m}{m} = \alpha l - 1 \geq \alpha l - \alpha l/2 = \alpha l/2.$$

We conclude that F' is an $\alpha/2$ -approximation for H . ◀

3 Algorithmic upper bounds for clique-width parameterization

Lemma 7 shows that LARGEST BOND on graphs of clique-width w cannot be solved in time $f(w) \times n^{o(w)}$ unless the ETH fails. Now, we show that given an expression tree of width at most w , LARGEST BOND can be solved in $f(w) \times n^{O(w)}$ time.

An expression tree \mathcal{T} is irredundant if for any join node $\eta(i, j)$, the vertices labeled by i and j are not adjacent in the graph associated with its child. It was shown by Courcelle and Olariu [7] that every expression tree \mathcal{T} of G can be transformed into an irredundant expression tree \mathcal{T} of the same width in time linear in the size of \mathcal{T} . Therefore, without loss of generality, we can assume that \mathcal{T} is irredundant.

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Our algorithm is based on dynamic programming over the expression tree of the input graph. We first describe what we store in the tables corresponding to the nodes in the expression tree.

Given a w -labeled graph G , two connected components of G has the same *type* if they have the same set of labels. Thus, a w -labeled graph G has at most $2^w - 1$ types of connected components.

Now, for every node X_ℓ of \mathcal{T} , denote by G_{X_ℓ} the w -labeled graph associated with this node, and let $L_1(X_\ell), \dots, L_w(X_\ell)$ be the sets of vertices of G_{X_ℓ} labeled with $1, \dots, w$, respectively. We define a table where each entry is of the form $c[\ell, s_1, \dots, s_w, r, e_1, \dots, e_{2^w-1}, d_1, \dots, d_{2^w-1}]$, such that: $0 \leq s_i \leq |L_i(X_\ell)|$ for $1 \leq i \leq w$; $0 \leq r \leq |E(G_{X_\ell})|$; $0 \leq e_i \leq \min\{2, |L_i(X_\ell)|\}$ for $1 \leq i \leq 2^w - 1$; and $0 \leq d_i \leq \min\{2, |L_i(X_\ell)|\}$ for $1 \leq i \leq 2^w - 1$.

Each entry of the table represents whether there is a partition V_1, V_2 of $V(G_{X_\ell})$ such that: $|V_1 \cap L_i(G_{X_\ell})| = s_i$; the cut-set of $[V_1, V_2]$ has size at least r ; $G_{X_\ell}[V_1]$ has e_i connected components of type i ; $G_{X_\ell}[V_2]$ has d_i connected components of type i , where $e_i = 2$ means that $G_{X_\ell}[V_1]$ has *at least* two connected components of type i . The same holds for d_i .

Notice that this table contains $f(w) \times n^{\mathcal{O}(w)}$ entries. If X_ℓ is the root node of \mathcal{T} (that is, $G = G_{X_\ell}$), then the size of the largest bond of G is equal to the maximum value of r for which the table for X_ℓ contains a valid entry (true value), such that there are j and k such that $e_i = 0, e_j = 1$ for $1 \leq i, j \leq 2^w - 1, i \neq j$; and $d_i = 0, d_k = 1$ for $1 \leq i, k \leq 2^w - 1, i \neq k$.

It is easy to see that we store enough information to compute a largest bond. Note that a w -labeled graph is connected if and only if it has exactly one type of connected components and exactly one component of such a type.

Now we provide the details of how to construct and update such tables. The construction for introduce nodes of \mathcal{T} is straightforward.

Relabel node: Suppose that X_ℓ is a relabel node $\rho(i, j)$, and let $X_{\ell'}$ be the child of X_ℓ . Then the table for X_ℓ contains a valid entry $c[\ell, s_1, \dots, s_w, r, e_1, \dots, e_{2^w-1}, d_1, \dots, d_{2^w-1}]$ if and only if the table for $X_{\ell'}$ contains an entry $c[\ell', s'_1, \dots, s'_w, r, e'_1, \dots, e'_{2^w-1}, d'_1, \dots, d'_{2^w-1}] = \text{true}$, where: $s_i = 0$; $s_j = s'_i + s'_j$; $s_p = s'_p$ for $1 \leq p \leq w, p \neq i, j$; $e_p = e'_p$ for any type that contain neither i nor j ; $e_p = 0$ for any type that contains i ; and for any type e_p that contains j , it holds that $e_p = \min\{2, e'_p + e'_q + e'_r\}$ where e'_q represent the set of labels $(C_p \setminus \{j\}) \cup \{i\}$, e'_r represent the set of labels $C_p \cup \{i\}$, and C_p is the set of labels associated to p . The same holds for d_1, \dots, d_{2^w-1} .

