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Leanness Computation: Small Values and Special Graph Classes *

David Coudert¹, Samuel Coulomb², and Guillaume Ducoffe³

¹Université Côte d'Azur, CNRS, Inria, I3S, Sophia Antipolis, France
²École Normale Supérieure, Paris, France
³National Institute for Research and Development in Informatics & University of Bucharest, București, România

Abstract

Let u and v be vertices in a connected graph G = (V, E). For any integer k such that $0 \le k \le d_G(u, v)$, the k-slice $S_k(u, v)$ contains all vertices x on a shortest uv-path such that $d_G(u, x) = k$. The leanness of G is the maximum diameter of a slice. This metric graph invariant has been studied under different names, such as "interval thinness" and "fellow traveler property". Graphs with leanness equal to 0, a.k.a. geodetic graphs, also have received special attention in Graph Theory. The practical computation of leanness in real-life complex networks has been studied recently (Mohammed et al., COMPLEX NETWORKS21). In this paper, we give a finer-grained complexity analysis of two related problems, namely: deciding whether the leanness of a graph G is at most some small value ℓ ; and computing the leanness on specific graph classes. We obtain improved algorithms in some cases, and time complexity lower bounds under plausible hypotheses.

Keywords: Leanness; Geodetic graphs; SETH-based lower bounds; Graph algorithms.

1 Introduction

The graph parameter *Leanness*, which is the main topic of this work, arises from Metric and Geometric Graph Theory [2]. For undefined graph terminology, see [4]. Unless stated otherwise, all graphs considered are finite, simple (they have neither loops nor multiple edges), undirected, unweighted and connected. Roughly, the leanness of a graph G is the smallest integer ℓ such that, for every source vertex s and every destination vertex t, two same-speed travelers on shortest st-paths always stay at distance at most ℓ to each other. See Sec. 2 for a formal definition of leanness, and for the required graph notations and terminology for this work. In what follows, let $\lambda(G)$ denote the leanness of a given graph G.

Related work. To our best knowledge, leanness of graphs has been first studied in Group Theory, under the insightful name of fellow traveler property [20]. Although the word problem is notoriously undecidable for general groups [29], it was shown in [20] that it can be solved in quadratic time on so-called automatic groups. A beautiful combinatorial characterization of automatic groups is given in [20], which implies their (infinite) Cayley graph satisfies the fellow traveler property (i.e., it has bounded leanness).

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Hyperbolicity is a metric tree-likeness parameter, first introduced by Gromov [21], which for graphs is tightly related to leanness. A metric space (X, d) is called δ -hyperbolic if it satisfies the following four-point condition for every $u, v, x, y \in X$: the two largest distance-sums amongst

$$d(u, v) + d(x, y)$$
, $d(u, x) + d(v, y)$ and $d(u, y) + d(v, x)$

must differ by at most 2δ . The hyperbolicity of (X, d) is the infimum of all values δ such that (X, d) is δ -hyperbolic. For a graph G, we denote by $\delta(G)$ its hyperbolicity. The value $\delta(G)$ is a lower bound on the smallest additive distortion for embedding G into an edge-weighted tree [21], with both values differing by at most some logarithmic factor [10, 21]. Moreover, there is empirical evidence that some complex networks have very small hyperbolicity [24]. In this respect, it has been argued in [24] that hyperbolicity helps in better classifying complex networks, while it also explains some of their observed properties [11].

On one hand, it can be easily verified from the definition that $\lambda(G) \leq 2\delta(G)$ for every graph G. Therefore, bounded hyperbolicity implies bounded leanness¹. However, on the other hand, cycles C_{4n+3} satisfy $\lambda(C_{4n+3}) = 0$ whereas $\delta(C_{4n+3}) = n$. Papasoglu proved a deeper relationship between hyperbolicity and leanness in geodesic metric spaces [31], which can be restated as follows for graphs: let H be the graph obtained from a graph G by subdividing once every edge; there exists a doubly-exponential function f such that, if the leanness of H is at most ℓ , then G must be $f(\ell)$ -hyperbolic. In particular, even though the hyperbolicity of G can be arbitrarily larger than its leanness, the hyperbolicity of G and the leanness of G are functionally equivalent. Even more strongly, the authors in [27] reported that all real-life networks G from [13] satisfy $\chi(G) = 2\delta(G)$. In the same way, it follows from [10, Proposition 10] that in modular graphs, pseudo-modular graphs and their respective subclasses, the leanness and the hyperbolicity can only differ by some small constant factor. For example, it was proved in [19] that $\chi(G) \leq 2\delta(G) \leq \chi(G) \leq$

According to [27], computing the leanness of a graph is a good heuristic for computing its hyperbolicity (and it always outputs a lower bound on the real value). However, while there has been substantial work toward practical hyperbolicity computation [5, 12, 13, 16, 17], comparatively little has been done for leanness computation. In [27], an algorithm in $\mathcal{O}(n^2m)$ time and $\mathcal{O}(n^2)$ space was proposed. We are not aware of any previous studies on the leanness in some graph classes (for such studies on the hyperbolicity, see [7, 10, 19, 25, 39]).

Graphs with leanness equal to zero are exactly the graphs such that there exists only one shortest path between every two vertices. They have been studied on their own under the different name of geodetic graphs [30]. Different constructions of geodetic graphs have been proposed in the literature [32]. Since their introduction by Ore in 1962, it has been asked repeatedly for a full characterization of these graphs. An algebraic characterization was proved in [28]. Combinatorial characterizations are known only for restricted subclasses, such as planar geodetic graphs [34]. Recently, Bodwin characterized the path systems that can be realized as unique shortest paths in a graph with arbitrary real edge weights [3].

Our Contributions. We address the complexity of computing the leanness of a graph, under two natural restrictions. First, in Sec. 3, we consider the recognition of graphs with leanness at most some constant ℓ . It is folklore that geodetic graphs can be recognized in polynomial time, by using a variation of breadth-first search (BFS). We present a different $\tilde{\mathcal{O}}(n^{\omega})$ -time algorithm for this problem, where $\omega < 2.371552$ [37] denotes the exponent for square matrix

¹We note that learness is often called *interval thinness* in prior works on hyperbolicity.

multiplication². By doing so, we improve the state of the art. The existence of an almost linear-time algorithm remains an open problem. We complement this result with conditional quadratic-time lower bounds for the recognition of graphs with leanness at most one (at most two, resp.).

Sec. 4, 5, 6 are devoted to the computation of leanness in restricted graph classes. Different types of grids are analyzed in Sec. 4, partly because of their relationship to hyperbolicity in subclasses of weakly modular graphs [8,19]. Other subclasses of planar graphs are considered in Sec. 5, namely outerplanar and bicyclic graphs. Finally, in Sec. 6, we study the cographs, chordal graphs, distance-hereditary graphs and bisplit graphs. As already noted in [27], the leanness is bounded on all these classes. We prove that the leanness can be computed in linear time on all these classes, except for bisplit graphs. Furthermore, under the Strong Exponential-Time Hypothesis (SETH) [23], there is no subquadratic-time algorithm for computing the leanness of bisplit graphs.

2 Definitions and notations

We now introduce the required notations and terminology for this paper. Recall that we only consider finite, undirected, unweighted and connected graphs. The graph G = (V, E) has n = |V| vertices and m = |E| edges. We denote N(u) the set of neighbors of vertex $u \in V$. Given two vertices $u, v \in V$, a uv-path of length $\ell \geq 0$ is a sequence of pairwise different vertices $(u = v_0, v_1, \dots, v_\ell = v)$ such that $\{v_i, v_{i+1}\} \in E$ for every i. The $distance \ d(u, v)$ is the minimum length of a uv-path in G. The $eccentricity \ ecc(u)$ is the maximum distance between vertex u and any other vertex $v \in V$, i.e., $ecc(u) = \max_{v \in V} d(u, v)$. The $diameter \ diam(G)$ is the maximum eccentricity of the graph, i.e., $diam(G) = \max_{u \in V} ecc(u)$.

