# Inapproximability of H-Transversal/Packing 

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#### Abstract

Given an undirected graph $G=\left(V_{G}, E_{G}\right)$ and a fixed pattern graph $H=\left(V_{H}, E_{H}\right)$ 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 $V_{G} \backslash S$ does not have $H$ as a subgraph, and the latter asks to find the maximum number of pairwise disjoint $k$-subsets $S_{1}, \ldots, S_{m} \subseteq V_{G}$ 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.


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## 1 Introduction

Given a collection of subsets $S_{1}, \ldots, S_{m}$ 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 $S_{i_{1}}, \ldots, S_{i_{m^{\prime}}}$ 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 $O\left(\frac{k}{\log k}\right)[34]$ 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).

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Given a large graph $G=\left(V_{G}, E_{G}\right)$ and a fixed graph $H=\left(V_{H}, E_{H}\right)$ 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 $\Omega(k / \operatorname{polylog}(k))$. 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 $k^{\epsilon}$ 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 $G=\left(V_{G}, E_{G}\right)$ and $H=\left(V_{H}, E_{H}\right)$ with $\left|V_{H}\right|=k$, we define the following problems.

- $H$-Transversal asks to find the smallest $F \subseteq V_{G}$ such that the subgraph of $G$ induced by $V_{G} \backslash F$ does not have $H$ as a subgraph.
- $H$-Packing asks to find the maximum number of pairwise disjoint $k$-subsets of $S_{1}, \ldots, S_{m}$ 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 $\mathrm{NP} \subseteq \mathrm{BPP}$, no polynomial time algorithm approximates $H$-Transversal within a factor better than $k-1$, and H-Packing within a factor better than $\Omega\left(\frac{k}{\log ^{7} k}\right)$.

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 $G=\left(V_{G}, E_{G}\right)$, find the smallest $F \subseteq V_{G}$ such that the subgraph induced by $V_{G} \backslash F$ 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 $\Omega\left(\frac{k}{\log ^{4} k}\right)$. 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 $\mathrm{NP} \subseteq$ BPP, no polynomial time algorithm approximates MIS-k-g within a factor of $\Omega\left(\frac{k}{\log ^{7} k}\right)$.

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 $\Omega\left(\frac{k}{\log ^{4} k}\right)$, 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 $V^{\prime} \subseteq V_{G}$ such that the subgraph induced by $V^{\prime}$ 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. $\left\{H_{1}, \ldots, H_{l}\right\}$-Transversal in our terminology) can be approximated with a constant ratio. They did not provide explicit constants (one trivial approximation ratio for $\left\{H_{1}, \ldots, H_{l}\right\}$ Transversal is $\left.\max \left(\left|V_{H_{1}}\right|, \ldots,\left|V_{H_{l}}\right|\right)\right)$, 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 $G \backslash F$ 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 $\mathrm{P} \neq \mathrm{NP}$ or NP $\nsubseteq B P P$.

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 $3,4, \ldots, O\left(\frac{\log n}{\log \log n}\right)$.

- Theorem 4. Fix an integer $k \geq 3$ and $\epsilon \in(0,1)$. Given a graph $G=\left(V_{G}, E_{G}\right)$ (directed or undirected), unless $\mathrm{NP} \subseteq \mathrm{BPP}$, 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 $O\left(\frac{\log n}{\log \log n}\right)$ (hidden constant in $O$ depends on $k$ and $\epsilon$ ).
- 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 $\epsilon$ 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.

$$
\min \sum_{v \in V_{G}} x_{v} \quad \text { subject to } \quad \sum_{v \in C} x_{v} \geq 1 \quad \forall \text { cycle } C, \quad \text { and } \quad 0 \leq x_{v} \leq 1 \quad \forall v \in V_{G}
$$

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 $O\left(n \log \log n / \log ^{0.1} n\right)=o(n)$ 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 $\mathrm{P} \neq \mathrm{NP}$.

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 EdgeDisjoint 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 $2 n$ 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 $P=\left(V_{P}, E_{P}\right)$ such that each $e \in E_{P}$ is a $k$-subset of $V_{P}$. We denote $e$ as an ordered $k$-tuple $e=\left(v^{1}, \ldots, v^{k}\right)$. 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 $i$ th (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 $\epsilon$.

### 2.3 MIS-k

Given a graph $G=\left(V_{G}, E_{G}\right)$, 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 $O\left(\frac{\log ^{3} k}{k}\right)$ contains an edge.