Union node: Suppose that X_ℓ is a union node with children $X_{\ell'}$ and $X_{\ell''}$. It holds that $c[\ell, s_1, \dots, s_w, r, e_1, \dots, e_{2^w-1}, d_1, \dots, d_{2^w-1}]$ equals true if and only if there are valid entries $c[\ell', s'_1, \dots, s'_w, r', e'_1, \dots, e'_{2^w-1}, d'_1, \dots, d'_{2^w-1}]$ and $c[\ell'', s''_1, \dots, s''_w, r'', e''_1, \dots, e''_{2^w-1}, d''_1, \dots, d''_{2^w-1}]$, having: $s_i = s'_i + s''_i$ for $1 \leq i \leq w$; $r' + r'' \geq r$; $e_k = \min\{2, e'_k + e''_k\}$, and $d_k = \min\{2, d'_k + d''_k\}$ for $1 \leq k \leq 2^w - 1$.

Join node: Finally, let X_ℓ be a join node $\eta(i, j)$ with the child $X_{\ell'}$. Remind that since the expression tree is irredundant then the vertices labeled by i and j are not adjacent in the graph $G_{X_{\ell'}}$. Therefore, the entry $c[\ell, s_1, \dots, s_w, r, e_1, \dots, e_{2^w-1}, d_1, \dots, d_{2^w-1}]$ equals true if and only if there is a valid entry $c[\ell', s_1, \dots, s_w, r', e'_1, \dots, e'_{2^w-1}, d'_1, \dots, d'_{2^w-1}]$ where

$$r' + s_i \times (|L_j(X_{\ell'})| - s_j) + s_j \times (|L_i(X_{\ell'})| - s_i) \geq r,$$

and $e_p = e'_p$, case p is associated to a type that contains neither i nor j ; $e_p = 1$, case p is associated to $C'_{i,j} \setminus \{i\}$, where $C'_{i,j}$ is the set of labels obtained by the union of the types of $G_{X_{\ell'}}$ with some connected component having either label i or label j ; $e_p = 0$, otherwise. The same holds for d_1, \dots, d_{2^w-1} .

The correctness of the algorithm follows from the description of the procedure. Since for each ℓ , there are $\mathcal{O}((n+1)^w \times m \times (3^{2^w-1})^2)$ entries, the running time of the algorithm is $f(w) \times n^{\mathcal{O}(w)}$. This algorithm together with Lemma 7 concludes the proof of the Theorem 17.

► **Theorem 17.** *LARGEST BOND cannot be solved in time $f(w) \times n^{\mathcal{O}(w)}$ unless ETH fails, where w is the clique-width of the input graph. Moreover, given an expression tree of width at most w , LARGEST BOND can be solved in time $f(w) \times n^{\mathcal{O}(w)}$.*

In order to extend this result to LARGEST *st*-BOND, it is enough to observe that given a tree expression \mathcal{T} of G with width w , it is easy to construct a tree expression \mathcal{T}' with width equals $w+2$, where no vertex of $V(G)$ has the same label than either s or t . Let $w+1$ be the label of s , and let $w+2$ be the label of t . By fixing, for each ℓ , $s_{w+1} = |L_{w+1}(X_\ell)|$ and $s_{w+2} = 0$, one can solve LARGEST *st*-BOND in time $f(w) \times n^{\mathcal{O}(w)}$.

4 Bounding the treewidth of G

In the remainder of this paper we deal with our main problems: LARGEST BOND and LARGEST *st*-BOND parameterized by the size of the solution (k). Inspired by the principle of preprocessing the input to obtain a kernel, we consider the strategy of preprocessing the input in order to bound the treewidth of the resulting instance.

We start our analysis with LARGEST BOND.

