# Toward a Dichotomy for Approximation of H-Coloring 

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#### Abstract

Given two (di)graphs $G, H$ and a cost function $c: V(G) \times V(H) \rightarrow \mathbb{Q}_{\geq 0} \cup\{+\infty\}$, in the minimum cost homomorphism problem, $\operatorname{MinHOM}(H)$, we are interested in finding a homomorphism $f: V(G) \rightarrow$ $V(H)$ (a.k.a $H$-coloring) that minimizes $\sum_{v \in V(G)} c(v, f(v))$. The complexity of exact minimization of this problem is well understood [35], and the class of digraphs $H$, for which the $\operatorname{MinHOM}(H)$ is polynomial time solvable is a small subset of all digraphs.

In this paper, we consider the approximation of MinHOM within a constant factor. In terms of digraphs, $\operatorname{MinHOM}(H)$ is not approximable if $H$ contains a digraph asteroidal triple (DAT). We take a major step toward a dichotomy classification of approximable cases. We give a dichotomy classification for approximating the $\operatorname{MinHOM}(H)$ when $H$ is a graph (i.e. symmetric digraph). For digraphs, we provide constant factor approximation algorithms for two important classes of digraphs, namely bi-arc digraphs (digraphs with a conservative semi-lattice polymorphism or min-ordering), and $k$-arc digraphs (digraphs with an extended min-ordering). Specifically, we show that: - Dichotomy for Graphs: $\operatorname{MinHOM}(H)$ has a $2|V(H)|$-approximation algorithm if graph $H$ admits a conservative majority polymorphims (i.e. $H$ is a bi-arc graph), otherwise, it is inapproximable; - $\operatorname{MinHOM}(H)$ has a $|V(H)|^{2}$-approximation algorithm if $H$ is a bi-arc digraph; - $\operatorname{MinHOM}(H)$ has a $|V(H)|^{2}$-approximation algorithm if $H$ is a $k$-arc digraph.

In conclusion, we show the importance of these results and provide insights for achieving a dichotomy classification of approximable cases. Our constant factors depend on the size of $H$. However, the implementation of our algorithms provides a much better approximation ratio. It leaves open to investigate a classification of digraphs $H$, where $\operatorname{MinHOM}(H)$ admits a constant factor approximation algorithm that is independent of $|V(H)|$.


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## 1 Introduction

For a digraph $D$, let $V(D)$ denote the vertex set of $D$, and let $A(D)$ denote the arcs of $D$. We denote the number of vertices of $D$ by $|D|$. Instead of $(u, v) \in A(D)$, we use the shorthand $u v \in A(D)$ or simply $u v \in D$. A graph $G$ is a symmetric digraph, that is, $x y \in A(G)$ if and only if (iff) $y x \in A(G)$. An edge is just a symmetric arc.

A homomorphism of a digraph $D$ to a digraph $H$ (a.k.a H-COLORING) is a mapping $f: V(D) \rightarrow V(H)$ such that for each arc $x y$ of $D, f(x) f(y)$ is an arc of $H$. We say mapping $f$ does not satisfy arc $x y$, if $f(x) f(y)$ is not an arc of $H$. The homomorphism problem for a fixed target digraph $H, \operatorname{HOM}(H)$, takes a digraph $D$ as input and asks whether there is a homomorphism from $D$ to $H$. Therefore, by fixing the digraph $H$ we obtain a class of problems, one problem for each digraph $D$. For example, $\operatorname{HOM}(H)$, when $H$ is an edge, is exactly the problem of determining whether the input graph $G$ is bipartite (i.e., the 2-Coloring problem). Similarly, if $V(H)=\{u, v, x\}, A(H)=\{u v, v u, v x, x v, u x, x u\}$, then $\operatorname{HOM}(H)$ is exactly the classical 3-Coloring problem. More generally, if $H$ is a clique on $k$ vertices, then $\operatorname{HOM}(H)$ is the K-Coloring problem. The H-Coloring problem can be considered within a more general framework, the constraint satisfaction problem (CSP). In the CSP associated with a finite relational structure $\mathbb{H}, \operatorname{CSP}(\mathbb{H})$, the question is whether there exists a homomorphism of a given finite relational structure to $\mathbb{H}$. Thus, the H-Coloring problem is a particular case of the CSP in which the involved relational structures are digraphs. A celebrated result due to Hell and Nesetril [31], states that, for graph $H, \operatorname{HOM}(H)$ is in P if $H$ is bipartite or contains a looped vertex, and that it is NP-complete for all other graphs $H$. See [9] for an algebraic proof of the same result, and $[12,55]$ for a dichotomy for $\operatorname{CSP}(\mathbb{H})$.

There are several natural optimization versions of the $\operatorname{HOM}(H)$ problem. One is to find a mapping $f: V(D) \rightarrow V(H)$ that maximizes (minimizes) number of satisfied (unsatisfied) arcs in $D$. This problem is known under the name of Max 2-Csp (Min 2-Csp). For example, the most basic Boolean Max 2-Csp problem is Max Cut where the target graph $H$ is an edge. This line of research has received a lot of attention in the literature and there are very strong results concerning various aspects of approximability Max 2-CsP and Min 2-Csp [2, 22, 28, 41, 45]. See [47] for a recent survey on this and approximation of Max k-Csp and Min k-Csp. We consider another natural optimization version of the $\mathrm{HOM}(H)$ problem, i.e., we are not only interested in the existence of a homomorphism, but want to find the "best homomorphism". The minimum cost homomorphism problem to $H$, denoted by $\operatorname{MinHOM}(H)$, for a given input digraph $D$, and a cost function $c(x, i), x \in V(D), i \in$ $V(H)$, seeks a homomorphism $f$ of $D$ to $H$ that minimizes the total cost $\sum_{x \in V(D)} c(x, f(x))$. The cost function $c$ can take non-negative rational values and positive infinity, that is $c: V(D) \times V(H) \rightarrow \mathbb{Q} \geq 0 \cup\{+\infty\}$. The MinHOM was introduced in [25], where it was motivated by a real-world problem in defence logistics. The MinHOM problem offers a natural and practical way to model and generalizes many optimization problems.