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

## $3 \quad \mathrm{H}$-Transversal

In this section, given a 2-connected graph $H=\left(V_{H}, E_{H}\right)$ with $k$ vertices, we give a reduction from $k$-HVC to $H$-Transversal. The simplest try will be, given a hypergraph $P=\left(V_{P}, E_{P}\right)$ (let $n=\left|V_{P}\right|, m=\left|E_{P}\right|$ ), to produce a graph $G=\left(V_{G}, E_{G}\right)$ where $V_{G}=V_{P}$, and for each hyperedge $e=\left(v^{1}, \ldots, v^{k}\right)$ add $\left|E_{H}\right|$ 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 $\left(v^{1}, \ldots, v^{k}\right)$, 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 $P=\left(V_{P}, E_{P}\right)$, construct an undirected graph $G=\left(V_{G}, E_{G}\right)$ such that

- $V_{G}=V_{P} \times[B]$. Let $n=\left|V_{P}\right|$ and $N=\left|V_{G}\right|=n B$. For $v \in V_{P}$, let $\operatorname{cloud}(v):=\{v\} \times[B]$ be the copy of $[B]$ associated with $v$.
- For each hyperedge $e=\left(v^{1}, \ldots, v^{k}\right)$, for $a B$ times, take $l^{1}, \ldots, l^{k}$ independently and uniformly from $[B]$. For each edge $(i, j) \in H(1 \leq i<j \leq k)$, add $\left(\left(v^{i}, l^{i}\right),\left(v^{j}, l^{j}\right)\right)$ to $E_{G}$. Each time we add $\left|E_{H}\right|$ edges isomorphic to $H$, and we have $a B$ 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 $V_{G} \backslash F$ 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 $\left(v^{1}, l^{1}\right), \ldots,\left(v^{k}, l^{k}\right)$ 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 $k d$ choices for each $v^{i}(i>1)$. The number of possibilities to choose such $\left(v^{1}, l^{1}\right), \ldots,\left(v^{k}, l^{k}\right)$ is bounded by $n(d k)^{k} B^{k}$. 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 $\left(\left(v^{1}, l^{1}\right), \ldots,\left(v^{k}, l^{k}\right)\right)$ contains a copy in this order - formally, for all $(i, j) \in E_{H},\left(\left(v^{i}, l^{i}\right),\left(v^{j}, l^{j}\right)\right) \in E_{G}$. For each $(i, j) \in E_{H}$, we call a pair $\left(\left(v^{i}, l^{i}\right),\left(v^{j}, l^{j}\right)\right) \in\binom{V_{G}}{2}$ 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 $\left|E_{H}\right|$ 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 $\left\{v^{1}, \ldots, v^{k}\right\}$. In this case, $\left(\left(v^{1}, l^{1}\right), \ldots,\left(v^{k}, l^{k}\right)\right)$ intersects $F$. (right case of Figure 1). When $\left|E_{H}\right|$ 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 $\left(\left(v^{1}, l^{1}\right), \ldots,\left(v^{k}, l^{k}\right)\right)$ cannot be covered by a single canonical copy. Then the probability that it forms a copy of $H$ is at most $\frac{(a d k)^{k^{2}}}{B^{k}}$.

Proof. Fix $2 \leq p \leq\left|E_{H}\right|$. Partition $\left|E_{H}\right|$ 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 $p^{\left|E_{H}\right|}$ possibilities to partition. For each $v \in V_{P}$, there are at most $d$ hyperedges containing $v$ and at most $a B d$ canonical copies intersecting cloud $(v)$. Therefore, all edges in one group can be covered simultaneously by at most $a B d$ copies of canonical copies. There are at most $(a B d)^{p}$ 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 $L=\left(V_{L}, E_{L}\right)$ which is supposed to cover them. Let $I^{\prime} \subseteq V_{G}$ be the set of vertices which are incident on the edges in $I$. Suppose $V_{L}=\left\{\left(u^{1}, l^{\prime 1}\right), \ldots,\left(u^{k}, l^{\prime k}\right)\right\}$, which is created by a hyperedge $f=\left(u^{1}, \ldots, u^{k}\right) \in E_{P}$. We calculate the probability that $L$ contains all edges in $I$ over the choice of labels $l^{\prime 1}, \ldots, l^{\prime k}$ for $L$. One necessary condition is that $\left\{v \mid(v, l) \in I^{\prime}\right.$ for some $\left.l \in[B]\right\}$ (i.e. the set $I^{\prime}$ projected to $V_{P}$ ) is contained in $f$. Otherwise, some vertices of $I^{\prime}$ cannot be covered by $L$. Another necessary condition is $v^{i} \neq v^{j}$ for any $\left(v^{i}, l^{i}\right) \neq\left(v^{j}, l^{j}\right) \in I^{\prime}$. Otherwise (i.e. $\left(v, l^{i}\right),\left(v, l^{j}\right) \in I^{\prime}$ for $\left.l^{i} \neq l^{j}\right)$, since $L$ gives only one label to each vertex in $f \subseteq V_{P},\left(v, l^{i}\right)$ and $\left(v, l^{j}\right)$ cannot be contained in $L$ simultaneously. Therefore, we have a nice characterization of $I^{\prime}$ : 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 $\left(v^{i}, l^{i}\right) \in I^{\prime}$, $l^{i}$ is equal to the label $L$ assigns to $v^{i}$, which is $B^{-\left|I^{\prime}\right|}$. Now we need the following lemma saying that the sum of $\left|I^{\prime}\right|$ 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, $\sum_{i=1}^{p}\left|I_{i}^{\prime}\right| \geq k+p$.