► **Definition 18.** *A graph H is called a minor of a graph G if H can be formed from G by deleting edges, deleting vertices, and by contracting edges. For each vertex v of H , the set of vertices of G that are contracted into v is called a branch set of H .*

► **Lemma 19.** *Let G be a simple connected undirected graph, and k be a positive integer. If G contains $K_{2,k}$ as a minor then G has a bond of size at least k .*

Proof. Let H be a minor of G isomorphic to $K_{2,k}$. Since G is connected and each branch set of H induces a connected subgraph of G , from H it is easy to construct a bond of G of size at least k . ◀

Combined with Lemma 19, the following results show that, without loss of generality, our study on k -bonds can be reduced to graphs of treewidth $\mathcal{O}(k)$.

► **Lemma 20.** *[3] Every graph $G = (V, E)$ contains $K_{2,k}$ as a minor or has treewidth at most $2k-2$.*

► **Lemma 21.** *[3] There is a polynomial-time algorithm that either concludes that the input graph G contains $K_{2,k}$ as a minor, or outputs a tree-decomposition of G of width at most $2k-2$.*

From Lemma 19 and Lemma 21 it follows that there is a polynomial-time algorithm that either concludes that the input graph G has a bond of size at least k , or outputs a tree-decomposition of G of width at most $2k-2$.

4.1 The *st*-bond case

Let $S \subseteq V(G)$ and let $\partial(S)$ be a bond of a connected graph G . Recall that a block is a 2-vertex-connected subgraph of G which is inclusion-wise maximal, and a block-cut tree of G is a tree whose vertices represent the blocks and the cut vertices of G , and there is an edge in the block-cut tree for each pair of a block and a cut vertex that belongs to that

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block. Then, $\partial(S)$ intersects at most one block of G . More precisely, for any two distinct blocks B_1 and B_2 of G , if $S \cap V(B_1) \neq \emptyset$ and $S \cap V(B_2) \neq \emptyset$, then either $V(B_2) \subseteq S$, or $V(B_2) \subseteq V \setminus S$. Indeed, if this is not the case, then either $G[S]$ or $G[V \setminus S]$ would be disconnected. Thus, to solve LARGEST st -BOND, it is enough to consider, individually, each block on the path between s and t in the block-cut tree of G . Also, if a block is composed of a single edge, then it is a bridge in G , which is not a solution for the problem unless $k = 1$. Thus, we may assume without loss of generality that G is 2-vertex-connected.

► **Lemma 22.** *Let G be a 2-vertex-connected graph. For all $v \in V(G) \setminus \{s, t\}$, there is an sv -path and a tv -path which are internally disjoint.*

► **Lemma 23.** *Let G be a 2-vertex-connected graph. If G contains $K_{2,2k}$ as a minor, then there exists $S \subseteq V(G)$ such that $\partial(S)$ is a bond of size at least k .*

Proof. Let G be a graph containing a $K_{2,2k}$ as a minor. If $k = 1$, the statement holds trivially, thus assume $k \geq 2$. Also, since G is connected, one can assume that this minor was obtained by contracting or removing edges only, and thus its branch sets contain all vertices of G . Let A and B be the branch sets corresponding to first side of $K_{2,2k}$, and let X_1, X_2, \dots, X_{2k} be the remaining branch sets.

First, suppose that s and t are in distinct branch sets. If this is the case, then there exist distinct indices $a, b \in \{1, \dots, 2k\}$ such that $s \in A \cup X_a$ and $t \in B \cup X_b$. Now observe that $G[A \cup X_a]$ and $G[B \cup X_b]$ are connected, which implies an st -bond with at least $2k - 1 \geq k$ edges. Now, suppose that s and t are in the same branch set. In this case, one can assume without loss of generality that $s, t \in A \cup X_{2k}$.

Define $U = A \cup X_{2k}$ and $Q = V(G) \setminus U$. Observe that $G[U]$ and $G[Q]$ are connected. Consider an arbitrary vertex v in the set Q . Since G is 2-vertex-connected, Lemma 22 implies that there exist an sv -path P_s and a tv -path P_t which are internally disjoint. Let P'_s and P'_t be maximal prefixes of P_s and P_t , respectively, whose vertices are contained in U .

