Inapproximability of H-Transversal/Packing

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──── **Abstract** ────

Given an undirected graph $G = (V_G, E_G)$ and a fixed pattern graph $H = (V_H, E_H)$ with k vertices, we consider the H-Transversal and H-Packing problems. The former asks to find the smallest $S \subseteq V_G$ such that the subgraph induced by $V_G \setminus S$ does not have H as a subgraph, and the latter asks to find the maximum number of pairwise disjoint k-subsets $S_1, ..., S_m \subseteq V_G$ such that the subgraph induced by each S_i has H as a subgraph.

We prove that if H is 2-connected, H-Transversal and H-Packing are almost as hard to approximate as general k-Hypergraph Vertex Cover and k-Set Packing, so it is NP-hard to approximate them within a factor of $\Omega(k)$ and $\widetilde{\Omega}(k)$ respectively. We also show that there is a 1-connected H where H-Transversal admits an $O(\log k)$-approximation algorithm, so that the connectivity requirement cannot be relaxed from 2 to 1. For a special case of H-Transversal where H is a (family of) cycles, we mention the implication of our result to the related Feedback Vertex Set problem, and give a different hardness proof for directed graphs.

1 Introduction

Given a collection of subsets $S_1, ..., S_m$ of the underlying set U, the *Set Transversal* problem asks to find the smallest subset of U that intersects every S_i, and the *Set Packing* problem asks to find the largest subcollection $S_{i_1}, ..., S_{i_{m'}}$ which are pairwise disjoint.[1] It is clear that optimum of the former is always at least that of the latter (i.e. weak duality holds). Studying the (approximate) reverse direction of the inequality (i.e. strong duality) as well as the complexity of both problems for many interesting classes of set systems is arguably the most studied paradigm in combinatorial optimization.

This work focuses on set systems where the size of each set is bounded by a constant k. With this restriction, Set Transversal and Set Packing are known as k-Hypergraph Vertex cover (k-HVC) and k-Set Packing (k-SP), respectively. This assumption significantly simplifies the problem since there are at most n^k sets. While there is a simple factor k-approximation algorithm for both problems, it is NP-hard to approximate k-HVC and k-SP within a factor less than $k - 1$ [21] and $O(\frac{k}{\log k})$ [34] respectively.

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[1] These problems are called many different names in the literature. Set Transversal is also called Hypergraph Vertex Cover, Set Cover (of the dual set system), and Hitting Set. Set Packing is also called Hypergraph Matching. We try to use Transversal / Packing unless another name is established in the literature (e.g. k-Hypergraph Vertex Cover).

Given a large graph $G = (V_G, E_G)$ and a fixed graph $H = (V_H, E_H)$ with k vertices, one of the natural attempts to further restrict set systems is to set $U = V_G$, and take the collection of subsets to be all *copies* of H in G (formally defined in the next subsection). This natural representation in graphs often results in a deeper understanding of the underlying structure and better algorithms, with Maximum Matching ($H = K_2$) being the most well-known example. Kirkpatrick and Hell [39] proved that Maximum Matching is essentially the only case where H-Packing can be solved exactly in polynomial time – unless H is the union of isolated vertices and edges, it is NP-hard to decide whether V_G can be partitioned into k-subsets each inducing a subgraph containing H. A similar characterization for the edge version (i.e. $U = E_G$) was obtained much later by Dor and Tarsi [22].

We extend these results by studying the approximability of H-Transversal and H-Packing. We use the term *strong inapproximability* to denote NP-hardness of approximation within a factor $\Omega(k/polylog(k))$. We give a simple sufficient condition that implies strong inapproximability – if H is 2-vertex connected, H-Transversal and H-Packing are almost as hard to approximate as k-HVC and k-SP. We also show that there is a 1-connected H where H-Transversal admits an $O(\log k)$-approximation algorithm, so 1-connectivity is not sufficient for strong inapproximability for H-Transversal. It is an interesting open problem whether 1-connectivity is enough to imply strong inapproximability of H-Packing, or there is a class of connected graphs where H-Packing admits a significantly nontrivial approximation algorithm (e.g. factor k^ϵ for some $\epsilon < 1$).

Our results give an unified answer to questions left open in many independent works studying a special case where H is a cycle or clique, and raises some new open questions. In the subsequent subsections, we state our main results, review related work, and state potential future directions.

1.1 Problems and Our Results

Given an undirected graphs $G = (V_G, E_G)$ and $H = (V_H, E_H)$ with $|V_H| = k$, we define the following problems.

- H-Transversal asks to find the smallest $F \subseteq V_G$ such that the subgraph of G induced by $V_G \setminus F$ does not have H as a subgraph.
- H-Packing asks to find the maximum number of pairwise disjoint k-subsets of $S_1, ..., S_m$ of V_G such that the subgraph induced by each S_i has H as a subgraph.

Our main result states that 2-connectivity of H is sufficient to make H-Transversal and H-Packing hard to approximate.

▶ **Theorem 1.** *If H is a 2-vertex connected with k vertices, unless* NP $\subseteq$ BPP, *no polynomial time algorithm approximates H-Transversal within a factor better than $k - 1$, and H-Packing within a factor better than $\Omega(\frac{k}{\log^7 k})$.*

Let k-Star denote $K_{1,k-1}$, the complete bipartite graph with 1 and $k - 1$ vertices on each side. The following theorem shows that k-Star Transversal admits a good approximation algorithm, so the assumption of 2-connectedness in Theorem 1 is required for strong inapproximability of H-Transversal.

▶ **Theorem 2.** *k-Star Transversal can be approximated within a factor of $O(\log k)$ in polynomial time.*

This algorithmic result matches $\Omega(\log k)$-hardness of k-Star Transversal via a simple reduction from Minimum Dominating Set on degree-k graphs [16]. This problem has the

following equivalent but more natural interpretation: given a graph $G = (V_G, E_G)$, find the smallest $F \subseteq V_G$ such that the subgraph induced by $V_G \setminus F$ has maximum degree at most $k - 2$. Our algorithm, which uses iterative roundings of 2-rounds of Sherali-Adams hierarchy of linear programming (LP) followed by a simple greedy algorithm for *Constrained Set Cover*, is also interesting in its own right, but we defer the details to Appendix A.

Our hardness results for transversal problems rely on hardness of k-HVC which is NP-hard to approximate within a factor better than $k - 1$ [21]. Our hardness results for packing problems rely on hardness of Maximum Independent Set on graphs with maximum degree k and girth strictly *greater than g* (MIS-k-g). Almost tight inapproximability of MIS on graphs with maximum degree k (MIS-k) is recently proved in Chan [11], which rules out an approximation algorithm with ratio better than $\Omega(\frac{k}{\log^4 k})$. We are able to extend his result to MIS-k-g with losing only a polylogarithmic factor. All applications in this work require $g = \Theta(k)$.

▶ **Theorem 3.** *For any constants k and g, unless* NP $\subseteq$ BPP, *no polynomial time algorithm approximates MIS-k-g within a factor of* $\Omega(\frac{k}{\log^7 k})$.

We remark that assuming the Unique Games Conjecture (UGC) slightly improves our hardness ratios through better hardness of k-HVC [38] and MIS-k [3], and even simplifies the proof for some problems (e.g. k-Clique Transversal) through structured hardness of k-HVC [5]. Indeed, an earlier (unpublished) version of this work [30] relied on the UGC to prove that MIS-k-k is hard to approximate within a factor of $\Omega(\frac{k}{\log^4 k})$, while only giving $\widetilde{\Omega}(\sqrt{k})$-factor hardness without it. Now that we obtain almost matching hardness, we focus on proving hardness results without the UGC.

1.2 Related Work and Special Cases

After the aforementioned work characterizing those pattern graphs H admitting the existence of a polynomial-time exact algorithm for H-Packing [39, 22], Lund and Yannakakis [45] studied the maximization version of H-Transversal (i.e. find the largest $V' \subseteq V_G$ such that the subgraph induced by V' does not have H as a subgraph), and showed it is hard to approximate within factor $2^{\log^{1/2-\epsilon} n}$ for any $\epsilon > 0$. They also mentioned the minimization version of two extensions of H-Transversal. The most general node-deletion problem is APX-hard for every nontrivial hereditary (i.e. closed under node deletion) property, and the special case where the property is characterized by a finite number of forbidden subgraphs (i.e. $\{H_1, ..., H_l\}$-Transversal in our terminology) can be approximated with a constant ratio. They did not provide explicit constants (one trivial approximation ratio for $\{H_1, ..., H_l\}$-Transversal is $\max(|V_{H_1}|, ..., |V_{H_l}|)$), and our result can be viewed as a quantitative extension of their inapproximability results for the special case of H-Transversal.

H-Transversal / Packing has been also studied outside the approximation algorithms community. The duality between our H-Transversal and H-Packing is closely related to the famous Erdős-Pósa property actively studied in combinatorics. The recent work of Jansen and Marx [36] considered problems similar to our H-Packing with respect to fixed-parameter tractability (FPT).

Many other works on H-Transversal / Packing focus on a special case where H is a cycle or clique. We define k-Cycle (resp. k-Clique) to be the cycle (resp. clique) on k vertices.