Proof. Let $t$ be the number of vertices contained in at least two $I_{i}^{\prime}$ s. Call them boundary vertices. Note that exactly $k-t$ vertices belongs to exactly one $I_{i}^{\prime}$. For $i=1, \ldots, p$, let $b_{i}$ be the number of boundary vertices in $\left|I_{i}^{\prime}\right|$. Since $\left(I_{i}^{\prime}, I_{i}\right)$ is a proper subgraph of $H$ and $H$ is 2 -vertex connected, $b_{i} \geq 2$ for each $i$. Therefore,

$$
\sum_{i=1}^{p}\left|I_{i}^{\prime}\right|=(k-t)+\max (2 p, 2 t) \geq k+p
$$

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

$$
(a B d)^{p} \prod_{q=1}^{p} B^{-\left|I_{q}^{\prime}\right|}=\frac{(a B d)^{p}}{B^{k+p}}=\frac{(a d)^{p}}{B^{k}} .
$$

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

$$
\sum_{p=2}^{\left|E_{H}\right|} p^{\left|E_{H}\right|} \frac{(a d)^{p}}{B^{k}} \leq \frac{(a d k)^{k^{2}}}{B^{k}}
$$

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

### 3.3 Soundness

The soundness claim above is easier to establish. By an averaging argument, a subset $I$ of $V_{G}$ of measure $2 \epsilon$ must contain $\epsilon B$ vertices from the clouds corresponding to a subset $S$ of measure $\epsilon$ 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 $B=\Omega\left(\log \left|E_{P}\right|\right)$, if every subset of $V_{P}$ of measure at least $\epsilon$ contains a hyperedge in the induced subgraph, with probability at least $3 / 4$, every subset of $V_{G}$ with measure $2 \epsilon$ contains a canonical copy of $H$.

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

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 $\exp \left(k B-a B \epsilon^{k}\right)=\exp (-B) \leq \frac{1}{4\left|E_{P}\right|}$ by taking $a$ large enough constant depending on $k$ and $\epsilon$, and $B=\Omega\left(\log \left|E_{P}\right|\right)$. 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 $2 \epsilon$. By an averaging argument, at least $\epsilon$ fraction of good vertices $v \in V_{P}$ satisfy that $\left|\operatorname{cloud}\left(v^{i}\right) \cap I\right| \geq \epsilon B$. 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 $M=\left(V_{M}, E_{M}\right)$ of MIS- $k$ - $k$, we take $G=\left(V_{G}, E_{G}\right)$ 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 $\operatorname{star}(v):=\left\{e \in V_{G}: v \in e\right\} . \operatorname{star}(v)$ induces a $k$-clique, and for $v, u \in V_{M}, \operatorname{star}(v)$ and $\operatorname{star}(u)$ 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 $M=\left(V_{M}, E_{M}\right)$ be an instance of MIS- $k$ - $k M$ with maximum degree $k$ and girth greater than $k$. First, let $G=\left(V_{G}=E_{M}, E_{G}\right)$ be the line graph of $M$. For each vertex $v \in V_{M}$ with degree strictly less than $k$, we add $k-\operatorname{deg}(v)$ new vertices to $V_{G}$. Let $\operatorname{star}(v) \subseteq V_{G}$ be the union of the edges of $M$ incident on $v$ and the newly added vertices for $v$. Note that $|\operatorname{star}(v)|=k$ for all $v \in V_{M}$. Add edges to $G$ to ensure that every $\operatorname{star}(v)$ induces a $k$-clique. For two vertices $u$ and $v$ of $M, \operatorname{star}(u)$ and $\operatorname{star}(v)$ 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 $\{\operatorname{star}(v)\}_{v \in S}$ 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 $\operatorname{star}(v)$ 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 $T \notin \operatorname{star}(u), T$ 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 $T \notin \operatorname{star}(v)$ gives another edge $\left(w^{\prime}, u\right)$ in $T$. If $w \neq w^{\prime},(w, v)$ and $\left(w^{\prime}, u\right)$ 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 $P_{1}^{\prime}, P_{2}^{\prime}$ in $M$ where each $P_{i}^{\prime}$ 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 $P_{1}^{\prime}, P_{2}^{\prime}$ together with $(u, v),(w, x)$. Since $\left|P_{1}^{\prime}\right|+\left|P_{2}^{\prime}\right|+2 \leq k$, 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 NP $\subseteq$ BPP, no polynomial time algorithm approximates MIS-k-g within a factor of $\Omega\left(\frac{k}{\log ^{7} k}\right)$.