We partition the set U into parts U_s and U_t such that $G[U_s]$ and $G[U_t]$ are connected. Since $G[U]$ is connected, there exists a tree T spanning U . Direct all edges of T towards s and partition U as follows. Every vertex in P'_s belongs to U_s and every vertex in P'_t belongs to U_t . For a vertex $u \notin V(P'_s \cup P'_t)$, let w be the first ancestor of u (accordingly to T) which is in $P'_s \cup P'_t$. Notice that w is well-defined since $u \in V(T)$ and the root of T is $s \in V(P'_s \cup P'_t)$. Then u belongs to U_s if $w \in V(P'_s)$, and u belongs to U_t if $w \in V(P'_t)$.

Observe that there are at least $2k - 1$ edges between U and Q , and thus there are at least k edges between U_s and Q , or between U_t and Q . Assume the former holds, as the other case is analogous. It follows that $G[U_s]$ and $G[U_t \cup Q]$ are connected and induce a bond of G with at least k edges. ◀

Lemma 21 and Lemma 23 imply that there is an algorithm that either concludes that the input graph G has a bond of size at least k , or outputs a tree-decomposition of an equivalent instance G' of width $\mathcal{O}(k)$.

► **Corollary 24.** *Given a graph G , vertices $s, t \in V(G)$, and an integer k , there exists a polynomial-time algorithm that either concludes that G has an st -bond of size at least k or outputs a subgraph G' of G together with a tree decomposition of G' of width equals $\mathcal{O}(k)$, such that G' has an st -bond of size at least k if and only if G has an st -bond of size at least k .*

Proof. Find a block-cut tree of G in linear time [5], and let B_s and B_t be the blocks of G that contain s and t , respectively. Remove each block that is not in the path from B_s to B_t in the block-cut tree of G . Let G' be the remaining graph. For each block B of G' , consider

the vertices s' and t' of B which are nearest to s and t , respectively. Using Lemmas 21 and 23 one can in polynomial time either conclude that B has an $s't'$ -bond, in which case G is a yes-instance, or compute a tree decomposition of B with width at most $\mathcal{O}(k)$.

Now, construct a tree decomposition of G' as follows. Start with the union of the tree decompositions of all blocks of G' . Next, create a bag $\{u\}$ for each cut vertex u of G' . Finally, for each cut vertex u and any bag corresponding to a block B connected through u , add an edge between $\{u\}$ and one bag of the tree decomposition of B containing u . Note that this defines a tree decomposition of G' and that each bag has at most $\mathcal{O}(k)$ vertices. ◀

Note that since k -bonds are solutions for CONNECTED MAX CUT, the results presented in this section naturally apply to such a problem as well.

5 Taking the treewidth as parameter

In the following, given a tree decomposition \mathcal{T} , we denote by ℓ one node of \mathcal{T} and by X_ℓ the vertices contained in the *bag* of ℓ . We assume w.l.o.g that \mathcal{T} is an extended version of a *nice* tree decomposition (see [8]), that is, we assume that there is a special root node r such that $X_r = \emptyset$ and all edges of the tree are directed towards r and each node ℓ has one of the following five types: *Leaf*; *Introduce vertex*; *Introduce edge*; *Forget vertex*; and *Join*. Moreover, define G_ℓ to be the subgraph of G which contains only vertices and edges that have been introduced in ℓ or in a descendant of ℓ .

The number of partitions of a set of k elements is the k -th *Bell number*, which we denote by $B(k)$ ($B(k) \leq k!$ [27]).

► **Theorem 25.** *Given a nice tree decomposition of G with width tw , one can find a bond of maximum size in time $2^{\mathcal{O}(tw \log tw)} \times n$ where n is the number of vertices of G .*

Proof. Let $\partial_G(U)$ be a bond of G , and $[U, V \setminus U]$ be the cut defined by such a bond. Set $S_U^\ell = U \cap X_\ell$. The removal of $\partial_G(U)$ partitions $G_\ell[U]$ into a set C_U^ℓ of connected components, and $G_\ell[V \setminus U]$ into a set $C_{V \setminus U}^\ell$ of connected components. Note that C_U^ℓ and $C_{V \setminus U}^\ell$ define partitions of S_U^ℓ and $X_\ell \setminus S_U^\ell$, denoted by ρ_1^ℓ and ρ_2^ℓ respectively, where the intersection of each connected component of C_U^ℓ with S_U^ℓ corresponds to one part of ρ_1^ℓ . The same holds for $C_{V \setminus U}^\ell$ with respect to $X_\ell \setminus S_U^\ell$ and ρ_2^ℓ .