- Example 1 ((Weighted) Minimum Vertex Cover). This problem can be seen as $\operatorname{MinHOM}(H)$ where $V(H)=\{0,1\}, E(H)=\{11,01\}$ and $c(u, 0)=0, c(u, 1)>0$ for every $u \in V(G)$. Note that $G$ and $H$ are graphs.
- Example 2 (List Homomorphism (LHOM)). LHOM $(H)$, seeks, for a given input digraph $D$ and lists $L(x) \subseteq V(H), x \in V(D)$, a homomorphism from $D$ to $H$ such that $f(x) \in L(x)$ for all $x \in V(D)$. This is equivalent to $\operatorname{MinHOM}(H)$ with $c(u, i)=0$ if $i \in L(u)$, otherwise $c(u, i)=+\infty$. This problem is also known as List H-Coloring and its complexity is fully understood due to series of results [ $5,8,10,11,18,33]$.

The MinHOM problem generalizes many other problems such as (Weighted) Min Ones [1, 15, 40], Min Sol [39, 53], a large class of bounded integer linear programs, retraction problems [19], Minimum Sum Coloring [4, 21, 44], and various optimum cost chromatic partition problems [27, 37, 38, 43].

A special case of MinHOM problem is where the cost function $c$ is choosen from a fixed set $\Delta$. This problem is denoted by $\operatorname{MinHOM}(H, \Delta)[14,53,54]$. The Valued Constrained Satisfaction Problems (VCSPs) is a generalization of this special case of the MinHOM problem. An instance of the VCSP is given by a collection of variables that must be assigned labels from a given domain with the goal to minimize the objective function that is given by the sum of cost functions, each depending on some subset of the variables [13]. Interestingly, a recent work by Cohen et al. [14] proved that VCSPs over a fixed valued constraint language are polynomial-time equivalent to $\operatorname{MinHOM}(H, \Delta)$ over a fixed digraph and a proper choice of $\Delta$.

Exact Minimization. The complexity of exact minimization of $\operatorname{MinHOM}(H)$ was studied in a series of papers, and complete complexity classifications were given in [23] for undirected graphs, in [35] for digraphs, and in [51] for more general structures. Certain minimum cost homomorphism problems have polynomial time algorithms [23, 24, 25, 35], but most are NP-hard. We remark that, the complexity of exact minimization of VCSPs is well understood $[42,52]$.

Approximation. For a minimization problem, an $\alpha$-approximation algorithm is a (randomized) polynomial-time algorithm that finds an approximate solution of cost at most $\alpha$ times the minimum cost. A constant ratio approximation algorithm is an $\alpha$-approximation algorithm for some constant $\alpha$. We say a problem is not approximable if there is no polynomial time approximation algorithm with a multiplicative guarantee unless $\mathrm{P}=\mathrm{NP}$. The approximability of MinHOM is fairly understood when we restrict the cost function to a fixed set $\Delta$, and further, we restrict it to take only finite values (not $\infty$ ). This setting is a special case of finite VCSPs, and there are strong approximation results on finite VCSPs. For finite VCSPs, Raghavendra [50] showed how to use the basic SDP relaxation to obtain a constant approximation. Moreover, he proved that the approximation ratio cannot be improved under Unique Game Conjecture (UGC). This constant is not explicit, but there is an algorithm that can compute it with any given accuracy in doubly exponential time. In another line of research, the power of so-called basic linear program (BLP) concerning constant factor approximation of finite VCSPs has been recently studied in [16, 17]. However, the approximability of VCSPs for constraint languages that are not finite-valued remains poorly understood, and [30,39] are the only results on approximation of VCSP for languages that have cost functions that can take infinite values.

Hell et al., [30] proved a dichotomy for approximating $\operatorname{MinHOM}(H)$ when $H$ is a bipartite graph by transforming the $\operatorname{MinHOM}(H)$ to a linear program, and rounding the fractional values to get a homomorphism to $H$.

- Theorem 3 (Dichotomy for bipartite graphs [30]). For a fixed bipartite graph $H$, $\operatorname{MinHOM}(H)$ admits a constant factor approximation algorithm if $H$ admits a min-ordering (complement of $H$ is a circular arc graph), otherwise $\operatorname{MinHOM}(H)$ is not approximable unless $P=N P$.

Beyond this, there is no result concerning the approximation of $\operatorname{MinHOM}(H)$. We go beyond bipartite case and present a constant factor approximation algorithm for bi-arc graphs (graphs with a conservative majority polymorphism). Designing an approximation
algorithm for $\operatorname{MinHOM}(H)$ when $H$ is a digraph is much more complex than when $H$ is a graph. We improve state-of-the-art by providing constant factor approximation algorithms for $\operatorname{MinHOM}(H)$ where $H$ belongs to these two important cases of digraphs, namely bi-arc digraphs (digraphs with a conservative semi-lattice polymorphism a.k.a min-ordering), and $k$-arc digraphs (digraphs with a $k$-min-ordering). To do so, we introduce new LPs that reflect the structural properties of the target (di)graph $H$ as well as new methods to round the fractional solutions and obtain homomorphisms to $H$. We will show our randomized rounding procedure can be de-randomized, and hence, we get a deterministic polynomial algorithm. Furthermore, we argue that our techniques can be used towards finding a dichotomy for the approximation of $\operatorname{MinHOM}(H)$.

### 1.1 Our Contributions

Most of the minimum cost homomorphism problems are NP-hard, therefore we investigate the approximation of $\operatorname{MinHoM}(H)$.

Approximating Minimum Cost Homomorphism to Digraph $H$.
Input: A digraph $D$ and a vertex-mapping costs $c(x, u), x \in V(D), u \in V(H)$, Output: A homomorphism $f$ of $D$ to $H$ with the total cost of $\sum_{x \in V(D)} c(x, f(x)) \leq$ $\alpha \cdot O P T$, where $\alpha$ is a constant.

Here, $O P T$ denotes the cost of a minimum cost homomorphism of $D$ to $H$. Moreover, we assume size of $H$ is constant. Recall that we approximate the cost over real homomorphisms, rather than approximating the maximum weight of satisfied constraints, as in, say, MAX Csp. One can show that if $\operatorname{LHOM}(H)$ is not polynomial time solvable then there is no approximation algorithm for $\operatorname{MinHOM}(H)[30,48]$.

- Observation 4. If $\operatorname{LHOM}(H)$ is not polynomial time solvable, then there is no approximation algorithm for $\operatorname{MinHOM}(H)$.