Proof. We reduce from MIS- $d$ to MIS- $k-g$ where $k=O\left(d \log ^{2} d\right)$. Given an instance $G_{0}=$ $\left(V_{G_{0}}, E_{G_{0}}\right)$ of MIS- $d$, we construct $G=\left(V_{G}, E_{G}\right)$ and $G^{\prime}=\left(V_{G^{\prime}}, E_{G^{\prime}}\right)$ by the following procedure:

- $V_{G}=V_{G_{0}} \times[B]$. As usual, let $\operatorname{cloud}(v)=\{v\} \times[B]$.
- For each edge $(u, v) \in E_{G_{0}}$, for $a$ times, add a random matching as follows.
- Take a random permutation $\pi:[B] \rightarrow[B]$.
- Add an edge $((u, i),(v, \pi(i))$ for all $i \in[B]$.
- Call the resulting graph $G$. To get the final graph $G^{\prime}$,
- 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 $O\left(n^{g}\right)$. Let $n=\left|V_{G_{0}}\right|, m=\left|E_{G_{0}}\right|, N=n B=\left|V_{G}\right| \geq\left|V_{G^{\prime}}\right|, M=m \cdot a B=\left|E_{G}\right| \geq\left|E_{G^{\prime}}\right|$. The maximum degree of $G$ and $G^{\prime}$ is at most ad. By construction, girth of $G^{\prime}$ 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 $\left|V_{G}\right|$ and $\left|V_{G^{\prime}}\right|$ are almost the same. Let $k^{\prime}$ be the length of a purported cycle. Choose $k^{\prime}$ vertices $\left(v^{1}, l^{1}\right), \ldots,\left(v^{k^{\prime}}, l^{k^{\prime}}\right)$ which satisfy

- $v^{1} \in V_{G_{0}}$ can be any vertex.
- For each $1 \leq i<k^{\prime},\left(v^{i}, v^{i+1}\right) \in E_{G_{0}}$.
- $l^{1}, \ldots, l^{k^{\prime}} \in B$ 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 $\left(v^{1}, l^{1}\right), \ldots,\left(v^{k^{\prime}}, l^{k^{\prime}}\right)$ is bounded by $n d^{k^{\prime}-1} B^{k^{\prime}}$. 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 $\left(\left(v^{1}, l^{1}\right), \ldots,\left(v^{k^{\prime}}, l^{k^{\prime}}\right)\right)$ and the subgraph induced by $\operatorname{cloud}(u) \cup \operatorname{cloud}(w)$. 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 $\operatorname{Pr}\left[e^{i} \mid e^{1}, \ldots, e^{i-1}\right]$ for each $e^{i}$.

- Claim 2. $\operatorname{Pr}\left[e^{i} \mid e^{1}, \ldots, e^{i-1}\right] \leq \frac{a}{B-i}$.

Proof. There are $a$ random matchings between $\operatorname{cloud}(u)$ and $\operatorname{cloud}(w)$, 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, $\operatorname{Pr}\left[e^{i} \mid e^{1}, \ldots, e^{i-1}\right] \leq \frac{a}{B-i}$.
The probability that all of $e^{1}, \ldots, e^{q}$ exist is at most

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

Since edges of $G_{0}$ are processed independently, the probability of success for one fixed purported cycle is $\left(\frac{a}{B-k^{\prime}}\right)^{k^{\prime}}$. The expected number of cycles of length $k^{\prime}$ is

$$
\begin{aligned}
& n d^{k^{\prime}-1} B^{k^{\prime}} \cdot\left(\frac{a}{B-k^{\prime}}\right)^{k^{\prime}}=n d^{k^{\prime}-1} a^{k^{\prime}}\left(1+\frac{k^{\prime}}{B-k^{\prime}}\right)^{k^{\prime}} \\
\leq & n d^{k^{\prime}-1} a^{k^{\prime}} \exp \left(\frac{k^{\prime 2}}{B-k^{\prime}}\right) \leq e n(a d)^{k^{\prime}}
\end{aligned}
$$

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

Hardness of MIS- $d$ states that it is NP-hard to distinguish the case $G_{0}$ has an independent set of measure $c:=\Omega\left(\frac{1}{\log d}\right)$ and the case where the maximum independent set has measure at most $s:=O\left(\frac{\log ^{3} d}{d}\right)$.