For a pair (x,y) of vertices of G, the interval I(x,y) is the set of vertices that lay on any shortest xy-path, i.e., $I(x,y) = \{u \in V : d(x,y) = d(x,u) + d(u,y)\}$. An interval can be divided into a set of slices $S_k(x,y)$, $k = 0, 1, \ldots, d(x,y)$, such that $S_k(x,y) = \{u \in I(x,y) : d(x,u) = k\}$. We observe that the slices of an interval I(x,y) can be constructed in time $\mathcal{O}(n)$ if the distance matrix of the graph is given, or in time $\mathcal{O}(n+m)$ otherwise. The diameter $\operatorname{diam}(S_k(x,y))$ of a slice is defined as $\max_{u,v \in S_k(x,y)} d(u,v)$. Let $\lambda(x,y) = \max_{0 \le k \le d(x,y)} \operatorname{diam}(S_k(x,y))$. We call the interval I(x,y) ℓ -lean if and only if $\lambda(x,y) \le \ell$ (see Fig. 1 for an illustration).

Definition 1. The learness $\lambda(G)$ of a graph G = (V, E) is defined as

$$\lambda(G) = \max_{x,y \in V} \lambda(x,y).$$

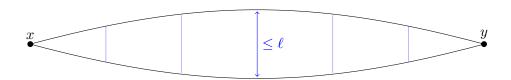


Figure 1: An ℓ -lean interval.

Roughly, the leanness of a graph measures the maximum distance between two shortest paths with same end-vertices. Clearly, we have $0 \le \lambda(G) \le \operatorname{diam}(G)$ for every graph G. These bounds are sharp. In particular, $\lambda(G) = 0$ if and only if there exists a unique shortest path

²The $\tilde{\mathcal{O}}()$ notation suppresses polylogarithmic factors.

between every two vertices of G. Graphs with learness equal to zero are called *geodetic graphs*. Examples of geodetic graphs are cliques, trees and odd cycles.

Let x, y be two vertices on different biconnected components of G. Then, all shortest xy-paths must cross some cut-vertex z. In this situation, every slice of interval I(x, y) must be a slice of I(x, z), or a slice of I(z, y). Therefore, $\lambda(x, y) = \max\{\lambda(x, z), \lambda(z, y)\}$. See Fig. 2 for an illustration. It follows from this observation that $\lambda(G)$ always equals the maximum leanness of its biconnected components. Furthermore, computing the biconnected components of a graph G can be done in linear time [22]. As a result, we can always assume in what follows that the graphs considered are biconnected.

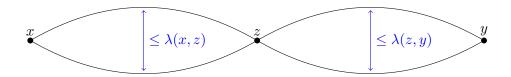


Figure 2: A cut-vertex z inside some interval I(x, y).

A vertex pair (x, y) is called a far-apart pair if and only if for every vertex z, we have d(x, z) < d(x, y) + d(y, z), and similarly d(y, z) < d(y, x) + d(x, z). The following observation will be often used in our analyzes:

Lemma 1. For every graph G = (V, E), there exists a far-apart pair (x, y) s.t. $\lambda(G) = \lambda(x, y)$.

Proof. Let $(x,y) \in V^2$ be such that $\lambda(G) = \lambda(x,y)$ and d(x,y) is maximized. Suppose by contradiction that (x,y) is not far-apart. By symmetry, we may assume the existence of some vertex z such that d(x,z) = d(x,y) + d(y,z). Then, $I(x,y) \subset I(x,z)$, and so, $\lambda(x,y) \leq \lambda(x,z)$. However, since d(x,z) > d(x,y), the latter contradicts the maximality of d(x,y).

The leanness of a graph G can be computed in $\mathcal{O}(n^4)$ time and $\mathcal{O}(n^2)$ space as follows: we precompute the distance matrix of G, then we iterate over all 4-tuples of vertices. A dynamic programming algorithm is presented in [27], that decreases the running time to $\mathcal{O}(n^2m)$. In what follows, we give a finer-grained complexity analysis for the recognition of graphs with small leanness, and for the computation of leanness in special graph classes.

3 Recognition of graphs with small leanness

In this section, we address the recognition of graphs with leanness at most two. A new algorithm is presented for deciding whether a graph is geodetic (Sec. 3.1). In Sec. 3.2, we prove conditional time complexity lower bounds for the recognition of graphs with leanness at most one (at most two, resp.).

3.1 Geodetic graphs

It is folklore that geodetic graphs can be recognized in polynomial time as follows:

• We consider each vertex u sequentially, and we compute a BFS with start vertex u. By doing so, we computed the distances d(u, v), for every vertex v. Then, if d(u, v) > 1, we check whether there exists a unique neighbour $w \in N(v)$ such that d(u, w) = d(u, v) - 1.

We summarize this discussion as follows:

Theorem 2. There is a combinatorial $\mathcal{O}(nm)$ -time algorithm for deciding whether a graph is geodetic.

Clearly, $\mathcal{O}(nm) = \mathcal{O}(n^3)$ for any graph. We now present a different algorithm that runs in $\tilde{\mathcal{O}}(n^{\omega})$ time, where $\omega < 2.371552$ [37] stands for the exponent of square matrix multiplication.

Theorem 3. There is an $\tilde{\mathcal{O}}(n^{\omega})$ -time algorithm for deciding whether a graph is geodetic.

Proof. We start pre-computing the distance matrix of G. This can be done in time $\mathcal{O}(n^{\omega})$ by using Seidel's algorithm [33]. The remaining steps of our algorithms, presented next, can be regarded as an adaptation of Seidel's algorithm to leanness computation.

First, we check all vertex pairs (u, v) such that d(u, v) is a power of two. For that, we consider each power 2^i , for $i = 1, 2, \ldots, \lfloor \log n \rfloor$, sequentially. Let A_i be the $(n \times n)$ -dimensional matrix such that $A_i[x, y] = 1$ if and only if $d(x, y) = 2^{i-1}$, and $A_i[x, y] = 0$ otherwise. – We note that A_1 is just the usual adjacency matrix of G. – Let $B_i = (A_i)^2$. If there exists some pair (u, v) such that $d(u, v) = 2^i$ and $B_i[u, v] > 1$, then we reject. Indeed, we claim that in this situation G is not geodetic. This is because $B_i[u, v] > 1$ implies the existence of two vertices $w, w' \in S_{2^{i-1}}(u, v)$, where $S_{2^{i-1}}(u, v) = \{x \in I(u, v) : d(u, x) = 2^{i-1}\}$ is a slice of the interval I(u, v), and so, of two shortest uv-paths.

Conversely, we claim that if we never reject during this phase, then there exists a unique shortest uv-path for every pair (u, v) such that d(u, v) is a power of two. Indeed, suppose by contradiction the existence of a pair (u, v) such that $d(u, v) = 2^i$ and there exist two shortest uv-paths. Without loss of generality, exponent i is minimized. Since we assume $B_i[u, v] = 1$, there exists a unique $w \in S_{2^{i-1}}(u, v)$. But then, let k be such that $0 < k < 2^i$ and $|S_k(u, v)| > 1$. Note that $k \neq 2^{i-1}$. Therefore, either $k < 2^{i-1}$ and there exist two shortest uw-paths, or $k > 2^{i-1}$ and there exist two shortest uv-paths. In either case, the latter contradicts the minimality of exponent i since we have $d(u, w) = d(w, v) = 2^{i-1}$.

Second, we again consider each power 2^i , for $i = 1, 2, ..., \lfloor \log n \rfloor$, sequentially. Then, we consider all pairs (u, v) such that $d(u, v) = (2j + 1)2^{i-1}$, for some integer $j \geq 1$. These are exactly the pairs (u, v) such that 2^{i-1} divides d(u, v) but 2^i does not divide d(u, v). If there exists such a pair (u, v) such that $|S_{2^{i-1}}(u, v)| > 1$, then G is not geodetic, and so, we reject. Otherwise (no such a pair exists for every i), we accept. This above condition, for every i, can be verified as follows:

- 1. Let G_i be the graph whose adjacency matrix equals A_i . In particular, $G_1 = G$. Note that G_i is a priori not connected. All connected components of G_i must be considered separately.
- 2. We compute all distances in G_i by using Seidel's algorithm. In what follows, let $d_i(u, v)$ denote the distance between two vertices that are in a same connected component of G_i . Note that if $d(u, v) = (2j + 1)2^{i-1}$, then $d_i(u, v) = 2j + 1$. Furthermore, all shortest uv-paths of G_i are subsets of shortest uv-paths of G (the latter might be false for other pairs (u', v') such that 2^{i-1} does not divide d(u', v')).
- 3. For r = 0, 1, 2, let $C_{i,r}$ be the $(n \times n)$ -dimensional matrix such that $C_{i,r}[x,y] = 1$ if and only if x, y are in a same connected component of G_i and $d_i(x,y) = r \pmod{3}$, otherwise $C_{i,r}[x,y] = 0$. We show next that it is sufficient to compute the three matrix products $A_iC_{i,r}$ in order to verify the condition. Specifically, for every pair (u,v) such that $d(u,v) = (2j+1)2^{i-1}$, we claim that $|S_{2^{i-1}}(u,v)| > 1$ if and only if, for the unique $r \in \{0,1,2\}$ such that $2j = r \pmod{3}$, we have $(A_iC_{i,r})[u,v] > 1$. This is because the vertices of $S_{2^{i-1}}(u,v)$ are exactly the neighbours of u in G_i that lie on some shortest

uv-path. These are exactly the neighbours w of u in G_i s.t. $d_i(w,v) = d_i(u,v) - 1 = 2j \pmod{3}$.