1.2.1 Cycles

The initial motivation for our work was to prove a super-constant factor inapproximability for the Feedback Vertex Set (FVS) problem without relying on the Unique Games Conjecture. Given a (directed) graph G, the FVS problem asks to find a subset F of vertices with the minimum cardinality that intersects every cycle in the graph (equivalently, the induced subgraph $G \setminus F$ is acyclic). One of Karp's 21 NP-complete problems, FVS has been a subject of active research for many years in terms of approximation algorithms and fixed-parameter tractability (FPT). For FPT results, see [8, 14, 18, 15] and references therein.

FVS on undirected graphs has a 2-approximation algorithm [4, 7, 17], but the same problem is not well-understood in directed graphs. The best approximation algorithm [48, 26, 25] achieves an approximation factor of $O(\log n \log \log n)$. The best hardness result follows from a simple approximation preserving reduction from Vertex Cover, which implies that it is NP-hard to approximate FVS within a factor of 1.36 [20]. Assuming UGC [37], it is NP-hard to approximate FVS in directed graphs within any constant factor [29, 50] (we give a simpler proof in [30]). The main challenge is to *bypass* the UGC and to show a super-constant inapproximability result for FVS assuming only $P \neq NP$ or $NP \not\subseteq BPP$.

By Theorem 1, we prove that k-Cycle Transversal is hard to approximate within factor $\Omega(k)$. In the full version of this work [31], we prove the following theorem that improves the result of Theorem 1 in the sense that in the completeness case, a small number of vertices not only intersect cycles of length exactly k, but intersect every cycle of length $3, 4, ..., O(\frac{\log n}{\log \log n})$.

▶ **Theorem 4.** *Fix an integer $k \geq 3$ and $\epsilon \in (0, 1)$. Given a graph $G = (V_G, E_G)$ (directed or undirected), unless $NP \subseteq BPP$, there is no polynomial time algorithm to tell apart the following two cases.*

- *Completeness: There exists $F \subseteq V_G$ with $\frac{1}{k-1} + \epsilon$ fraction of vertices that intersects every cycle of at most length $O(\frac{\log n}{\log \log n})$ (hidden constant in O depends on k and ϵ).*
- *Soundness: Every subset F with less than $1 - \epsilon$ fraction of vertices does not intersect at least one cycle of length k. Equivalently, any subset with more than ϵ fraction of vertices has a cycle of length exactly k in the induced subgraph.*

This can be viewed as some (modest) progress towards showing inapproximability of FVS in the following sense. Consider the following standard linear programming (LP) relaxation for FVS.

$$\min \sum_{v \in V_G} x_v \quad \text{subject to} \quad \sum_{v \in C} x_v \geq 1 \quad \forall \text{ cycle } C , \quad \text{and} \quad 0 \leq x_v \leq 1 \;\; \forall v \in V_G$$

The integrality gap of the above LP is upper bounded by $O(\log n)$ for undirected graphs [6] and $O(\log n \log \log n)$ for directed graphs [26]. Suppose in the completeness case, there exists a set of measure c that intersects every cycle of length at most $\log^{1.1} n$ (or any number bigger than the known integrality gaps). If we remove these vertices and consider the above LP on the remaining subgraphs, since every cycle is of length at least $\log^{1.1} n$, setting $x_v = 1/\log^{1.1} n$ is a feasible solution, implying that the optimal solution to the LP is at most $n/\log^{1.1} n$. Since the integrality gap is at most $O(\log n \log \log n)$, we can conclude that the remaining cycles can be hit by at most $O(n \log \log n / \log^{0.1} n) = o(n)$ vertices, extending the completeness result to every cycle. Thus, improving our result to hit cycles of length $\omega(\log n \log \log n)$ in the completeness case will prove a factor-$\omega(1)$ inapproximability of FVS.

Another interesting aspect about Theorem 4 is that it also holds for undirected graphs. This should be contrasted with the fact that undirected graphs admit a 2-approximation

algorithm for FVS, suggesting that to overcome $\log n$-cycle barrier mentioned above, some properties of directed graphs must be exploited. Towards developing a directed graph specific approach, we also present a different reduction technique called *labeling gadget*. It has an additional advantage of being derandomized and assumes only $\mathsf{P} \neq \mathsf{NP}$.

For cycles of bounded length, Kortsarz et al. [41] studied k-Cycle Edge Transversal, and suggested a $(k-1)$-approximation algorithm as well as proved that improving the ratio 2 for K_3 will have the same impact on Vertex Cover, refuting the Unique Games Conjecture [38].

For the dual problem of packing cycles of any length, called Vertex-Disjoint Cycle Packing (VDCP), the results of [42, 28] imply that the best approximation factor by any polynomial time algorithm lies between $\Omega(\sqrt{\log n})$ and $O(\log n)$. In a closely related problem Edge-Disjoint Cycle Packing (EDCP), the same papers showed that $\Theta(\log n)$ is the best possible. In directed graphs the vertex and edge version have the same approximability, the best known algorithms achieves $O(\sqrt{n})$-approximation while the best hardness result remains $\Omega(\log n)$.

Variants of k-Cycle Packing have also been considered in the literature. Rautenbach and Regen [47] studied k-Cycle Edge Packing on graphs with girth k and small degree. Chalermsook et al. [10] studied a variant of k-Cycle Packing on directed graphs for $k \geq n^{1/2}$ where we want to pack as many disjoint cycles of length at most k as possible, and proved that it is NP-hard to approximate within a factor of $n^{1/2-\epsilon}$. This matches the algorithm implied by [42].

1.2.2 Cliques

Minimum Maximal (resp. Maximum) Clique Transversal asks to find the smallest subset of vertices that intersects every maximal (resp. maximum) clique in the graph. In mathematics, Tuza [51] and Erdős et al. [24] started to estimate the size of the smallest such set depending on structure of graphs. See the recent work of Shan et al. [49] and references therein. In computer science, exactly computing the smallest set on special classes of graphs appears in many works [32, 44, 12, 23, 43].

Both the edge and vertex version of k-Clique Packing also have been studied actively both in mathematics and computer science. In mathematics, the main focus of research is lower bounding the maximum number of edge or vertex-disjoint copies of K_k in very dense graphs (note that even K_3 does not exist in $K_{n,n}$ which has $2n$ vertices and n^2 edges). See the recent paper [52] or the survey [53] of Yuster. The latter survey also mentions approximation algorithms, including APX-hardness and the general approximation algorithm for k-Set Packing which now achieves $\frac{k+1+\epsilon}{3}$ for the vertex version and $\frac{\binom{k}{2}+1+\epsilon}{3}$ for the edge version [19]. Feder and Subi [27] considered H-Edge Packing and showed APX-hardness when H is k-cycle or k-clique. Chataigner et al. [13] considered an interesting variant where we want to pack vertex-disjoint cliques of any size to maximize the total number of edges of the packed cliques, and proved APX-hardness and a 2-approximation algorithm. Exact algorithms for special classes of graphs have been considered in [9, 33, 35, 40].

1.3 Open Problems

For H-Transversal, 1-connectivity is not sufficient for strong hardness, because k-Star Transversal admits an $O(\log k)$-approximation algorithm by Theorem 2. It is open whether 1-connectivity is sufficient or not for such strong hardness for H-Packing. k-Star Packing is at least as hard as MIS-k by a trivial reduction, but the approximability of k-Path Packing appears to be still unknown. Whether k-Path Transversal admits a factor $o(k)$ approximation

algorithm is also an intriguing question. For directed acyclic graphs, Svensson [50] proved that it is Unique Games-hard to approximate k-Path Transversal within a factor better than k.

The approximability of H-Edge Transversal and H-Edge Packing is less understood than the vertex versions. Proving tight characterizations for the edge versions similar to Theorem 1 is an interesting open problem.

1.4 Organization

The rest of the main body is devoted to proving Theorem 1 for H-Transversal / Packing and Theorem 3 for MIS-k-g. Section 2 recalls and extends previous hardness results for the problems we reduce from; Sections 3 and 4 prove hardness of H-Transversal and H-Packing respectively. Appendix A gives an $O(\log k)$-approximation algorithm for k-Star Transversal, proving Theorem 2. Theorem 4 is proved in the full version of this work [31].

2 Preliminary

2.1 Notation

A k-uniform hypergraph is denoted by $P = (V_P, E_P)$ such that each $e \in E_P$ is a k-subset of V_P. We denote e as an *ordered k-tuple* $e = (v^1, \ldots, v^k)$. The ordering can be chosen arbitrarily given P, but should be fixed throughout. If v indicates a vertex of some graph, we use a superscript v^i to denote another vertex of the same graph, and e^i to denote the ith (hyper)edge. For an integer m, let $[m] = \{1, 2, \ldots, m\}$. Unless otherwise stated, the *measure* of $F \subseteq V$ is obtained under the uniform measure on V, which is simply $\frac{|F|}{|V|}$.

2.2 k-HVC

An instance of k-HVC consists of a k-uniform hypergraph P, where the goal is to find a set $C \subseteq V_P$ with the minimum cardinality such that it intersects every hyperedge. The result of Dinur, Guruswami, Khot and Regev [21] states that

▶ **Theorem 5** ([21]). *Given a k-uniform hypergraph ($k \geq 3$) and $\epsilon > 0$, it is NP-hard to tell apart the following cases:*
- *Completeness: There exists a vertex cover of measure $\frac{1+\epsilon}{k-1}$.*
- *Soundness: Every vertex cover has measure at least $1 - \epsilon$.*

Therefore, it is NP-hard to approximate k-HVC within a factor $k - 1 + 2\epsilon$.

Moreover, the above result holds even when the degree of a hypergraph is bounded by d depending on k and ϵ.