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 $I^{\prime}=I \cap V_{G^{\prime}}$. $I^{\prime}$ is independent in both $G$ and $G^{\prime}$, and the measure of $I^{\prime}$ in $G^{\prime}$ is at least the measure of $I^{\prime}$ in $G$, which is at least $c-1 / d^{2}=\Omega\left(\frac{1}{\log d}\right)$.

Soundness. Suppose that every subset of $V_{G_{0}}$ of measure at least $s$ contains an edge. Say a graph is $(\beta, \alpha)$-dense if we take $\beta$ fraction of vertices, at least $\alpha$ fraction of edges lie within the induced subgraph. We also say a bipartite graph is $(\beta, \alpha)$-bipartite dense if we take $\beta$ fraction of vertices from each side, at least $\alpha$ fraction of edges lie within the induced subgraph.

- Claim 3. For $a=O\left(\frac{\log (1 / s)}{s}\right)$ and $B=O\left(\frac{\log m}{s}\right)$ the following holds with probability at least 3/4: For every $(u, w) \in E_{G_{0}}$, the bipartite graph between $\operatorname{cloud}(u)$ and $\operatorname{cloud}(w)$ is $\left(\epsilon, \epsilon^{2} / 8\right)$-bipartite dense for all $\epsilon \geq s$.

Proof. Fix $(u, w)$, and $\epsilon \in[s, 1]$, and $X \subseteq \operatorname{cloud}(u)$ and $Y \subseteq \operatorname{cloud}(w)$ 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. $\operatorname{Pr}\left[X_{1}=1\right]=\epsilon$, and $\operatorname{Pr}\left[X_{i}=1 \mid X_{1}, \ldots, X_{i-1}\right] \geq \epsilon / 2$ 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 $\exp \left(\frac{a \epsilon^{2} B}{32}\right)$. By union bound over all possibilities of choosing $X$ and $Y$, the probability that the bipartite graph is not $\left(\epsilon, \epsilon^{2} / 8\right)$-bipartite dense is

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

by taking $a=O\left(\frac{\log (1 / s)}{s}\right)$ and $B=O\left(\frac{\log m}{s}\right)$. A union bound over all possible choices of $\epsilon$ ( $B$ possibilities) and $m$ edges of $E_{0}$ implies the claim.

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

Proof. Fix a subset $S$ of measure $4 s \log (1 / s)$. For a vertex $v$ of $G_{0}$, let $\mu(v):=\frac{|\operatorname{cloud}(v) \cap S|}{B}$. Note that $\mathbb{E}_{v}[\mu(v)]=4 s \log (1 / s)$. Partition $V_{G_{0}}$ into $t+1$ buckets $B_{0}, \ldots, B_{t}(t:=$ $\left\lceil\log _{2}(1 / s)\right\rceil$ ), such that $B_{0}$ contains $v$ such that $\mu(v) \leq s$, and for $i \geq 1, B_{i}$ contains $v$ such that $\mu(v) \in\left(2^{i-1} s, 2^{i} s\right]$. Denote

$$
\mu\left(B_{i}\right):=\frac{\sum_{v \in B_{i}} \mu(v)}{\left|V_{G_{0}}\right|} .
$$

Clearly $\mu\left(B_{0}\right) \leq s$. Pick $i \in\{1, \ldots, t\}$ with the largest $\mu\left(B_{i}\right)$. We have $\mu\left(B_{i}\right) \geq 2 s$ since $\mathbb{E}_{v}[\mu(v)] \geq 4 s \log (1 / s)$. Let $\gamma=2^{i-1} s$. All vertices of $B_{i}$ has $\mu(v) \in[\gamma, 2 \gamma]$, so $\left|B_{i}\right| \geq(s / \gamma) n$.

Since $G_{0}$ has no independent set with more than $n s$ vertices, Turán's Theorem says that the subgraph of $G_{0}$ induced by $B_{i}$ has at least $\frac{\left|B_{i}\right|}{2}\left(\frac{\left|B_{i}\right|}{n s}-1\right)=\Omega\left(\frac{s}{\gamma^{2}} n\right)$ edges. This is at least $\Omega\left(\frac{s}{d \gamma^{2}}\right)$ 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 $\Omega\left(\frac{s}{d \gamma^{2}}\right) \cdot \frac{\gamma^{2}}{8}=\Omega\left(\frac{s}{d}\right)$ fraction of edges inside the subgraph induced by $S$.