The complexity of the LHOM problems for graphs, digraphs, and relational structures (with arity two and higher) have been classified in $[18,33,10]$ respectively. $\operatorname{LHOM}(H)$ is polynomial time solvable if the digraph $H$ does not contain a digraph asteroidal triple $(D A T)^{1}$ as an induced sub-digraph, and NP-complete when $H$ contains a DAT [33].
$\operatorname{MinHOM}(H)$ is polynomial time solvable when digraph $H$ admits a $k$-min-max-ordering, a subclass of DAT-free digraphs, and otherwise, NP-complete [35, 34]. Here, in this paper, we take an important step towards closing the gap between DAT-free digraphs and the one that admit a $k$-min-max-ordering. First, we consider digraphs that admit a min-ordering. Digraphs that admit a min-ordering have been studied under the name of bi-arc digraphs [36] and signed interval digraphs [29]. Deciding if digraph $H$ has a min-ordering and finding a min-ordering of $H$ is in P [36]. We provide a constant factor approximation algorithm for $\operatorname{MinHOM}(H)$ where $H$ admits a min-ordering.

- Theorem 5 (Digraphs with a min-ordering). If digraph $H$ admits a min-ordering, then $\operatorname{MinHOM}(H)$ has a constant factor approximation algorithm.

Sections 4,5 are dedicated to the proof of Theorem 5. In section 6, we turn our attention to digraphs with $k$-min-orderings, for integer $k>1$. They are also called digraphs with extended $X$-underbar $[3,26,46]$. It was shown in [26] that if $H$ has the $X$-underbar property,

[^0]then the $\operatorname{HOM}(H)$ problem is polynomial time solvable. In Lemma 21, we show that if $H$ admits a $k$-min-ordering, then $H$ is a DAT-free digraph, and provide a simple proof that $\operatorname{LHOM}(H)$ is polynomial time solvable. Finally, we have the following theorem.

- Theorem 6 (Digraphs with a $k$-min-ordering). If digraph $H$ admits a $k$-min-ordering for some integer $k>1$, then $\operatorname{MinHOM}(H)$ has a constant factor approximation algorithm.

Considering graphs, Feder et al., [18] proved that $\operatorname{LHOM}(H)$ is polynomial time solvable if $H$ is a bi-arc graph, and is NP-complete otherwise. In the same paper, they showed graph $H$ is a bi-arc graph iff it admits a conservative majority polymorphism. In Section 7, we show that the same dichotomy classification holds in terms of approximation.

- Theorem 7 (Dichotomy for graphs). There exists a constant factor approximation algorithm for $\operatorname{MinHOM}(H)$ if $H$ is a bi-arc graph, otherwise $\operatorname{MinHOM}(H)$ is inapproximable.

In section 8, we give a concrete plan of how to solve the general case. By combining the approach for obtaining the dichotomy in the graph case, together with the idea of getting an approximation algorithm for digraphs admitting a min-ordering, we might be able to achieve a constant factor approximation algorithm for $\operatorname{MinHOM}(H)$ when $H$ is DAT-free.

Our constant factors depend on the size of $H$. However, the implementation of the LP and the ILP would yield a small integrality gap (details in the full version [49]). This indicates perhaps a better analysis of the performance of our algorithm is possible.

- Open Problem 8. For which digraphs $\operatorname{MinHOM}(H)$ is approximable within a constant factor independent of size of $H$ ?


## 2 Preliminaries and Definitions

Complexity and approximation of the minimum cost homomorphism problems, and in general the constraint satisfaction problems, are often studied under the existence of polymorphisms [6]. A polymorphism of $H$ of arity $k$ is a mapping $f$ from the set of $k$-tuples over $V(H)$ to $V(H)$ such that if $x_{i} y_{i} \in A(H)$ for $i=1,2, \ldots, k$, then $f\left(x_{1}, x_{2}, \ldots, x_{k}\right) f\left(y_{1}, y_{2}, \ldots, y_{k}\right) \in$ $A(H)$. If $f$ is a polymorphism of $H$ we also say that $H$ admits $f$. A polymorphism $f$ is idempotent if it satisfies $f(x, x, \ldots, x)=x$ for all $x \in V(H)$, and is conservative if $f\left(x_{1}, x_{2}, \ldots, x_{k}\right) \in\left\{x_{1}, x_{2}, \ldots, x_{k}\right\}$. A conservative semi-lattice polymorphism is a conservative binary polymorphism that is associative, idempotent, commutative. A conservative majority polymorphism $\mu$ of $H$ is a conservative ternary polymorphism such that $\mu(x, x, y)=\mu(x, y, x)=\mu(y, x, x)=x$ for all $x, y \in V(H)$.

A conservative semi-lattice polymorphism of $H$ naturally defines a binary relation $x \leq y$ on the vertices of $H$ by $x \leq y$ iff $f(x, y)=x$; by associative, the relation $\leq$ is a linear order on $V(H)$, which we call a min-ordering of $H$.

- Definition 9. The ordering $v_{1}<v_{2}<\cdots<v_{n}$ of $V(H)$ is a
- min-ordering iff $u v \in A(H), u^{\prime} v^{\prime} \in A(H)$ and $u<u^{\prime}, v^{\prime}<v$ implies that $u v^{\prime} \in A(H)$;
- max-ordering iff $u v \in A(H), u^{\prime} v^{\prime} \in A(H)$ and $u<u^{\prime}, v^{\prime}<v$ implies that $u^{\prime} v \in A(H)$;
- min-max-ordering iff $u v \in A(H), u^{\prime} v^{\prime} \in A(H)$ and $u<u^{\prime}, v^{\prime}<v$ implies that $u v^{\prime}, u^{\prime} v \in$ $A(H)$.
For bipartite graph $H=(B, W)$ let $\vec{H}$ be the digraph obtained by orienting all the edges of $H$ from $B$ to $W$. If $\vec{H}$ admits a min-ordering then we say $H$ admits a min-ordering. It is worth mentioning that, a bipartite graph $H$ admits a conservative majority, iff it admits a min-ordering [30]. Moreover, the complement of $H$ is a circular arc graphs with clique cover two [18].
- Definition 10. Let $H=(V, E)$ be a digraph that admits a homomorphism $f: V(H) \rightarrow \overrightarrow{C_{k}}$ (here $\overrightarrow{C_{k}}$ is the induced directed cycle on $\{0,1,2, \ldots, k-1\}$ ( i.e., arc set $\{(01,12,23, \ldots,(k-$ $2)(k-1),(k-1) 0\})$. Let $V_{i}=f^{-1}(i), 0 \leq i \leq k-1$.
- Ak-min-ordering of $H$ is a linear ordering $<$ of the vertices of $H$, so that $<$ is a minordering on the subgraph induced by any two circularly consecutive $V_{i}, V_{i+1}$ (subscript addition modulo $k$ ).
- A $k$-min-max-ordering of $H$ is a linear ordering $<$ of the vertices of $H$, so that $<$ is a min-max-ordering on the subgraph induced by any two circularly consecutive $V_{i}, V_{i+1}$ (subscript addition modulo $k$ ).