At each step i, we call Seidel's algorithm once, and compute $\mathcal{O}(1)$ matrix products. Since there are $\tilde{\mathcal{O}}(1)$ steps, the running time of the algorithm is in $\tilde{\mathcal{O}}(n^{\omega})$.

Finally, let us prove correctness of this algorithm. Suppose by contradiction the existence of two shortest uv-paths, for some pair (u,v). Recall that d(u,v) cannot be a power of two. Therefore, $d(u,v)=(2j+1)2^{i-1}$, for some $i,j\geq 1$. Without loss of generality, exponent i is maximized. There exists a unique $w\in S_{2^{i-1}}(u,v)$ because otherwise we would have rejected during the second phase of the algorithm. Furthermore, since $d(u,w)=2^{i-1}$, there exists a unique shortest uw-path. This implies the existence of two shortest wv-path, and $d(w,v)=2^ij$ is not a power of two. But then, $2^ij=(2t+1)2^{i+s-1}$, for some $s,t\geq 1$, thus contradicting the maximality of exponent i.

The existence of an almost linear-time algorithm is left as an intriguing open question.

3.2 Time complexity lower bounds

We were unsuccessful in establishing lower bounds for the recognition of geodetic graphs. However, in what follows, we do prove conditional lower bounds for the recognition of graphs with leanness at most one (at most two, resp.).

Let $\mathcal{H} = (X, R)$ be a 3-uniform hypergraph (i.e., $R \subseteq {V \choose 3}$). A 4-hyperclique in \mathcal{H} is a vertex subset $X = \{x_1, x_2, x_3, x_4\}$ such that every 3-subset of X is a hyperedge. The so-called 3-uniform 4-hyperclique hypothesis posits that detecting a 4-hyperclique in a 3-uniform hypergraph of n nodes requires $n^{4-o(1)}$ time [26]. Evidence for this conjecture is that its refutation would imply faster algorithms for well-studied problems such as MAX-3-SAT [38].

Proposition 4. Under the 3-uniform 4-hyperclique hypothesis, the recognition of n-vertex graphs G s.t. $\lambda(G) \leq 1$ requires $\Omega(n^{2-o(1)})$ time. The result holds even if G has $\mathcal{O}(n^{3/2})$ edges.

Proof. It was proved in [18] that detecting an induced C_4 in an n-vertex $\mathcal{O}(n^{3/2})$ -edge graph G requires $\Omega(n^{2-o(1)})$ time under the 3-uniform 4-hyperclique hypothesis. Let G' be obtained from G by adding a universal vertex. Then, $\lambda(G') \leq 1$ if and only if G is C_4 -free.

The STRONG EXPONENTIAL-TIME HYPOTHESIS (SETH) posits that for every $\varepsilon > 0$, there exists some integer k such that k-SAT cannot be solved in $(2-\varepsilon)^n$ time [23]. SETH-based lower bounds have gained momentum in the nascent field of "Fine-Grained Complexity in P". These conditional lower bounds are often achieved via an intermediate problem called DISJOINTSET. In the latter problem, we are given two families of n sets and the goal is to determine whether there exists two disjoint sets, with one set in each family. Under the SETH, it was proved in [6,38] that DISJOINTSET requires $\Omega(n^{2-o(1)})$ time, even if the universe of both families only contains $n^{o(1)}$ elements. We will come back to the SETH in Sec. 6.4.

Proposition 5. Under the SETH, the recognition of n-vertex graphs G s.t. $\lambda(G) \leq 2$ requires $\Omega(n^{2-o(1)})$ time. The result holds even if G has $n^{1+o(1)}$ edges.

Proof. Under the SETH, deciding whether an n-vertex graph G = (V, E) has diameter two or three requires $\Omega(n^{2-o(1)})$ time, even if G only has $n^{1+o(1)}$ edges (such a graph can be constructed from any instance of DISJOINTSET) [6]. Let G' be constructed from G as follows:

• The vertex set of G' is $V \cup V_x \cup V_y \cup \{x, y, z\}$ where V_x, V_y are disjoint copies of V;

- $N(x) = V_x$, $N(y) = V_y$, and $N(z) = V_x \cup V_y$;
- V_x and V_y are independent sets;
- G'[V] = G;
- finally, for every $v \in V$, we add two edges $\{v_x, v\}, \{v, v_y\}$.

Note that we can construct G' from G in $n^{1+o(1)}$ time. Furthermore, it was proved in [6, Sec. 3.3] that $\delta(G') = \operatorname{diam}(G)/2$, where $\delta(G')$ denotes the hyperbolicity of G' (see Sec. 1). Recall that $\lambda(G') \leq 2\delta(G')$, and so, $\lambda(G') \leq \operatorname{diam}(G)$. This is in fact an equality because $V \subseteq S_2(x,y)$ and G is an isometric sugraph of G'. As a result, $\lambda(G') \leq 2$ iff $\operatorname{diam}(G) = 2$, which, under SETH, requires $\Omega(n^{2-o(1)})$ time to decide.

4 Grid variants

In this section, we establish closed-form formulas for the leanness of different types of grids. We refer to Fig. 3, 4, 5, 7 for illustrations. We stress that grids often appear as an obstruction to small hyperbolicity or leanness in various graph classes. For example, it follows from [8, Proposition 9.10] that for median graphs G, $\lambda(G) \leq 2\ell$ if and only if every isometric square grid of G has side at most ℓ .

Square grids

The $p \times q$ grid is the graph G = (V, E), where $V = \llbracket 0, p - 1 \rrbracket \times \llbracket 0, q - 1 \rrbracket$ and $E = \{(u, v) \in V^2 : |u_1 - v_1| + |u_2 - v_2| = 1\}$. The distance between two vertices u and v is the Manhattan distance $d(u, v) = |u_1 - v_1| + |u_2 - v_2|$.

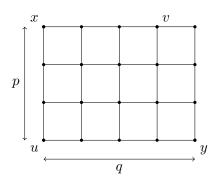


Figure 3: The 4×5 grid.

Proposition 6. The $p \times q$ grid G has learness $\lambda(G) = 2\min\{p,q\} - 2$.

Proof. Without loss of generality, suppose $p \le q$. The only far-apart pairs are the two pairs of opposite corners $\{x,y\} = \{(0,0),(p-1,q-1)\}$ and $\{x',y'\} = \{(0,q-1),(p-1,0)\}$. By Lemma 1, $\lambda(G) = \max\{\lambda(x,y),\lambda(x',y')\}$. Because of symmetry, they both have the same leanness, so we just need to find the leanness of one of them, say $\{x,y\}$. Let k be such that $0 \le k \le p+q-2$.

$$S_k(x,y) = \{(i,j) \in V : i+j=k\} = \{(i,k-i) : 0 \le i < p, \ 0 \le k-i < q\}$$

Let u = (i, k - i) and v = (j, k - j) be two vertices in the slice $S_k(x, y)$.

$$d(u,v) = |j-i| + |k-i-(k-j)| = 2|j-i| \le 2(p-1)$$

Hence, $\lambda(G) \leq 2(p-1)$. Then, taking u = (p-1,0) and v = (0,p-1), we do have $u,v \in S_k(x,y)$ and d(u,v) = 2(p-1), so $\lambda(G) \geq 2(p-1)$. Thus, we have proven that $\lambda(G) = 2(p-1)$.

Cylinder grids

The cylinder grid of size $p \times q$ is given by connecting the two sides of size p from the $p \times q$ grid. Formally, it means adding the edges $\{\{(k,0),(k,q-1)\}:0\leq k< p\}$. It can also be seen as replacing [0, q-1] with $\mathbb{Z}/q\mathbb{Z}$.