Sparsification. Recall that $G^{\prime}$ 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, $G^{\prime}$ has an independent set of measure at least $c-1 / d^{2}=\Omega\left(\frac{1}{\log d}\right)$. In the soundness case, $G$ is $\left(4 s \log (1 / s), \Omega\left(\frac{s}{d}\right)\right)$-dense, so $G^{\prime}$ is $(\beta, \alpha)$-dense where $\beta:=\Omega\left(\frac{\log ^{4} d}{d}\right), \alpha:=\Omega\left(\frac{\log ^{3} d}{d^{2}}\right)$. Using density of $G^{\prime}$, we sparsify $G^{\prime}$ again keep each edge of $G^{\prime}$ by probability $\frac{k n}{\left|E_{G^{\prime}}\right|}$ so that the expected total number of edges is $k n$.

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

$$
n \cdot \max _{\beta \in\left[\beta_{0}, 1\right]}\{\exp (-\alpha k n / 32) \cdot \exp (2 \beta \log (1 / \beta) n)\} \leq n \cdot e^{-n}
$$

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

Combining all these results, we have a graph with small degree $10 k=O\left(d \log ^{2} d\right)$ and girth strictly greater than $g$, where it is NP-hard to approximate MIS within a factor of $\frac{c-\frac{1}{d^{2}}}{\beta}=\Omega\left(\frac{d}{\log ^{5} d}\right)=\Omega\left(\frac{k}{\log ^{7} k}\right)$. Therefore, it is NP-hard to approximate MIS- $k$ - $g$ within a factor of $\Omega\left(\frac{k}{\log ^{7} k}\right)$.
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## 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 $G=\left(V_{G}, E_{G}\right)$ 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 $V_{G} \backslash F$ 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 $V_{G} \backslash F$ has small degree (at most $2 k$ ) 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 $\left\{x_{v}\right\}_{v \in V_{G}} \cup$ $\left\{x_{u, v}\right\}_{u, v \in V_{G}}$. 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.

$$
\begin{array}{lll}
\text { minimize } & \sum_{v \in V_{G}} x_{v} & \\
\text { subject to } & 0 \leq x_{u, v}, x_{v} \leq 1 & \forall u, v \in V_{G} \\
& x_{u, v} \leq x_{u} & \forall u, v \in V_{G} \\
& x_{u}+x_{v}-x_{u, v} \leq 1 & \forall u, v \in V_{G} \\
& \sum_{v:(u, v) \in E_{G}}\left(x_{v}-x_{u, v}\right) \geq(\operatorname{deg}(u)-k+2)\left(1-x_{u}\right) & \forall u \in V_{G}
\end{array}
$$

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: $\operatorname{Pr}[\alpha(u)=\alpha(v)=1]:=x_{u, v}, \operatorname{Pr}[\alpha(u)=0, \alpha(v)=1]:=x_{v}-x_{u, v}, \operatorname{Pr}[\alpha(u)=$ $1, \alpha(v)=0]:=x_{u}-x_{u, v}, \operatorname{Pr}[\alpha(u)=\alpha(v)=0]:=1-x_{u}-x_{v}+x_{u, v}$.

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 $\operatorname{deg}(u)-k+2$ 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 $V_{G} \backslash S$ - 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 $\operatorname{SA}(S)$.
- $S \leftarrow\left\{v: x_{v} \geq \frac{1}{\alpha}\right\}$, 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 $V_{G} \backslash S$ is small.
- The remaining solution has small fractional values $-x_{v}<\frac{1}{\alpha}$ for all $v \in V_{G} \backslash S$.

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 $\mathrm{SA}(\emptyset)$. When the above procedure terminates, $|S| \leq \alpha$ Frac.

Proof. Assume that the above loop iterated $l$ times, and for $i=0, \ldots, l$, let $S_{i}$ be $S$ after the $i$ th loop such that $S_{0}=\emptyset, \ldots, S_{l}=S$. We use induction from the last iteration. Let $\mathrm{Frac}_{i}$ be the optimal fractional solution to $\mathrm{SA}\left(S_{i}\right)$ minus $\left|S_{i}\right|$ such that $\mathrm{Frac}=\mathrm{Frac}_{0}$.

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

$$
\left|S_{l}\right|-\left|S_{l-1}\right|=\left|\left\{v \notin S_{l-1}: x_{v} \geq \frac{1}{\alpha}\right\}\right| \leq \alpha \operatorname{Frac}_{l-1}
$$

For $i=l-2, l-1, \ldots, 0$, we show that $\left|S_{l}\right|-\left|S_{i}\right| \leq \alpha \operatorname{Frac}_{i}$. Let $x$ be the optimal fraction solution to $\mathrm{SA}\left(S_{i}\right)$, and $x^{\prime}$ be the solution obtained by partially rounding $x$ in the following way.