## 3 LP for Digraphs with a min-max-ordering

Before presenting the LP, we give a procedure to modify lists associated to the vertices of $D$. To each vertex $x \in D$, associate a list $L(x)$ that initially contains $V(H)$. Think of $L(x)$ as the set of possible images for $x$ in a homomorphism from $D$ to $H$. Apply the arc consistency procedure as follows. Take an arbitrary arc $x y \in A(D)(y x \in A(D))$ and let $a \in L(x)$. If there is no out-neighbor (in-neighbor) of $a$ in $L(y)$ then remove $a$ from $L(x)$. Repeat this until a list becomes empty or no more changes can be made. Note that if we end up with an empty list after arc consistency then there is no homomorphism of $D$ to $H$.

Let $a_{1}, a_{2}, a_{3}, \ldots, a_{p}$ be a min-max-ordering $<$ of the target digraph $H$. Define $\ell^{+}(i)$ to be the smallest subscript $j$ such that $a_{j}$ is an out-neighbor of $a_{i}$ (and $\ell^{-}(i)$ to be the smallest subscript $j$ such that $a_{j}$ is an in-neighbor of $a_{i}$ ).

Consider the following linear program. For every vertex $v$ of $D$ and every vertex $a_{i}$ of $H$ define variable $v_{i}$. Moreover, define variable $v_{p+1}$ for every $v \in D$ whose value is set to zero.

$$
\begin{array}{rll}
\min & \sum_{v, i} c\left(v, a_{i}\right)\left(v_{i}-v_{i+1}\right) & \\
\text { subject to: } & v_{i} \geq 0 & \\
& v_{1}=1 & \\
& v_{p+1}=0 & \\
& v_{i+1} \leq v_{i} & \text { if } a_{i} \notin L(v) \\
& v_{i+1}=v_{i} & \forall u v \in A(D) \\
& u_{i} \leq v_{l+(i)} & \forall u v \in A(D)
\end{array}
$$

Let $\mathcal{S}$ denote the set of constraints of the above LP, then:

- Theorem 11. If digraph $H$ admits a min-max-ordering, then there is a one-to-one correspondence between homomorphisms of $D$ to $H$ and integer solutions of $\mathcal{S}$.

Proof. For homomorphism $f: D \rightarrow H$, if $f(v)=a_{t}$ we set $v_{i}=1$ for all $i \leq t$, otherwise we set $v_{i}=0$. We set $v_{1}=1$ and $v_{p+1}=0$ for all $v \in V(D)$. Now all the variables are nonnegative and we have $v_{i+1} \leq v_{i}$. Note that if $a_{i} \notin L(v)$ then $f(v) \neq a_{i}$ and we have $v_{i}-v_{i+1}=0$. It remains to show that $u_{i} \leq v_{l^{+}(i)}$ for every $u v$ arc in $D$. Suppose for contradiction that $u_{i}=1$ and $v_{l^{+}(i)}=0$ and let $f(u)=a_{r}$ and $f(v)=a_{s}$. This implies that $u_{r}=1$, whence $i \leq r$; and $v_{s}=1$, whence $s<l^{+}(i)$. Since $a_{i} a_{l^{+}(i)}$ and $a_{r} a_{s}$ both are arcs of $H$ with $i \leq r$ and $s<l^{+}(i)$, the fact that $H$ has a min-ordering implies that $a_{i} a_{s}$ must also be an arc of $H$, contradicting the definition of $l^{+}(i)$. The proof for $v_{i} \leq u_{l^{-}(i)}$ is analogous.

Conversely, if there is an integer solution for $\mathcal{S}$, we define a homomorphism $f$ as follows: we let $f(v)=a_{i}$ when $i$ is the largest subscript with $v_{i}=1$. We prove that this is indeed a homomorphism by showing that every $\operatorname{arc}$ of $D$ is mapped to an arc of $H$. Let $u v$ be an arc
of $D$ and assume $f(u)=a_{r}, f(v)=a_{s}$. We show that $a_{r} a_{s}$ is an arc in $H$. Observe that $1=u_{r} \leq v_{l^{+}(r)} \leq 1$ and $1=v_{s} \leq u_{l^{-}(s)} \leq 1$, therefore we must have $v_{l^{+}(r)}=u_{l^{-}(s)}=1$. Since $r$ and $s$ are the largest subscripts such that $u_{r}=v_{s}=1$ then $l^{+}(r) \leq s$ and $l^{-}(s) \leq r$. Since $a_{r} a_{l^{+}(r)}$ and $a_{l^{-}(s)} a_{s}$ are arcs of $H$, we must have the arc $a_{r} a_{s}$, as $H$ admits a maxordering. Furthermore, $f(v)=a_{i}$ iff $v_{i}=1$ and $v_{i+1}=0$, so, $c\left(v, a_{i}\right)$ contributes to the sum iff $f(v)=a_{i}$.

We have translated the minimum cost homomorphism problem to a linear program. In fact, this linear program corresponds to a minimum cut problem in an auxiliary network, and can be solved by network flow algorithms [23, 48]. In [30], a similar result to Theorem 11 was proved for the $\operatorname{MinHOM}(\mathrm{H})$ problem on undirected graphs when target graph $H$ is bipartite and admits a min-max-ordering. We shall enhance the above system $\mathcal{S}$ to obtain an approximation algorithm for the case where $H$ is only assumed to admit a min-ordering.

## 4 LP for Digraphs with a min-ordering

In the rest of the paper assume lists are not empty. Moreover, non-empty lists guarantee a homomorphism when $H$ admits a min-ordering.

- Lemma 12. [32] Let $H$ be a digraph that admits a min-ordering. If all the lists are non-empty after arc consistency, then there exists a homomorphism from $D$ to $H$.

Suppose $a_{1}, a_{2}, \ldots, a_{p}$ is a min-ordering of $H$. Let $E^{\prime}$ denote the set of all the pairs $\left(a_{i}, a_{j}\right)$ such that $a_{i} a_{j}$ is not an arc of $H$, but there is an arc $a_{i} a_{j^{\prime}}$ of $H$ with $j^{\prime}<j$ and an $\operatorname{arc} a_{i^{\prime}} a_{j}$ of $H$ with $i^{\prime}<i$. Let $E=A(H)$ and define $H^{\prime}$ to be the digraph with vertex set $V(H)$ and arc set $E \cup E^{\prime}$. Note that $E$ and $E^{\prime}$ are disjoint sets.