2.3 MIS-k

Given a graph $G = (V_G, E_G)$, a subset $S \subseteq V_G$ is *independent* if the subgraph induced by S does not contain any edge. The Maximum Independent Set (MIS) problem asks to find the largest independent set, and MIS-k indicates the same problem where G is promised to have maximum degree at most k. The recent result of Chan [11] implies

▶ **Theorem 6** ([11]). *Given a graph G with maximum degree at most k, it is NP-hard to tell apart the following cases:*
- *Completeness: There exists an independent set of measure $\Omega(1/(\log k))$.*
- *Soundness: Every subset of vertices of measure $O(\frac{\log^3 k}{k})$ contains an edge.*

Therefore, it is NP-hard to approximate MIS-k within a factor $\Omega(\frac{k}{\log^4 k})$.

3   H-Transversal

In this section, given a 2-connected graph $H = (V_H, E_H)$ with k vertices, we give a reduction from k-HVC to H-Transversal. The simplest try will be, given a hypergraph $P = (V_P, E_P)$ (let $n = |V_P|, m = |E_P|$), to produce a graph $G = (V_G, E_G)$ where $V_G = V_P$, and for each hyperedge $e = (v^1, \ldots, v^k)$ add $|E_H|$ edges that form a *canonical copy* of H to E_G. While the soundness follows directly (if $F \subseteq V_P$ contains a hyperedge, the subgraph induced by F contains H), the completeness property does not hold since edges that belong to different canonical copies may form an unintended non-canonical copy. To prevent this, a natural strategy is to replace each vertex by a set of many vertices (call it a *cloud*), and for each hyperedge $(v^1, \ldots, v^k)$, add many canonical copies on the k clouds (each copy consists of one vertex from each cloud). If we have too many canonical copies, soundness works easily but completeness is hard to show due to the risk posed by non-canonical copies, and in the other extreme, having too few canonical copies could result in the violation of the soundness property. Therefore, it is important to control the structure (number) of canonical copies that ensure both completeness and soundness at the same time.

Our technique, which we call *random matching*, proceeds by creating a carefully chosen number of *random* copies of H for each hyperedge to ensure both completeness and soundness. We remark that properties of random matchings are also used to bound the number of short non-canonical paths in inapproximability results for edge-disjoint paths on undirected graphs [2, 1]. The details in our case are different as we create *many* copies of H based on a *hypergraph*.

Fix $\epsilon > 0$, apply Theorem 5, let $c := \frac{1+\epsilon}{k-1}, s := 1 - \epsilon$ be the measure of the minimum vertex cover in the completeness and soundness case respectively, and $d := d(k, \epsilon)$ be the maximum degree of hard instances. Let a and B be integer constants greater than 1, which will be determined later. Lemma 7 and 9 with these parameters imply the first half of Theorem 1.

3.1   Reduction

Without loss of generality, assume that $V_H = [k]$. Given a hypergraph $P = (V_P, E_P)$, construct an undirected graph $G = (V_G, E_G)$ such that

- $V_G = V_P \times [B]$. Let $n = |V_P|$ and $N = |V_G| = nB$. For $v \in V_P$, let $\mathsf{cloud}(v) := \{v\} \times [B]$ be the copy of $[B]$ associated with v.
- For each hyperedge $e = (v^1, \ldots, v^k)$, for aB times, take $l^1, \ldots, l^k$ independently and uniformly from $[B]$. For each edge $(i, j) \in H$ ($1 \le i < j \le k$), add $((v^i, l^i), (v^j, l^j))$ to E_G. Each time we add $|E_H|$ edges isomorphic to H, and we have aB of such copies of H per each hyperedge. Call such copies *canonical*.

3.2   Completeness

The next lemma shows that if P has a small vertex cover, G also has a small H-Transversal.

▶ **Lemma 7.** *Suppose P has a vertex cover C of measure c. For any $\epsilon > 0$, with probability at least $3/4$, there exists a subset $F \subseteq V_G$ of measure at most $c + \epsilon$ such that the subgraph induced by $V_G \setminus F$ has no copy of H.*

Proof. Let $F = C \times [B]$. We consider the expected number of copies of H that avoid F and argue that a small fraction of additional vertices intersect all of these copies. Choose k vertices $(v^1, l^1), \ldots, (v^k, l^k)$ which satisfy
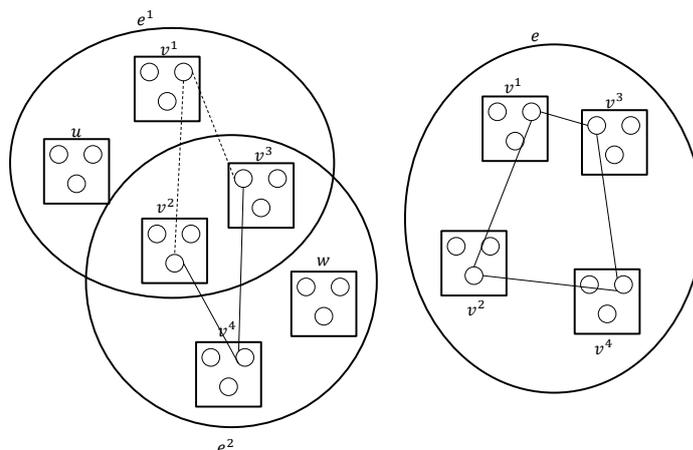
Figure 1 Two examples where $k = 4$ and H is a 4-cycle. On the left, purported edges are divided into two groups (dashed and solid edges). Each copy of canonical cycle should match the labels of three vertices to ensure it covers 2 designated edges (6 labels total). On the right, one canonical copy can cover all the edges, and it only needs to match the labels of four vertices (4 labels total).

- $v^1 \in V_P$ can be any vertex.
- $l^1, \ldots, l^k \in B$ can be arbitrary labels.
- For each $(i, j) \in E_H$, there must be a hyperedge of P containing both i and j.

There are n possible choices for v^1, B choices for each l^i, and at most kd choices for each v^i $(i > 1)$. The number of possibilities to choose such $(v^1, l^1), \ldots, (v^k, l^k)$ is bounded by $n(dk)^k B^k$. Note that no other k-tuple of vertices induce a connected graph and contain a copy of H. Further discard the tuple when two vertices are the same.

We calculate the probability that the subgraph induced by $((v^1, l^1), \ldots, (v^k, l^k))$ contains a copy *in this order* – formally, for all $(i, j) \in E_H$, $((v^i, l^i), (v^j, l^j)) \in E_G$. For each $(i, j) \in E_H$, we call a pair $((v^i, l^i), (v^j, l^j)) \in \binom{V_G}{2}$ a *purported edge*. For a set of purported edges, we say that this set can be *covered by a single canonical copy* if one copy of canonical copy of H can contain all purported edges with nonzero probability. Suppose that all $|E_H|$ purported edges can be covered by a single canonical copy of H. It is only possible when there is a hyperedge whose k vertices are exactly $\{v^1, \ldots, v^k\}$. In this case, $((v^1, l^1), \ldots, (v^k, l^k))$ intersects F. (right case of Figure 1). When $|E_H|$ purported edges have to be covered by more than one canonical copy, some vertices must be covered by more than one canonical copy, and each canonical copy covering the same vertex should give the same label to that vertex. This redundancy makes it unlikely to have all k edges exist at the same time. (left case of Figure 1). The below claim formalizes this intuition.

▶ **Claim 1.** *Suppose that $((v^1, l^1), \ldots, (v^k, l^k))$ cannot be covered by a single canonical copy. Then the probability that it forms a copy of H is at most $\frac{(adk)^{k^2}}{B^k}$.*

Proof. Fix $2 \le p \le |E_H|$. Partition $|E_H|$ purported edges into p nonempty *groups* $I_1, \ldots, I_p$ such that each group can be covered by a single canonical copy of H. There are at most $p^{|E_H|}$ possibilities to partition. For each $v \in V_P$, there are at most d hyperedges containing v and at most aBd canonical copies intersecting $\mathsf{cloud}(v)$. Therefore, all edges in one group can be covered simultaneously by at most aBd copies of canonical copies. There are at most $(aBd)^p$ possibilities to assign a canonical copy to each group. Assume that one canonical

copy is responsible for exactly one group. This is without loss of generality since if one canonical copy is responsible for many groups, we can merge them and this case can be dealt with smaller p.

Focus on one group I of purported edges, and one canonical copy $L = (V_L, E_L)$ which is supposed to cover them. Let $I' \subseteq V_G$ be the set of vertices which are incident on the edges in I. Suppose $V_L = \{(u^1, l'^1), \ldots, (u^k, l'^k)\}$, which is created by a hyperedge $f = (u^1, \ldots, u^k) \in E_P$. We calculate the probability that L contains all edges in I over the choice of labels $l'^1, \ldots, l'^k$ for L. One necessary condition is that $\{v | (v, l) \in I' \text{ for some } l \in [B]\}$ (i.e. the set I' projected to V_P) is contained in f. Otherwise, some vertices of I' cannot be covered by L. Another necessary condition is $v^i \neq v^j$ for any $(v^i, l^i) \neq (v^j, l^j) \in I'$. Otherwise (i.e. $(v, l^i), (v, l^j) \in I'$ for $l^i \neq l^j$), since L gives only one label to each vertex in $f \subseteq V_P$, (v, l^i) and (v, l^j) cannot be contained in L simultaneously. Therefore, we have a nice characterization of I': It consists of at most one vertex from the cloud of each vertex in f.