- $x_{v}^{\prime}=1$ if $v \in S_{i}$. Otherwise, $x_{v}^{\prime}=x_{v}$.
- $x_{u, v}^{\prime}=x_{u}^{\prime}\left(v \in S_{i}\right), x_{v}^{\prime}\left(u \in S_{i}\right)$, or $x_{u, v}$ otherwise.

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

$$
\left|S_{i}\right|+\sum_{v \notin S_{i}, x_{v}<\frac{1}{\alpha}} x_{v} \geq\left|S_{i}\right|+\mathrm{Frac}_{i+1}
$$

which implies

$$
\mathrm{Frac}_{i}=\sum_{v \notin S_{i}, x_{v} \geq \frac{1}{\alpha}} x_{v}+\sum_{v \notin S_{i}, x_{v}<\frac{1}{\alpha}} x_{v} \geq \frac{1}{\alpha}\left(\left|S_{i+1}\right|-\left|S_{i}\right|\right)+\mathrm{Frac}_{i+1}
$$

Finally, we have

$$
\begin{array}{r}
\left|S_{l}\right|-\left|S_{i}\right| \\
=\left(\left|S_{l}\right|-\left|S_{i+1}\right|\right)+\left(\left|S_{i+1}\right|-\left|S_{i}\right|\right) \\
\leq \alpha \operatorname{Frac}_{i+1}+\left(\left|S_{i+1}\right|-\left|S_{i}\right|\right) \\
\leq \alpha \operatorname{Frac}_{i},
\end{array}
$$

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

- Lemma 13. After the termination, every vertex has degree at most $2 k$ in the subgraph induced by $V_{G} \backslash S$.

Proof. We prove that at least one vertex is added to $S$ if the subgraph induced by $V_{G} \backslash S$ has a vertex of degree more than $2 k$. Fix one such iteration, and let $S_{1}$ and $S_{2}$ be $S$ before and after the iteration respectively. Let $G^{\prime}$ be the subgraph of $G$ induced by $V_{G} \backslash S_{1}$. If the subgraph induced by $V_{G} \backslash S_{2}$ does not have any vertex with degree more than $2 k$, we are done. Otherwise, fix one such vertex $u \in V_{G} \backslash S_{2}$. Note that the degree of $u$ in $G^{\prime}$ is also more than $2 k$.

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 $\mathrm{SA}\left(S_{1}\right)$ and consider the following constraint for $u$.

$$
\sum_{v:(u, v) \in E_{G}}\left(x_{v}-x_{u, v}\right) \geq(\operatorname{deg}(u)-k+2)\left(1-x_{u}\right)
$$

Let $\operatorname{Nbr}(u)$ and $\operatorname{Nbr}^{\prime}(u)$ be the set of neighbors of $u$ in $G$ and $G^{\prime}$ respectively, and $\operatorname{deg}^{\prime}(u)=$ $\left|\operatorname{Nbr}^{\prime}(u)\right|$. Note that $\operatorname{Nbr}^{\prime}(u)=\operatorname{Nbr}(u) \backslash S_{1}$, and for $v \in \operatorname{Nbr}(u) \cap S_{1}, x_{v}=1$ and $x_{u, v}=x_{u}$. Therefore, the above constraint is equivalent to

$$
\begin{aligned}
& \sum_{v: \operatorname{Nbr}(u) \cap S_{1}}\left(1-x_{u}\right)+\sum_{v: \mathrm{Nbr}^{\prime}(u)}\left(x_{v}-x_{u, v}\right) \geq(\operatorname{deg}(u)-k+2)\left(1-x_{u}\right) \\
\Leftrightarrow & \sum_{v: \operatorname{Nbr}^{\prime}(u)}\left(x_{v}-x_{u, v}\right) \geq\left(\operatorname{deg}^{\prime}(u)-k+2\right)\left(1-x_{u}\right) .
\end{aligned}
$$

The fact that $u \notin S_{2}$ implies that $x_{u}<\frac{1}{\alpha}$, which implies

$$
\begin{aligned}
& \sum_{v \in \operatorname{Nbr}^{\prime}(u)} x_{v} \\
\geq & \sum_{v \in \operatorname{Nbr}^{\prime}(u)}\left(x_{v}-x_{u, v}\right) \geq\left(1-\frac{1}{\alpha}\right)\left(\operatorname{deg}^{\prime}(u)-k\right)=\quad\left(1-\frac{1}{\alpha}\right) \operatorname{deg}^{\prime}(u)\left(1-\frac{k}{\operatorname{deg}^{\prime}(u)}\right) .
\end{aligned}
$$