- Observation 13. The ordering $a_{1}, a_{2}, \cdots, a_{p}$ is a min-max-ordering of $H^{\prime}$.
- Observation 14. Let $e=a_{i} a_{j} \in E^{\prime}$. Then $a_{i}$ does not have any out-neighbor in $H$ after $a_{j}$, or $a_{j}$ does not have any in-neighbor in $H$ after $a_{i}$.

Observation 14 easily follows from the fact that $H$ has a min-ordering. Since $H^{\prime}$ has a min-max-ordering, we can form system of linear inequalities $\mathcal{S}$, for $H^{\prime}$ as described in Section 3. Homomorphisms of $D$ to $H^{\prime}$ are in a one-to-one correspondence with integer solutions of $\mathcal{S}$, by Theorem 11. However, we are interested in homomorphisms of $D$ to $H$, not $H^{\prime}$. Therefore, we shall add further inequalities to $\mathcal{S}$ to ensure that we only admit homomorphisms from $D$ to $H$, i.e., avoid mapping arcs of $D$ to the arcs in $E^{\prime}$.

For every arc $e=a_{i} a_{j} \in E^{\prime}$ and every arc $u v \in A(D)$, by Observation 14, two of the following set of inequalities will be added to $\mathcal{S}$ (i.e. either (C8), (C11) or (C9), (C10) or (C9), (C11)).

$$
\begin{array}{ll}
v_{j} \leq u_{s}+\sum_{t<i, a_{t} a_{j} \in E, a_{t} \in L(u)}\left(u_{t}-u_{t+1}\right) & \text { if } a_{s} \in L(u) \text { is the first in-neighbour of } a_{j} \text { after } a_{i} \\
v_{j} \leq v_{j+1}+\sum_{t<i, a_{t} a_{j} \in E, a_{t} \in L(u)}\left(u_{t}-u_{t+1}\right) & \text { if } a_{j} \text { has no in-neighbour after } a_{i} \\
u_{i} \leq v_{s}+\sum_{t<j, a_{i} a_{t} \in E, a_{t} \in L(v)}\left(v_{t}-v_{t+1}\right) & \text { if } a_{s} \in L(v) \text { is the first out-neighbour of } a_{i} \text { after } a_{j} \\
u_{i} \leq u_{i+1}+\sum_{t<j, a_{i} a_{t} \in E, a_{t} \in L(v)}^{(C 10)} \tag{C11}
\end{array}
$$

Additionally, for every pair $(x, y) \in V(D) \times V(D)$ consider a list $L(x, y)$ of possible pairs $(a, b), a \in L(x)$ and $b \in L(y)$. Perform pair consistency procedure as follows. Consider three vertices $x, y, z \in V(D)$. For $(a, b) \in L(x, y)$ if there is no $c \in L(z)$ such that $(a, c) \in L(x, z)$
and $(c, b) \in L(z, y)$ then remove $(a, b)$ from $L(x, y)$. Repeat this until a pair list becomes empty or no more changes can be made. Here, we assume that after pair consistency procedure no pair list is empty, as otherwise there is no homomorphism of $D$ to $H$. Therefore, by pair consistency, add the following constraints for every $u, v$ in $V(D)$ and $a_{i} \in L(u)$ :

$$
\begin{equation*}
u_{i}-u_{i+1} \leq \sum_{\substack{j \\\left(a_{i}, a_{j}\right) \in L(u, v)}}\left(v_{j}-v_{j+1}\right) \tag{C12}
\end{equation*}
$$

- Lemma 15. If H admits a min-ordering, then there is a one-to-one correspondence between homomorphisms of $D$ to $H$ and integer solutions of the extended system $\mathcal{S}$.


## 5 Approximation for Digraphs with a min-ordering

In what follows, we describe an overview of our approximation algorithm for $\operatorname{MinHOM}(H)$ where the fixed digraph $H$ has a min-ordering. We encourage the reader to see Algorithm 1 while reading this section. An overview of the proofs of the correctness and approximation bound are postponed for the later subsections (further details in the full version [49]).

Let $D$ be the input digraph together with a cost function $c$. Let $a_{1}, \ldots, a_{p}$ be a min ordering of the vertices of $H$. The algorithm, first constructs digraph $H^{\prime}$ from $H$ as in Section 4. By Observation $13, a_{1}, \ldots, a_{p}$ is a min-max-ordering for $H^{\prime}$. By Lemma 15, the integral solutions of the extended LP are in one-to-one correspondence to homomorphisms from $D$ to $H$. At this point, our algorithm will minimize the cost function over extended $\mathcal{S}$ in polynomial time using a linear programming algorithm. This will generally result in a fractional solution (Even though the original system $\mathcal{S}$ is known to be totally unimodular [23, 48] and hence have integral optima, we have added inequalities, and hence lost this advantage). We will obtain an integer solution by a randomized procedure called rounding. Choose, uniformly at random, a random variable $X \in[0,1]$, and define the rounded values $u_{i}^{\prime}=1$ when $u_{i} \geq X\left(u_{i}\right.$ is the returned value by the LP), and $u_{i}^{\prime}=0$ otherwise. It is easy to check that the rounded values satisfy the original inequalities, i.e., correspond to a homomorphism $f$ of $D$ to $H^{\prime}$.

Now the algorithm will once more modify the solution $f$ to become a homomorphism from $D$ to $H$, i.e., to avoid mapping the arcs of $D$ to the arcs in $E^{\prime}$. This will be accomplished by another randomized procedure, which we call Shift. We choose, uniformly at random, another random variable $Y \in[0,1]$, which will guide the shifting. Let $F$ denote the set of all arcs in $E^{\prime}$ to which some arcs of $D$ are mapped by $f$. If $F$ is empty, we need no shifting. Otherwise, let $a_{i} a_{j}$ be an arc of $F$. Since $F \subseteq E^{\prime}$, Observation 14 implies that either $a_{j}$ has no in-neighbor after $a_{i}$ or $a_{i}$ has no out-neighbor after $a_{j}$. Suppose the first case happens (the shifting process is similar in the other case).