Proposition 7. The $p \times q$ cylinder grid has leanness

$$\lambda = \begin{cases} \min\left\{2(p-1), q-1\right\} & \text{if } q \text{ is odd} \\ \min\left\{q, 2\left\lfloor\frac{2(p-1)+q}{4}\right\rfloor\right\} & \text{if } q \text{ is even} \end{cases}$$

Proof. By Lemma 1 we only need to consider far-apart pairs. It was proved in [15, Lemma 55] that the set of far-apart pairs is $\{\{(0,i),(p-1,j)\}:0\leqslant i,j< q,\ j-i\equiv\pm\left\lfloor\frac{q}{2}\right\rfloor\pmod{q}\}.$ Because of symmetry, we may only look at one of them, say $\{x,y\} = \{(0,0), (p-1, \lfloor \frac{q}{2} \rfloor)\}.$

If q is odd, then the graph induced by I(x,y) is the $p \times \frac{q+1}{2}$ grid. By Proposition 6, we get $\lambda(G) = \min\{2p, q+1\} - 2 = \min\{2(p-1), q-1\}.$

If q is even, then the graph induced by I(x,y) is the whole graph G. Furthermore, G is bipartite. Then, any two vertices in a same slice will always be in the same partite set, and therefore the distance between such vertices must be even. As the diameter of G is $p-1+\frac{q}{2}$, we obtain that $\lambda(G) \leq 2\lfloor \frac{p-1+q/2}{2} \rfloor$. Let $u = (u_1, u_2)$ and $v = (v_1, v_2)$ be two vertices on a same slice $S_k(x, y)$. Note that

 $k = d(x, u) = u_1 + \min\{u_2, q - u_2\} \le u_1 + \frac{q}{2}$. In particular, $k - \frac{q}{2} \le u_1 \le k$. Hence,

$$d(u,v) = |u_1 - v_1| + \max\{|u_2 - v_2|, q - |u_2 - v_2|\} \le |u_1 - v_1| + q/2 \le q.$$

Altogether combined, we obtain $\lambda(G) \leq \min\left\{q, 2\left\lfloor\frac{p-1+q/2}{2}\right\rfloor\right\}$. We end up proving the above is always an equality. If $\frac{q}{2} \leq p-1$, then $u=(\frac{q}{2},0)$ and $v=(0,\frac{q}{2})$ are in $S_{q/2}(x,y)$, and $\mathrm{d}(u,v)=q$. Else, $p-1\leq r=\left\lfloor\frac{p-1+q/2}{2}\right\rfloor\leq\frac{q}{2}$, vertices u=(0,q-r) and v=(p-1,r-p+1) are in $S_r(x,y)$, and $\mathrm{d}(u,v)=2r$.

Torus grids

The torus grid of size $p \times q$ is obtained from the $p \times q$ cylinder grid by adding all edges in $\{\{(0,k),(p-1,k)\}:0\leq k< q\}.$

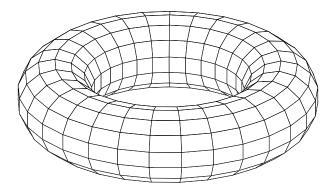


Figure 4: A torus grid.

Proposition 8. The $p \times q$ torus grid has leanness

$$\lambda = \begin{cases} \min{\{p,q\} - 1} & \textit{if p and q are odd} \\ \min{\left\{q, 2 \left\lfloor \frac{p+q}{4} \right\rfloor \right\}} & \textit{if p is odd and q is even} \\ \min{\left\{p, 2 \left\lfloor \frac{p+q}{4} \right\rfloor \right\}} & \textit{if p is even and q is odd} \\ 2 \left\lfloor \frac{p+q}{4} \right\rfloor & \textit{if p and q are even} \end{cases}$$

Proof. A pair $\{(i,j),(k,l)\}$ is far-apart if and only if $k-i \equiv \pm \lfloor \frac{p}{2} \rfloor \pmod{p}$ and $l-j \equiv \pm \lfloor \frac{q}{2} \rfloor \pmod{q}$. By Lemma 1, it is enough to find the leanness of one of these pairs, say the pair $\{x,y\} = \{(0,0),(\lfloor \frac{p}{2} \rfloor,\lfloor \frac{q}{2} \rfloor)\}$.

If p and q are odd, then the graph induced by I(x,y) is the $\frac{p+1}{2} \times \frac{q+1}{2}$ grid. By Proposition 6, $\lambda(G) = \min\{p+1, q+1\} - 2 = \min\{p, q\} - 1$.

If p is odd and q is even, then the graph induced by I(x,y) is the $\frac{p+1}{2} \times q$ cylinder grid. By Proposition 7, $\lambda(G) = \min\left\{q, 2\left\lfloor\frac{p+q-1}{4}\right\rfloor\right\}$. Note that $\left\lfloor\frac{p+q-1}{4}\right\rfloor = \left\lfloor\frac{p+q}{4}\right\rfloor$ because p+q is odd. The case when p is even and q is odd is dealt with similarly.

Finally, assume both p and q are even. Then, $\operatorname{diam}(G)=(p+q)/2$. Furthermore, G is bipartite. Since every slice of I(x,y) must be fully contained in one partite set, the diameter of each slice must be even. Hence $\lambda(G)$ must be even, and so, $\lambda(G) \leq 2 \left \lfloor \frac{p+q}{4} \right \rfloor$. Without loss of generality we assume from now on $p \leq q$. Let $u = (0, \left \lfloor \frac{p+q}{4} \right \rfloor)$ and $v = (\frac{p}{2}, \left \lfloor \frac{p+q}{4} \right \rfloor - \frac{p}{2})$. Observe that $u, v \in S_{\left \lfloor \frac{p+q}{4} \right \rfloor}(x,y)$. Therefore, $\lambda(G) \geq \lambda(x,y) \geq \operatorname{d}(u,v) = 2 \left \lfloor \frac{p+q}{4} \right \rfloor$.

King's grids

The King's grid of size $p \times q$ is obtained from the $p \times q$ grid by adding both diagonals in every square, i.e., by adding all edges in $\{\{(i,j),(i+1,j+1)\}: 0 \leq i < p-1,\ 0 \leq j < q-1\} \cup \{\{(i,j),(i-1,j+1)\}: 0 < i < p,\ 0 \leq j < q-1\}$. It is an example of Helly graph.

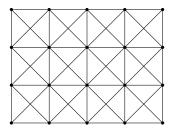


Figure 5: A King's grid.

Proposition 9. The
$$p \times q$$
 King's grid has learness $\lambda = \begin{cases} p-2 & \text{if } p=q \text{ are even} \\ \min\{p,q\}-1 & \text{otherwise} \end{cases}$

Proof. Recall the definition of hyperbolicity in Sec. 1. We denote $\delta(G)$ the hyperbolicity of the King's grid G of size $p \times q$. By [19, Theorem 2], we have $2\delta(G) - 1 \le \lambda(G) \le 2\delta(G)$. Furthermore, if $\lambda(G) = 2\delta(G) - 1$, then necessarily $\delta(G)$ is an integer and G contains an isometric $H_3^{\delta(G)-1}$ (see [19] and Figure 6 for the definition of the graphs H_1^k , H_2^k and H_3^k).

Without loss of generality, $p \leq q$. We first consider the case p < q. There are two subcases, depending on the parity of p:

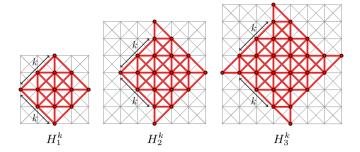


Figure 6: The graphs H_1^k , H_2^k and H_3^k from [19]. They isometrically embed in the King's grids of respective sizes $(2k+1) \times (2k+1)$, $(2k+2) \times (2k+3)$ and $(2k+4) \times (2k+4)$.

- Subcase p is even. The maximum k s.t. G contains an isometric H_1^k (H_2^k and H_3^k , resp.) are $k = \frac{p}{2} 1$ ($k = \frac{p}{2} 1$ and $k = \frac{p}{2} 2$, resp.). By [19, Theorem 3], we obtain $\delta(G) = \frac{p-1}{2}$. Since $\delta(G)$ is not an integer, we get $\lambda(G) = 2\delta(G) = p 1$.
- Subcase p is odd. The maximum k s.t. G contains an isometric H_1^k (H_2^k and H_3^k , resp.) are $k = \frac{p-1}{2}$ ($k = \frac{p-1}{2} 1$ and $k = \frac{p-1}{2} 2$, resp.). By [19, Theorem 3], we obtain $\delta(G) = \frac{p-1}{2}$. Since G does not contain an isometric $H_3^{\frac{p-1}{2}-1}$, we get $\lambda(G) = 2\delta(G) = p-1$.