The probability that L contains I is at most the probability that for each $(v^i, l^i) \in I'$, l^i is equal to the label L assigns to v^i, which is $B^{-|I'|}$. Now we need the following lemma saying that the sum of $|I'|$ is large, which relies on 2-connectivity of H.

▶ **Lemma 8.** *Fix $p \geq 2$. For any partition $I_1, \ldots, I_p$ of purported edges into p non-empty groups, $\sum_{i=1}^{p} |I_i'| \geq k + p$.*

Proof. Let t be the number of vertices contained in at least two I_i's. Call them *boundary vertices*. Note that exactly $k - t$ vertices belongs to exactly one I_i'. For $i = 1, \ldots, p$, let b_i be the number of boundary vertices in $|I_i'|$. Since (I_i', I_i) is a proper subgraph of H and H is 2-vertex connected, $b_i \geq 2$ for each i. Therefore,

$$\sum_{i=1}^{p} |I_i'| = (k - t) + \max(2p, 2t) \geq k + p. \qquad \blacktriangleleft$$

We conclude that for each partition, the probability of having all the edges is at most

$$(aBd)^p \prod_{q=1}^{p} B^{-|I_q'|} = \frac{(aBd)^p}{B^{k+p}} = \frac{(ad)^p}{B^k} .$$

The probability that $((v^1, l^1), \ldots, (v^k, l^k))$ forms a copy is therefore bounded by

$$\sum_{p=2}^{|E_H|} p^{|E_H|} \frac{(ad)^p}{B^k} \leq \frac{(adk)^{k^2}}{B^k} . \qquad \blacktriangleleft$$

Therefore, the expected number of copies that avoid F is bounded by $n(kd)^k B^k \cdot \frac{(adk)^{k^2}}{B^k}$. With probability at least $3/4$, the number of such copies is at most $4n(adk)^{2k^2}$. Let $B \geq \frac{4(adk)^{2k^2}}{\epsilon}$. Then these copies of H can be covered by at most $\epsilon n B = \epsilon N$ vertices. $\qquad \blacktriangleleft$

3.3 Soundness

The soundness claim above is easier to establish. By an averaging argument, a subset I of V_G of measure 2ϵ must contain ϵB vertices from the clouds corresponding to a subset S of measure ϵ in V_P. There must be a hyperedge e contained within S, and the chosen parameters ensure that one of the canonical copies corresponding to e is likely to lie within I.

▶ **Lemma 9.** *For $a = a(k, \epsilon)$ and $B = \Omega(\log |E_P|)$, if every subset of V_P of measure at least ϵ contains a hyperedge in the induced subgraph, with probability at least $3/4$, every subset of V_G with measure 2ϵ contains a canonical copy of H.*

Proof. We want to show that the following property holds for every hyperedge $e = (v^1, \ldots, v^k)$: if a subset of vertices $I \subseteq V_G$ has at least ϵ fraction of vertices from each $\mathsf{cloud}(v^i)$, then I will contain a canonical copy. Fix $A^1 \subseteq \mathsf{cloud}(v^1), \ldots, A^k \subseteq \mathsf{cloud}(v^k)$ be such that for each i, $|A^i| \geq \epsilon B$. There are at most 2^{kB} ways to choose such A's. The probability that one canonical copy associated with e is not contained in $(v^1, A^1) \times \cdots \times (v^k, A^k)$ is at most $1 - \epsilon^k$. The probability that none of canonical copy associated with e is contained in $(v^1, A^1) \times \cdots \times (v^k, A^k)$ is $(1 - \epsilon^k)^{aB} \leq \exp(-aB\epsilon^k)$.

By union bound over all $A^1, \ldots, A^k$, the probability that there exists $A^1, \ldots, A^k$ containing no canonical copy is at most $\exp(kB - aB\epsilon^k) = \exp(-B) \leq \frac{1}{4|E_P|}$ by taking a large enough constant depending on k and ϵ, and $B = \Omega(\log |E_P|)$. Therefore, with probability at least $3/4$, the desired property holds for all hyperedges.

Let I be a subset of V_G of measure at least 2ϵ. By an averaging argument, at least ϵ fraction of *good* vertices $v \in V_P$ satisfy that $|\mathsf{cloud}(v^i) \cap I| \geq \epsilon B$. By the soundness property of P, there is a hyperedge e contained in the subgraph induced by the good vertices, and the above property for e ensures that I contains a canonical copy. ◀

4   H-Packing and MIS-k-g

Given a 2-connected graph H, the reduction from MIS-k-k to H-Packing is relatively straightforward. Here we assume that hard instances of MIS-k-k are indeed k-regular for simplicity. Given an instance $M = (V_M, E_M)$ of MIS-k-k, we take $G = (V_G, E_G)$ to be its *line graph* – $V_G = E_M$, and $e, f \in V_G$ are adjacent if and only if they share an endpoint as edges of M.

For each vertex $v \in V_M$, let $\mathsf{star}(v) := \{e \in V_G : v \in e\}$. $\mathsf{star}(v)$ induces a k-clique, and for $v, u \in V_M$, $\mathsf{star}(v)$ and $\mathsf{star}(u)$ share one vertex if u and v are adjacent, and share no vertex otherwise. Given an independent set S of M, we can find $|S|$ pairwise disjoint stars in G, which gives $|S|$ vertex-disjoint copies of H. On the other hand, 2-connectivity of H and large girth of M implies that any copy of H must be entirely contained in one star, proving that many disjoint copies of H in G also give a large independent set of M with the same cardinality, completing the reduction from MIS-k-k to H-Packing. The following theorem formalizes the above intuition.

▶ **Lemma 10.** *For a 2-connected graph H with k vertices, there is an approximation-preserving reduction from MIS-k-k to H-Packing.*

Proof. Let $M = (V_M, E_M)$ be an instance of MIS-k-k M with maximum degree k and girth greater than k. First, let $G = (V_G = E_M, E_G)$ be the line graph of M. For each vertex $v \in V_M$ with degree strictly less than k, we add $k - \mathsf{deg}(v)$ new vertices to V_G. Let $\mathsf{star}(v) \subseteq V_G$ be the union of the edges of M incident on v and the newly added vertices for v. Note that $|\mathsf{star}(v)| = k$ for all $v \in V_M$. Add edges to G to ensure that every $\mathsf{star}(v)$ induces a k-clique. For two vertices u and v of M, $\mathsf{star}(u)$ and $\mathsf{star}(v)$ share exactly one vertex if u and v are adjacent in M, and share no vertex otherwise.

Let S be an independent set of M. The $|S|$ stars $\{\mathsf{star}(v)\}_{v \in S}$ are pairwise disjoint and each induces a k-clique, so G contains at least $|S|$ disjoint copies of H.

We claim that any k-subset of V_G that induces a 2-connected subgraph must be $\mathsf{star}(v)$ for some v. Assume towards contradiction, let T be a k-subset inducing a 2-connected subgraph

of G that cannot be contained in a single star. We first show T must contain two disjoint edges of M. Take any $(u, v) \in T$. Since $T \notin \mathsf{star}(u)$, T contains an edge of M not incident on u. If it is not incident on v either, we are done. Otherwise, let (w, v) be this edge. The same argument from $T \notin \mathsf{star}(v)$ gives another edge (w', u) in T. If $w \neq w'$, (w, v) and (w', u) are disjoint. Otherwise, w, u, v form a triangle in M, contradicting a large girth. Let (u, v), (w, x) be two disjoint edges of M in contained in T.

Since the subgraph of G vertex-induced by T is 2-connected, there are two internally vertex-disjoint paths P_1, P_2 in G from (u, v) to (w, x). The sum of the two lengths is at most k, where the length of a path is defined to be the number of edges. By considering the internal vertices of P_i (edges of M) and deleting unnecessary portions, we have two edge-disjoint paths P_1', P_2' in M where each P_i' connects $\{u, v\}$ and $\{w, x\}$, with length at most the length of P_i minus one. There is a cycle in M consists only of the edges of P_1', P_2' together with $(u, v), (w, x)$. Since $|P_1'| + |P_2'| + 2 \leq k$, it contradicts that M has girth strictly greater than k. ◄

We prove that MIS-k-g is also hard to approximate by a reduction from MIS-d ($d = \widetilde{\Omega}(k)$), using a slightly different *random matching* idea. Given a degree-d graph with possibly small girth, we replace each vertex by a cloud of B vertices, and replace each edge by a copies of random matching between the two clouds. While maintaining the soundness guarantee, we show that there are only a few small cycles, and by deleting a vertex from each of them and *sparsifying* the graph we obtain a hard instance for MIS-k-g. Note that g does not affect the inapproximability factor but only the runtime of the reduction.

▶ **Theorem 11** (Restatement of Theorem 3). *For any constants k and g, unless $\mathsf{NP} \subseteq \mathsf{BPP}$, no polynomial time algorithm approximates MIS-k-g within a factor of $\Omega(\frac{k}{\log^7 k})$.*

Proof. We reduce from MIS-d to MIS-k-g where $k = O(d \log^2 d)$. Given an instance $G_0 = (V_{G_0}, E_{G_0})$ of MIS-d, we construct $G = (V_G, E_G)$ and $G' = (V_{G'}, E_{G'})$ by the following procedure:

- $V_G = V_{G_0} \times [B]$. As usual, let $\mathsf{cloud}(v) = \{v\} \times [B]$.
- For each edge $(u, v) \in E_{G_0}$, for a times, add a *random matching* as follows.
    - Take a random permutation $\pi : [B] \to [B]$.
    - Add an edge $((u, i), (v, \pi(i))$ for all $i \in [B]$.
- Call the resulting graph G. To get the final graph G',
    - For any cycle of length at most g, delete an arbitrary vertex from the cycle. Repeat until there is no cycle of length at most g.