We define a table for which an entry $c[\ell, S, \rho_1, \rho_2]$ is the size of a largest cut-set (partial solution) of the subgraph G_ℓ , where S is the subset of X_ℓ to the left part of the bond, $X_\ell \setminus S$ is the subset to the right part, and ρ_1, ρ_2 are the partitions of S and $X_\ell \setminus S$ representing, after the removal of the partial solution, the intersection with the connected components to the left and to the right, respectively. If there is no such a partial solution then $c[\ell, S, \rho_1, \rho_2] = -\infty$.

For the case that S is empty, two special cases may occur: either $U \cap V(G_\ell) = \emptyset$, in which case there are no connected components in C_U^ℓ , and thus $\rho_1 = \emptyset$; or C_U^ℓ has only one connected component which does not intersect X_ℓ , i.e., $\rho_1 = \{\emptyset\}$, this case means that the connected component in C_U^ℓ was completely forgotten. Analogously, we may have $\rho_2 = \emptyset$ and $\rho_2 = \{\emptyset\}$. Note that we do not need to consider the case $\{\emptyset\} \subsetneq \rho_i$ since it would imply in a disconnected solution. The largest bond of a connected graph G corresponds to the root entry $c[r, \emptyset, \{\emptyset\}, \{\emptyset\}]$.

To describe a dynamic programming algorithm, we only need to present the recurrence relation for each node type.

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Leaf: In this case, $X_\ell = \emptyset$. There are a few combinations for ρ_1 and ρ_2 : either $\rho_1 = \emptyset$, or $\rho_1 = \{\emptyset\}$, and either $\rho_2 = \emptyset$, or $\rho_2 = \{\emptyset\}$. Since for this case G_ℓ is empty, there can be no connected components, so having $\rho_1 = \emptyset$ and $\rho_2 = \emptyset$ is the only feasible choice.

$$c[\ell, S, \rho_1, \rho_2] = \begin{cases} 0 & \text{if } S = \emptyset, \rho_1 = \emptyset \text{ and } \rho_2 = \emptyset, \\ -\infty & \text{otherwise.} \end{cases}$$

Introduce vertex: We have only two possibilities in this case, either v is an isolated vertex to the left ($v \in S$) or it is an isolated vertex to the right ($v \notin S$). Thus, a partial solution on ℓ induces a partial solution on ℓ' , excluding v from its part.

$$c[\ell, S, \rho_1, \rho_2] = \begin{cases} c[\ell', S \setminus \{v\}, \rho_1 \setminus \{\{v\}\}, \rho_2] & \text{if } \{v\} \in \rho_1, \\ c[\ell', S, \rho_1, \rho_2 \setminus \{\{v\}\}] & \text{if } \{v\} \in \rho_2, \\ -\infty & \text{if } \{v\} \notin \rho_1 \cup \rho_2. \end{cases}$$

Introduce edge: In this case, either the edge $\{u, v\}$ that is being inserted is incident with one vertex of each side, or the two endpoints are at the same side. In the former case, a solution on ℓ corresponds to a solution on ℓ' with the same partitions, but with value increased. In the latter case, edge $\{u, v\}$ may connect two connected components of a partial solution on ℓ' .

$$c[\ell, S, \rho_1, \rho_2] = \begin{cases} c[\ell', S, \rho_1, \rho_2] + 1 & \text{if } u \in S \text{ and } v \notin S \text{ or } u \notin S \text{ and } v \in S, \\ \max_{\rho'_1} \{c[\ell', S, \rho'_1, \rho_2]\} & \text{if } u \in S \text{ and } v \in S, \\ \max_{\rho'_2} \{c[\ell', S, \rho_1, \rho'_2]\} & \text{if } u \notin S \text{ and } v \notin S. \end{cases}$$

Here, ρ'_1 spans over all refinements of ρ_1 such that the union of the parts containing u and v results in the partition ρ_1 . The same holds for ρ'_2 .