Consider a vertex $v$ in $D$ such that $f(v)=a_{j}$ (i.e. $v_{j}^{\prime}=1$ and $v_{j+1}^{\prime}=0$ ) and $v$ has an in-neighbor $u$ in $D$ with $f(u)=a_{i}$ (i.e. $u_{i}^{\prime}=1$ and $u_{i+1}^{\prime}=0$ ). For such a vertex $v$, let $S_{v}=\left\{a_{t_{1}}, a_{t_{2}}, \ldots, a_{t_{k}}\right\}$ be the set of all vertices $a_{t}$ with $t<j$ such that $a_{i} a_{t} \in E$ and $a_{t} \in L(v)$. Suppose $S_{v}$ consists of $a_{t}$ with subscripts $t$ ordered as $t_{1}<t_{2}<\cdots<t_{k}$.

- Lemma 16. During procedure Shift, the set of indices $t_{1}<\cdots<t_{k}$ considered in Line 6 of the Algorithm 1 is non-empty.

By Lemma 16, $S_{v}$ is not empty. The algorithm now selects one vertex from this set as follows. Let $P_{v, t}=\frac{v_{t}-v_{t+1}}{P_{v}}$, where $P_{v}=\sum_{\substack{t<j \\ a_{i} a_{t} \in E, a_{t} \in L(v)}}\left(v_{t}-v_{t+1}\right)$.

Note that $P_{v}>0$ because of constraints (C9) and (C10). Then $a_{t_{q}}$ is selected if $\sum_{p=1}^{q} P_{v, t_{p}}<Y \leq \sum_{p=1}^{q+1} P_{v, t_{p}}$. Thus a concrete $a_{t}$ is selected with probability $P_{v, t}$, which is
proportional to the difference of the fractional values $v_{t}-v_{t+1}$. When the selected vertex is $a_{t}$, we shift the image of the vertex $v$ from $a_{j}$ to $a_{t}$, and set $v_{r}^{\prime}=1$ if $r \leq t$, else set $v_{r}^{\prime}=0$. Note that $a_{t}$ is before $a_{j}$ in the min-ordering. Now we might need to shift images of the neighbors of $v$. In this case, repeat the shifting procedure for neighbors of $v$. This processes continues in a Breadth-first search (BFS) like manner, until no more shift is required (Figure 1 gives an illustration). Note that a vertex might be visited multiple times in procedure shift while a pair $\left(v, a_{i}\right) \in V(D) \times V(H)$ is considered at most one time.


Figure 1 Two examples. In the right example, the target digraph is $H_{1}$ and input is $D_{1}$. Digraphs $D_{1}$ and $H_{1}$ both can be view as bipartite graphs and $1,2,3,4,5,6,7$ is a min-ordering of $H$. When $x, y$ are mapped to 3 and $w$ is mapped to 6 then the algorithm should shift the image of $w$ from 6 to 5 and since 35 is an arc there is no need to shift the image of $y$. In the left example, the target digraph is $H$ and the input is $D .1,2,3,4,5,6,7,8$ is a min-ordering of $H$ and 24 is a missing arc. Suppose $x$ is mapped to $2, y$ to $4, w$ to $7, z$ to $8, u$ to 5 and $v$ to 2 . Then we should shift the image of $y$ to 3 and then $w$ to 6 and $z$ to 6 and then $u$ to 3 and $v$ to one of the 1,2 .

We remark that the images of vertices in $D$ are always shifted towards smaller elements in their lists. Lemma 17 shows that this shifting modifies the homomorphism $f$, and hence, the corresponding values of the variables. Namely, $v_{t+1}^{\prime}, \ldots, v_{j}^{\prime}$ are reset to 0 , keeping all other values the same. Note that these modified values still satisfy the original set of constraints $\mathcal{S}$, i.e., the modified mapping is still a homomorphism.

- Lemma 17. Procedure SHIFT, in polynomial time, returns a homomorphism of $D$ to $H^{\prime}$.

We repeat the same process for the next $v$ with these properties, until no edge of $D$ is mapped to an edge in $E^{\prime}$. Each iteration involves at most $|V(H)| \cdot|V(D)|$ shifts. After at most $\left|E^{\prime}\right|$ iterations, no edge of $D$ is mapped to an edge in $F$ and we no longer need to shift. Next theorem follows from Lemma 16 and 17.

- Theorem 18. Our algorithm, in polynomial time, returns a homomorphism of $D$ to $H$.


### 5.1 Analyzing the Approximation Ratio

We now claim that the cost of this homomorphism is at most $|V(H)|^{2}$ times the minimum cost of a homomorphism. Let $w$ denote the value of the objective function with the fractional optimum $u_{i}, v_{j}$, and $w^{\prime}$ denote the value of the objective function with the final values $u_{i}^{\prime}, v_{j}^{\prime}$, after the rounding and all the shifting. Also, let $w^{*}$ be the minimum cost of a homomorphism of $D$ to $H$. Obviously, $w \leq w^{*} \leq w^{\prime}$.