Note that the step of eliminating the small cycles can be implemented trivially in time $O(n^g)$. Let $n = |V_{G_0}|, m = |E_{G_0}|, N = nB = |V_G| \geq |V_{G'}|, M = m \cdot aB = |E_G| \geq |E_{G'}|$. The maximum degree of G and G' is at most ad. By construction, girth of G' is at least $g + 1$.

Girth Control. We calculate the expected number of small cycles in G, and argue that the number of these cycles is much smaller than the total number of vertices, so that $|V_G|$ and $|V_{G'}|$ are almost the same. Let k' be the length of a purported cycle. Choose k' vertices $(v^1, l^1), \ldots, (v^{k'}, l^{k'})$ which satisfy

- $v^1 \in V_{G_0}$ can be any vertex.
- For each $1 \leq i < k'$, $(v^i, v^{i+1}) \in E_{G_0}$.
- $l^1, \ldots, l^{k'} \in B$ can be arbitrary labels.

There are n possible choices for v^1, B choices for each l^i, and d choices for each v^i $(i > 1)$. The number of possibilities to choose such $(v^1, l^1), \ldots, (v^{k'}, l^{k'})$ is bounded by $nd^{k'-1}B^{k'}$. Without loss of generality, assume that no vertices appear more than once.

For each edge $e = (u, w) \in G_0$, consider the intersection of the purported cycle $((v^1, l^1), ..., (v^{k'}, l^{k'}))$ and the subgraph induced by $\mathsf{cloud}(u) \cup \mathsf{cloud}(w)$. It is a bipartite graph with the maximum degree 2. Suppose there are q purported edges $e^1, \ldots, e^q$ (ordered arbitrarily) in this bipartite graph. By slightly abusing notation, let e^i also denote the event that e^i exists in G. The following claim upper bounds $\Pr[e^i | e^1, \ldots, e^{i-1}]$ for each e^i.

▶ **Claim 2.** $\Pr[e^i | e^1, \ldots, e^{i-1}] \leq \frac{a}{B-i}$.

Proof. There are a random matchings between $\mathsf{cloud}(u)$ and $\mathsf{cloud}(w)$, and for each $j < i$, there is at least one random matching including e^j. We fix one random matching and calculate the probability that the random matching contains e^i, conditioned on the fact that it already contains some of $e^1, \ldots, e^{i-1}$.

If there is e^j $(j < i)$ that shares a vertex with e^i, e^i cannot be covered by the same random matching with e^j. If a random matching covers p of $e^1, \ldots, e^{i-1}$ which are disjoint from e^i, the probability that e^i is covered by that random matching is $\frac{1}{B-p}$, and this is maximized when $p = i - 1$.

By a union bound over the a random matchings, $\Pr[e^i | e^1, \ldots, e^{i-1}] \leq \frac{a}{B-i}$. ◀

The probability that all of $e^1, \ldots, e^q$ exist is at most

$$\prod_{i=1}^{q} \frac{a}{B-i} \leq \left(\frac{a}{B-q}\right)^q \leq \left(\frac{a}{B-k'}\right)^q .$$

Since edges of G_0 are processed independently, the probability of success for one fixed purported cycle is $(\frac{a}{B-k'})^{k'}$. The expected number of cycles of length k' is

$$nd^{k'-1}B^{k'} \cdot \left(\frac{a}{B-k'}\right)^{k'} = nd^{k'-1}a^{k'}\left(1 + \frac{k'}{B-k'}\right)^{k'}$$

$$\leq nd^{k'-1}a^{k'}\exp\left(\frac{k'^2}{B-k'}\right) \leq en(ad)^{k'}$$

by taking $B - k' \geq k'^2$. Summing over $k' = 1, \ldots, g$, the expected number of cycles of length up to g, is bounded by $eg(ad)^g n$. Take $B \geq 4d^2 \cdot eg(ad)^g$. Then with probability at least $3/4$, the number of cycles of length at most g is at most $\frac{Bn}{d^2}$. By taking $1/d^2$ fraction of vertices away (one for each short cycle), we have a girth at least $g + 1$, which implies $\left(1 - \frac{1}{d^2}\right)|V_G| \leq |V_{G'}| \leq |V_G|$.

Hardness of MIS-d states that it is NP-hard to distinguish the case G_0 has an independent set of measure $c := \Omega(\frac{1}{\log d})$ and the case where the maximum independent set has measure at most $s := O(\frac{\log^3 d}{d})$.

Completeness. Let I_0 be an independent set of G_0 of measure c. Then $I = I_0 \times [B]$ is also an independent set of G of measure c. Let $I' = I \cap V_{G'}$. I' is independent in both G and G', and the measure of I' in G' is at least the measure of I' in G, which is at least $c - 1/d^2 = \Omega(\frac{1}{\log d})$.

Soundness.   Suppose that every subset of V_{G_0} of measure at least s contains an edge. Say a graph is (β, α)-*dense* if we take β fraction of vertices, at least α fraction of edges lie within the induced subgraph. We also say a bipartite graph is (β, α)-*bipartite dense* if we take β fraction of vertices from each side, at least α fraction of edges lie within the induced subgraph.

▶ **Claim 3.** *For $a = O(\frac{\log(1/s)}{s})$ and $B = O(\frac{\log m}{s})$ the following holds with probability at least 3/4: For every $(u, w) \in E_{G_0}$, the bipartite graph between $\mathsf{cloud}(u)$ and $\mathsf{cloud}(w)$ is $(\epsilon, \epsilon^2/8)$-bipartite dense for all $\epsilon \geq s$.*

Proof. Fix (u, w), and $\epsilon \in [s, 1]$, and $X \subseteq \mathsf{cloud}(u)$ and $Y \subseteq \mathsf{cloud}(w)$ be such that $|X| = |Y| = \epsilon B$. The possibilities of choosing X and Y is

$$\binom{B}{\epsilon B}^2 \leq \exp(O(\epsilon \log(1/\epsilon)B))$$

Without loss of generality, let $X = Y = [\epsilon B]$. In one random matching, let X_i ($i \in [\epsilon B]$) be the random variable indicating whether vertex $(u, i) \in X$ is matched with a vertex in Y or not. $\Pr[X_1 = 1] = \epsilon$, and $\Pr[X_i = 1 | X_1, \ldots, X_{i-1}] \geq \epsilon/2$ for $i \in [\epsilon B/2]$ and any $X_1, \ldots, X_{i-1}$. Therefore, the expected number of edges between X and Y is at least $\epsilon^2 B/4$. With a random matchings, the expected number is at least $a\epsilon^2 B/4$. By Chernoff bound, the probability that it is less than $a\epsilon^2 B/8$ is at most $\exp(\frac{a\epsilon^2 B}{32})$. By union bound over all possibilities of choosing X and Y, the probability that the bipartite graph is not $(\epsilon, \epsilon^2/8)$-bipartite dense is

$$\exp(\epsilon \log(1/\epsilon)B) \cdot \exp\left(-\frac{a\epsilon^2 B}{32}\right) \leq \frac{1}{4mB}$$

by taking $a = O(\frac{\log(1/s)}{s})$ and $B = O\left(\frac{\log m}{s}\right)$. A union bound over all possible choices of ϵ (B possibilities) and m edges of E_0 implies the claim.   ◀

▶ **Claim 4.** *With the parameters a and B above, G is $(4s \log(1/s), \Omega(\frac{s}{d}))$-dense.*

Proof. Fix a subset S of measure $4s \log(1/s)$. For a vertex v of G_0, let $\mu(v) := \frac{|\mathsf{cloud}(v) \cap S|}{B}$. Note that $\mathbb{E}_v[\mu(v)] = 4s \log(1/s)$. Partition V_{G_0} into $t + 1$ buckets $B_0, \ldots, B_t$ ($t := \lceil \log_2(1/s) \rceil$), such that B_0 contains v such that $\mu(v) \leq s$, and for $i \geq 1$, B_i contains v such that $\mu(v) \in (2^{i-1}s, 2^i s]$. Denote

$$\mu(B_i) := \frac{\sum_{v \in B_i} \mu(v)}{|V_{G_0}|} \ .$$

Clearly $\mu(B_0) \leq s$. Pick $i \in \{1, \ldots, t\}$ with the largest $\mu(B_i)$. We have $\mu(B_i) \geq 2s$ since $\mathbb{E}_v[\mu(v)] \geq 4s \log(1/s)$. Let $\gamma = 2^{i-1}s$. All vertices of B_i has $\mu(v) \in [\gamma, 2\gamma]$, so $|B_i| \geq (s/\gamma)n$.

Since G_0 has no independent set with more than ns vertices, Turán's Theorem says that the subgraph of G_0 induced by B_i has at least $\frac{|B_i|}{2}(\frac{|B_i|}{ns} - 1) = \Omega(\frac{s}{\gamma^2}n)$ edges. This is at least $\Omega(\frac{s}{d\gamma^2})$ fraction of the total number of edges.