Forget vertex: In this case, either the forgotten vertex v is in the left side of the partial solution induced on ℓ , or is in the right side. Thus, v must be in the connected component which contains some part of ρ_1 , or some part of ρ_2 . We select the possibility that maximizes the value

$$c[\ell, S, \rho_1, \rho_2] = \max_{\rho'_1, \rho'_2} \{c[\ell', S \cup \{v\}, \rho'_1, \rho_2], c[\ell', S, \rho_1, \rho'_2]\}.$$

Here, ρ'_1 spans over all partitions obtained from ρ_1 by adding v in some part of ρ_1 (if $\rho_1 = \{\emptyset\}$ then $\rho'_1 = \{v\}$). The same holds for ρ'_2 .

Join: This node represents the join of two subgraphs $G_{\ell'}$ and $G_{\ell''}$ and $X_\ell = X_{\ell'} = X_{\ell''}$. By counting the bond edges contained in $G_{\ell'}$ and in $G_{\ell''}$, each edge is counted at least once, but edges in X_ℓ are counted twice. Thus

$$c[\ell, S, \rho_1, \rho_2] = \max \{c[\ell', S, \rho'_1, \rho'_2] + c[\ell'', S, \rho''_1, \rho''_2]\} - |\{\{u, v\} \in E, u \in S, v \in X_\ell \setminus S\}|.$$

In this case, we must find the best combination between the two children. Namely, for $i \in \{1, 2\}$, we consider combinations of ρ'_i with ρ''_i which merge into ρ_i . If $\rho_i = \{\emptyset\}$ then either $\rho'_i = \{\emptyset\}$ and $\rho''_i = \emptyset$; or $\rho'_i = \emptyset$ and $\rho''_i = \{\emptyset\}$. Also, if $\rho_i = \emptyset$ then $\rho'_i = \emptyset$ and $\rho''_i = \emptyset$.

The running time of the dynamic programming algorithm can be estimated as follows. The number of nodes in the decomposition is $\mathcal{O}(tw \times n)$ [8]. For each node ℓ , the parameters ρ_1 and ρ_2 induce a partition of X_ℓ ; the number of partitions of X_ℓ is given by the corresponding Bell number, $B(|X_\ell|) \leq B(tw + 1)$. Each such a partition ρ corresponds to a number of choice of parameter S that corresponds to a subset of the parts of ρ ; thus the number of choices for S is not larger than $2^{|\rho|} \leq 2^{|X_\ell|} \leq 2^{tw+1}$. Therefore, we conclude that the table size is at most $\mathcal{O}(B(tw + 1) \times 2^{tw} \times tw \times n)$. Since each entry can be computed in $2^{\mathcal{O}(tw \log tw)}$ time, the total complexity is $2^{\mathcal{O}(tw \log tw)} \times n$. The correctness of the recursive formulas is straightforward. ◀

The reason for the $2^{\mathcal{O}(tw \log tw)}$ dependence on treewidth is because we enumerate all partitions of a bag to check connectivity. However, one can obtain single exponential-time dependence by modifying the presented algorithm using techniques based on Gauss elimination, as described in [8, Chapter 11] for STEINER TREE.

► **Theorem 26.** LARGEST *st*-BOND is fixed-parameter tractable when parameterized by treewidth.

Proof. The solution of LARGEST *st*-BOND can be found by a dynamic programming as presented in Theorem 25 where we add s and t in all the nodes and we fix $s \in S$ and $t \notin S$. ◀

Finally, the following holds.

► **Corollary 27.** LARGEST BOND and LARGEST *st*-BOND are fixed-parameter tractable when parameterized by the size of the solution, k .

Proof. Follows from Lemma 19, Lemma 21, Corollary 24, Theorem 25 and Theorem 26. ◀

6 Infeasibility of polynomial kernels

An or-composition for LARGEST BOND parameterized by k can be done from the disjoint union of ℓ inputs, by selecting exactly one vertex of each input graph and contracting them into a single vertex. Now, let $(G_1, k, s_1, t_1), (G_2, k, s_2, t_2), \dots, (G_\ell, k, s_\ell, t_\ell)$ be ℓ instances of LARGEST *st*-BOND parameterized by k . An or-composition for LARGEST *st*-BOND parameterized by k can be done from the disjoint union of G_1, G_2, \dots, G_ℓ , by contracting t_i, s_{i+1} into a single vertex, $1 \leq i \leq \ell - 1$, and setting $s = s_1$ and $t = t_\ell$.

► **Theorem 28.** LARGEST BOND and LARGEST *st*-BOND do not admit polynomial kernel unless $NP \subseteq coNP/poly$.

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