Finally, let us assume p = q. If p = q = 2, then G is a clique, and so, $\delta(G) = \lambda(G) = 0$. Thus, from now on we assume $p \ge 3$. Again, there are two subcases, depending on the parity of p = q:

- Subcase p=q is even. The maximum k s.t. G contains an isometric H_1^k (H_2^k and H_3^k , resp.) are $k=\frac{p}{2}-1$ ($k=\frac{p}{2}-2$ and $k=\frac{p}{2}-2$, resp.). By [19, Theorem 3], we obtain $\delta(G)=\frac{p}{2}-1$. In particular, $\lambda(G)\leqslant p-2$. Since the $(p-1)\times q$ King's grid is an isometric subgraph of G with leanness p-2, $\lambda(G)\geqslant p-2$.
- Subcase p=q is odd. The maximum k s.t. G contains an isometric H_1^k (H_2^k and H_3^k , resp.) are $k=\frac{p-1}{2}$ ($k=\frac{p-1}{2}-1$ and $k=\frac{p-1}{2}-2$, resp.). By [19, Theorem 3], we obtain $\delta(G)=\frac{p-1}{2}$. Since G does not contain an isometric $H_3^{\frac{p-1}{2}-1}$, we get $\lambda(G)=2\delta(G)=p-1$.

Triangular grids

Lastly, the triangular grid of size $p \times q$ is obtained from the $p \times q$ grid by adding all edges in $\{\{(i,j),(i+1,j+1)\}:0\leq i< p-1,\ 0\leq j< q-1\}.$

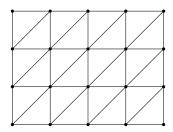


Figure 7: A triangular grid.

Proposition 10. The $p \times q$ triangular grid has learness $\lambda = \min\{p, q\} - 1$.

Proof. Recall the definition of the hyperbolicity $\delta(G)$ in Sec. 1. As proved in [15, Lemma 51], $\delta(G) = \frac{\min\{p,q\}-1}{2}$. Hence, $\lambda(G) \leq 2\delta(G) = \min\{p,q\}-1$. Without loss of generality, $p \leq q$. Let $x = (p-1,0), \ y = (0,q-1)$ and $u = (0,0), \ v = (p-1,p-1)$. Notice that d(x,y) = p+q-2. Furthermore, we have $u, v \in S_{p-1}(x,y)$. As a result, $\lambda(G) \geq \lambda(x,y) \geq d(u,v) = p-1$.

5 Planar graphs

In this section, we analyze the learness for subclasses of planar graphs.

5.1 Outerplanar graphs

An outerplanar graph is a planar graph in which all vertices belong to the outer face. Notice that outerplanar graphs form a hereditary class. Hence, we may restrict our study on leanness to biconnected outerplanar graphs. We refer to [35] for basic properties of outerplanar graphs. In particular, the weak dual graph of a biconnected outerplanar graph (that is obtained from its classic dual graph by removing the universal vertex corresponding to the outer face) is a tree.

We start with an important lemma for what follows:

Lemma 11. In a biconnected outerplanar graph, every slice is contained in a face.

Proof. Let $x,y \in V$ be arbitrary, and let $u,v \in S_k(x,y)$ for some k s.t. $0 < k < \operatorname{d}(x,y)$. Suppose by contradiction there is no face containing both u and v. Then, there exists some cut-edge $\{a,b\} \in E$ such that u and v are in two different connected components of $G - \{a,b\}$. Denote by $x = u_0, u_1, \cdots u_l = y$ and $x = v_0, v_1, \cdots v_l = y$ two shortest xy-paths with $u = u_k$ and $v = v_k$. The path $u_k u_{k-1} \cdots u_1 x v_1 v_2 \cdots v_k$ must go through a or b, so, there exists some p such that $0 \le p < k$ and $\{u_p, v_p\} \cap \{a, b\} \neq \emptyset$. W.l.o.g. suppose $u_p = a$. Similarly, there exists some q such that $k < q \le l$ and $\{u_q, v_q\} \cap \{a, b\} \neq \emptyset$. Then, $\min\{\operatorname{d}(u_p, u_q), \operatorname{d}(u_p, v_q)\} \leq \operatorname{d}(a, b) = 1$. However, since we have p < k < q, $\operatorname{d}(u_p, v_q) \geq \operatorname{d}(u_p, u_q) = q - p \geq 2$. A contradiction. Hence, there must be a face containing both u and v.

The concept of extraction is now introduced:

Definition 2. Let G be a biconnected outerplanar graph and C be one of its faces. The extraction of C, denoted \hat{C} , is a copy of C where every edge $\{u, v\}$ of C is turned into a triangle if u and v are equidistant to some vertex in $G - (C - \{u, v\})$.

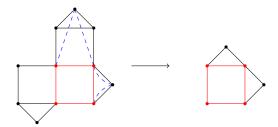


Figure 8: Extraction of face C (the red face).

This concept was first introduced in [12], but under the different name of "sunshine graph". In [12, Lemma 27], a linear-time algorithm is presented in order to compute the respective extractions of all faces in a biconnected outerplanar graph.

Theorem 12. Let \mathcal{F} denote the set of all (inner) faces in a biconnected outerplanar graph G. Then, the learness of G is $\lambda(G) = \max_{C \in \mathcal{F}} \lambda(\hat{C})$.

Proof. Let $x = u_0, u_1, \dots u_{l-1}, u_l = y$ and $x = v_0, v_1, \dots v_{l-1}, v_l = y$ be two shortest xy-paths for some $x, y \in V$. We pick some k such that 0 < k < d(x, y) and $d(u_k, v_k)$ is maximized. By Lemma 11, vertices u_k and v_k are contained in a face C. Let us prove that $d(u_k, v_k) \leq \lambda(\hat{C})$. For convenience, in what follows, for every $w \in C$, we denote by \hat{w} its copy in \hat{C} . Consider minimal indices i and j such that $u_i, v_i \in C$. We set vertex x' and index p as follows:

- If $u_i = v_j$, then $i = d(u_0, u_i) = d(v_0, v_j) = j$. We set $x' = \hat{u}_i$ and p = i.
- Else if i = j, vertices u_{i-1} and v_{i-1} are not in C, by minimality of i and j. Therefore, there is a face $C' \neq C$ containing u_{i-1} and v_{i-1} . Note that $C \cap C' = \{u_i, v_j\}$ must be a cut-edge. Since vertex x is equidistant to u_i, v_j and it is a vertex of $G (C \{u_i, v_j\})$, edge $\{u_i, v_j\}$ has been replaced in \hat{C} by a triangle. We set x' as the third vertex in this triangle, $x' \neq \hat{u}_i, \hat{u}_j$, and we set p = i 1.
- Otherwise, we may assume wlog that i < j. As in the previous case, it follows from the minimality of i, j that there exists some face $C' \neq C$ such that $u_{i-1}, v_{j-1} \in C'$ and $C \cap C' = \{u_i, v_j\}$ is a cut-edge. In particular, we get j = i + 1. Then, $x, u_1, \ldots u_i, v_j, v_{j+1}, \ldots y$ is also a shortest path. We set $x' = \hat{u}_i$ and p = j 1 = i.

By reverting the two shortest xy-paths, we can define vertex $y' \in \hat{C}$ and index q in a similar way as above. By doing so, we get the two shortest paths $x', \hat{u}_{p+1}, \cdots \hat{u}_k, \cdots \hat{u}_{q-1}, y'$ and $x', \hat{v}_{p+1}, \cdots \hat{v}_k, \cdots \hat{v}_{q-1}, y'$ in \hat{C} . As a result, $\lambda(\hat{C}) \geq \lambda(x', y') \geq d_{\hat{C}}(\hat{u}_k, \hat{v}_k) = d_G(u_k, v_k)$. This implies $\lambda(G) \leq \max_{C \in \mathcal{F}} \lambda(\hat{C})$.