For each of these edges, by Claim 3, at least $\gamma^2/8$ fraction of the edges from the bipartite graph connecting the clouds of its two endpoints, lie in the subgraph induced by S (since $\gamma \geq s$). Overall, we conclude that there are at least $\Omega(\frac{s}{d\gamma^2}) \cdot \frac{\gamma^2}{8} = \Omega(\frac{s}{d})$ fraction of edges inside the subgraph induced by S.   ◀

Sparsification.   Recall that G' is obtained from G by deleting at most $\frac{1}{d^2}$ fraction of vertices to have girth greater than g. In the completeness case, G' has an independent set of measure at least $c - 1/d^2 = \Omega(\frac{1}{\log d})$. In the soundness case, G is $(4s\log(1/s), \Omega(\frac{s}{d}))$-dense, so G' is (β, α)-dense where $\beta := \Omega(\frac{\log^4 d}{d}), \alpha := \Omega(\frac{\log^3 d}{d^2})$. Using density of G', we sparsify G' again – keep each edge of G' by probability $\frac{kn}{|E_{G'}|}$ so that the expected total number of edges is kn.

Fix a subset $S \subseteq V_{G'}$ of measure β. Since there are at least α fraction of edges in the subgraph induced by S, the expected number of picked edges in this subgraph is at least αkn. By Chernoff bound, the probability that it is less than $\frac{\alpha kn}{8}$ is at most $\exp(-\frac{\alpha kn}{32})$. By union bound over all sets of measure exactly β (there are at most $\binom{n}{n\beta} \leq \exp(2\beta \log(1/\beta)n)$ of them), and over all possible values of β (there are at most n possible sizes), the desired property fails with probability at most

$$n \cdot \max_{\beta \in [\beta_0, 1]} \left\{ \exp(-\alpha kn/32) \cdot \exp(2\beta \log(1/\beta)n) \right\} \leq n \cdot e^{-n}$$

when $k = O(\frac{\beta \log(1/\beta)}{\alpha}) = O(d\log^2 d)$. In the last step we remove all the vertices of degree more than $10k$. Since the expected degree of each vertex is at most $2k$, the expected fraction of deleted vertices is $\exp(-\Omega(k)) \ll \beta$.

Combining all these results, we have a graph with small degree $10k = O(d\log^2 d)$ and girth strictly greater than g, where it is NP-hard to approximate MIS within a factor of $\frac{c - \frac{1}{d^2}}{\beta} = \Omega(\frac{d}{\log^5 d}) = \Omega(\frac{k}{\log^7 k})$. Therefore, it is NP-hard to approximate MIS-k-g within a factor of $\Omega(\frac{k}{\log^7 k})$. ◀

------- **References** -------

1   Matthew Andrews, Julia Chuzhoy, Venkatesan Guruswami, Sanjeev Khanna, Kunal Talwar, and Lisa Zhang. Inapproximability of edge-disjoint paths and low congestion routing on undirected graphs. *Combinatorica*, 30(5):485–520, 2010.

2   Matthew Andrews and Lisa Zhang. Logarithmic hardness of the undirected edge-disjoint paths problem. *Journal of the ACM*, 53(5):745–761, 2006.

3   P. Austrin, S. Khot, and M. Safra. Inapproximability of vertex cover and independent set in bounded degree graphs. In *Proceedings of the 24th annual IEEE Conference on Computational Complexity*, CCC'09, pages 74–80, 2009.

4   V. Bafna, P. Berman, and T. Fujito. Constant ratio approximations of the weighted feedback vertex set problem for undirected graphs. In *Proceedings of the 6th International Symposium on Algorithms and Computation*, ISAAC'95, pages 142–151, 1995.

5   N. Bansal and S. Khot. Inapproximability of hypergraph vertex cover and applications to scheduling problems. In *Proceedings of the 37th International Colloquium on Automata, Languages and Programming*, ICALP'10, pages 250–261, 2010.

6   R. Bar-Yehuda, D. Geiger, J. Naor, and R. Roth. Approximation algorithms for the feedback vertex set problem with applications to constraint satisfaction and bayesian inference. *SIAM Journal on Computing*, 27(4):942–959, 1998.

7   A. Becker and D. Geiger. Optimization of Pearl's method of conditioning and greedy-like approximation algorithms for the vertex feedback set problem. *Artificial Intelligence*, 83(1):167–188, 1996.

8   H. L. Bodlaender. On disjoint cycles. *International Journal of Foundations of Computer Science*, 5(01):59–68, 1994.

9   A. Brandstädt, V. Chepoi, and F. Dragan. Clique r-domination and clique r-packing problems on dually chordal graphs. *SIAM J. on Discrete Mathematics*, 10(1):109–127, 1997.

10 Parinya Chalermsook, Bundit Laekhanukit, and Danupon Nanongkai. Pre-reduction graph products: Hardnesses of properly learning DFAs and approximating EDP on DAGs. In *Proc. of the 55rd Annual Symp. on Foundations of Computer Science (FOCS'14)*. IEEE, 2014.

11 S. Chan. Approximation resistance from pairwise independent subgroups. In *Proc. of the 45th Annual ACM Symp. on Symposium on Theory of Computing*, STOC'13, pages 447–456, 2013.

12 Maw-Shang Chang, Ton Kloks, and Chuan-Min Lee, editors. *Graph-Theoretic Concepts in Computer Science*, volume 2204 of *Lecture Notes in Computer Science*. Springer, 2001.

13 F. Chataigner, G. Manić, Y. Wakabayashi, and R. Yuster. Approximation algorithms and hardness results for the clique packing problem. *Discrete Applied Mathematics*, 157(7):1396–1406, 2009.

14 J. Chen, Y. Liu, S. Lu, B. O'sullivan, and I. Razgon. A fixed-parameter algorithm for the directed feedback vertex set problem. *Journal of the ACM*, 55(5):21:1–21:19, 2008.

15 R. Chitnis, M. Cygan, M. Hajiaghayi, and D. Marx. Directed subset feedback vertex set is fixed-parameter tractable. In *Proceedings of the 39th International Colloquium on Automata, Languages, and Programming*, ICALP'12, pages 230–241, 2012.

16 Miroslav Chlebík and Janka Chlebíková. Approximation hardness of dominating set problems in bounded degree graphs. *Information and Computation*, 206(11):1264–1275, 2008.

17 F. A. Chudak, M. X. Goemans, D. S. Hochbaum, and D. P. Williamson. A primal-dual interpretation of two 2-approximation algorithms for the feedback vertex set problem in undirected graphs. *Operations Research Letters*, 22(4–5):111–118, 1998.

18 M. Cygan, M. Pilipczuk, M. Pilipczuk, and J. O. Wojtaszczyk. Subset feedback vertex set is fixed-parameter tractable. In *Proceedings of the 38th International Colloquim on Automata, Languages and Programming*, ICALP'11, pages 449–461, 2011.

19 Marek Cygan. Improved approximation for 3-dimensional matching via bounded pathwidth local search. In *Proceedings of the 54th annual IEEE symposium on Foundations of Computer Science*, FOCS'13, pages 509–518, 2013.

20 I. Dinur and S. Safra. On the hardness of approximating minimum vertex cover. *Annals of Mathematics*, pages 439–485, 2005.

21 Irit Dinur, Venkatesan Guruswami, Subhash Khot, and Oded Regev. A new multilayered PCP and the hardness of hypergraph vertex cover. *SIAM Journal on Computing*, 34(5):1129–1146, 2005.

22 D. Dor and M. Tarsi. Graph decomposition is NP-complete: A complete proof of Holyer's conjecture. *SIAM Journal on Computing*, 26(4):1166–1187, 1997.

23 Guillermo Durán, Min Chih Lin, Sergio Mera, and Jayme L Szwarcfiter. Algorithms for finding clique-transversals of graphs. *Annals of Operations Research*, 157(1):37–45, 2008.

24 Paul Erdős, Tibor Gallai, and Zsolt Tuza. Covering the cliques of a graph with vertices. *Discrete Mathematics*, 108(1):279–289, 1992.

25 G. Even, J. Naor, S. Rao, and B. Schieber. Divide-and-conquer approximation algorithms via spreading metrics. *Journal of the ACM*, 47(4):585–616, 2000.

26 G. Even, J. Naor, B. Schieber, and M. Sudan. Approximating minimum feedback sets and multicuts in directed graphs. *Algorithmica*, 20:151–174, 1998.

27 Tomas Feder and Carlos Subi. Packing edge-disjoint triangles in given graphs. *Electronic Colloquium on Computational Complexity (ECCC)*, 2012. TR12-013.

28 Z. Friggstad and M. R. Salavatipour. Approximability of packing disjoint cycles. In *Proc. of the 18th Int'l Conf. on Algorithms and Computation*, ISAAC'07, pages 304–315, 2007.

29 V. Guruswami, R. Manokaran, and P. Raghavendra. Beating the random ordering is hard: Inapproximability of maximum acyclic subgraph. In *Proceedings of the 49th annual IEEE symposium on Foundations of Computer Science*, FOCS'08, pages 573–582, 2008.

30 Venkatesan Guruswami and Euiwoong Lee. Inapproximability of feedback vertex set for bounded length cycles. *Electronic Colloquium on Computational Complexity (ECCC)*, TR 14-006, 2014.

31 Venkatesan Guruswami and Euiwoong Lee. Inapproximability of H-transversal/matching. *arXiv preprint arXiv:1506.06302*, 2015.