Therefore, there is one $v \in \operatorname{Nbr}^{\prime}(u)$ with $x_{v} \geq\left(1-\frac{1}{\alpha}\right)\left(1-\frac{k}{\operatorname{deg}^{\prime}(u)}\right) \geq \frac{9}{10} \cdot \frac{1}{2}>\frac{1}{\alpha} \cdot v$ 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 $\alpha$ times the optimal solution and the subgraph induced by $V_{G} \backslash S$ has maximum degree at most $2 k$. As above, let $G^{\prime}$ be the subgraph induced by $V_{G} \backslash S, \operatorname{Nbr}(u), \operatorname{Nbr}^{\prime}(u)$ be the neighbors of $u$ in $G$ and $G^{\prime}$ respectively, and $\operatorname{deg}(u)=|\operatorname{Nbr}(u)|, \operatorname{deg}^{\prime}(u)=\left|\operatorname{Nbr}^{\prime}(u)\right|$. The remaining task is to find a small subset $F \subseteq V_{G} \backslash S$ such that the subgraph of $G^{\prime}$ (and $\left.G\right)$ induced by $V_{G} \backslash(S \cup F)$ 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 $U=\left\{e_{1}, \ldots, e_{n}\right\}$, a collection of subsets $\mathcal{C}=$ $\left\{C_{1}, \ldots, C_{m}\right\}$, 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 $d$ th harmonic number) times the optimal solution to the LP, where $d$ is the maximum set size.

$$
\begin{array}{lll}
\text { minimize } & \sum_{C \in \mathcal{C}} z_{C} & \\
\text { subject to } & \sum_{C: e \in C} z_{C} \geq r_{e} & e \in U \\
& 0 \leq z_{C} \leq 1 & C \in \mathcal{C}
\end{array}
$$

Our remaining problem, $k$-Star Transversal on $G^{\prime}$, can be thought as an instance of Constrained Set Cover in the following way: $U:=\left\{u \in V_{G} \backslash S: \operatorname{deg}^{\prime}(u) \geq k-1\right\}$ with $r_{u}:=\operatorname{deg}^{\prime}(u)-k+2$, and for each $v \in V_{G} \backslash S$, add $\operatorname{Nbr}^{\prime}(v) \cap U$ 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:

$$
\begin{array}{cl}
\text { minimize } & \sum_{v \in V_{G} \backslash S} z_{v} \\
\text { subject to } & \sum_{v: v \in \operatorname{Nbr}^{\prime}(u)} z_{v} \geq \operatorname{deg}^{\prime}(u)-k+2 \\
& 0 \leq z_{v} \leq 1
\end{array} \quad u \in U
$$

Consider the last iteration of the first phase where we solved $\mathrm{SA}(S)$. Let $x$ be the optimal solution to $\operatorname{SA}(S)$ and Frac $:=\sum_{v} x_{v}-|S|$. Note that $x_{v}<\frac{1}{\alpha}$ when $v \notin S$. Define $\left\{y_{v}\right\}_{v \in V \backslash S}$ such that $y_{v}:=2 x_{v}$.

- Lemma 15. $\left\{y_{v}\right\}$ 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 $u \in U$,

$$
\sum_{v: v \in \operatorname{Nbr}^{\prime}(u)} y_{v} \geq \operatorname{deg}^{\prime}(u)-k+2
$$

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

$$
\begin{aligned}
& \sum_{v: \operatorname{Nbr}^{\prime}(u)}\left(x_{v}-x_{u, v}\right) \geq\left(\operatorname{deg}^{\prime}(u)-k+2\right)\left(1-x_{u}\right) \\
\Rightarrow & \sum_{v: \operatorname{Nbr}^{\prime}(u)} x_{v} \geq\left(\operatorname{deg}^{\prime}(u)-k+2\right)\left(1-x_{u}\right) \\
\Rightarrow & \sum_{v: \operatorname{Nbr}^{\prime}(u)} 2 x_{v} \geq \operatorname{deg}^{\prime}(u)-k+2,
\end{aligned}
$$

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 Frac, and the greedy algorithm gives a $k$-Star Transversal $F$ with $|F| \leq 2 \cdot$ Frac $\cdot H_{2 k}$. Since Frac is at most the size of the optimal $k$-Star Transversal for $G^{\prime}$ (and clearly $\left.G\right),|S \cup F|$ is at most $O(\log k)$ times the size of the smallest $k$-Star Transversal of $G$.