Conversely, let $x', y' \in \hat{C}$ be arbitrary, for some face C of G. Let us call triangle vertex any vertex $w \in \hat{C}$ such that $\hat{u}\hat{v}w$ is a triangle and $\{u,v\} \in E(C)$. Observe that any internal vertex in a shortest path is not a triangle vertex. In particular, the vertices in $I(x',y') \setminus \{x',y'\}$ cannot be triangle vertices. If x' is a triangle vertex, and $N(x') = \{\hat{u}_1,\hat{v}_1\}$, then there exists a vertex x of $G - (C - \{u_1,v_1\})$ that is equidistant to u_1,v_1 . We replace edge $x'\hat{u}_1$ ($x'\hat{v}_1$, resp.) by a shortest xu_1 -path in G (a shortest xv_1 -path, resp.). We proceed similarly for y' if it is a triangle vertex. By doing so, all shortest x'y'-paths in \hat{C} can be extended into shortest xy-paths in G. Furthermore, every slice $S_{k'}(x',y')$, $0 < k' < d_{\hat{C}}(x',y')$, must be contained in some slice $S_k(x,y)$ in G. As a result, $\lambda(\hat{C}) \leq \lambda(G)$.

We complete Theorem 12 with a closed-form formula for the leanness of extractions. Namely:

Lemma 13. Let C be a face of length 4p+r, $0 \le r \le 3$, in a biconnected outerplanar graph G.

- If r is odd, then $\lambda(\hat{C}) = \begin{cases} 0 & \text{if } C = \hat{C} \\ 2p + \lfloor \frac{r}{2} \rfloor & \text{otherwise.} \end{cases}$
- If r=2, then $\lambda(\hat{C})=\begin{cases} 2p+1 & \text{if there are two diametrically opposed triangles in } \hat{C} \\ 2p & \text{otherwise.} \end{cases}$
- Else, $\lambda(\hat{C}) = 2p$.

Proof. By Lemma 11, $\lambda(\hat{C})$ is at most the maximum diameter of its faces. Therefore, $\lambda(\hat{C}) \leq \text{diam}(C)$. If r = 0, then $\lambda(\hat{C}) \geq \lambda(C) = \text{diam}(C) = 2p$. Thus from now on we assume r > 0. If furthermore $C = \hat{C}$, then

$$\lambda(\hat{C}) = \lambda(C) = \begin{cases} 0 & \text{if } r \text{ is odd} \\ 2p & \text{otherwise.} \end{cases}$$

Therefore, we also assume for the remainder of the proof $C \neq \hat{C}$.

First we assume r is odd. Let uvx be some triangle in \hat{C} that replaces edge $\{u,v\} \in E(C)$. Since C is odd, there exists a unique $y \in C$ such that $d(y,u) = d(y,v) = \dim(C) = 2p + \lfloor \frac{r}{2} \rfloor$. Then, $I(x,y) = \{x\} \cup C$. Furthermore, $S_{p+1}(x,y)$ is reduced to some diametral pair of C. Hence, $\lambda(\hat{C}) \geq \dim(C) = 2p + \lfloor \frac{r}{2} \rfloor$ in this case.

Assume r=2. As before, let uvx be some triangle in \hat{C} that replaces edge $\{u,v\} \in E(C)$. A vertex y of C cannot be equidistant to u,v because C is bipartite. Therefore, for every y of C, $\lambda(x,y) \leq \lambda(C)$. Now, let yu'v' be some triangle in \hat{C} that replaces edge $\{u',v'\} \in E(C)$. Without loss of generality, $G - \{\{u,v\},\{u',v'\}\}\}$ is made of a uu'-path and a vv'-path. If d(u,u') < d(v,v'), then $\lambda(x,y) = \lambda(u,u') \leq \lambda(C)$. Otherwise, d(u,u') = d(v,v') = 2p. In the latter situation, $I(x,y) \setminus \{x,y\} = C$. Furthermore, $S_p(x,y)$ is reduced to some diametral pair of C. Hence, $\lambda(\hat{C}) \geq \text{diam}(C) = 2p + 1$ in this case.

Corollary 14. The leanness of an outerplanar graph can be computed in linear time.

5.2 Bicyclic graphs

A bicyclic graph is a graph with m = n + 1 edges. Bicyclic graphs form a hereditary subclass of planar graphs. Furthermore, the nontrivial biconnected components of a bicyclic graph can only be of two different types, namely: they are either a cycle, or the union of three internally disjoint ab-paths for some vertices a, b.

For non-cycle biconnected components, the following lemma explains how to compute their leanness:

Lemma 15. Let G = (V, E) be the union of 3 disjoint ab-paths of respective lengths $f \leq g \leq h$. Then, we can compute $\lambda(G)$ in linear time.

Proof. Notice G contains three cycles C_1, C_2, C_3 of respective lengths $f + g \leq f + h \leq g + h$. Furthermore, C_1, C_2 must be isometric. This implies $\lambda(G) \geqslant \max\{\lambda(C_1), \lambda(C_2)\}$. We next give a closed-form formula for this lower bound:

Claim 1.
$$\max\{\lambda(C_1), \lambda(C_2)\} = \begin{cases} 2 \left\lfloor \frac{f+h}{4} \right\rfloor & \text{if } f, h \text{ are of same parity} \\ 2 \left\lfloor \frac{f+g}{4} \right\rfloor & \text{if } f, g \text{ and } h \text{ are of different parity} \\ 0 & \text{otherwise.} \end{cases}$$

Let us first consider f,h to be of same parity. Then, $\lambda(C_2)=2\left\lfloor\frac{f+h}{4}\right\rfloor$ because C_2 is even. If f,h are of different parity, then $\lambda(C_1)=0$ because C_1 is odd; otherwise, $\lambda(C_1)=2\left\lfloor\frac{f+g}{4}\right\rfloor\leqslant\lambda(C_2)$. Assume now f,h are of different parity. Then, $\max\{\lambda(C_1),\lambda(C_2)\}=\lambda(C_1)$ because C_2 is odd. In particular, $\lambda(C_1)=2\left\lfloor\frac{f+g}{4}\right\rfloor$ if f,g are of same parity, else $\lambda(C_1)=0$. \diamond

Case f = g = h. In order to compute $\lambda(G)$, by Lemma 1, it suffices to consider far-apart pairs (x,y) of G. For any $(x,y) \neq (a,b)$, let $x,y \in C_i$ for some i. Since (x,y) must be far-apart, $d(x,y) = \text{diam}(C_i) = f$. In particular, $\{x,y\} \cap \{a,b\} = \emptyset$. Furthermore, $I(x,y) \subseteq C_i$ because there is no shortest xy-path that goes by a,b. Hence, $\lambda(x,y) \leq \lambda(C_i) = 2 \left\lfloor \frac{f}{2} \right\rfloor$. Since we also have $\lambda(a,b) = 2 \left\lfloor \frac{f}{2} \right\rfloor$, it follows that $\lambda(G) = \lambda(a,b) = 2 \left\lfloor \frac{f}{2} \right\rfloor$.

Case f < h. Recall that $\lambda(G) \ge \max\{\lambda(C_1), \lambda(C_2)\}$. Conversely, we prove that for most pairs x, y of vertices we have $\lambda(x, y) \le \max\{\lambda(C_1), \lambda(C_2)\}$.

Claim 2. For $x, y \in V$, $\lambda(x, y) \leq \max\{\lambda(G_1), \lambda(G_2)\}$ if one of the following conditions hold:

1.
$$x, y \in C_1$$
;

- 2. $x, y \in C_2$;
- 3. $x \in C_1 \setminus C_2$, and min $\{d(x, a), d(x, b)\} < \frac{g-f}{2}$;
- 4. $d(x, a) + d(a, y) \neq d(x, b) + d(b, y)$.

In particular, $\lambda(G) = \max\{\lambda(C_1), \lambda(C_2)\}\$ if g, h are of different parity.

If $x, y \in C_1$ then $\lambda(x, y) \leq \lambda(C_1)$ because C_1 is a convex subgraph of G. In particular, $\lambda(a, b) \leq \lambda(C_1)$. Assume now $x, y \in C_2$. Either $I(x, y) \subseteq C_2$, or f = g and there exists a shortest xy-path that goes by a, b. Hence, $\lambda(x, y) \leq \max\{\lambda(C_2), \lambda(a, b)\} \leq \max\{\lambda(C_1), \lambda(C_2)\}$.