32 Venkatesan Guruswami and C. Pandu Rangan. Algorithmic aspects of clique-transversal and clique-independent sets. *Discrete Applied Mathematics*, 100(3):183–202, 2000.

33 Venkatesan Guruswami, C. Pandu Rangan, M. S. Chang, G. J. Chang, and C. K. Wong. The K_r-packing problem. *Computing*, 66(1):79–89, 2001.

34 Elad Hazan, Shmuel Safra, and Oded Schwartz. On the complexity of approximating *k*-set packing. *Computational Complexity*, 15(1):20–39, 2006.

35 P. Hell, S. Klein, L.T. Nogueira, and F. Protti. Packing r-cliques in weighted chordal graphs. *Annals of Operations Research*, 138(1):179–187, 2005.

36 Bart M. P. Jansen and Dániel Marx. Characterizing the easy-to-find subgraphs from the viewpoint of polynomial-time algorithms, kernels, and turing kernels. In *Proceedings of the 26th Annual ACM-SIAM Symposium on Discrete Algorithms*, pages 616–629, 2015.

37 Subhash Khot. On the power of unique 2-prover 1-round games. In *Proceedings of the 34th annual ACM Symposium on Theory of Computing*, STOC'02, pages 767–775, 2002.

38 Subhash Khot and Oded Regev. Vertex cover might be hard to approximate to within $2-\epsilon$. *Journal of Computer and System Sciences*, 74(3):335–349, 2008.

39 D. Kirkpatrick and P. Hell. On the complexity of general graph factor problems. *SIAM Journal on Computing*, 12(3):601–609, 1983.

40 Ton Kloks. Packing interval graphs with vertex-disjoint triangles. *arXiv preprint arXiv:1202.1041*, 2012.

41 Guy Kortsarz, Michael Langberg, and Zeev Nutov. Approximating maximum subgraphs without short cycles. *SIAM Journal on Discrete Mathematics*, 24(1):255–269, 2010.

42 M. Krivelevich, Z. Nutov, M. R. Salavatipour, J. V. Yuster, and R. Yuster. Approximation algorithms and hardness results for cycle packing problems. *ACM Transactions on Algorithms*, 3(4), November 2007.

43 Chuan-Min Lee. Weighted maximum-clique transversal sets of graphs. *ISRN Discrete Mathematics*, 2011, 2012.

44 CM Lee, MS Chang, and SC Sheu. The clique transversal and clique independence of distance hereditary graphs. In *Proceedings of the 19th Workshop on Combinatorial Mathematics and Computation Theory, Taiwan*, pages 64–69, 2002.

45 Carsten Lund and Mihalis Yannakakis. The approximation of maximum subgraph problems. In *Automata, Languages and Programming*, volume 700 of *Lecture Notes in Computer Science*, pages 40–51. Springer, 1993.

46 Sridhar Rajagopalan and Vijay V. Vazirani. Primal-dual rnc approximation algorithms for set cover and covering integer programs. *SIAM J. on Computing*, 28(2):525–540, 1998.

47 Dieter Rautenbach and Friedrich Regen. On packing shortest cycles in graphs. *Information Processing Letters*, 109(14):816–821, 2009.

48 P. D. Seymour. Packing directed circuits fractionally. *Combinatorica*, 15(2):281–288, 1995.

49 Erfang Shan, Zuosong Liang, and Liying Kang. Clique-transversal sets and clique-coloring in planar graphs. *European Journal of Combinatorics*, 36:367–376, 2014.

50 Ola Svensson. Hardness of vertex deletion and project scheduling. In *Approximation, Randomization, and Combinatorial Optimization. Algorithms and Techniques*, volume 7408 of *Lecture Notes in Computer Science*, pages 301–312. Springer, 2012.

51 Zsolt Tuza. Covering all cliques of a graph. In S.T. Hedetniemi, editor, *Topics on Domination*, volume 48 of *Annals of Discrete Mathematics*, pages 117–126. Elsevier, 1991.

52   R. Yuster. Edge-disjoint cliques in graphs with high minimum degree. *SIAM Journal on Discrete Mathematics*, 28(2):893–910, 2014.

53   Raphael Yuster. Combinatorial and computational aspects of graph packing and graph decomposition. *Computer Science Review*, 1(1):12–26, 2007.

A **Approximation Algorithm for k-Star Transversal**

In this section, we show that k-Star Transversal admits an $O(\log k)$-approximation algorithm, matching the $\Omega(\log k)$-hardness obtained via a simple reduction from Minimum Dominating Set on degree-$(k-1)$ graphs [16], and proving Theorem 2. Let $G = (V_G, E_G)$ be the instance of k-Star Transversal. This problem has a natural interpretation that it is equivalent to finding the smallest $F \subseteq V_G$ such that the subgraph induced by $V_G \setminus F$ has maximum degree at most $k - 2$. Our algorithm consists of two phases.

1. Iteratively solve 2-rounds of Sherali-Adams linear programming (LP) hierarchy and put vertices with a large fractional value in the transversal. If this phase terminates with a partial transversal F, the remaining subgraph induced by $V_G \setminus F$ has small degree (at most $2k$) and the LP solution to the last iteration is *highly fractional.*

2. We reduce the remaining problem to *Constrained Set Multicover* and use the standard greedy algorithm. While the analysis of the greedy algorithm for Constrained Set Multicover is used as a black-box, low degree of the remaining graph and high fractionality of the LP solution imply that the analysis is almost tight for our problem as well.

A.1 Iterative Sherali-Adams

Given G, 2-rounds of Sherali-Adams hierarchy of LP relaxation has variables $\{x_v\}_{v \in V_G} \cup \{x_{u,v}\}_{u,v \in V_G}$. An integral solution $y : V_G \mapsto \{0, 1\}$, where $y(v) = 1$ indicates that v is picked in the transvesal, naturally gives a feasible solution to the hierarchy by $x_v = y_v$, $x_{u,v} = y_u y_v$. Consider the following relaxation for k-Star Transversal.

$$
\begin{aligned}
\text{minimize} \quad & \sum_{v \in V_G} x_v \\
\text{subject to} \quad & 0 \leq x_{u,v}, x_v \leq 1 && \forall u, v \in V_G \\
& x_{u,v} \leq x_u && \forall u, v \in V_G \\
& x_u + x_v - x_{u,v} \leq 1 && \forall u, v \in V_G \\
& \sum_{v:(u,v) \in E_G} (x_v - x_{u,v}) \geq (\mathsf{deg}(u) - k + 2)(1 - x_u) && \forall u \in V_G
\end{aligned}
$$

The first three constraints are common to any 2-rounds of Sherali-Adams hierarchy, and ensure that for any $u, v \in V_G$, the *local distribution* on four assignments $\alpha : \{u, v\} \mapsto \{0, 1\}$ forms a valid distribution. In other words, the following four numbers are nonnegative and sum to 1: $\Pr[\alpha(u) = \alpha(v) = 1] := x_{u,v}$, $\Pr[\alpha(u) = 0, \alpha(v) = 1] := x_v - x_{u,v}$, $\Pr[\alpha(u) = 1, \alpha(v) = 0] := x_u - x_{u,v}$, $\Pr[\alpha(u) = \alpha(v) = 0] := 1 - x_u - x_v + x_{u,v}$.

The last constraint is specific to k-Star Transversal, and it is easy to see that it is a valid relaxation: Given a feasible integral solution $y : V_G \mapsto \{0, 1\}$, the last constraint is vacuously satisfied when $y_u = x_u = 1$, and if not, it requires that at least $\mathsf{deg}(u) - k + 2$ vertices should be picked in the transversal so that there is no copy of k-Star in the induced subgraph centered on u. The first phase proceeds as the following.

- Let $S \leftarrow \emptyset$.
- Repeat the following until the size of S does not increase in one iteration.
  - Solve the above Sherali-Adams hierarchy for $V_G \setminus S$ – it means to solve the above LP with additional constraints $x_v = 1$ for all $v \in S$, which also implies $x_{u,v} = x_u$ for $v \in S, u \in V_G$. Denote this LP by $\mathsf{SA}(S)$.
  - $S \leftarrow \{v : x_v \geq \frac{1}{\alpha}\}$, where $\alpha := 10$.

We need to establish three properties from the first phase:
- The size of S is close to that of the optimal k-Star Transversal.
- Maximum degree of the subgraph induced by $V_G \setminus S$ is small.
- The remaining solution has small fractional values – $x_v < \frac{1}{\alpha}$ for all $v \in V_G \setminus S$.

The final property is satisfied by the procedure. The following two lemmas establish the other two properties.

▶ **Lemma 12.** *Let* Frac *be the optimal value of* $\mathsf{SA}(\emptyset)$. *When the above procedure terminates,* $|S| \leq \alpha \, \mathsf{Frac}$.

Proof. Assume that the above loop iterated l times, and for $i = 0, ..., l$, let S_i be S after the ith loop such that $S_0 = \emptyset, ..., S_l = S$. We use induction from the last iteration. Let Frac_i be the optimal fractional solution to $\mathsf{SA}(S_i)$ minus $|S_i|$ such that $\mathsf{Frac} = \mathsf{Frac}_0$.