We now show that the expected value of $w^{\prime}$ is at most a constant times $w$. Let us focus on the contribution of one summand, say $v_{t}^{\prime}-v_{t+1}^{\prime}$, to the calculation of the cost. In any integer solution, $v_{t}^{\prime}-v_{t+1}^{\prime}$ is either 0 or 1 . The probability that $v_{t}^{\prime}-v_{t+1}^{\prime}$ contributes to $w^{\prime}$ is the probability of the event that $v_{t}^{\prime}=1$ and $v_{t+1}^{\prime}=0$. This can happen in the following:

```
Algorithm 1 Approximation \(\operatorname{MinHOM}(H)\).
    procedure Approx-MinHOM \((D, H)\)
        Construct \(H^{\prime}\) from \(H\) (as in Section 3)
        Let \(u_{i} \mathrm{~s}\) be the (fractional) values returned by the extended LP
        Choose a random variable \(X \in[0,1]\), and \(\forall u_{i}\) s : if \(X \leq u_{i}\) let \(u_{i}^{\prime}=1\), else let \(u_{i}^{\prime}=0\)
        Let \(f(u)=a_{i}\) where \(i\) is the largest subscript with \(u_{i}^{\prime}=1 \triangleright f\) is a homomorphism
    from \(D\) to \(H^{\prime}\)
        Choose a random variable \(Y \in[0,1]\)
        while \(\exists u v \in A(D)\) such that \(f(u) f(v) \in A\left(H^{\prime}\right) \backslash A(H)\) do
            if \(f(v)\) does not have an in-neighbor after \(f(u)\) then \(\operatorname{Shift}(f, v)\)
            else if \(f(u)\) does not have an out-neighbor after \(f(v)\) then \(\operatorname{Shift}(f, u)\)
        return \(f \quad \triangleright f\) is a homomorphism from \(D\) to \(H\)
```

```
Algorithm 2 The Shifting Procedure.
    procedure \(\operatorname{Shift}(f, x)\)
        Let Q be a Queue, \(Q\). .enqueue \((x)\)
        while Q is not empty do
            \(v \leftarrow Q\).dequeue ()
            for \(u v \in A(D)\) with \(f(u) f(v) \notin A(H)\) or \(v u \in A(D)\) with \(f(v) f(u) \notin A(H)\) do
                        \(\triangleright\) Here we assume the first condition holds, the other case is similar
                        \(\triangleright\) Further, we assume \(f(v)\) does not have an in-neighbor after \(f(u)\)
                Let \(t_{1}<\cdots<t_{k}\) be indices so that \(a_{t_{j}}<f(v), a_{t_{j}} \in L(v), f(u) a_{t_{j}} \in A(H)\)
                    Let \(P_{v} \leftarrow \sum_{j=1}^{j=k}\left(v_{t_{j}}-v_{t_{j}+1}\right)\) and \(P_{v, t} \leftarrow\left(v_{t}-v_{t+1}\right) / P_{v}\)
                    if \(\sum_{p=1}^{q} P_{v, t_{p}}<Y \leq \sum_{p=1}^{q+1} P_{v, t_{p}}\) then
                    \(f(v) \leftarrow a_{t_{q}}\), set \(v_{i}^{\prime}=1\) for \(1 \leq i \leq t_{q}\), and set \(v_{i}^{\prime}=0\) for \(t_{p}<i\)
                    for \(v z \in A(D)(z v \in A(D))\) with \(f(v) f(z) \notin A(H)(f(z) f(v) \notin A(H))\) do
                        Q.enqueue \((z)\)
        return \(f \quad \triangleright f\) is a homomorphism from \(D\) to \(H^{\prime}\)
```

1. $v$ is mapped to $a_{t}$ by rounding, and is not shifted away. In other words, we have $v_{t}^{\prime}=1$ and $v_{t+1}^{\prime}=0$ after rounding, and these values don't change by procedure Shift.
2. $v$ is first mapped to some $a_{j}, j>t$, by rounding, and then re-mapped to $a_{t}$ by procedure Shift.

- Lemma 19. The expected contribution of one summand, say $v_{t}^{\prime}-v_{t+1}^{\prime}$, to the expected cost of $w^{\prime}$ is at most $|V(H)|^{2} c\left(v, a_{t}\right)\left(v_{t}-v_{t+1}\right)$.
- Theorem 20. Algorithm 1 returns a homomorphism with expected cost $|V(H)|^{2} \cdot O P T$. The algorithm can be de-randomized to obtain a deterministic $|V(H)|^{2}$-approximation algorithm.


## 6 Approximation for Digraphs with a k-min-ordering

Digraphs admitting $k$-min-ordering $(k>1)$ do not admit a min-ordering or a conservative majority polymorphism. However, this does not rule out the possibility of a constant factor approximation algorithm. We show that they are in fact DAT-free digraphs, and they admit a nice geometric representation (see the full version [49]).

- Lemma 21. Let $H$ be a digraph that admits a $k$-min-ordering. Then $H$ is DAT-free, and $\operatorname{LHOM}(H)$ is polynomial time solvable.

Let $H$ be a digraph with a $k$-min-ordering $(k>1)$ and partition $V_{0}, V_{1}, \ldots, V_{k-1}$ of its vertices, and let $<$ be a $k$-min-ordering of $V(H)$. It is easy to argue that the input digraph $D$ must be homomorphic to $\overrightarrow{C_{k}}$ otherwise there is no homomorphism from $D$ to $H$. Therefore, we assume (for some $0 \leq \ell \leq k-1$ ), $L(u) \subseteq V_{i}$ for every $u \in U_{i+\ell}, 0 \leq i \leq k-1$. Now the LP is designed according to the lists $L$. Since $<$ is a min-ordering of $V_{i} \cup V_{i+1}$, the constraints are very similar to the ones in Section 3. The conclusion of this section is the following:

- Theorem 22. There is a (deterministic) $|V(H)|^{2}$-approximation algorithm for $\operatorname{MinHOM}(H)$ when the target digraph $H$ admits a $k$-min-ordering, $k>1$.


## 7 A Dichotomy for Graphs

Feder and Vardi [20] proved that if a graph $H$ admits a conservative majority polymorphism, then $\operatorname{LHOM}(\mathrm{H})$ is polynomial time solvable. Later, Feder et al., [18] showed that $\mathrm{LHOM}(\mathrm{H})$ is polynomial time solvable iff $H$ is a bi-arc graph. Hence, by Observation 4, the problem is inapproximable beyond bi-arc graphs. A bi-arc graph is represented by a pair of families of arcs on a circle with specific conditions (exact definition is given in the full version [49]). Note that in a bi-arc graph a vertex may have a self-loop.

- Theorem 23 ([7, 18]). A graph admits a conservative majority polymorphism iff it is a bi-arc graph.
- Definition $24\left(G^{*}\right)$. Let $G=(V, E)$ be a graph. Let $G^{*}$ be a bipartite graph with partite sets $V, V^{\prime}$ where $V^{\prime}$ is a copy of $V$. Two vertices $u \in V$, and $v^{\prime} \in V^{\prime}$ of $G^{*}$ are adjacent in $G^{*}$ iff $u v$ is an edge of $G$.

A circular arc graph is a graph that is the intersection graph of a family of arcs on a circle. A bipartite graph whose complement is a circular arc graph, is called a co-circular arc graph. Note that co-circular arc graphs are irreflexive, meaning no vertex has a loop.

- Lemma 25. Let $H^{*}$ be the bipartite graph constructed from a bi-arc graph $H$. Then $H^{*}$ is a co-circular arc graph and $H^{*}$ admits a min-ordering.

Let $H$ be a bi-arc graph, with vertex set $I$, and let $H^{*}=\left(I, I^{\prime}\right)$ be the bipartite graph constructed from $H$. Let $a_{1}, a_{2}, \ldots, a_{p}$ be an ordering of the vertices in $I$ and $b_{1}, b_{2}, \ldots, b_{p}$ be an ordering of the vertices of $I^{\prime}$. Note that each $a_{i}$ has a copy $b_{\pi(i)}$ in $\left\{b_{1}, b_{2}, \ldots, b_{n}\right\}$ where $\pi$ is a permutation on $\{1,2,3, \ldots, p\}$. By Lemma 25 , let us assume $a_{1}, a_{2}, \ldots, a_{p}, b_{1}, b_{2}, \ldots, b_{p}$ is a min-ordering for $H^{*}$.