Since C_1, C_2 cover G, up to symmetries, we are left considering the pairs x, y such that $x \in C_1 \setminus C_2, y \in C_2 \setminus C_1$. In particular, x is on the second shortest ab-path, of length g, while g must be on the longest ab-path, of length g. Assume g0 as g1, or equivalently, g - g2, or equivalently, g - g3, or equivalently, g - g4, and g5 as g6. This implies every shortest g7 as g8, as g9, and g9, we get g9, we get g9, we get g9, and g9, and g9, and g9, and g9, and g9, and g9, we get g9, we get g9, we get g9, which is a sum of g9, and g9, and

Assume $d(x, a) + d(a, y) \neq d(x, b) + d(b, y)$. Without loss of generality, d(x, a) + d(y, a) < d(x, b) + d(y, b). Then again, every shortest xy-path must go by a, so, $\lambda(x, y) = \max\{\lambda(x, a), \lambda(y, a)\} \leq \max\{\lambda(C_1), \lambda(C_2)\}$. Finally, note that if none of the four conditions of our claim hold, then g + h = d(x, a) + d(y, a) + d(x, b) + d(y, b) must be even. \diamond

The two above claims imply a closed-form formula for $\lambda(G)$ if g,h are of different parity, namely:

$$\lambda(G) = \begin{cases} 2 \left\lfloor \frac{f+h}{4} \right\rfloor & \text{if } f, h \text{ are of different parity than } g \\ 2 \left\lfloor \frac{f+g}{4} \right\rfloor & \text{if } f, g \text{ are of different parity than } h. \end{cases}$$

From now on, we assume g, h to be of same parity.

Claim 3. Let $(x_{\text{lim}}, y_{\text{lim}})$ be the unique pair of vertices in $(C_1 \setminus C_2) \times (C_2 \setminus C_1)$ so that $d(x_{\text{lim}}, a) = \left\lceil \frac{g-f}{2} \right\rceil$ and $d(x_{\text{lim}}, y_{\text{lim}}) = \frac{g+h}{2}$. Then, $\lambda(G) = \max\{\lambda(C_1), \lambda(C_2), \lambda(x_{\text{lim}}, y_{\text{lim}})\}$.

Assume the existence of a pair (x, y) such that $\lambda(x, y) > \max\{\lambda(C_1), \lambda(C_2)\}$ (else, we are done). We deduce from the previous claim that, up to symmetries:

- $x \in C_1 \setminus C_2, y \in C_2 \setminus C_1;$
- $\frac{g-f}{2} \leqslant \mathrm{d}(x,a) \leqslant \frac{g+f}{2};$
- $d(x,y) = \frac{g+h}{2}.$

By symmetry, we may further assume $d(x,a) \leq d(x,b)$. Then, $\frac{g-f}{2} \leq d(x,a) \leq \frac{g}{2}$.

There are at most three shortest xy-paths, namely: one going by a and not by b, one going by b but not by a, and one going by both a and b. Furthermore, the shortest xy-path that goes by a, b exists if and only if d(x, b) = g - d(x, a) = f + d(x, a) (the other two shortest paths always exist, because $a, b \in I(x, y)$, and they cover C_3).

Assume for what follows $d(x,a) > \left\lceil \frac{g-f}{2} \right\rceil$. Then, there are two shortest xy-paths. Let k be such that 0 < k < d(x,y) and $diam(S_k(x,y))$ is maximized. We have that $S_k(x,y) = \{u,v\}$, with u,a and v,b being on a same shortest xy-path respectively. Let us pick $x' \in S_1(x,a)$, $y' \in S_1(y,b)$. Note that x',y' are unique. Furthermore, we still have $\frac{g-f}{2} \leqslant d(x',a) \leqslant \frac{g}{2}$ and $d(x',y') = \frac{g+h}{2}$. Observe that $u',v' \in S_k(x',y')$ where $u' \in S_1(u,y)$, $v' \in S_1(v,x)$. Since all induced u'v'-paths in G are at least as long as the corresponding induced uv-paths, $\lambda(x',y') \geqslant d(u',v') \geqslant d(u,v) = \lambda(x,y)$.

The remainder of the proof consists in computing $\lambda(x_{\lim}, y_{\lim})$. For that, let P_a be the shortest $x_{\lim}y_{\lim}$ -path that goes by a and does not go by b. Let P_b be defined similarly. For every k such that $0 < k < \frac{g+h}{2}$, let $u_k \in P_a$, $v_k \in P_b$ be at distance k to x_{\lim} , and let $d_k = d(u_k, v_k)$.

Claim 4.
$$\lambda(x_{\lim}, y_{\lim}) \le \max\{\lambda(C_1)\} \cup \{d_k : 0 < k < (g+h)/2\}.$$

The claim is trivial if there are only two shortest $x_{\lim}y_{\lim}$ -paths. So, assume $d(x_{\lim}, a) = \frac{g-f}{2}$, and let P_{ab} be the shortest $x_{\lim}y_{\lim}$ -path that goes by a, b. For k such that $\frac{g-f}{2} < k < \frac{g+f}{2}$, let $w_k \in P_{ab}$ be at distance k to x_{\lim} . Observe that $v_k, u_k \in S_k(x_{\lim}, b)$, so, $d(v_k, u_k) \leq \lambda(x, b) \leq \lambda(C_1)$. Furthermore, $d(u_k, v_k) \geq d(u_k, w_k)$ because all induced $u_k v_k$ -paths are at least as long as the corresponding $u_k w_k$ -paths. Hence, $d(x_k, y_k) \leq d(x_k, y_k) \leq d(x$

In order to compute d_k , we need to consider the ab-paths that contain u_k, v_k .

• Assume $k \leq d(x_{\lim}, a)$. Then, u_k, v_k both lie on the second ab-path, of length g. We have $d_k = \min\{2k, f + g - 2k\}$. Since the function $t \mapsto \min\{2t, f + g - 2t\}$ is maximized at t = (f + g)/4, we get

$$d_k \leqslant \ell_1 = \begin{cases} 2 \operatorname{d}(x_{\lim}, a) = 2 \left\lceil \frac{g - f}{2} \right\rceil & \text{if } \operatorname{d}(x_{\lim}, a) \leqslant \left\lfloor \frac{g + f}{4} \right\rfloor \\ \max \left(2 \left\lfloor \frac{f + g}{4} \right\rfloor, f + g - 2 \left\lceil \frac{f + g}{4} \right\rceil \right) & \text{else.} \end{cases}$$

• Assume $d(x_{\lim}, a) < k \le d(x_{\lim}, b)$. Now, u_k is on the third ab-path, of length h. We have $d_k = \min\{2k, g+h-2k, f+g-2d(x_{\lim}, a)\} = \min\{2k, g+h-2k, f+g-2\left\lceil\frac{g-f}{2}\right\rceil\}$. Since the function $t \mapsto \min\{2t, g+h-2t\}$ is maximized at $t = \frac{g+h}{4}$, we get

$$d_k \leqslant \ell_2 = \begin{cases} \min\left\{2\operatorname{d}(x_{\lim}, b), f + g - 2\left\lceil\frac{g - f}{2}\right\rceil\right\} & \text{if } \operatorname{d}(x_{\lim}, b) \leqslant \left\lfloor\frac{g + h}{4}\right\rfloor \\ = \min\left\{2\left\lfloor\frac{g + f}{2}\right\rfloor, f + g - 2\left\lceil\frac{g - f}{2}\right\rceil\right\} & \text{if } \operatorname{d}(x_{\lim}, b) \leqslant \left\lfloor\frac{g + h}{4}\right\rfloor \\ \min\left\{\max\left(2\left\lfloor\frac{g + h}{4}\right\rfloor, g + h - 2\left\lceil\frac{g + h}{4}\right\rceil\right), f + g - 2\left\lceil\frac{g - f}{2}\right\rceil\right\} & \text{else.} \end{cases}$$

• Otherwise, $k > d(x_{\lim}, b)$. Now, v_k is also on the third ab-path. We have $d_k = \min\{2k - g + f, g + h - 2k\}$. Since the function $t \mapsto \min\{2t - g + f, g + h - 2t\}$ is maximized at $t = \frac{2g + h - f}{4}$, we get

$$d_k \leqslant \ell_3 = \begin{cases} g + h - 2 \operatorname{d}(x_{\lim}, b) = g + h - 2 \left\lfloor \frac{g + f}{2} \right\rfloor & \text{if } \operatorname{d}(x_{\lim}, b) \geqslant \left\lceil \frac{2g + h - f}{4} \right\rceil \\ \max \left(2 \left\lfloor \frac{2g + h - f}{4} \right\rfloor - g + f, g + h - 2 \left\lceil \frac{2g + h - f}{4} \right\rceil \right) & \text{else.} \end{cases}$$

Finally,
$$\lambda(x_{\lim}, y_{\lim}) = \max\{\ell_1, \ell_2, \ell_3\}.$$

Corollary 16. The learness of a bicyclic graph can be computed in linear time.