We first establish $|S_l| - |S_{l-1}| \leq \alpha \, \mathsf{Frac}_{l-1}$. This is easy to see because, when x is the optimal fraction solution to $\mathsf{SA}(S_{l-1})$,

$$|S_l| - |S_{l-1}| = |\{v \notin S_{l-1} : x_v \geq \frac{1}{\alpha}\}| \leq \alpha \, \mathsf{Frac}_{l-1} \,.$$

For $i = l-2, l-1, ..., 0$, we show that $|S_l| - |S_i| \leq \alpha \, \mathsf{Frac}_i$. Let x be the optimal fraction solution to $\mathsf{SA}(S_i)$, and x' be the solution obtained by *partially rounding* x in the following way.
- $x'_v = 1$ if $v \in S_i$. Otherwise, $x'_v = x_v$.
- $x'_{u,v} = x'_u$ ($v \in S_i$), x'_v ($u \in S_i$), or $x_{u,v}$ otherwise.

It is easy to check that it is a feasible solution to $\mathsf{SA}(S_{i+1})$ (intuitively, *rounding up* only helps feasibility), so its value is

$$|S_i| + \sum_{v \notin S_i, x_v < \frac{1}{\alpha}} x_v \geq |S_i| + \mathsf{Frac}_{i+1},$$

which implies

$$\mathsf{Frac}_i = \sum_{v \notin S_i, x_v \geq \frac{1}{\alpha}} x_v + \sum_{v \notin S_i, x_v < \frac{1}{\alpha}} x_v \geq \frac{1}{\alpha}(|S_{i+1}| - |S_i|) + \mathsf{Frac}_{i+1} \,.$$

Finally, we have

$$|S_l| - |S_i|$$
$$= (|S_l| - |S_{i+1}|) + (|S_{i+1}| - |S_i|)$$
$$\leq \alpha \, \mathsf{Frac}_{i+1} + (|S_{i+1}| - |S_i|)$$
$$\leq \alpha \, \mathsf{Frac}_i,$$

where the first inequality follows from the induction hypothesis. This completes the induction.
◀

▶ **Lemma 13.** *After the termination, every vertex has degree at most $2k$ in the subgraph induced by $V_G \setminus S$.*

Proof. We prove that at least one vertex is added to S if the subgraph induced by $V_G \setminus S$ has a vertex of degree more than $2k$. Fix one such iteration, and let S_1 and S_2 be S before and after the iteration respectively. Let G' be the subgraph of G induced by $V_G \setminus S_1$. If the subgraph induced by $V_G \setminus S_2$ does not have any vertex with degree more than $2k$, we are done. Otherwise, fix one such vertex $u \in V_G \setminus S_2$. Note that the degree of u in G' is also more than $2k$.

We show that at least one neighbor v of u satisfies $v \notin S_1$ but $v \in S_2$. Let x be the optimal fractional solution to $\mathsf{SA}(S_1)$ and consider the following constraint for u.

$$\sum_{v:(u,v) \in E_G} (x_v - x_{u,v}) \geq (\deg(u) - k + 2)(1 - x_u).$$

Let $\mathsf{Nbr}(u)$ and $\mathsf{Nbr}'(u)$ be the set of neighbors of u in G and G' respectively, and $\deg'(u) = |\mathsf{Nbr}'(u)|$. Note that $\mathsf{Nbr}'(u) = \mathsf{Nbr}(u) \setminus S_1$, and for $v \in \mathsf{Nbr}(u) \cap S_1$, $x_v = 1$ and $x_{u,v} = x_u$. Therefore, the above constraint is equivalent to

$$\sum_{v:\mathsf{Nbr}(u) \cap S_1} (1 - x_u) + \sum_{v:\mathsf{Nbr}'(u)} (x_v - x_{u,v}) \geq (\deg(u) - k + 2)(1 - x_u)$$

$$\Leftrightarrow \sum_{v:\mathsf{Nbr}'(u)} (x_v - x_{u,v}) \geq (\deg'(u) - k + 2)(1 - x_u).$$

The fact that $u \notin S_2$ implies that $x_u < \frac{1}{\alpha}$, which implies

$$\sum_{v \in \mathsf{Nbr}'(u)} x_v$$

$$\geq \sum_{v \in \mathsf{Nbr}'(u)} (x_v - x_{u,v}) \geq (1 - \frac{1}{\alpha})(\deg'(u) - k) = \quad (1 - \frac{1}{\alpha}) \deg'(u)(1 - \frac{k}{\deg'(u)}).$$

Therefore, there is one $v \in \mathsf{Nbr}'(u)$ with $x_v \geq (1 - \frac{1}{\alpha})(1 - \frac{k}{\deg'(u)}) \geq \frac{9}{10} \cdot \frac{1}{2} > \frac{1}{\alpha}$. v satisfies $v \notin S_1$ but $v \in S_2$. ◀

A.2 Constrained Set Multicover

The first phase returns a set S whose size is at most α times the optimal solution and the subgraph induced by $V_G \setminus S$ has maximum degree at most $2k$. As above, let G' be the subgraph induced by $V_G \setminus S$, $\mathsf{Nbr}(u), \mathsf{Nbr}'(u)$ be the neighbors of u in G and G' respectively, and $\deg(u) = |\mathsf{Nbr}(u)|, \deg'(u) = |\mathsf{Nbr}'(u)|$. The remaining task is to find a small subset $F \subseteq V_G \setminus S$ such that the subgraph of G' (and G) induced by $V_G \setminus (S \cup F)$ has no vertex of degree at least $k - 1$. We reduce the remaining problem to the *Constrained Set Multicover* problem defined below.

▶ **Definition 14.** Given an set system $U = \{e_1, ..., e_n\}$, a collection of subsets $\mathcal{C} = \{C_1, ..., C_m\}$, and a positive integer r_e for each $e \in U$, the Constrained Set Multicover problem asks to find the smallest subcollection (each set must be used at most once) such that each element e is covered by at least r_e times.

Probably the most natural greedy algorithm does the following:
- Pick a set C with the largest cardinality (ties broken arbitrarily).

- Set $r_e \leftarrow r_e - 1$ for $e \in C$. If $r_e = 0$, remove it from U. For each $C \in \mathcal{C}$, let $C \leftarrow C \cap U$.
- Repeat while U is nonempty.

Constrained Set Cover has the following standard LP relaxation, and Rajagopalan and Vazirani [46] showed that the greedy algorithm gives an integral solution whose value is at most H_d (i.e. the dth harmonic number) times the optimal solution to the LP, where d is the maximum set size.

$$\text{minimize} \quad \sum_{C \in \mathcal{C}} z_C$$
$$\text{subject to} \quad \sum_{C:e\in C} z_C \geq r_e \qquad\qquad e \in U$$
$$0 \leq z_C \leq 1 \qquad\qquad C \in \mathcal{C}$$

Our remaining problem, k-Star Transversal on G', can be thought as an instance of Constrained Set Cover in the following way: $U := \{u \in V_G \setminus S : \deg'(u) \geq k - 1\}$ with $r_u := \deg'(u) - k + 2$, and for each $v \in V_G \setminus S$, add $\mathsf{Nbr}'(v) \cap U$ to $\mathcal{C}$. Intuitively, this formulation requires at least r_u neighbors be picked in the transversal whether u is picked or not. This is not a valid reduction because the optimal solution of the above formulation can be much more than the optimal solution of our problem. However, at least one direction is clear (any feasible solution to the above formulation is feasible for our problem), and it suffices to show that the above LP admits a solution whose value is close to the optimum of our problem. The LP relaxation of the above special case of Constrained Set Cover is the following:

$$\text{minimize} \quad \sum_{v \in V_G \setminus S} z_v$$
$$\text{subject to} \quad \sum_{v:v\in\mathsf{Nbr}'(u)} z_v \geq \deg'(u) - k + 2 \qquad\qquad u \in U$$
$$0 \leq z_v \leq 1 \qquad\qquad v \in V_G \setminus S$$

Consider the last iteration of the first phase where we solved $\mathsf{SA}(S)$. Let x be the optimal solution to $\mathsf{SA}(S)$ and $\mathsf{Frac} := \sum_v x_v - |S|$. Note that $x_v < \frac{1}{\alpha}$ when $v \notin S$. Define $\{y_v\}_{v \in V \setminus S}$ such that $y_v := 2x_v$.

▶ **Lemma 15.** $\{y_v\}$ *is a feasible solution to the above LP for Constrained Set Cover.*

Proof. By construction $0 \leq y_v < \frac{2}{\alpha}$, so it suffices to check for each $u \in U$,

$$\sum_{v:v\in\mathsf{Nbr}'(u)} y_v \geq \deg'(u) - k + 2.$$

Fix $u \in U$. Recall that Sherali-Adams constraints on x imply that

$$\sum_{v:\mathsf{Nbr}'(u)} (x_v - x_{u,v}) \geq (\deg'(u) - k + 2)(1 - x_u)$$
$$\Rightarrow \sum_{v:\mathsf{Nbr}'(u)} x_v \geq (\deg'(u) - k + 2)(1 - x_u)$$
$$\Rightarrow \sum_{v:\mathsf{Nbr}'(u)} 2x_v \geq \deg'(u) - k + 2,$$

where the last line follows from the fact that $1 - \frac{1}{\alpha} > \frac{1}{2}$. ◀

Therefore, Constrained Set Cover LP admits a feasible solution of value $2\,\mathsf{Frac}$, and the greedy algorithm gives a k-Star Transversal F with $|F| \leq 2 \cdot \mathsf{Frac} \cdot H_{2k}$. Since Frac is at most the size of the optimal k-Star Transversal for G' (and clearly G), $|S \cup F|$ is at most $O(\log k)$ times the size of the smallest k-Star Transversal of G.