Let $G$ be the input graph with vertex set $V$ and a cost function $c$. Construct $G^{*}$ from $G$ with vertex set $V \cup V^{\prime}$ as in Definition 24. Now construct an instance of the $\operatorname{MinHOM}\left(H^{*}\right)$ for the input graph $G^{*}$ and set $c\left(v^{\prime}, b_{\pi(i)}\right)=c\left(v, a_{i}\right)$ for $v \in V$ and $v^{\prime} \in V^{\prime}$. Further, make $H^{*}$ a digraph by orienting all its edges from $I$ to $I^{\prime}$, and similarly make $G^{*}$ a digraph by
orienting all its edges from $V$ to $V^{\prime}$. Note that, by construction of $H^{*}$ and $G^{*}$, there exists a homomorphism $f: G \rightarrow H$ with cost $\mathfrak{C}$ iff there exists homomorphism $f^{*}: G^{*} \rightarrow H^{*}$ with cost $2 \mathfrak{C}$ such that, if $f^{*}(v)=a_{i}$ then $f^{*}\left(v^{\prime}\right)=b_{j}$ with $j=\pi(i)$.

We first perform the arc consistency and pair consistency procedures for the vertices in $G^{*}$. Note that if $L(u)$ contains element $a_{i}$ then $L\left(u^{\prime}\right)$ contains $b_{\pi(i)}$ and when $L\left(u^{\prime}\right)$ contains some $b_{j}$ then $L(u)$ contains $a_{\pi^{-1}(j)}$. Next, we define the system of linear equations $\widehat{S^{*}}$ with the same construction as in Sections 3, 4. Further, we add the following constraint to $\widehat{S^{*}}$. The full set of constraints in $\widehat{S^{*}}$ is presented in the extended version.

$$
u_{i}-u_{i+1}=u_{\pi(i)}^{\prime}-u_{\pi(i)+1}^{\prime} \quad \forall u, u^{\prime} \in G^{*}, \forall a_{i}, b_{\pi(i)} \in H^{*}
$$

- Lemma 26. If $H$ is a bi-arc graph, then there is a one-to-one correspondence between homomorphisms from $G$ to $H$ and integer solutions of $\widehat{S^{*}}$.

Once again we round an optimal fractional solution of $\widehat{S^{*}}$, using random variable $X \in[0,1]$. Let $\mathcal{F}$ be a mapping form $V\left(G^{*}\right)$ to $V\left(H^{*}\right)$ obtained after rounding using $X$. We give an algorithm that modifies $\mathcal{F}$ and achieves a homomorphism $f: G \rightarrow H$ (i.e. an integral solution that satisfies $\widehat{S^{*}}$ ). The algorithm deploys a shifting procedure that first uses a random variable $Y$ to shift the images of some of the vertices of $V\left(G^{*}\right)$ to obtain a homomorphism $f$ from $G^{*}$ to $H^{*}$. Second, it applies a breadth-first search function to make $f$ consistent on $V$ and $V^{\prime}$; meaning that $f(u)=a_{i}, u \in V$ iff $f\left(u^{\prime}\right)=b_{\pi(i)}, u^{\prime} \in V^{\prime}$. The proof of the following theorems and de-randomization of the algorithm appear in the full version [49].

- Theorem 27. There exists a randomized algorithm that modifies $\mathcal{F}$ and obtain a homomorphism $f: G \rightarrow H$. Moreover, the expected cost of the homomorphism returned by this algorithm is at most $2|V(H)| \cdot O P T$.
- Theorem 28. If $H$ admits a conservative majority polymorphism, then $\operatorname{MinHOM}(H)$ has a (deterministic) $2|V(H)|$-approximation algorithm, otherwise it is inapproximable.


## 8 Beyond majority and min-ordering (DAT-free cases)

This section offers a view of moving forward to get a dichotomy classification for constant approximability of $\operatorname{MinHOM}(H)$. We believe the class of DAT-free digraphs is the right boundary between the approximable cases, and the ones that do not admit any approximation.

- Conjecture 29. $\operatorname{MinHOM}(H)$ admits a constant approximation polynomial time algorithm when $H$ is a DAT-free digraph, otherwise, $\operatorname{MinHOM}(H)$ is not approximable.

For digraph $D=(V, A)$, let $D^{*}$ be a bipartite digraph with partite sets $V, V^{\prime}$ where $V^{\prime}$ is a copy of $V$. There is an arc in $D^{*}$ from $u \in V$ to $v^{\prime} \in V^{\prime}$ iff $u v$ is an arc of $D$. In what follows, we give a road map for solving the conjecture. Let us start off by making a connection between homomorphisms from $D$ to a DAT-free target digraph $H$, and the homomorphisms from $D^{*}$ to $H^{*}$.

- Proposition 30. Let $D, H$ be two digraphs and let $D, H, L$ (here $L$ are the lists) be an instance of the $\operatorname{LHOM}(H)$. Suppose $H$ is DAT-free. Then $H^{*}$ admits a min-ordering, and $\operatorname{LHOM}\left(H^{*}\right)$ is polynomial time solvable for instance $D^{*}, H^{*}$ where $L^{*}\left(v^{\prime}\right)=\left\{a^{\prime} \mid a \in L(v)\right\}$ and $L^{*}(v)=L(v)$ for every $v \in V(D)$.

Similar to Lemmas 15 and 26, we can obtain set of constraints $\widehat{S^{*}}$ such that there is a one-to-one correspondence between homomorphisms from $D$ to $H$ and integer solutions of $\widehat{S^{*}}$ (details in the full version [49]). Our primary challenge would be finding a rounding
procedure to obtain a homomorphism from $D$ to $H$. We believe there is a need to deploy the shift procedure in min-ordering case (Section 3), as well as, the shifting procedure in the majority case (Section 7). This essentially means obtaining a new way of solving a list homomorphism from $D$ to $H$ when $H$ is a DAT-free, using the bi-partition method. The calculation should work out; yielding a constant bound between the fractional value of the LP and the integral value obtained by rounding. Notice that in the majority case the symmetry of the arcs is heavily used in our argument, whereas in the digraph case we no longer have this property in hand.

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[^0]:    1 The definition of DAT is rather technical and it is not necessary to fully understand it in this paper.