6 Other classes

In this last section, we consider some nonplanar graph classes. Unlike the graphs in Sec. 4 & 5, all graphs considered in what follows have bounded learness. We obtain linear-time algorithms for computing their learness, for all graph classes considered except for bisplit graphs. For the latter class, a conditional quadratic lower bound is proved assuming the SETH (see Sec. 6.4).

6.1 Chordal graphs

Recall that a graph is chordal if and only if every induced cycle has length three. A block graph is a graph whose biconnected components are cliques. Note that block graphs are a subclass of both chordal graphs and geodetic graphs.

Proposition 17. For a chordal graph G = (V, E), we have

$$\lambda(G) = \begin{cases} 0 & \text{if } G \text{ is a block graph} \\ 1 & \text{else.} \end{cases}$$

In particular, the leanness of a chordal graph can be computed in linear time.

Proof. It has been proved in [9] that every slice in a chordal graph G is a clique. Hence, $\lambda(G) \leq 1$. Furthermore, a necessary condition for having $\lambda(G) = 0$ is that G must be diamond-free. Diamond-free chordal graphs are exactly the block graphs [1], so, they are geodetic graphs. \square

6.2 Cographs

Cographs can be recursively defined as follows:

- the one-vertex graph is a cograph;
- the disjoint union of two cographs is also a cograph;
- and the complement of a cograph is also a cograph.

The join of two graphs G_1, G_2 , denoted $G_1 \oplus G_2$, is obtained by adding all possible edges between $V(G_1)$ and $V(G_2)$. Notice that the join of two cographs is also a cograph. In fact, every connected cograph with n > 1 vertices must be the join of two cographs.

An equivalent definition of cographs is that they are exactly the P_4 -free graphs [14]. Therefore, $\lambda(G) \leq \text{diam}(G) \leq 2$ for every cograph G. Next, we present a characterization of the leanness on cographs, that can be verified in linear time.

Proposition 18. For a cograph G = (V, E), we have

$$\begin{cases} \lambda(G) \leqslant 1 & \text{if } G \text{ is chordal} \\ \lambda(G) = 2 & \text{otherwise.} \end{cases}$$

Proof. Since G is P_4 -free, every induced cycle in G has length at most four. So, if G is C_4 -free, then G is chordal, so $\lambda(G) \leq 1$. Otherwise, $\lambda(G) \geq \lambda(C_4) = 2$. Since we also have $\lambda(G) \leq \operatorname{diam}(G) \leq 2$, we obtain $\lambda(G) = 2$.

Recall that chordal graphs can be recognized in linear time [36]. Therefore:

Corollary 19. The learness of a cograph can be computed in linear time.

6.3 Distance-hereditary graphs

A graph is called *distance-hereditary* if every induced subgraph is also isometric (i.e., distance-preserving). Our next result shows that the leanness of distance-hereditary graphs can be characterized in the exact same way as for cographs.

Proposition 20. For a distance-hereditary graph G = (V, E), we have

$$\begin{cases} \lambda(G) \leqslant 1 & \text{if G is chordal} \\ \lambda(G) = 2 & \text{otherwise}. \end{cases}$$

In particular, the leanness of G can be computed in linear time.

Proof. For every $k \geqslant 5$, we have $\operatorname{diam}(P_{k-1}) = k-2 > \left\lfloor \frac{k}{2} \right\rfloor = \operatorname{diam}(C_k)$. This implies that every induced cycle in a distance-hereditary graph G must have length at most four. In particular, if G is C_4 -free, then it is chordal, so, $\lambda(G) \leqslant 1$. Otherwise, $\lambda(G) \geqslant 2$. In order to complete the proof, we show that $\lambda(G) \leqslant 2$ for every distance-hereditary graph G. By contradiction, let G = (V, E) be a minimum-size distance-hereditary graph such that $\lambda(G) \geqslant 3$. There is no degree-one vertex $v \in V$ because otherwise $\lambda(G) = \lambda(G \setminus v)$, thus contradicting the minimality of G. By [1, Theorem 1], there exist twin vertices u, v in G, i.e., we have $N(u) \setminus \{v\} = N(v) \setminus \{u\}$. However, in this situation $\lambda(G) \leqslant \max\{\lambda(G \setminus v), \operatorname{d}(u, v)\} \le \max\{\lambda(G \setminus v), 2\}$. The latter either contradicts the minimality of G, or that $\lambda(G) \geqslant 3$.

We stress that distance-hereditary graphs are a superclass of cographs. Hence, our result for distance-hereditary graphs subsumes the one for cographs. Nevertheless, the proof for cographs is, in our opinion, slightly simpler.

6.4 Lower bound on bisplit graphs

A split graph is a graph where the vertices can be bipartitioned in a clique and a stable set. A bisplit graph is a graph where the vertices can be bipartitioned in a biclique (i.e., complete bipartite subgraph) and an independent set. We stress that split graphs (bisplit graphs, resp.) have diameter at most three (at most four, resp.). Hence, the leanness of split graphs and bisplit graphs is bounded. Furthermore, split graphs are a special case of chordal graphs, so, their leanness can be computed in linear time. Perhaps surprisingly, we prove next that the situation is different for bisplit graphs.

Recall the Strong Exponential-Time Hypothesis (SETH) and the DISJOINTSET problem were introduced in Sec. 3.2.

Theorem 21. Under the SETH, deciding whether a bisplit graph G = (V, E) has learness 2 or 4 requires $\Omega(n^{2-o(1)})$ time. The result holds even if G has $n^{1+o(1)}$ edges.

Proof. Let $A, B \subset \mathcal{P}(C)$ be some instance of DISJOINTSET, where A, B are families of n sets over come common universe $C, |C| = n^{o(1)}$. Recall that our objective is to decide whether there exist $a, \in A, b \in B$ such that $a \cap b = \emptyset$.

The graph G = (V, E) is constructed as follows (see Fig. 9).

- $V = A \cup B \cup C \cup C' \cup \{u, v, x, y, z\}$, where C' is a disjoint copy of C and u, v, x, y, z are fresh new vertices. For every $c \in C$, let us denote c' the corresponding element in C';
- For every set $a \in A$ and every element $c \in C$, we add the two edges ac, ac' if and only if $c \in a$. We proceed similarly for every set $b \in B$ and every element $c \in C$;
- We add all possible edges between: u and A, v and B, x and C, y and C';
- Finally, vertex z is adjacent to u, v and to every vertex of $C \cup C'$.

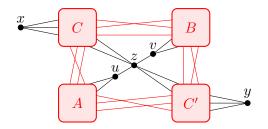


Figure 9: Reduction from DISJOINTSET.

Note that we can construct G from A,B,C in $n^{1+o(1)}$ time. Furthermore, G is a bisplit graph: $\{x,y\}\cup A\cup B$ is an independent set, and there is a complete bipartite subgraph with respective partite sets $\{z\}$ and $\{u,v\}\cup C\cup C'$. The graph G is also bipartite: its partite sets are $\{x,y,z\}\cup A\cup B$ and $\{u,v\}\cup C\cup C'$. Therefore, $\lambda(G)$ must be even. Furthermore, $\lambda(G)\leqslant \mathrm{diam}(G)\leqslant 2\operatorname{ecc}(z)=4$. Hence, we are left deciding whether $\lambda(G)=2$ or $\lambda(G)=4$.

Assume there exist $a \in A$, $b \in B$ such that $a \cap b = \emptyset$. By construction of G, $S_2(x,y) = A \cup B \cup \{z\}$. Therefore, $\lambda(G) \geqslant \lambda(x,y) \geqslant d(a,b) = 4$. Conversely, assume that $a \cap b \neq \emptyset$ for every $a \in A$, $b \in B$. Then, x, y is the only pair of vertices such that d(x,y) = 4. Since we have $\lambda(x,y) = \text{diam}(S_2(x,y)) = \text{diam}(A \cup B \cup \{z\}) = 2$, it follows that $\lambda(G) = 2$.

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